# Diffusive stability of Turing patterns via normal forms 

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#### Abstract

We investigate dynamics near Turing patterns in reaction-diffusion systems posed on the real line. Linear analysis predicts diffusive decay of small perturbations. We construct a "normal form" coordinate system near such Turing patterns which exhibits an approximate discrete conservation law. The key ingredients to the normal form is a conjugation of the reaction-diffusion system on the real line to a lattice dynamical system. At each lattice site, we decompose perturbations into neutral phase shifts and normal decaying components. As an application of our normal form construction, we prove nonlinear stability of Turing patterns with respect to perturbations that are small in $L^{1} \cap L^{\infty}$, with sharp rates, recovering and slightly improving on results in [21, 10].


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## 1 Introduction

Turing predicted that the simple interplay of reaction and diffusion can lead to stable, spatially periodic patterns [22]. His ideas proved quite influential in the general area of pattern formation, where one seeks to understand the formation and dynamics of self-organized spatio-temporal structures. One can easily envision simple reaction-diffusion systems with two species that exhibit diffusion-driven instabilities of spatially homogeneous equilibria. Typical examples are activator-inhibitor systems such as the Gray-Scott or the Gierer-Meinhard equation; see for instance [15, 16]. Perturbations of the homogeneous unstable equilibrium grow exponentially at an initial stage, with fastest growth for distinct spatial wavenumbers. This wavenumber is roughly independent of boundary conditions in large enough domains. As a final result, one often finds a spatially periodic pattern, up to narrow, exponentially localized boundary layers. In order to understand such nonlinear spatially periodic patterns and the process of wavenumber selection, one is therefore naturally led to considering reaction-diffusion systems on idealized unbouned domains.

To fix ideas, consider

$$
\mathbf{u}_{t}=D \triangle \mathbf{u}+\mathbf{f}(\mathbf{u})
$$

for $\mathbf{u}(t, x) \in \mathbb{R}^{n}$, with $x \in \mathbb{R}^{N}$, with smooth reaction-kinetics $\mathbf{f}$ and positive diagonal diffusion matrix $D=\operatorname{diag}\left(d_{j}\right)>0$. Here, and in the following, the term "smooth" refers to functions with sufficiently many derivatives. In many circumstances, one can show that there exist families of spatially periodic striped solutions,

$$
\mathbf{u}(t, x)=\mathbf{u}_{\star}\left(k x_{1} ; k\right), \quad \mathbf{u}_{\star}(\xi ; k)=\mathbf{u}_{\star}(\xi+2 \pi ; k)
$$

parameterized by the spatial wavenumber $k>0$. In fact, such families occur for an open class of reaction-diffusion systems, including but not limited to systems of activator-inhibitor type mentioned above.
As a first predictor on the stability of such solutions with respect to perturbations, one analyzes the linearization,

$$
\begin{equation*}
\mathbf{v}_{t}=D \triangle \mathbf{v}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}(k x ; k)\right) \mathbf{v} . \tag{1.1}
\end{equation*}
$$

It turns out that, again for open classes of reaction-diffusion systems including the above examples, solutions to this linear equation are bounded for bounded initial data, for an open subset of patterns $\mathbf{u}_{\star}(\cdot ; k)$ in the family. We refer to such patterns as linearly stable Turing patterns. We will discuss detailed assumptions that guarantee such linear stability later in this section.

The presence of a family of patterns, parameterized by the wavenumber, and, even more obviously, by translations of the pattern in $x$, implies that solutions to (1.1) with general initial conditions will not decay. More explicitly, $\mathbf{v}(t, x)=\partial_{x} \mathbf{u}_{\star}(k x ; k)$ and $\mathbf{v}(t, x)=\frac{\mathrm{d}}{\mathrm{d} k} \mathbf{u}_{\star}(k x ; k)$ are constant in time and solve (1.1).

In fact, one can show that under typical assumptions, initial conditions $\mathbf{v}(t=0, x) \in L^{1}\left(\mathbb{R}^{N}, \mathbb{R}^{n}\right)$ will give rise to diffusive decay, $\sup _{x}|\mathbf{v}(t, x)| \leqslant C t^{-N / 2}$. Such algebraic decay is in general not strong enough to ensure nonlinear decay in dimensions $N \leqslant 3$. The simplest example is the nonlinear heat equation

$$
u_{t}=\triangle u+u^{2}
$$

which exhibits blowup of arbitrarily small, smooth, positive initial data at finite time in dimensions $N \leqslant$ $3[6,3]$. In the seminal paper [21], Schneider recognized that diffusive decay near Turing patterns is not
altered by the presence of nonlinear terms due to cancellations in a Bloch-wave expansion. He studied the most difficult case, $N=1$, where diffusion is weak and nonlinearity potentially most dangerous, in the specific example of the Swift-Hohenberg equation. His proof has later been generalized, simplified, and adapted; see $[23,7,8,11,9,10,4,19]$. Our focus here is, again, on the one-dimensional case, in a general reaction-diffusion setting. Our goal is to find coordinates that show explicitly why nonlinear terms do not alter linear decay near Turing patterns. Going back to the scalar heat equation, the interaction of nonlinear terms with diffusion can be categorized as relevant, critical, or irrelevant; $[1,2]$. Explicitly, in the heat equation $u_{t}=u_{x x}+f\left(u, u_{x}, u_{x x}\right)$,
(i) Nonlinear terms such as $f\left(u, u_{x}, u_{x x}\right)=u u_{x x}, u_{x}^{2}, u^{p}$, where $p>3$ are irrelevant;
(ii) Nonlinear terms such as $f\left(u, u_{x}, u_{x x}\right)=u u_{x}, u^{3}$ are critical;
(iii) Nonlinear terms such as $f(u)=u^{2}$ are relevant.

Without pretending to fully explain this phenomenon, notice that, for $L^{1}$-initial data, assuming Gaussian decay, we find $u_{x x} \sim t^{-3 / 2}$ in $L^{\infty}$. Irrelevant nonlinear terms decay with rate $t^{-\alpha}, \alpha>3 / 2$, critical terms have $\alpha=3 / 2$, and relevant terms have $\alpha<3 / 2$.
Perturbations $\mathbf{v}$ of Turing patterns solve a system

$$
\mathbf{v}_{t}=\partial_{x x} \mathbf{v}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}(x)\right) \mathbf{v}+\mathbf{g}(x, \mathbf{v}),
$$

where $\mathbf{g}(x, \mathbf{v})=\mathrm{O}\left(|\mathbf{v}|^{2}\right)$. Note that from here on, we fix the wavenumber $k=1$, without loss of generality, and write $\mathbf{u}_{\star}(x):=\mathbf{u}_{\star}(x ; 1)$. In particular, the nonlinearity $\mathbf{g}$ has potentially dangerous quadratic terms. Roughly speaking, our goal is to find coordinates in which the nonlinearity involves at least two "derivatives", which according to the numerology for the scalar heat equation would be sufficient to guarantee nonlinear decay. The reason to hope for derivatives is the presence of a conservation law associated with the translation symmetry, which in turn generates the neutral decay in the linearization.

To be precise, we now consider reaction diffusion systems

$$
\begin{equation*}
\mathbf{u}_{t}=D \partial_{x x} \mathbf{u}+\mathbf{f}(\mathbf{u}), \tag{1.2}
\end{equation*}
$$

where $\mathbf{u}, \mathbf{f} \in \mathbb{R}^{n}, x \in \mathbb{R}, t \in(0,+\infty), D \in \mathbb{R}^{n \times n}$ is a diagonal matrix with strictly positive diagonal entries and $\mathbf{f}$ is smooth. Firstly, we assume the existence of a Turing pattern of the system.

Hypothesis 1.1 (existence) The system of ordinary differential equations $D \partial_{x x} \mathbf{u}+\mathbf{f}(\mathbf{u})=0$ possesses a smooth periodic even solution $\mathbf{u}_{\star}$.

Without loss of generality, we assume that the period is $2 \pi$. Our aim is to study nonlinear stability of this temporal equilibrium under general small non-periodic perturbations. To this end, we introduce an initial condition

$$
\begin{equation*}
\mathbf{u}(0, x)=\mathbf{u}_{\star}(x)+\mathbf{v}^{0}(x) . \tag{1.3}
\end{equation*}
$$

Then assuming that $\mathbf{u}(t, x)=\mathbf{u}_{\star}(x)+\mathbf{v}(t, x)$ is a solution to (1.2) with the given initial condition (1.3), we have

$$
\left\{\begin{array}{l}
\mathbf{v}_{t}=A \mathbf{v}+\mathbf{g}(x, \mathbf{v})  \tag{1.4}\\
\mathbf{v}(0)=\mathbf{v}^{0}
\end{array}\right.
$$

where

$$
\begin{array}{cccc}
A: X^{1} & \longrightarrow & X  \tag{1.5}\\
\mathbf{v} & \longmapsto D \partial_{x x} \mathbf{v}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v} .
\end{array}
$$

Here we define

$$
\begin{equation*}
X=\left(L^{1}(\mathbb{R})\right)^{n} \cap\left(L^{\infty}(\mathbb{R})\right)^{n}, \quad X^{1}=\left(W^{2,1}(\mathbb{R})\right)^{n} \cap\left(W^{2, \infty}(\mathbb{R})\right)^{n}, \tag{1.6}
\end{equation*}
$$

with norms

$$
\|\cdot\|_{X}=\|\cdot\|_{L^{1}}+\|\cdot\|_{L^{\infty}}, \quad\|\cdot\|_{X^{1}}=\|\cdot\|_{W^{2,1}}+\|\cdot\|_{W^{2, \infty}} .
$$

Note that from now on, we suppress $n$ and $\mathbb{R}$ if there is no ambiguity. Moreover, $\mathbf{g}: \mathbb{T}_{2 \pi} \times \mathbb{R}^{n} \rightarrow \mathbb{R}^{n}$ is smooth, $\mathbf{g}(x, \mathbf{v})=\mathbf{f}\left(\mathbf{u}_{\star}+\mathbf{v}(x)\right)-\mathbf{f}\left(\mathbf{u}_{\star}\right)-\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v}$, so that $\mathbf{g}(x, 0) \equiv 0$ and $\partial_{\mathbf{v}} \mathbf{g}(x, 0) \equiv 0$.
According to Bloch wave decomposition, let us introduce the family of Bloch operators, for $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$,

$$
\begin{array}{ccc}
B(\sigma):\left(H^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n} & \longrightarrow & \left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n} \\
\mathbf{v} & \longmapsto D\left(\partial_{x}+\mathrm{i} \sigma\right)^{2} \mathbf{v}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v} \tag{1.7}
\end{array}
$$

For further reading on Bloch wave decomposition and Bloch operators, we refer to Section 6.2 and [17]. Note that one obtains $B(\sigma)$ formally by applying $A$ to functions of the form $\mathbf{u}=\mathrm{e}^{\mathrm{i} \sigma x} \mathbf{v}$.

Hypothesis 1.2 (spectral stability) The family of Bloch wave operators $B(\sigma)$ has the following properties.
(i) $\operatorname{spec}(B(\sigma)) \bigcap\{R e \lambda \geqslant 0\}=\emptyset$, for $\sigma \neq 0$;
(ii) $\operatorname{spec}(B(0)) \bigcap\{R e \lambda \geqslant 0\}=\{0\}$ and 0 is simple with $\operatorname{span}\left\{\mathbf{u}_{\star}^{\prime}\right\}$ as its eigenspace;
(iii) Near $\sigma=0$, the only eigenvalue $\lambda$ is a smooth function of $\sigma$ and the expression of $\lambda(\sigma)$ reads: $\lambda(\sigma)=-d \sigma^{2}+\mathrm{O}\left(|\sigma|^{3}\right)$, where $d>0$ is a constant.

Remark 1.3 The expansion in (iii) is a consequence of the simplicity of $\lambda=0$ at $\sigma=0$ and the evenness of $\mathbf{u}_{\star}$. In fact, we have an "explicit" expression for d; see Section 6.4.

Given the above hypotheses, we can state our main result.
Theorem 1 (nonlinear stability) Assume Hpotheses 1.1 and 1.2 hold. There are $C, \sigma>0$ so that, for any $\left\|\mathbf{v}^{0}\right\|_{X}<\sigma$, where $X=\left(L^{1}(\mathbb{R})\right)^{n} \cap\left(L^{\infty}(\mathbb{R})\right)^{n}$, the solution $\mathbf{v}(t)$ to the system (1.4) exists for time $t \in[0, \infty)$ and satisfies the estimate

$$
\begin{equation*}
\|\mathbf{v}(t)\|_{\left(L^{\infty}(\mathbb{R})\right)^{n}} \leqslant C \frac{\left\|\mathbf{v}^{0}\right\|_{X}}{(1+t)^{\frac{1}{2}}} . \tag{1.8}
\end{equation*}
$$

The rest of the paper contains three main contributions. First, we construct normal form coordinates, where the neutral mode is represented by a discrete phase $\theta_{j}$, which decays according to a linear discrete diffusion equation $\dot{\theta}_{j}=d\left(\theta_{j+1}-2 \theta_{j}+\theta_{j-1}\right)$. The idea is to capture the leading order dynamics of perturbations using an ansatz of the type $\mathbf{u}(t, x)=\mathbf{u}_{\star}\left(x-\theta_{j}\right)+\mathbf{w}_{j}(t, x)$ on intervals $x \in[2 \pi(j-$ $1 / 2), 2 \pi(j+1 / 2)$ ], where $\mathbf{w}_{j}(t, x)$ lies in a linear strong stable fiber. The coordinate change mimics the much simpler coordinate change in [4], where strong stable fibers of a temporally periodic, but spatially homogeneous solution were straightened out.

Our second main contribution are decay estimates for the linearization in these coordinates. In particular, we show that the $\mathbf{w}_{j}$ indeed decay with higher algebraic rate than the $\theta_{j}$.
Our third main contribution is the computation of nonlinear terms in the new coordinate systems. Leading nonlinear terms turn out to involve discrete derivatives, associated with the discrete translational symmetry near the periodic pattern. Similarly to the scalar case, these discrete derivatives render the nonlinearity irrelevant. From a different view point, dependence on derivatives, only, indicates the presence of a conservation law: An equation $u_{t}=u_{x x}+f\left(u_{x}\right)$ can be rewritten as $v_{t}=v_{x x}+(f(v))_{x}$, for $v=u_{x}$, and the gain in decay is now clear from an integration by parts in the variation of constant formula. An analogous observation applies to the $\theta-\mathbf{W}$ system, where discrete derivatives in the nonlinearity reflect a discrete conservation law.

Together, these observations quite readily imply a nonlinear stability result-Theorem 1 as shown above. The remainder of this paper is organized as follows. In Section 2, we construct the normal form. Section 3 contains linear estimates in Fourier-Bloch space. Section 4 converts those decay estimates into $L^{p}-L^{q}$ decay estimates in physical space. Section 5 contains the proof of the nonlinear stability result. We relegate a detailed description of the nonlinearity, and the spectral properties and the analytic semigroup results of the linear operator to the appendix.

Notation Throughout we will use the following notation.

- $(\cdot, \cdot)$ is the standard inner product on $\mathbb{R}^{n}$ given by

$$
(\mathbf{u}, \mathbf{v})=\sum_{j=1}^{n} u_{j} v_{j}, \text { for any } \mathbf{u}=\left\{u_{j}\right\}_{j=1}^{n}, \mathbf{v}=\left\{v_{j}\right\}_{j=1}^{n} \in \mathbb{R}^{n}
$$

- $\langle\cdot, \cdot\rangle$ is the standard inner product on the Hilbert space $\left(L^{2}(-\pi, \pi)\right)^{n}$ given by

$$
\langle\mathbf{u}, \mathbf{v}\rangle=\int_{-\pi}^{\pi}(\mathbf{u}(x), \mathbf{v}(x)) \mathrm{d} x, \text { for any } \mathbf{u}, \mathbf{v} \in\left(L^{2}(-\pi, \pi)\right)^{n} .
$$

- $\langle\langle\cdot, \cdot\rangle\rangle$ is the standard inner products on $\left(\ell^{2}\right)^{n}$, or the $\left(\ell^{p}\right)^{n}-\left(\ell^{q}\right)^{n}$ pairing, given by

$$
\left\langle\langle\underline{\mathbf{u}}, \underline{\mathbf{v}}\rangle=\sum_{j \in \mathbb{Z}}\left(\mathbf{u}_{j}, \mathbf{v}_{j}\right), \text { for any } \mathbf{u}=\left\{u_{j}\right\}_{j \in \mathbb{Z}}, \mathbf{v}=\left\{v_{j}\right\}_{j \in \mathbb{Z}} .\right.
$$

We denote the Euclidean norm in Euclidean spaces as $|\cdot|$, the norm in a general Banach space $\mathscr{X}$ as $\|\cdot\|_{\mathscr{X}}$, and the norm of a linear operator from a Banach space $\mathscr{X}$ to $\mathscr{Y}$ as $\|\cdot \cdot\|_{\mathscr{X} \rightarrow \mathscr{Y}}$. For the case $\mathscr{Y}=\mathscr{X}$, the last norm notation simply becomes $\|\|\cdot\|\|_{\mathscr{X}}$.

## 2 Normal form

As we pointed out in the introduction, the linear system for the perturbation

$$
\begin{equation*}
\mathbf{v}_{t}=\partial_{x x} \mathbf{v}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}(x)\right) \mathbf{v} \tag{2.1}
\end{equation*}
$$

is expected to exhibit diffusive decay for the linear part. This weak decay is not obviously strong enough to conclude nonlinear decay because of quadratic and cubic terms in the nonlinearity. Our approach here converts the system (2.1) into an infinite-dimensional lattice dynamical system for $\underline{\mathbf{V}}=\left\{\mathbf{V}_{j}\right\}_{j \in \mathbb{Z}}$, where $\mathbf{V}_{j}=\left(\theta_{j}, \mathbf{W}_{j}\right) \in \mathbb{R} \times L^{p}(-\pi, \pi)$ for all $j \in \mathbb{Z}$. Here, the scalar component $\theta_{j}$ of $\mathbf{V}_{j}$ measures local shifts of the primary periodic pattern, and the infinite-dimensional component, $\mathbf{W}_{j}$, represents local complements. In such a representation, one expects diffusive decay of $\theta_{j}$ and faster decay of $\mathbf{W}_{j}$. We will make this precise in Section 4. In fact, the linear asymptotics of $\theta_{j}$ are equivalent to the discrete diffusion

$$
\dot{\theta}_{j}=d\left(\theta_{j+1}-2 \theta_{j}+\theta_{j-1}\right)
$$

The key idea is that in this lattice system, nonlinear terms in the $\theta$-equations involve discrete derivatives, $\theta_{j+1}-\theta_{j}$, rather than $\theta_{j}$ alone. Roughly speaking, we expect $\theta$-dependence to disappear when $\theta_{j}=\theta_{j+1}$ for all $j \in \mathbb{Z}$ due to shift invariance of the original system. Just like in the continuous scalar heat equation, these derivatives decay faster, so that terms like $\left(\theta_{j+1}-\theta_{j}\right)^{2}$ are now irrelevant, that is, they do not alter linear diffusive decay.
In summary, we will find a system, where the linear part exhibits diffusive decay, and where nonlinearities are explicitly irrelevant. In this sense, our transformation has eliminated lower-order terms in the system, that turn out not to contribute to leading order dynamics. The term normal form alludes to this elimination of lower-order terms by comparing with normal form theory in ODEs, where coordinate changes are used to simplify equations and systems at least locally, mostly through removing lower-order terms in the Taylor jet of the equation or system.
The remainder of this section is organized as follows. We discuss local well-posedness and "choppingup", the first key step in the transformation to a lattice system in Section 2.1. The ultimate transformation towards a quasilinear lattice dynamical system is constructed in Section 2.2. Key steps involve separation of the neutral phase $\theta_{j}$ and a smoothing procedure at the chopping boundaries.

### 2.1 Well-posedness: spatially extended system and lattice system

We first show local-in-time well-posedness of the system (1.4) on the space $X=L^{1} \cap L^{\infty}$.
Lemma 2.1 The initial value problem of the semi-linear parabolic system (1.4) is locally well-posed in $X$. To be precise, the following assertions hold:
(i) (existence and uniqueness) For any given $\mathbf{v}^{0} \in X$, there exists some $T>0$, depending only on $\left\|\mathbf{v}^{0}\right\|_{X}$, such that the system (1.4) admits a unique mild solution

$$
\mathbf{v} \in C^{0}\left([0, T],\left(L^{1}(\mathbb{R})\right)^{n}\right) \cap C^{0}\left((0, T],\left(L^{\infty}(\mathbb{R})\right)^{n}\right)
$$

Here a mild solution solves the integral-equation variant of (1.4).
(ii) (regularity) The solution $\mathbf{v}(t, x)$ to (1.4) is smooth for $t \in(0, T]$. Moreover, there exists $C>0$ such that, for all $t \in(0, T]$,

$$
\|\mathbf{v}(t)\|_{H^{2}} \leqslant C t^{-1}\left\|\mathbf{v}^{0}\right\|_{X}
$$

Proof. The existence and uniqueness follow directly from [5] and [12]. To show that $\|\mathbf{v}(t)\|_{H^{2}} \leqslant$ $C t^{-1}\left\|\mathbf{v}^{0}\right\|_{X}$, we first note that for any $T_{0} \in(0, T)$, there exists $C\left(T_{0}\right)>0$ such that

$$
\begin{equation*}
\|\mathbf{v}(t)\|_{H^{2}} \leqslant C\left(T_{0}\right)\left\|\mathbf{v}^{0}\right\|_{L^{2}}, \text { for all } t \in\left(T_{0}, T\right) \tag{2.2}
\end{equation*}
$$

Moreover, by [12, Thm. 7.1.5], for every $\mathbf{v}^{0} \in L^{2}$, there are $T_{1}>0$ and $C\left(T_{1}\right)>0$ such that

$$
\|\mathbf{v}(t / 2)\|_{H^{1}} \leqslant C\left(T_{1}\right)(t / 2)^{-1 / 2}\left\|\mathbf{v}^{0}\right\|_{L^{2}}, \quad\|\mathbf{v}(t)\|_{H^{2}} \leqslant C\left(T_{1}\right)(t / 2)^{-1 / 2}\|\mathbf{v}(t / 2)\|_{H^{1}}, \text { for all } t \in\left(0, T_{1}\right),
$$

which implies that

$$
\begin{equation*}
\|\mathbf{v}(t)\|_{H^{2}} \leqslant \frac{C\left(T_{1}\right)}{2} t^{-1}\left\|\mathbf{v}^{0}\right\|_{L^{2}}, \text { for all } t \in\left(0, T_{1}\right) \tag{2.3}
\end{equation*}
$$

Combining (2.2) and (2.3), we conclude our proof.
Remark 2.2 By Lemma 2.1, we can assume without loss of generality that, in the proof of Theorem 1, the initial perturbation is small in $X \cap H^{2}$.

Now suppose that $\mathbf{v}(t, x)$ is a solution to (1.4), close to 0 . In particular, $\mathbf{v}(t, x)$ is close to 0 on all intervals $[2 \pi(j-1 / 2), 2 \pi(j+1 / 2)], j \in \mathbb{Z}$. Then instead of solving (1.4), we claim that it is equivalent to solve the infinite-dimensional system, for all $j \in \mathbb{Z}$,

$$
\left\{\begin{array}{l}
\partial_{t} \mathbf{v}_{j}=D \partial_{x x} \mathbf{v}_{j}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v}_{j}+\mathbf{g}\left(x, \mathbf{v}_{j}\right)  \tag{2.4}\\
\mathbf{v}_{j}(t, \pi)=\mathbf{v}_{j+1}(t,-\pi) \\
\partial_{x} \mathbf{v}_{j}(t, \pi)=\partial_{x} \mathbf{v}_{j+1}(t,-\pi)
\end{array}\right.
$$

In order to justify the well-posedness of (2.4), we first introduce the chopped space

$$
\begin{equation*}
X_{\mathrm{ch}}=\ell^{1}\left(\mathbb{Z},\left(L^{1}(-\pi, \pi)\right)^{n}\right) \cap \ell^{\infty}\left(\mathbb{Z},\left(L^{\infty}(-\pi, \pi)\right)^{n}\right) \tag{2.5}
\end{equation*}
$$

with the norm defined as

$$
\|\underline{\mathbf{w}}\|_{X_{\mathrm{ch}}}=\sum_{j \in \mathbb{Z}}\left\|\mathbf{w}_{j}\right\|_{L^{1}}+\sup _{j \in \mathbb{Z}}\left\|\mathbf{w}_{j}\right\|_{L^{\infty}}, \text { for any } \underline{\mathbf{w}}=\left\{\mathbf{w}_{j}\right\}_{j \in \mathbb{Z}} \in X_{\mathrm{ch}} .
$$

We then consider the chopping map

$$
\begin{array}{cccc}
\mathscr{T}_{\mathrm{ch}}: & X_{\mathrm{ch}} & \longrightarrow & X \\
\underline{\mathbf{v}} & \longmapsto & \mathscr{T}_{\mathrm{ch}}(\underline{\mathbf{v}}), \tag{2.6}
\end{array}
$$

where $X$ is defined in (1.6) and $\mathscr{T}_{\text {ch }}(\underline{\mathbf{v}})(2 \pi j+x)=\mathbf{v}_{j}(x)$, for all $x \in[-\pi, \pi]$ and $j \in \mathbb{Z}$. It is not hard to see that $\mathscr{T}_{\text {ch }}$ is an isomorphism and thus we have the diagram

$$
\begin{array}{ccc}
X^{1} & \xrightarrow{A} & X \\
\mathscr{T}_{\mathrm{ch}} \uparrow & & \mathscr{T}_{\mathrm{ch}} \uparrow \\
X_{\mathrm{ch}}^{1} & \xrightarrow{A_{\mathrm{ch}}} & X_{\mathrm{ch}},
\end{array}
$$

where $X_{\mathrm{ch}}^{1}:=\mathscr{T}_{\text {ch }}^{-1}\left(X^{1}\right)$ and

$$
\begin{array}{rccc}
A_{\mathrm{ch}}: X_{\mathrm{ch}}^{1} & \longrightarrow & X_{\mathrm{ch}} \\
\underline{\mathbf{v}} & \longmapsto & \mathscr{T}_{\mathrm{ch}}^{-1} A \mathscr{T}_{\mathrm{ch}} \mathbf{v} . \tag{2.7}
\end{array}
$$

More specifically, $\left(A_{\mathrm{ch}} \underline{\mathbf{v}}\right)_{j}=D \partial_{x x} \mathbf{v}_{j}+\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v}_{j}$. To describe $X_{\mathrm{ch}}^{1}$, we define

$$
\begin{gathered}
\widetilde{\mathscr{D}}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right):=\ell^{1}\left(W^{2,1}(-\pi, \pi)\right) \cap \ell^{\infty}\left(W^{2, \infty}(-\pi, \pi)\right), \\
\mathscr{D}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right):=\left\{\underline{\mathbf{v}} \in \widetilde{\mathscr{D}}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right) \mid \mathbf{v}_{j}^{(k)}(t, \pi)=\mathbf{v}_{j+1}^{(k)}(t,-\pi), t \geqslant 0, j \in \mathbb{Z}, k=0,1\right\} .
\end{gathered}
$$

Lemma 2.3 We have $X_{\mathrm{ch}}^{1}=\mathscr{D}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right)$.
Proof. From the definition, we find $X_{\mathrm{ch}}^{1} \subseteq \mathscr{D}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right)$. We only need to show that for any given $\underline{\mathbf{v}} \in \mathscr{D}\left(A_{\mathrm{ch}}, X_{\mathrm{ch}}\right)$, we have $\mathbf{v}=\mathscr{T}_{\mathrm{ch}}(\underline{\mathbf{v}}) \in X^{1}$. In fact, for arbitrary $\mathbf{w} \in C_{c}^{\infty}$, we obtain

$$
\left\langle\mathbf{v}, \mathbf{w}^{\prime}\right\rangle_{L^{2}(\mathbb{R})}=\sum_{j \in \mathbb{Z}}\left\langle\mathbf{v}_{j}(x), \mathbf{w}^{\prime}(2 \pi j+x)\right\rangle=-\sum_{j \in \mathbb{Z}}\left\langle\mathbf{v}_{j}^{\prime}(x), \mathbf{w}(2 \pi j+x)\right\rangle=-\left\langle\mathscr{T}_{\mathrm{ch}}\left(\left\{\mathbf{v}_{j}^{\prime}\right\}_{j \in \mathbb{Z}}\right), \mathbf{w}\right\rangle_{L^{2}(\mathbb{R})},
$$

which shows that $\mathbf{v}^{\prime}=\mathscr{T}_{\mathrm{ch}}\left(\left\{\mathbf{v}_{j}^{\prime}\right\}_{j \in \mathbb{Z}}\right) \in X$. Similarly, we have $\mathbf{v}^{\prime \prime} \stackrel{\text { a.e. }}{=} \mathscr{T}_{\mathrm{ch}}\left(\left\{\mathbf{v}_{j}^{\prime \prime}\right\}_{j \in \mathbb{Z}}\right) \in X$.
In all, we conclude that our initial value problem for a spatially extended system (1.4) is equivalent to an initial value problem for a lattice system as follows.

$$
\left\{\begin{array}{l}
\partial_{t \underline{\mathbf{v}}}=A_{\mathrm{ch}} \underline{\mathbf{v}}+\mathbf{G}(\underline{\mathbf{v}}), \quad x \in(-\pi, \pi), t>0,  \tag{2.8}\\
\mathbf{v}_{j}^{(k)}(t, \pi)=\mathbf{v}_{j+1}^{(k)}(t,-\pi), \quad k=0,1, j \in \mathbb{Z}, t \geqslant 0, \\
\mathbf{v}_{j}(0, x)=\mathbf{v}^{0}(2 \pi j+x), \quad x \in[-\pi, \pi], j \in \mathbb{Z},
\end{array}\right.
$$

where $\mathbf{G}(\underline{\mathbf{v}})=\left\{\mathbf{g}\left(x, \mathbf{v}_{j}\right)\right\}_{j \in \mathbb{Z}}$.
Remark 2.4 For any solution $\mathbf{v}$ to (2.8), we have all higher matching boundary conditions, that is, $\partial_{x}^{m} \mathbf{v}_{j}(t, \pi)=\partial_{x}^{m} \mathbf{v}_{j+1}(t,-\pi)$, for all $t>0, j \in \mathbb{Z}$ and $m \in \mathbb{Z}^{+}$.

### 2.2 Lattice system: phase decomposition and boundary-condition matching

We start with sketching the construction of the normal form step by step without rigorous justification. We first decompose each $2 \pi$-long piece $\mathbf{v}_{j}(x)=\mathbf{v}(2 \pi j+x)$ into a linearly neutral phase and a stable phase and then match the boundary conditions for the stable phase. This two-step smooth phase decomposition procedure will be summarized and justified rigorously in a lemma at the end of this section.
We now decompose each $\mathbf{v}_{j}$ according to

$$
\left\{\begin{array}{l}
\mathbf{v}_{j}(x)=\mathbf{w}_{j}(x)+\mathbf{u}_{\star}\left(x-\theta_{j}\right)-\mathbf{u}_{\star}(x) \\
\left\langle\mathbf{w}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)\right\rangle=0
\end{array}\right.
$$

where $\mathbf{u}_{\text {ad }}$ is an element in the kernel of the adjoint operator of $B(0)$ with $\left\langle\mathbf{u}_{\star}^{\prime}, \mathbf{u}_{\mathrm{ad}}\right\rangle=1$. Substituting this expression into (1.4), we can therefore formally derive a system for $\theta_{j}$ and $\mathbf{w}_{j}$, which takes the explicit form

$$
\left\{\begin{align*}
\dot{\theta}_{j}= & \frac{1}{-1+\left\langle\mathbf{w}_{j}(x), \mathbf{u}_{\mathrm{ad}}^{\prime}\left(x-\theta_{j}\right)\right\rangle}\left[-\left(\mathbf{w}_{j}(\pi)-\mathbf{w}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)\right)\right.  \tag{2.9}\\
& \left.+\left(\partial_{x} \mathbf{w}_{j}(\pi)-\partial_{x} \mathbf{w}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)\right)+\left\langle\tilde{\mathbf{g}}\left(\theta_{j}, \mathbf{w}_{j}\right), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)\right\rangle\right] \\
\dot{\mathbf{w}}_{j}= & D \partial_{x x} \mathbf{w}_{j}+\mathbf{u}_{\star, \theta}\left(x-\theta_{j}\right) \dot{\theta}_{j}+\mathbf{f}\left(\mathbf{w}_{j}+\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right)-\mathbf{f}\left(\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right)
\end{align*}\right.
$$

with boundary conditions

$$
\left\{\begin{array}{l}
\partial_{x}^{m} \mathbf{w}_{j}(\pi)-\partial_{x}^{m} \mathbf{w}_{j+1}(-\pi)=\mathbf{u}_{\star}^{(m)}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}^{(m)}\left(\pi-\theta_{j}\right), \text { for } m=0,1 \\
\left\langle\mathbf{w}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)\right\rangle=0,
\end{array}\right.
$$

where

$$
\tilde{\mathbf{g}}\left(\theta_{j}, \mathbf{w}_{j}\right)=\mathbf{f}\left(\mathbf{w}_{j}+\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right)-\mathbf{f}\left(\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right)-\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right) \mathbf{w}_{j} .
$$

Remark 2.5 In the second equation of (2.9), $\dot{\theta}_{j}$ represents the right hand side of the first equation.
Note that $\mathbf{w}_{j}$ is in a codimension-one subspace depending on $\theta_{j}$. More formally, we mapped every $\mathbf{v}_{j}$ into a vector bundle. Also, the boundary conditions are now nonlinear. These facts generate technical difficulties so that we find it easier to work with a further modified system, where, for all $j \in \mathbb{Z}$, we substitute

$$
\begin{equation*}
\mathbf{w}_{j}(x)=\mathbf{W}_{j}(x)+\mathbf{H}\left(x, \theta_{j-1}, \theta_{j}, \theta_{j+1}, \mathbf{W}_{j}\right) \tag{2.10}
\end{equation*}
$$

For simplicity, we denote $\mathbf{H}_{j}(x)=\mathbf{H}\left(x, \theta_{j-1}, \theta_{j}, \theta_{j+1}, \mathbf{W}_{j}\right)$. In the new coordinates $\underline{\mathbf{V}}=(\underline{\theta}, \underline{\mathbf{W}})$, where $\underline{\theta}=\left\{\theta_{j}\right\}_{j \in \mathbb{Z}}$ and $\underline{\mathbf{W}}=\left\{\mathbf{W}_{j}\right\}_{j \in \mathbb{Z}}$, we will have again "homogeneous matching boundary conditions" and all $\mathbf{W}_{j}$ 's are in a fixed codimension- 1 subspace, that is, for all $j \in \mathbb{Z}$,

$$
\begin{gather*}
\partial_{x}^{m} \mathbf{W}_{j}(\pi)-\partial_{x}^{m} \mathbf{W}_{j+1}(-\pi)=0, \text { for } m=0,1,  \tag{2.11}\\
\left\langle\mathbf{W}_{j}(x), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle=0 . \tag{2.12}
\end{gather*}
$$

We now construct $\underline{\mathbf{H}}=\left\{\mathbf{H}_{j}(x)\right\}_{j \in \mathbb{Z}}$ explicitly in the form

$$
\begin{equation*}
\mathbf{H}_{j}=\mathbf{H}_{j}^{1}+\mathbf{H}_{j}^{2}, \tag{2.13}
\end{equation*}
$$

where $\mathbf{H}_{j}^{1}$ accomplishes "homogeneous matching boundary conditions" (2.11) and $\mathbf{H}_{j}^{2}$ corrects so that every $\mathbf{W}_{j}$ is perpendicular to $\mathbf{u}_{\text {ad }}$ (2.12). First, we construct $\mathbf{H}_{j}^{1}$. In order to accomplish (2.11), one readily verifies that we need

$$
\partial_{x}^{m} \mathbf{H}_{j}^{1}(\pi)-\partial_{x}^{m} \mathbf{H}_{j+1}^{1}(-\pi)=\mathbf{u}_{\star}^{(m)}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}^{(m)}\left(\pi-\theta_{j}\right), \text { for } m=0,1,
$$

which can be achieved by choosing

$$
\left\{\begin{array}{l}
\mathbf{H}_{j}^{1}(x)=\frac{1}{2}\left(\mathbf{u}_{\star}\left(x-\theta_{j+1}\right)-\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right), \text { for } x \sim \pi \\
\mathbf{H}_{j}^{1}(x)=\frac{1}{2}\left(\mathbf{u}_{\star}\left(x-\theta_{j-1}\right)-\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right), \text { for } x \sim-\pi .
\end{array}\right.
$$

In light of this observation, we let

$$
\begin{equation*}
\mathbf{H}_{j}^{1}(x)=\frac{1}{2} \phi(x)\left(\mathbf{u}_{\star}\left(x-\theta_{j+1}\right)-\mathbf{u}_{\star}\left(x-\theta_{j-1}\right)\right)+\frac{1}{4}\left(\mathbf{u}_{\star}\left(x-\theta_{j+1}\right)+\mathbf{u}_{\star}\left(x-\theta_{j-1}\right)-2 \mathbf{u}_{\star}\left(x-\theta_{j}\right)\right), \tag{2.14}
\end{equation*}
$$

where $\phi$ is a smooth odd, increasing function on $[-\pi, \pi]$ such that

$$
\phi(x)= \begin{cases}\frac{1}{2}, & \text { for } x>\frac{\pi}{2} \\ -\frac{1}{2}, & \text { for } x<-\frac{\pi}{2}\end{cases}
$$

To be specific, we can choose

$$
\phi(x)=\left[\eta * \chi_{[0, \infty)}\right](x) \cdot \chi_{[-\pi, \pi]}(x)-\frac{1}{2},
$$

where $\chi_{J}$ is the characteristic function of the interval $J$ and $\eta$ is a smooth nonnegative even mollifier such that

$$
\int_{\mathbb{R}} \eta(x) \mathrm{d} x=1, \text { and }|\eta(x)|=0, \text { for all }|x|>\frac{\pi}{2}
$$

In order to keep $\mathbf{H}_{j}$ identical with $\mathbf{H}_{j}^{1}$ near $\pm \pi, \mathbf{H}_{j}^{2}$ has to be 0 near $\pm \pi$. We first note that there exists an odd function $\psi \in\left(C_{c}^{\infty}(-\pi, \pi)\right)^{n}$ such that $\left\langle\psi, \mathbf{u}_{\text {ad }}\right\rangle=1$ since $\left(C_{c}^{\infty}(-\pi, \pi)\right)^{n}$ is dense in $\left(L^{2}(-\pi, \pi)\right)^{n}$ and $\left\langle\mathbf{u}_{\star}^{\prime}, \mathbf{u}_{\text {ad }}\right\rangle=1$. We then define

$$
\begin{equation*}
\mathbf{H}_{j}^{2}=c_{j} \psi\left(x-\theta_{j}\right), \tag{2.15}
\end{equation*}
$$

where

$$
\begin{equation*}
c_{j}=-\left\langle\mathbf{H}_{j}^{1}, \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)\right\rangle-\left\langle\mathbf{W}_{j}, \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle . \tag{2.16}
\end{equation*}
$$

Noting that $\theta_{j}$ and $\mathbf{W}_{j}$ are small, this concludes the construction of $\mathbf{H}_{j}$.
Defining $X_{\mathrm{ch}}^{\perp}=\left\{\underline{\mathbf{v}} \in X_{\mathrm{ch}} \mid\left\langle\mathbf{v}_{j}, \mathbf{u}_{\mathrm{ad}}\right\rangle=0\right.$, for all $\left.j \in \mathbb{Z}\right\}$, where $X_{\mathrm{ch}}$ is defined in (2.5), we summarize the "smooth phase decomposition" procedure, denoted as $\mathscr{T}_{\text {phd }}$, in the following lemma.

Lemma 2.6 The "smooth phase decomposition" operator $\mathscr{T}_{\text {phd }}$, as constructed above, is a smooth local diffeomorphism. More precisely, there are two neighborhoods of zero $\mathscr{U} \in X_{\mathrm{ch}}, \mathscr{V} \in \ell^{1} \times X_{\mathrm{ch}}^{\perp}$ such that the nonlinear transformation

$$
\begin{array}{cccc}
\mathscr{T}_{\text {phd }}: & \mathscr{V} & \longrightarrow & \mathscr{U} \\
& \underline{\mathbf{V}}=(\underline{\theta}, \underline{\mathbf{W}}(x)) & \longmapsto & \left.\longmapsto \mathbf{W}_{j}(x)+\mathbf{H}_{j}(x)+\mathbf{u}_{\star}\left(x-\theta_{j}\right)-\mathbf{u}_{\star}(x)\right\}_{j \in \mathbb{Z}}
\end{array}
$$

is invertible with $\mathscr{T}_{\text {phd }}$ and $\mathscr{T}_{\text {phd }}^{-1}$ smooth. Its derivative at the origin is

$$
\begin{array}{rccc}
\mathscr{L}_{\text {phd }}:=\mathscr{T}_{\text {phd }}^{\prime}(0): & \ell^{1} \times X_{\text {ch }}^{\perp} & \longrightarrow & X_{\mathrm{ch}}  \tag{2.17}\\
\underline{\mathbf{V}}=(\underline{\theta}, \underline{\mathbf{W}}) & \longmapsto & \underline{\mathbf{W}}+\underline{\mathbf{E}} * \underline{\theta},
\end{array}
$$

where $\mathbf{E}$ is defined in (2.18) below.
Proof. We claim that
(i) $\mathscr{T}_{\text {phd }}(0)=0$;
(ii) $\mathscr{T}_{\text {phd }}$ is $C^{\infty}$;
(iii) $\mathscr{T}_{\text {phd }}^{\prime}(0)$, denoted as $\mathscr{L}_{\text {phd }}$, is an invertible bounded linear operator.

Property (i) is straightforward. As for (ii), $\mathscr{T}_{\text {phd }}$ is smooth with respect to $\underline{\mathbf{W}}$ due to the fact that $\mathscr{T}_{\text {phd }}$ is linear in $\underline{\mathbf{W}}$ for fixed $\underline{\theta}$. On the other hand, the smoothness of $\mathscr{T}_{\text {phd }}$ with respect to $\underline{\theta}$ can be readily reduced to the smoothness of the mapping $\underline{\theta} \mapsto\left\{\mathbf{u}_{*}\left(x-\theta_{j}\right)-\mathbf{u}_{*}(x)\right\}_{j \in \mathbb{Z}}$. A direct calculation shows that, for given $m \in \mathbb{Z}^{+}$, the $m$ th-derivative mapping at $\underline{\theta}$ is $\underline{\eta} \mapsto\left\{\frac{1}{m!} \mathbf{u}_{*}^{(m)}\left(\theta_{j}-x\right) \eta_{j}\right\}_{j \in \mathbb{Z}}$. We now only have to show that (iii) is true. In fact, the linear part of $\left\{\mathbf{H}_{j}+\mathbf{u}_{\star}\left(x-\theta_{j}\right)-\mathbf{u}_{\star}(x)\right\}_{j \in \mathbb{Z}}$ with respect to $(\underline{\theta}, \underline{\mathbf{W}})$ around $(0,0)$ is $\underline{\mathbf{E}} * \underline{\theta}=\left\{\sum_{k \in \mathbb{Z}} \mathbf{E}_{j-k} \theta_{k}\right\}_{j \in \mathbb{Z}}$, where $\underline{\mathbf{E}}=\left\{\mathbf{E}_{j}\right\}_{j \in \mathbb{Z}}$ with

$$
\mathbf{E}_{j}= \begin{cases}\frac{1}{4} \psi(x)-\left(\frac{1}{4}+\frac{1}{2} \phi(x)\right) \mathbf{u}_{\star}^{\prime}(x), & j=-1,  \tag{2.18}\\ -\frac{1}{2}\left(\psi(x)+\mathbf{u}_{\star}^{\prime}(x)\right), & j=0, \\ \frac{1}{4} \psi(x)-\left(\frac{1}{4}-\frac{1}{2} \phi(x)\right) \mathbf{u}_{\star}^{\prime}(x), & j=1, \\ 0, & \text { others }\end{cases}
$$

Then we have the linear phase decomposition operator

$$
\begin{array}{cccc}
\mathscr{L}_{\mathrm{phd}}: & \ell^{1} \times X_{\mathrm{ch}}^{\perp} & \longrightarrow & X_{\mathrm{ch}} \\
\underline{\mathbf{V}}=(\underline{\theta}, \underline{\mathbf{W}}) & \longmapsto & \underline{\mathbf{W}}+\underline{\mathbf{E}} * \underline{\theta} .
\end{array}
$$

Moreover, through direct calculation, it is not hard to obtain the bounded inverse of $\mathscr{L}_{\text {phd }}$

$$
\begin{array}{cccc}
\mathscr{L}_{\mathrm{phd}}^{-1}: & X_{\mathrm{ch}} & \longrightarrow & \ell^{1} \times X_{\mathrm{ch}}^{\perp} \\
\underline{\mathbf{v}} & \longmapsto & (F \underline{\mathbf{v}}, \underline{\mathbf{v}}-\mathbf{E} * F \underline{\mathbf{v}}),
\end{array}
$$

where

$$
\begin{array}{cccc}
F: & X_{\mathrm{ch}} & \longrightarrow & \ell^{1}  \tag{2.19}\\
\underline{\mathbf{v}=}=\left\{\mathbf{v}_{j}\right\}_{j \in \mathbb{Z}} & \longmapsto & \left\{-\left\langle\mathbf{v}_{j}, \mathbf{u}_{\mathrm{ad}}\right\rangle\right\}_{j \in \mathbb{Z}} .
\end{array}
$$

By (i), (ii) and the inverse function theorem, the conclusion of the lemma follows.
Remark 2.7 The above lemma still holds when replacing $X_{\mathrm{ch}}$ with $\mathscr{T}_{\mathrm{ch}}\left(X \cap H^{2}\right)$ and the proof is similar.
In the new coordinates, the system contains lengthy expressions. We therefore introduce some simplifying notation first.

$$
\begin{array}{cccc}
\delta_{+}: & \mathbb{C}^{\mathbb{Z}} & \longrightarrow & \mathbb{C}^{\mathbb{Z}} \\
& \underline{x}=\left\{x_{j}\right\}_{j \in \mathbb{Z}} & \longmapsto & \left\{x_{j+1}-x_{j}\right\}_{j \in \mathbb{Z}} . \\
\delta_{-}: & \mathbb{C}^{\mathbb{Z}} & \longrightarrow & \mathbb{C}^{\mathbb{Z}}  \tag{2.20}\\
& \underline{x} & \longmapsto & \left\{x_{j}-x_{j-1}\right\}_{j \in \mathbb{Z}} . \\
\Gamma: & \left(C\left([-\pi, \pi], \mathbb{R}^{n}\right)\right)^{\mathbb{Z}} & \longrightarrow & \mathbb{R}^{\mathbb{Z}} \\
& \underline{\mathbf{v}} & \longmapsto & \left\{\left(\mathbf{v}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)\right\}_{j \in \mathbb{Z}} .
\end{array}
$$

Now, sorting out the linear terms, our lattice system is

$$
\begin{equation*}
\binom{\underline{\dot{\theta}}}{\underline{\dot{\mathbf{W}}}}=A_{\mathrm{nf}}\binom{\underline{\theta}}{\underline{\mathbf{W}}}+\binom{\mathbf{N}^{\theta}(\underline{\theta}, \underline{\mathbf{W}})}{\mathbf{N}^{\mathbf{w}}(\underline{\theta}, \underline{\mathbf{W}})} \tag{2.21}
\end{equation*}
$$

with boundary-matching and phase-decomposition conditions (2.11), (2.12), where $\mathbf{N}^{\theta / \mathrm{w}}$ represent the nonlinear terms of the system and

$$
A_{\mathrm{nf}}=\mathscr{L}_{\mathrm{phd}}^{-1} A_{\mathrm{ch}} \mathscr{L}_{\mathrm{phd}}=\binom{F}{\mathrm{id}-\mathbf{E} * F} A_{\mathrm{ch}}\left(\begin{array}{ll}
\mathbf{E} * & \mathrm{id}
\end{array}\right)=\left(\begin{array}{cc}
0 & \delta_{+} \Gamma  \tag{2.22}\\
A_{\mathrm{ch}} \mathbf{E} * & A_{\mathrm{ch}}-\mathbf{E} * \delta_{+} \Gamma
\end{array}\right),
$$

where $A_{\mathrm{ch}}$ is the linear operator acting on the chopped variables; see (2.7).
Remark 2.8 (i) Due to the fact that $\mathscr{T}_{\mathrm{ch} / \mathrm{nf}}$ are isomorphisms, $A_{\mathrm{nf}}=\mathscr{L}_{\mathrm{phd}}^{-1} \mathscr{T}_{\mathrm{ch}}^{-1} A \mathscr{T}_{\mathrm{ch}} \mathscr{L}_{\mathrm{phd}}$ shares many properties with $A$. For example, $A_{\mathrm{nf}}$ is sectorial in $\ell^{1} \times X_{\mathrm{ch}}^{\perp}$ since $A$ is sectorial in $X$. Here we use the definition of a sectorial operator from [12] which does not require the operator to have a dense domain.
(ii) We relegate the detailed estimates of the nonlinear terms to Lemma 6.1 in the appendix since expressions are lengthy. We have, roughly,

$$
\left\{\begin{array}{l}
\left|\mathbf{N}^{\theta}\right| \sim\left|\left(\delta_{+} \underline{\theta}\right)^{2}\right|+\left|\underline{\theta}^{3}\right|\left|\delta_{+} \underline{\theta}\right|+(|\underline{\theta}|+|\underline{\mathbf{W}}|)\left(|\underline{\mathbf{W}}|+\left|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right|\right)+\left|\underline{\mathbf{W}^{2}}\right| \\
\left|\mathbf{N}^{\mathbf{w}}\right| \sim|\underline{\theta}|\left|\delta_{+} \underline{\theta}\right|+(|\underline{\theta}|+|\underline{\mathbf{W}}|)\left(|\underline{\mathbf{W}}|+\left|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right|\right)+\left|\underline{\mathbf{W}^{2}}\right| .
\end{array}\right.
$$

(iii) Since the branch of continuous spectrum connected to $\lambda=0$ may intersect the branches of continuous spectrum in $\operatorname{Re} \lambda<0$, it is in general not clear how to globally separate neutral from stable modes even linearly. Phase decompositions have been achieved globally in the case of weak pulse interaction, that is, in the regime where $\mathbf{u}_{\star}(x)$ is close to a homoclinic orbit in the ordinary differential system $D \mathbf{u}_{x x}+\mathbf{f}(\mathbf{u})=0$; see [18] for a linear analysis and [24] for a nonlinear reduction.

## 3 Linear Fourier-Bloch estimates

In Section 3 and Section 4, we derive linear diffusive decay in our linear normal form

$$
\underline{\dot{\mathbf{V}}}=A_{\mathrm{nf}} \underline{\mathbf{V}}
$$

To illustrate the idea, we again use the linear heat equation

$$
u_{t}(t, x)=\triangle u(t, x)
$$

In order to obtain the diffusive decay on $\mathrm{e}^{\triangle t}$, we apply the Fourier transform and obtain the "diagonalized" equation

$$
\widehat{u}_{t}(t, k)=-k^{2} \widehat{u}(t, k)
$$

Then we have that $|\widehat{u}(t, k)|=\mathrm{e}^{-k^{2} t}|\widehat{u}(0, k)|$, for all $t>0$ and $k \in \mathbb{R}$, which, combined with Young's inequality, will give us diffusive decay for the scalar heat equation.

In light of this procedure, we exploit Fourier transforms and the Bloch wave decomposition of $A$ to construct an isomorphism diagram, from which we obtain a direct integral representation of $A_{\mathrm{nf}}$, that is, $\widehat{A}_{\mathrm{nf}}=\int_{-1 / 2}^{1 / 2} \widehat{A}_{\mathrm{nf}}(\sigma) \mathrm{d} \sigma$. Unlike the explicit expression of $\mathrm{e}^{-k^{2} t}$, the estimates on $\mathrm{e}^{\widehat{A}_{\mathrm{nf}}(\sigma) t}$ are more intricate and their derivation will occupy most of this section.

To show the conjugacy between the linear normal form and its counterpart in a Fourier-Bloch space, we build a commutative isomorphism diagram involving the underlying spaces for these two operators, the linear operator $A$ and its Bloch wave decomposition. To this end, we recall the definitions of the linearized operator $A$ in (1.5), the chopping operator $\mathscr{T}_{\text {ch }}$ in (2.6), and the linear phase decomposition operator $\mathscr{L}_{\text {phd }}$ in (2.17) from above. We now consider these operators on $L^{2} / \ell^{2}$-based spaces, that is, with new notation,

$$
\begin{array}{cccc}
\widetilde{A}: & \left(H^{2}(\mathbb{R})\right)^{n} & \longrightarrow & \left(L^{2}(\mathbb{R})\right)^{n} \\
\widetilde{\mathscr{T}}_{\text {ch }}: & \ell^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) & \longrightarrow & \left(L^{2}(\mathbb{R})\right)^{n}  \tag{3.1}\\
\widetilde{\mathscr{L}}_{\text {phd }}: & \ell^{2} \times \ell_{\perp}^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) & \longrightarrow & \ell^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right),
\end{array}
$$

where $\mathbb{T}_{\alpha}=\mathbb{R} / \alpha \mathbb{Z}$ is the one-dimensional torus of length $\alpha$ and

$$
\ell_{\perp}^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right)=\left\{\underline{\mathbf{w}} \in \ell^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \mid\left\langle\mathbf{w}_{j}, \mathbf{u}_{\mathrm{ad}}\right\rangle=0, \text { for all } j \in \mathbb{Z}\right\}
$$

We write $\widehat{\mathbf{u}}=\int_{\mathbb{R}} \mathbf{u}(x) \mathrm{e}^{-\mathrm{i} k x} \mathrm{~d} x$ and introduce several Fourier transform variants as follows:

$$
\begin{align*}
& \mathscr{F}: \quad \ell^{2} \quad \longrightarrow \quad L^{2}\left(\mathbb{T}_{1}\right) \\
& \underline{\theta}=\left\{\theta_{j}\right\}_{j \in \mathbb{Z}} \quad \longmapsto \quad \sum_{j \in \mathbb{Z}} \theta_{j} \mathrm{e}^{-\mathrm{i} 2 \pi j \sigma}, \\
& \mathscr{F}_{n}: \quad\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n} \quad \longrightarrow \quad\left(\ell^{2}\right)^{n} \\
& \mathbf{u}(x) \quad \longmapsto \quad \underline{\mathbf{u}}=\left\{\int_{-\pi}^{\pi} \mathbf{u}(x) \mathrm{e}^{-\mathrm{i} \ell x} \mathrm{~d} x\right\}_{\ell \in \mathbb{Z}}, \\
& \mathscr{F}_{\text {ch }}: \quad \ell^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \quad \longrightarrow \quad L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)  \tag{3.2}\\
& \underline{\mathbf{u}}(x)=\left\{\mathbf{u}_{j}(x)\right\}_{j \in \mathbb{Z}} \longmapsto \underline{\widehat{\mathbf{u}}}(\sigma)=\left\{\sum_{j \in \mathbb{Z}} \int_{\mathbb{T}_{2 \pi}} \mathbf{u}_{j}(x) \mathrm{e}^{-\mathrm{i}(\sigma+\ell)(2 \pi j+x)}\right\}_{\ell \in \mathbb{Z}}, \\
& \mathscr{F}_{\mathrm{nf}}: \quad \ell^{2} \times \ell_{\perp}^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \quad \longrightarrow \quad L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) \\
& (\underline{\theta}, \underline{\mathbf{u}})^{T} \quad \longmapsto \quad\left(\mathscr{F}(\underline{\theta}), \mathscr{F}_{\mathrm{ch}}(\underline{\mathbf{u}})\right)^{T},
\end{align*}
$$

where

$$
L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)=\left\{\underline{\mathbf{w}} \in L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)\left|\left\langle\underline{\mathbf{w}}(\sigma), \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle\right\rangle=0, \text { for all } \sigma \in \mathbb{T}_{1}\right\},
$$

We then have a commutative diagram of isomorphisms as follows,

$$
\begin{align*}
& \left(L^{2}(\mathbb{R})\right)^{n} \quad \stackrel{\widetilde{\mathscr{T}}_{\text {ch }}}{\leftrightarrows} \ell^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \stackrel{\widetilde{\mathscr{P}}_{\mathrm{hh}}}{\longleftarrow} \quad \ell^{2} \times \ell_{\perp}^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \\
& \downarrow \mathscr{B}^{-1} \quad \downarrow \mathscr{F}_{\text {ch }} \quad \downarrow \mathscr{F}_{\text {nf }}  \tag{3.3}\\
& L^{2}\left(\mathbb{T}_{1},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \stackrel{\widehat{\mathscr{T}}_{\text {ch }}}{\leftrightarrows} \quad L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) \quad \stackrel{\widehat{\mathscr{P}}_{\text {phd }}}{\leftrightarrows} L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right),
\end{align*}
$$

where $\mathscr{B}^{-1}$ is the inverse of the direct integral defined in (6.2), Section 6.2 and

$$
\begin{array}{cccc}
\widehat{\mathscr{T}}_{\text {ch }}: & L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) & \longrightarrow & L^{2}\left(\mathbb{T}_{1},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \\
& \underline{\mathbf{u}}(\sigma)=\left\{\mathbf{u}_{j}(\sigma)\right\}_{j \in \mathbb{Z}} & \longmapsto & \mathbf{u}(\sigma)=(2 \pi)^{\frac{1}{2}} \mathscr{F}_{n}^{-1} \underline{\mathbf{u}}(\sigma), \\
\widehat{\mathscr{L}}_{\text {phd }}: & L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) & \longrightarrow & L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) \\
& (\theta(\sigma), \underline{\mathbf{w}}(\sigma)) & \longmapsto & \theta(\sigma) \underline{\mathbf{E}}(\sigma)+\underline{\mathbf{w}}(\sigma) .
\end{array}
$$

Here we have

$$
\begin{equation*}
\underline{\widehat{\mathbf{E}}}(\sigma)=\mathscr{F}_{\mathrm{Fh}}(\underline{\mathbf{E}}), \tag{3.4}
\end{equation*}
$$

with $\underline{\mathbf{E}}$ defined in (2.18). The inverse of $\widehat{\mathscr{L}}_{\text {phd }}$, which will be used later, has the expression

$$
\begin{aligned}
\widehat{\mathscr{L}}_{\text {phd }}^{-1}: L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) & \longrightarrow \\
\underline{\mathbf{w}}(\sigma) & \longmapsto\left(\widehat{F}\left(\underline{\mathbb{W}_{1}}(\sigma)\right), \underline{\mathbf{w}}(\sigma)-\underline{\mathbf{E}^{2}}(\sigma)\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)\right. \\
& \longrightarrow \overline{\mathbf{w}}(\sigma))),
\end{aligned}
$$

where

$$
\begin{array}{rlc}
\widehat{F}: \quad L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right) & \longrightarrow & L^{2}\left(\mathbb{T}_{1}\right) \\
\underline{\mathbf{w}}(\sigma) & \longmapsto & -\left\langle\left\langle\underline{\mathbf{w}}(\sigma), \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle\right\rangle .
\end{array}
$$

We now use tildes for operators in physical space and hats for their conjugates in Fourier space. The index "ch" refers to the chopped operators, the index "phd" refers to the smooth phase decomposition operators, and the index " ff " refers to the normal form operators. We then define

$$
\begin{gather*}
\widetilde{A}_{\mathrm{ch}}:=\widetilde{\mathscr{T}}^{-1} \widetilde{A} \widetilde{\mathscr{T}}_{\mathrm{ch}},  \tag{3.5}\\
\widehat{A}_{\mathrm{ch}}:=\widehat{\mathscr{T}}^{-1} \widehat{A} \widehat{\mathscr{T}}_{\mathrm{ch}}  \tag{3.6}\\
\widetilde{A}_{\mathrm{nf}}:=\widetilde{\mathscr{L}}_{\mathrm{phd}}^{-1} \widetilde{\mathscr{T}}_{\mathrm{ch}}^{-1} \widetilde{A} \widetilde{\mathscr{T}}_{\mathrm{ch}} \widetilde{\mathscr{L}}_{\mathrm{phd}}, \\
\widehat{A}_{\mathrm{nf}}:=\widehat{\mathscr{L}}_{\mathrm{phd}}^{-1} \widehat{\mathscr{T}}_{\mathrm{ch}}^{-1} \widehat{A} \widehat{\mathscr{T}}_{\mathrm{ch}} \widehat{\mathscr{L}}_{\mathrm{phd}},
\end{gather*}
$$

where, according to the Bloch wave decomposition from Theorem 2 in the appendix, we have

$$
\begin{equation*}
\mathscr{B}^{-1} \widetilde{A} \mathscr{B}=\widehat{A}=\int_{-\frac{1}{2}}^{\frac{1}{2}} B(\sigma) \mathrm{d} \sigma, \tag{3.7}
\end{equation*}
$$

with $B(\sigma)$ defined in (1.7). Therefore, by the commutative diagram (3.3) and the equivalence relations in (3.6), (3.7), we find the conjugacy

$$
\begin{equation*}
\widetilde{A}_{\mathrm{nf}}=\mathscr{F}_{\mathrm{nf}}^{-1} \widehat{A}_{\mathrm{nf}} \mathscr{F}_{\mathrm{nf}} . \tag{3.8}
\end{equation*}
$$

Just as we pointed out at the beginning of this section, based on this conjugacy, in order to obtain estimates on $\mathrm{e}^{\widetilde{A}_{\text {n }} t}$, we only need to derive estimates on $\mathrm{e}^{\widehat{A}_{\mathrm{n}} t}$. To this end, we first derive an explicit direct integral expression of $\widehat{A}_{\text {nf }}$. From the equivalence relations in (3.5),(3.7), it is straightforward to see that

$$
\begin{equation*}
\widehat{A}_{\mathrm{ch}}=\int_{-\frac{1}{2}}^{\frac{1}{2}} \widehat{A}_{\mathrm{ch}}(\sigma) \mathrm{d} \sigma, \text { with } \widehat{A}_{\mathrm{ch}}(\sigma):=\mathscr{F}_{n} B(\sigma) \mathscr{F}_{n}^{-1}, \text { for all } \sigma \in[-1 / 2,1 / 2] \tag{3.9}
\end{equation*}
$$

Moreover, for any given $(\theta(\sigma), \underline{\mathbf{w}}(\sigma)) \in L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)$ and fixed $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$, by definition, we have,

$$
\left.\begin{array}{rl}
\left(\widehat{A}_{\mathrm{nf}}\binom{\theta}{\underline{\mathbf{w}}}\right)(\sigma) & =\left(\begin{array}{c}
\widehat{\mathscr{L}}_{\mathrm{phd}}^{-1} \widehat{\mathscr{T}}_{\mathrm{ch}}^{-1} \widehat{A} \widehat{\mathscr{T}}_{\mathrm{ch}} \widehat{\mathscr{L}}_{\mathrm{phd}}\binom{\theta}{\underline{\mathbf{w}}}
\end{array}\right)(\sigma) \\
& =\binom{\widehat{F}(\sigma)}{\mathrm{id}-\underline{\widehat{\mathbf{E}}}(\sigma) \widehat{F}(\sigma)} \widehat{A}_{\mathrm{ch}}(\sigma)(\underline{\widehat{\mathbf{E}}}(\sigma)  \tag{3.10}\\
\mathrm{id}
\end{array}\right)\binom{\theta(\sigma)}{\underline{\mathbf{w}}(\sigma)} .
$$

where

$$
\begin{array}{rllc}
\widehat{F}(\sigma):\left(\ell^{2}\right)^{n} & \longrightarrow & \mathbb{C} \\
\underline{\mathbf{w}} & \longmapsto & -\left\langle\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle\right\rangle,  \tag{3.11}\\
R(\sigma):\left(\ell^{1}\right)^{n} & \longrightarrow & \mathbb{C} \\
\underline{\mathbf{w}} & \longmapsto \mathrm{i} \frac{\sin \pi \sigma}{\pi}\left(\sum_{\ell}(-1)^{\ell} \mathbf{w}_{\ell}, D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right) .
\end{array}
$$

We now conclude that

$$
\begin{equation*}
\widehat{A}_{\mathrm{nf}}=\int_{-\frac{1}{2}}^{\frac{1}{2}} \widehat{A}_{\mathrm{nf}}(\sigma) \mathrm{d} \sigma=\int_{-\frac{1}{2}}^{\frac{1}{2}} \mathscr{L}_{\mathrm{phd}}(\sigma)^{-1} \widehat{A}_{\mathrm{ch}}(\sigma) \mathscr{L}_{\mathrm{phd}}(\sigma) \mathrm{d} \sigma, \tag{3.12}
\end{equation*}
$$

where

$$
\begin{array}{rlrl}
\mathscr{L}_{\mathrm{phd}}(\sigma): \mathbb{C} \times\left(\ell^{2}\right)^{n}(\sigma) & \longrightarrow & \left(\ell^{2}\right)^{n} \\
(\theta, \underline{\mathbf{w}}) & \longmapsto \theta \underline{\widehat{\mathbf{E}}}(\sigma)+\underline{\mathbf{w}} .
\end{array}
$$

Here $\left(\ell^{2}\right)^{n}(\sigma)=\left\{\underline{\mathbf{w}} \in\left(\ell^{2}\right)^{n}\left|\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle\right\rangle=0\right\}$. We also recall that $\widehat{A}_{\mathrm{ch}}(\sigma)$ is defined in (3.9) and $\underline{\widehat{\mathbf{E}}}(\sigma)$ defined in (3.4).

Remark 3.1 We note that for any $\underline{\mathbf{u}} \in L^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)$ and $\underline{\mathbf{v}} \in \mathscr{D}\left(\widehat{A}_{\mathrm{ch}}\right)$,

$$
(\widehat{F} \underline{\mathbf{u}})(\sigma)=\widehat{F}(\sigma) \underline{\mathbf{u}}(\sigma), \quad\left(\mathscr{F} \delta_{+} \Gamma \mathscr{F}_{\mathrm{ch}}^{-1} \underline{\mathbf{v}}\right)(\sigma)=R(\sigma) \underline{\mathbf{v}}(\sigma), \text { for a.e. } \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right] .
$$

In addition, for any $(\theta, \underline{\mathbf{w}}) \in L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1},\left(\ell^{2}\right)^{n}\right)$,

$$
\left(\widehat{\mathscr{L}}_{\mathrm{phd}}\binom{\theta}{\underline{\mathbf{w}}}\right)(\sigma)=\mathscr{L}_{\mathrm{phd}}(\sigma)\binom{\theta(\sigma)}{\underline{\mathbf{w}}(\sigma)} \text {, for a.e. } \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right] .
$$

We now consider the family of linear systems,

$$
\begin{equation*}
\binom{\dot{\theta}}{\underline{\dot{\mathbf{w}}}}=\widehat{A}_{\mathrm{nf}}(\sigma)\binom{\theta}{\underline{\mathbf{w}}} \text {, for all } \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right] \text {. } \tag{3.13}
\end{equation*}
$$

While we obtained these operators based on $L^{2} / \ell^{2}$ spaces, we can also consider them on $L^{q} / \ell^{q}$-based spaces. To be more precise, we first define a family of projections

$$
\begin{align*}
\widetilde{P}_{q}(\sigma): & Y_{q} \\
& \longrightarrow  \tag{3.14}\\
\underline{\mathbf{w}} & \longmapsto \underline{\mathbf{w}}-\frac{1}{2 \pi}\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\star}^{\prime}\right),
\end{align*}
$$

where

$$
Y_{q}= \begin{cases}\left(\ell^{q}\right)^{n}, & \text { for } 1 \leqslant q<\infty  \tag{3.15}\\ \left(\ell_{0}^{\infty}\right)^{n}, & \text { for } q=\infty\end{cases}
$$

Here we have $\ell_{0}^{\infty}=\left\{x \in \ell^{\infty}\left|\lim _{|n| \rightarrow \infty}\right| x_{n} \mid=0\right\}$ with the supremum norm. For any $q \in[1, \infty]$, the projection $P_{q}(\sigma)$ is well-defined. In fact, $\mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right), \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\star}^{\prime}\right) \in Y_{1}$ since $\mathbf{u}_{\mathrm{ad}}( \pm \pi)=\mathbf{u}_{\star}^{\prime}( \pm \pi)=0$. We now denote $\widetilde{Y}_{q, \mathrm{~s}}(\sigma)=\operatorname{Rg} \widetilde{P}_{q}(\sigma)$, and, in the following lemma, define $\widehat{A}_{\mathrm{nf}}(\sigma)$ on $L^{q} / \ell^{q}$-based space.

Lemma 3.2 For $q \in[1, \infty]$ and $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$,

$$
\begin{array}{rlc}
\mathscr{L}_{\text {phd }}(\sigma): \mathbb{C} \times \widetilde{Y}_{q, \mathrm{~s}}(\sigma) & \longrightarrow & Y_{q} \\
(\theta, \underline{\mathbf{w}}) & \longmapsto \theta \underline{\widehat{\mathbf{E}}(\sigma)+\underline{\mathbf{w}}}
\end{array}
$$

is uniformly bounded and invertible with its inverse

$$
\begin{array}{rlc}
\mathscr{L}_{\text {phd }}(\sigma)^{-1}: Y_{q} & \longrightarrow & \left.\left.\mathbb{C} \times \widetilde{Y}_{q, \mathbf{s}}(\sigma) \widehat{(\widehat{F}}\right) \underline{\mathbf{v}}-\underline{\widehat{E}}(\sigma) \widehat{F}(\sigma) \underline{\mathbf{v}}\right) . \\
\underline{\mathbf{v}} & \longmapsto(\sigma)
\end{array}
$$

Moreover,

$$
\widehat{A}_{\mathrm{nf}}(\sigma): \mathbb{C} \times\left(\widetilde{Y}_{q, \mathrm{~s}}(\sigma) \cap \mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)\right) \rightarrow \mathbb{C} \times \widetilde{Y}_{q, \mathrm{~s}}(\sigma)
$$

is well-defined and sectorial. Here $\mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)=\left\{\underline{\mathbf{w}} \in Y_{q} \mid\left\{\left(1+m^{2}\right) \mathbf{w}_{m}\right\}_{m \in \mathbb{Z}} \in Y_{q}\right\}$ is the domain of $\widehat{A}_{\mathrm{ch}}(\sigma)$ in $Y_{q}$.

Proof. The assertions for $\mathscr{L}_{\text {phd }}(\sigma)$ are straightforward. In order to show that $\widehat{A}_{\text {nf }}(\sigma)$ is well-defined, we recall the definition of $\widehat{A}_{\text {nf }}(\sigma)$ in (3.10), which indicates that we only need to show

$$
\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\widehat{\mathbf{E}}}(\sigma) \in \widetilde{Y}_{q, \mathrm{~s}}(\sigma), \quad \operatorname{Rg}\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\underline{\widehat{\mathbf{E}}}(\sigma) R(\sigma)\right) \subseteq \widetilde{Y}_{q, \mathrm{~s}}(\sigma) .
$$

We claim that $\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\widehat{\mathbf{E}}}(\sigma) \in \operatorname{Rg}\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\underline{\widehat{\mathbf{E}}}(\sigma) R(\sigma)\right)$. In fact, recall the definition of $R(\sigma)$ in (3.11) and define $\mathbf{E}(\sigma, x):=\left(\sum_{j} \mathbf{E}_{j}(x) \mathrm{e}^{-\mathrm{i} 2 \pi j \sigma}\right) \mathrm{e}^{-\mathrm{i} \sigma x} \in\left(C^{\infty}\right)^{n}\left(\mathscr{T}_{2 \pi}\right)$, we have

$$
R(\sigma) \underline{\widehat{\mathbf{E}}}(\sigma)=2 \mathrm{i} \sin \pi \sigma\left(\mathbf{E}(\sigma, \pi), D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)=0
$$

which means that $\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\widehat{\mathbf{E}}}(\sigma)=\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\underline{\widehat{\mathbf{E}}}(\sigma) R(\sigma)\right)(\underline{\widehat{\mathbf{E}}}(\sigma)) \in \operatorname{Rg}\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\underline{\widehat{\mathbf{E}}}(\sigma) R(\sigma)\right)$.
We now only have to show $\operatorname{Rg}\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\underline{\widehat{\mathbf{E}}}(\sigma) R(\sigma)\right) \subseteq \widetilde{Y}_{q, \mathrm{~s}}(\sigma)$. Actually, for any $\underline{\mathbf{w}} \in \mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)$ with finitely many nonzero components, we have

$$
\begin{aligned}
\left\langle\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\mathbf{w}}-\widehat{\mathbf{E}}(\sigma) R(\sigma) \underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle & =\left\langle\left\langle\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle+2 \pi R(\sigma) \underline{\mathbf{w}}\right. \\
& =2 \pi\left\langle A\left(\mathrm{e}^{\mathrm{i} \sigma x} \mathscr{F}_{n}^{-1} \underline{\mathbf{w}}\right), \mathbf{u}_{\mathrm{ad}}\right\rangle+\left.2 \pi\left(\mathrm{e}^{\mathrm{i} \sigma x} \mathscr{F}_{n}^{-1} \underline{\mathbf{w}}, D \mathbf{u}_{\mathrm{ad}}^{\prime}(x)\right)\right|_{-\pi} ^{\pi} \\
& =2 \pi\left\langle\mathrm{e}^{\mathrm{i} \sigma x} \mathscr{F}_{n}^{-1} \underline{\mathbf{w}}, B^{*}(0) \mathbf{u}_{\mathrm{ad}}\right\rangle=0,
\end{aligned}
$$

and $\left\{\underline{\mathbf{w}} \in D_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right) \mid \underline{\mathbf{w}}\right.$ has finite many nonzero elements $\}$ is dense in $\mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)$ under the graph norm of $\widehat{A}_{\mathrm{ch}}(\sigma)$. Therefore, $\widehat{A}_{\mathrm{nf}}(\sigma)$ is well-defined.
Next, $\widehat{A}_{\mathrm{nf}}(\sigma)$ is sectorial, due to the facts that $\widehat{A}_{\mathrm{nf}}(\sigma)=\mathscr{L}_{\mathrm{phd}}(\sigma)^{-1} \widehat{A}_{\mathrm{ch}}(\sigma) \mathscr{L}_{\mathrm{phd}}(\sigma)$ and $\widehat{A}_{\mathrm{ch}}(\sigma)$ is sectorial (for details, see Section 6.3 in the appendix).

Now we are ready to obtain the estimates for the time evolution of system (3.13), for any given $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$. Our discussion is split into the case $\sigma$ close to 0 and the case $\sigma$ away from 0 .
For the case $\sigma \sim 0$, the derivation of the estimate relies on a diagonalized normal form, that is, a complete separation of the netural and stable phase. First, we notice that $\operatorname{spec}\left(\widehat{A}_{\text {nf }}(\sigma)\right)=\operatorname{spec}\left(\widehat{A}_{\text {ch }}(\sigma)\right)$ is independent of the choice of $q \in[1, \infty]$ and $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$, which we will prove in Proposition 6.4. Moreover, for $\sigma$ sufficiently small, there is a unique continuation of the eigenvalue 0 , denoted as $\lambda(\sigma)$. The set $\Lambda_{1}:=\{\lambda(\sigma)\}$ is a spectral set; see Section 6.4, 6.5 for detailed treatment. Hence, let

$$
\begin{array}{rlc}
P_{q}(\sigma): & Y_{q} & \longrightarrow  \tag{3.16}\\
& \underline{\mathbf{w}} & \left.\longmapsto \underline{\mathbf{w}}-\frac{1}{2 \pi}\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathbf{e}^{*}(\sigma)\right)\right\rangle\right\rangle \mathscr{F}_{n}(\mathbf{e}(\sigma))
\end{array}
$$

be the spectral projection associated with $\Lambda_{2}:=\operatorname{spec}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right) \backslash\{\lambda(\sigma)\}$. Here $\mathbf{e}(\sigma)$ (respectively, $\mathbf{e}^{*}(\sigma)$ ) is the eigenvector of the Bloch wave operator $B(\sigma)$ (respectively, the adjoint operator $B^{*}(\sigma)$ ) according to $\lambda(\sigma)$ with

$$
\begin{equation*}
\mathbf{e}(0)=\mathbf{u}_{\star}^{\prime}, \quad \mathbf{e}^{*}(0)=\mathbf{u}_{\mathrm{ad}}, \quad\left\langle\mathbf{e}(\sigma), \mathbf{e}^{*}(\sigma)\right\rangle=1 . \tag{3.17}
\end{equation*}
$$

We refer to Section 6.4 in the appendix for more details on $\mathbf{e}(\sigma)$ and $\mathbf{e}^{*}(\sigma)$. We now denote

$$
\begin{align*}
Y_{q, \mathrm{c}}(\sigma)=\operatorname{span}\{\mathbf{e}(\sigma)\}, & Y_{q, \mathrm{~s}}(\sigma)=\operatorname{Rg} P_{q}(\sigma), \\
\left.\widehat{A}_{\mathrm{ch}}(\sigma)\right|_{Y_{q, \mathrm{c}}(\sigma)}=\widehat{A}_{\mathrm{c}}(\sigma), & \left.\widehat{A}_{\mathrm{ch}}(\sigma)\right|_{Y_{q, \mathrm{~s}}}=\widehat{A}_{\mathrm{s}}(\sigma) . \tag{3.18}
\end{align*}
$$

We then introduce the following diagonalized operator

$$
\widehat{A}_{\mathrm{dg}}(\sigma)=\left(\begin{array}{cc}
\lambda(\sigma) & 0  \tag{3.19}\\
0 & \widehat{A}_{\mathrm{s}}(\sigma)
\end{array}\right) .
$$

It is not hard to conclude that for $\sigma$ sufficiently small,

$$
\widehat{A}_{\mathrm{dg}}(\sigma): \mathbb{C} \times\left(Y_{q, \mathrm{~s}}(\sigma) \cap \mathscr{D}^{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)\right) \rightarrow \mathbb{C} \times Y_{q, \mathrm{~s}}(\sigma)
$$

is a well-defined operator.
The key step here is to find an invertible bounded linear transformation

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)=\left(\begin{array}{ll}
\widehat{T}_{00}(\sigma) & \widehat{T}_{01}(\sigma)  \tag{3.20}\\
\widehat{T}_{10}(\sigma) & \widehat{T}_{11}(\sigma)
\end{array}\right): \mathbb{C} \times \widetilde{Y}_{q, \mathrm{~s}}(\sigma) \rightarrow \mathbb{C} \times Y_{q, \mathrm{~s}}(\sigma)
$$

such that $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma) \widehat{A}_{\mathrm{nf}}(\sigma)=\widehat{A}_{\mathrm{dg}}(\sigma) \widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$. We note that the choice of $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$ is not unique since there are nontrivial invertible operators that commute with $\widehat{A}_{\mathrm{dg}}(\sigma)$.

Lemma 3.3 For $\sigma$ sufficiently small (that is, $|\sigma| \leqslant \gamma_{0}$ ) and $q \in[1, \infty]$,

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)=\left(\begin{array}{cc}
\mu(\delta) & \left.S(\sigma)\right|_{\widetilde{Y}_{q, \mathrm{~s}}(\sigma)}  \tag{3.21}\\
P_{q}(\sigma) \underline{\widehat{\mathbf{E}}}(\sigma) & \left.P_{q}(\sigma)\right|_{\widetilde{Y}_{q, \mathrm{~s}}(\sigma)}
\end{array}\right)
$$

satisfies the relation $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma) \widehat{A}_{\mathrm{nf}}(\sigma)=\widehat{A}_{\mathrm{dg}}(\sigma) \widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$. Here we have that $\mu(\sigma)=-\frac{1}{2 \pi}\left\langle\left\langle\underline{\mathbf{\mathbf { E }}}(\sigma), \mathscr{F}_{n}\left(\mathbf{e}^{*}(\sigma)\right)\right\rangle\right\rangle$ and

$$
\begin{aligned}
S(\sigma): & Y_{q} \\
& \longrightarrow \\
& \underline{\mathbf{w}}
\end{aligned}>-\frac{1}{2 \pi}\left\langle\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(\mathbf{e}^{*}(\sigma)\right)\right\rangle\right\rangle .
$$

Moreover, we have

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}^{-1}=\left(\widehat{T}_{00}-\widehat{T}_{01} \widehat{T}_{11}^{-1} \widehat{T}_{10}\right)^{-1}\left(\begin{array}{c}
1  \tag{3.22}\\
-\widehat{T}_{11}^{-1} \widehat{T}_{10}
\end{array} \quad\left(\widehat{T}_{00}-\widehat{T}_{01} \widehat{T}_{11}^{-1} \widehat{T}_{10}\right) \widehat{T}_{11}^{-1} \widehat{T}_{11}^{-1}+\widehat{T}_{11}^{-1} \widehat{T}_{10} \widehat{T}_{01} \widehat{T}_{11}^{-1}\right),
$$

in which we suppress $\sigma$-dependence for simplicity.

Proof. We recall from (3.10) that

$$
\widehat{A}_{\mathrm{nf}}(\sigma)=\binom{\widehat{F}(\sigma)}{\operatorname{id}-\underline{\widehat{\mathbf{E}}}(\sigma) \widehat{F}(\sigma)} \widehat{A}_{\mathrm{ch}}(\sigma)(\underline{\widehat{\mathbf{E}}}(\sigma) \mathrm{id}) .
$$

Therefore, in order to find a $\widehat{\mathscr{T}_{\mathrm{d}}}$ as required, we only need to find an invertible bounded linear operator

$$
\widehat{\mathscr{T}}_{\text {int }}(\sigma)=\binom{\widehat{\mathscr{T}_{1}}(\sigma)}{\widehat{\mathscr{T}}_{2}(\sigma)}: Y_{q} \longrightarrow \mathbb{C} \times Y_{q, \mathrm{~s}}(\sigma)
$$

such that

$$
\widehat{\mathscr{T}}_{\mathrm{int}}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma)=\widehat{A}_{\mathrm{dg}}(\sigma) \widehat{\mathscr{T}}_{\mathrm{int}}(\sigma),
$$

which is equivalent to

$$
\left\{\begin{array}{l}
\widehat{\mathscr{T}_{1}}(\sigma)\left(\lambda(\sigma)-\widehat{A}_{\mathrm{ch}}(\sigma)\right)=0, \\
\widehat{\mathscr{T}}_{2}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma)-\widehat{A}_{\mathrm{ch}}(\sigma) \widehat{\mathscr{T}}_{2}(\sigma)=0 .
\end{array}\right.
$$

While the choice of $\widehat{\mathscr{T}}_{1 / 2}(\sigma)$ satisfying the above equation is apparently not unique, we choose that $\widehat{\mathscr{T}}_{1}(\sigma)=S(\sigma)$ and $\widehat{\mathscr{T}_{2}}(\sigma)=P_{q}(\sigma)$. As a result, we have

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)=\binom{\widehat{\mathscr{T}_{1}}(\sigma)}{\widehat{\mathscr{T}}_{2}(\sigma)}\left(\begin{array}{ll}
\widehat{\mathbf{E}}(\sigma) & \left.\mathrm{id}\right|_{\widetilde{\tilde{Y}}_{q, \mathrm{~s}}(\sigma)}
\end{array}\right)=\left(\begin{array}{cc}
\mu(\delta) & \left.S(\sigma)\right|_{\widetilde{Y}_{q, s}(\sigma)} \\
P_{q}(\sigma) \widehat{\widehat{\mathbf{E}}}(\sigma) & \left.P_{q}(\sigma)\right|_{\widehat{Y}_{q, s}(\sigma)}
\end{array}\right) .
$$

To show that $\widehat{\mathscr{T}_{\mathrm{dg}}}(\sigma)^{-1}$ in (3.22) is correct, we only need to verify that

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)^{-1} \widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)=\left(\begin{array}{cc}
\mathrm{id} & 0 \\
0 & \left.\mathrm{id}\right|_{\widetilde{Y}_{q, \mathrm{~s}}(\sigma)}
\end{array}\right), \quad \widehat{\mathscr{T}} \mathrm{dg}(\sigma) \widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)^{-1}=\left(\begin{array}{cc}
\mathrm{id} & 0 \\
0 & \left.\mathrm{id}\right|_{Y_{q, \mathrm{~s}}(\sigma)}
\end{array}\right),
$$

which is clearly true.
Based on this lemma, we now derive the estimate for $\mathrm{e}^{\widehat{A}_{\text {nf }}(\sigma) t}$ when $\sigma$ is close to zero. We first introduce new notation $M(t, \sigma):=\mathrm{e}^{\widehat{A}_{\text {nf }} t}$ and $\mathscr{M}(t):=\mathrm{e}^{\widetilde{A}_{\text {nf }} t}$ with

$$
\begin{align*}
M(t, \sigma) & =\left(\begin{array}{ll}
M_{00}(t, \sigma) & M_{01}(t, \sigma) \\
M_{10}(t, \sigma) & M_{11}(t, \sigma)
\end{array}\right):  \tag{3.23}\\
\mathscr{M}(t) & =\left(\begin{array}{ll}
\mathscr{M}_{00}(t) & \mathscr{M}_{01}(t) \\
\mathscr{M}_{10}(t) & \mathscr{M}_{11}(t)
\end{array}\right): \\
\ell^{2} \times \widetilde{Y}_{q, \mathrm{~s}}(\sigma)\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) & \longrightarrow
\end{align*} \mathbb{C} \times \overparen{Y}_{q, \mathrm{~s}}^{2} \times \ell_{\perp}^{2}\left(\mathbb{Z},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) .
$$

To make sense of the derivatives and Taylor expansions with respect to $\sigma$ of entries in $\mathscr{T}_{\mathrm{dg}}(\sigma)$, we extend $\widehat{T}_{01}(\sigma)$ and $\widehat{T}_{11}(\sigma)$ continuously as operators on $Y_{q}$, that is,

$$
\begin{equation*}
\widehat{T}_{01}(\sigma)=S(\sigma) \widetilde{P}_{q}(\sigma), \quad \widehat{T}_{11}(\sigma)=P_{q}(\sigma) \widetilde{P}_{q}(\sigma) \tag{3.24}
\end{equation*}
$$

The same argument applies to operators $\widehat{\mathscr{T}}_{\mathrm{dg}}^{-1}(\sigma)$ and $M(t, \sigma)$.
Lemma 3.4 For $\sigma$ sufficiently small (that is, $|\sigma| \leqslant \gamma_{0}$ ) and $q \in[1, \infty]$, there exist positive constants $C(q)$ and $\tilde{d}$ such that, for all $t \geqslant 0$,

$$
\left(\begin{array}{cc}
\left|M_{00}(t, \sigma)\right| & \left\|M_{01}(t, \sigma)\right\|_{Y_{q} \rightarrow \mathbb{C}}  \tag{3.25}\\
\left\|\left\|M_{10}(t, \sigma)\right\|_{\mathbb{C} \rightarrow Y_{q}}\right. & \left\|M_{11}(t, \sigma)\right\|_{Y_{q}}
\end{array}\right) \leqslant C(q)\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{1+t}} \\
\frac{1}{\sqrt{1+t}} & \frac{1}{1+t}
\end{array}\right) \mathrm{e}^{-\widetilde{d} \sigma^{2} t} .
$$

Moreover, we have a higher regularity result for $M_{11}(t, \sigma)$, that is, for any given $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right], q \in[1, \infty]$ and $\alpha>0$, there exists $C(q, \alpha)>0$ such that

$$
\left\|M_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha)\left[\left(1+t^{-\alpha}\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}+\frac{1}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\right] .
$$

Proof. The idea is to evaluate $M(t, \sigma)=\mathrm{e}^{\widehat{A}_{\text {nf }}(\sigma) t}$ based on $\mathrm{e}^{\widehat{A}_{\mathrm{nf}}(\sigma) t}=\widehat{\mathscr{T}} \mathrm{dg}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\mathrm{dg}}(\sigma) t} \widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$. We first state the following estimate from Proposition 6.7: For all $q \in[1, \infty]$ and $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right]$, there exists a constant $C(q)>0$ such that

$$
\left|\mathrm{e}^{\lambda(\sigma) t}\right| \leqslant C(q) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}, \quad\left\|\mid \mathrm{e}^{\widehat{A_{\mathrm{s}}}(\sigma) t}\right\|_{q} \leqslant C(q) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}
$$

To obtain estimates on $\widehat{\mathscr{T}}_{\text {dg }}$ and its inverse, we start by computing the Taylor expansions of entries in $\widehat{\mathscr{T}_{\mathrm{dg}}}(\sigma)$. A straightforward calculation using (3.17), (3.4) and (2.18) shows that

$$
\left.\begin{array}{rlrl}
\mathbf{e}(\sigma) & =\mathbf{u}_{\star}^{\prime}+\mathrm{i} \sigma \mathbf{e}_{1}+\mathcal{O}\left(\sigma^{2}\right), & & \mathbf{e}^{*}(\sigma)=\mathbf{u}_{\mathrm{ad}}+\mathrm{i} \sigma \mathbf{e}_{1}^{*}+\mathcal{O}\left(\sigma^{2}\right), \\
\mathrm{e}^{-\mathrm{i} \sigma x} & =1-\mathrm{i} \sigma x+\mathcal{O}\left(\sigma^{2}\right), & & \underline{\mathbf{E}}(\sigma)
\end{array}\right)=\mathscr{F}_{n}\left(-\mathbf{u}_{\star}^{\prime}-2 \pi \mathrm{i} \sigma \phi \mathbf{u}_{\star}^{\prime}+\mathrm{i} \sigma x \mathbf{u}_{\star}^{\prime}\right)+\mathcal{O}\left(\sigma^{2}\right), ~ l
$$

where $\mathbf{e}_{1}$ (respectively, $\mathbf{e}_{1}^{*}$ ) is even and nonzero due to the fact that $B(0) \mathbf{e}_{1}=-2 D \mathbf{u}_{\star}^{\prime \prime}$ (respectively, $B^{*}(0) \mathbf{e}_{1}^{*}=-2 D \mathbf{u}_{\mathrm{ad}}^{\prime}$ ). Then, plugging these expansions into $\widehat{\mathscr{T}_{\mathrm{d}}}$, and using (3.24), we obtain

$$
\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)=\left(\begin{array}{cc}
\mu(\sigma) & S(\sigma) \widetilde{P}_{q}(\sigma)  \tag{3.26}\\
P_{q}(\sigma) \underline{\mathbf{\mathrm { E }}}(\sigma) & P_{q}(\sigma) \widetilde{P}_{q}(\sigma)
\end{array}\right)=\left(\begin{array}{cc}
1 & -2 \pi \mathrm{i} \sigma \Psi \\
-2 \pi \mathrm{i} \sigma \mathscr{F}_{n}(\Phi) & P_{q}(0)
\end{array}\right)+\left(\begin{array}{cc}
\mathcal{O}\left(\sigma^{2}\right) & \mathcal{O}\left(\sigma^{2}\right) \\
\mathcal{O}\left(\sigma^{2}\right) & \mathcal{O}(\sigma)
\end{array}\right),
$$

where $\Phi(x)=-\frac{x}{2 \pi} \mathbf{u}_{\star}^{\prime}+\phi \mathbf{u}_{\star}^{\prime}-\frac{\mathbf{e}_{1}}{2 \pi}$ and

$$
\begin{array}{rlcc}
\Psi: & Y_{q} & \longrightarrow & \mathbb{C} \\
& \underline{\mathbf{w}} & \longmapsto \frac{1}{4 \pi^{2}}\left\langle\left\langle\underline{\mathbf{w}}, \mathscr{F}_{n}\left(x \mathbf{u}_{\mathrm{ad}}+\mathbf{e}_{1}^{*}\right)\right\rangle\right\rangle
\end{array}
$$

Therefore, for any $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right], q \in[1, \infty]$, there exist positive constants $C(q)$ and $\widetilde{d}$ such that, for all $(\theta, \underline{\mathbf{w}}) \in \mathbb{C} \times Y_{q}$, we have the following estimate.

$$
\begin{aligned}
\left(\begin{array}{cc}
\left|M_{00}(t, \sigma) \theta\right| & \left|M_{01}(t, \sigma) \underline{\mathbf{w}}\right| \\
\left\|M_{10}(t, \sigma) \theta\right\|_{Y_{q}} & \left\|M_{11}(t, \sigma) \underline{\mathbf{w}}\right\|_{Y_{q}}
\end{array}\right) & \leqslant C(q)\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right)\left(\begin{array}{cc}
\mathrm{e}^{-\frac{d}{2} \sigma^{2} t} & 0 \\
0 & \mathrm{e}^{-\frac{\gamma_{1}}{2} t}
\end{array}\right)\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right)\binom{|\theta|}{\|\underline{\mathbf{w}}\|_{Y_{q}}} \\
& \leqslant C(q)\left[\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & |\sigma|^{2}
\end{array}\right) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}+\left(\begin{array}{cc}
|\sigma|^{2} & |\sigma| \\
|\sigma| & 1
\end{array}\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}\right]\binom{|\theta|}{\|\underline{\mathbf{w}}\|_{Y_{q}}} \\
& \leqslant C(q) \mathrm{e}^{-\tilde{d} \sigma^{2} t}\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{1+t}} \\
\frac{1}{\sqrt{1+t}} & \frac{1}{1+t}
\end{array}\right)\binom{|\theta|}{\|\underline{\mathbf{w}}\|_{Y_{q}}}
\end{aligned}
$$

Using (3.20), (3.22) and (3.23), we now expand $M_{11}(t, \sigma)$ and obtain

$$
\begin{aligned}
\left\|M_{11}(t, \sigma) \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}} \leqslant & C\left(\left\|\widehat{T}_{11}(\sigma)^{-1} \widehat{T}_{10}(\sigma)\right\|_{Y_{q}^{\alpha}}\left|\mathrm{e}^{\lambda(\sigma) t} \widehat{T}_{01}(\sigma) \underline{\mathbf{W}}\right|+\left\|\widehat{T}_{11}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\mathrm{s}}}(\sigma) t \widehat{T}_{11}(\sigma) \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}}+\right. \\
& \left.\left\|\widehat{T}_{11}^{-1}(\sigma) \widehat{T}_{10}(\sigma)\right\|_{Y_{q}^{\alpha}}\left|\widehat{T}_{01}(\sigma) \widehat{T}_{11}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) t} \widehat{T}_{11}(\sigma) \underline{\mathbf{W}}\right|\right) \\
\leqslant & C(q, \alpha)\left[|\sigma|^{2} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}+\left(t^{-\alpha}+1\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}+|\sigma|^{2} \mathrm{e}^{-\frac{\gamma_{1}}{2} t}\right]\|\underline{\mathbf{W}}\|_{Y_{q}} \\
\leqslant & C(q, \alpha)\left[\left(t^{-\alpha}+1\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}+\frac{1}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\right]\|\underline{\mathbf{W}}\|_{Y_{q}}
\end{aligned}
$$

where in the second inequality we used (3.26), and Proposition 6.7.
Remark 3.5 We point out that, in the above lemma, the estimate for $M_{10}(\sigma)$ can not be improved, since $\mathscr{F}_{n}(\Phi) \neq 0$. In fact, due to the fact that $\Phi(x) \in\left(\mathbf{C}^{\infty}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}$ and $\phi \mathbf{u}_{\star}^{\prime}$ is a nonzero even function, we have

$$
B(0) \Phi=\frac{1}{2 \pi}\left[-2 D \mathbf{u}_{\star}^{\prime \prime}-B(0) \mathbf{e}_{1}+2 \pi B(0)\left(\phi \mathbf{u}_{\star}^{\prime}\right)\right]=B(0)\left(\phi \mathbf{u}_{\star}^{\prime}\right) \neq 0
$$

On the other hand, the estimate for $M_{01}(\sigma)$ can be improved given suitable additional assumptions. For example, if we assume that $\mathbf{u}_{\mathrm{ad}}^{\prime}( \pm \pi)=0$, then $x \mathbf{u}_{\mathrm{ad}}+\mathbf{e}_{1}^{*}$ is zero, which leads to a better estimate.

For the case $\sigma$ away from 0 , we have the following result.
Lemma 3.6 For $\sigma$ away from zero (i.e., for $\gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}$ ) and $q \in[1, \infty]$, there exist constants $C(q), \gamma_{2}>0$ such that

$$
\begin{equation*}
\|M M(t, \sigma)\| \|_{\mathbb{C} \times Y_{q}} \leqslant C(q) \mathrm{e}^{-\gamma_{2} t} \tag{3.27}
\end{equation*}
$$

Moreover, we also have a higher regularity estimate for $M_{11}(t, \sigma)$, that is, for any given $\gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}$, $q \in[1, \infty]$ and $\alpha>0$, there exists $C(q, \alpha)>0$ such that

$$
\mid\left\|M_{11}(t, \sigma)\right\| \|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha)\left(1+t^{-\alpha}\right) \mathrm{e}^{-\gamma_{2} t}
$$

Proof. Recall that $\mathrm{e}^{\widehat{A}_{\text {nf }}(\sigma) t}=\mathscr{L}_{\text {phd }}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma) t} \mathscr{L}_{\text {phd }}(\sigma)$. The inequality (3.27) is true due to the uniform boundedness of $\mathscr{T}(\sigma)$ in Lemma 3.2 and the fact that $\left\|\left\|\mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t} \mid\right\|_{Y_{q}} \leqslant C(q) \mathrm{e}^{-\gamma_{2} t}\right.$, for $\sigma$ away
from 0, in Proposition 6.7. Moreover, by the expressions of $\mathscr{L}_{\text {phd }}(\sigma)$ and its inverse in Lemma 3.2, we have $M_{11}(t, \sigma)=(\operatorname{id}-\underline{\widehat{E}}(\sigma) \widehat{F}(\sigma)) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t}$. Applying Proposition 6.7, we conclude that

$$
\left\|\left\|M_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}}=\right\|\left\|(\mathrm{id}-\underline{\widehat{E}}(\sigma) \widehat{F}(\sigma)) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t}\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha)\left(1+t^{-\alpha}\right) \mathrm{e}^{-\gamma_{2} t} .
$$

Lemma 3.4 and 3.6 give the following proposition.
Proposition 3.7 (Fourier-Bloch estimates) For any $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty]$, there exist constants $C(q), c>0$ such that $\widehat{A}_{\mathrm{nf}}(\sigma)$ is sectorial and

$$
\left(\begin{array}{cc}
\left|M_{00}(t, \sigma)\right| & \left\|\mid M_{01}(t, \sigma)\right\|_{Y_{q} \rightarrow \mathbb{C}}  \tag{3.28}\\
\left\|\mid M_{10}(t, \sigma)\right\|_{\mathbb{C} \rightarrow Y_{q}} & \left\|M_{11}(t, \sigma)\right\|_{Y_{q}}
\end{array}\right) \leqslant C(q)\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{t+1}} \\
\frac{1}{\sqrt{t+1}} & \frac{1}{t+1}
\end{array}\right) \mathrm{e}^{-c \sigma^{2} t}, \text { for all } t \geqslant 0 .
$$

Moreover, we have a higher regularity estimate on $M_{11}(t, \sigma)$, that is, for any $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty]$ and $\alpha>0$, there exist constants $C(q, \alpha), \gamma>0$ such that

$$
\begin{equation*}
\left\|M_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha)\left(\left(1+t^{-\alpha}\right) \mathrm{e}^{-\gamma t}+\frac{1}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\right), \text { for all } t>0 . \tag{3.29}
\end{equation*}
$$

We also need the Fourier-Bloch estimates for the derivative $\partial_{\sigma} M(t, \sigma)$ in the following lemma.
Proposition 3.8 (Fourier-Bloch estimates for derivatives) For any $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty]$ and $\beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$, there exist positive constants $C(q, \beta)$ and $\widetilde{c}$ such that, for all $t \geqslant 0$,

$$
\left(\begin{array}{cc}
\left|\left(\partial_{\sigma} M\right)_{00}(t, \sigma)\right| & \left\|\left(\partial_{\sigma} M\right)_{01}(t, \sigma)\right\|_{Y_{q} \rightarrow \mathbb{C}}  \tag{3.30}\\
\left\|\left\|\left(\partial_{\sigma} M\right)_{10}(t, \sigma)\right\|_{\mathbb{C} \rightarrow Y_{q}}\right. & \left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\| \|_{Y_{q}}
\end{array}\right) \leqslant C(q, \beta)\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{1+t}} \\
\frac{1}{\sqrt{1+t}} & \frac{1}{1+t}
\end{array}\right)\left(t^{\frac{1}{2}}+t^{1-\beta}\right) \mathrm{e}^{-\widetilde{c} \sigma^{2} t} .
$$

Moreover, we have a higher regularity estimate on $\left(\partial_{\sigma} M\right)_{11}(t, \sigma)$, that is, for $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty]$, $\beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$ and $\alpha \in(0,1)$, there exist $C(q, \alpha, \beta)>0$ and $\widetilde{\gamma}>0$ such that

$$
\left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha, \beta)\left(\frac{t^{\frac{1}{2}}+t^{1-\beta}}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}+\left(t^{\frac{1}{2}-\alpha}+t^{1-\beta}\right) \mathrm{e}^{-\widetilde{\gamma} t}\right) \text {, for all } t>0 \text {. }
$$

Proof. On the one hand, we take the partial derivative of the following system with respect to $\sigma$

$$
\binom{\theta(t, \sigma)}{\underline{\widehat{\mathbf{W}}}(t, \sigma)}=M(t, \sigma)\binom{\theta(0, \sigma)}{\underline{\widehat{\mathbf{W}}}(0, \sigma)},
$$

and obtain

$$
\binom{\partial_{\sigma} \theta(t, \sigma)}{\partial_{\sigma} \underline{\widehat{\mathbf{W}}}(t, \sigma)}=M(t, \sigma)\binom{\partial_{\sigma} \theta(0, \sigma)}{\partial_{\sigma} \underline{\widehat{\mathbf{W}}}(0, \sigma)}+\left(\partial_{\sigma} M(t, \sigma)\right)\binom{\theta(0, \sigma)}{\underline{\widehat{\mathbf{W}}}(0, \sigma)} .
$$

On the other hand, we have that

$$
\binom{\underline{\dot{\theta}}(t, \sigma)}{\underline{\hat{\mathbf{W}}}(t, \sigma)}=\widehat{A}_{\mathrm{nf}}(\sigma)\binom{\underline{\hat{\theta}}(t, \sigma)}{\underline{\widehat{\mathbf{W}}}(t, \sigma)} .
$$

Taking the partial derivative with respect to $\sigma$, the equation becomes

$$
\binom{\left(\partial_{\sigma} \underline{\theta}\right)(t, \sigma)}{\left(\partial_{\sigma} \underline{\widehat{\mathbf{W}}}\right)(t, \sigma)}=\widehat{A}_{\mathrm{nf}}(\sigma)\binom{\partial_{\sigma} \underline{\theta}(t, \sigma)}{\partial_{\sigma} \underline{\widehat{\mathbf{W}}}(t, \sigma)}+\widehat{A}_{\mathrm{nf}}^{\prime}(\sigma)\binom{\underline{\theta}(t, \sigma)}{\underline{\widehat{\mathbf{W}}}(t, \sigma)},
$$

for which the variation of constant formula gives

$$
\binom{\partial_{\sigma} \theta(t, \sigma)}{\partial_{\sigma} \underline{\widehat{\mathbf{W}}}(t, \sigma)}=M(t, \sigma)\binom{\partial_{\sigma} \theta(0, \sigma)}{\partial_{\sigma} \underline{\widehat{\mathbf{W}}}(0, \sigma)}+\int_{0}^{t} M(t-s, \sigma) \widehat{A}_{\mathrm{nf}}^{\prime}(\sigma) M(s, \sigma)\binom{\theta(0, \sigma)}{\widehat{\widehat{\mathbf{W}}}(0, \sigma)} \mathrm{d} s
$$

Therefore, one has

$$
\begin{align*}
& \partial_{\sigma} M(t, \sigma)=\int_{0}^{t} M(t-s, \sigma) \widehat{A}_{\text {nf }}^{\prime}(\sigma) M(s, \sigma) \mathrm{d} s \\
= & \int_{0}^{t} \mathscr{L}_{\text {phd }}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma)(t-s)} \mathscr{L}_{\text {phd }}(\sigma)\left(\mathscr{L}_{\text {phd }}(\sigma)^{-1} \widehat{A}_{\text {ch }}(\sigma) \mathscr{L}_{\text {phd }}(\sigma)\right)^{\prime} \mathscr{L}_{\text {phd }}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma) s} \mathscr{L}_{\text {phd }}(\sigma) \mathrm{d} s  \tag{3.31}\\
= & \int_{0}^{t} \mathscr{L}_{\mathrm{phd}}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma)(t-s)} \mathscr{N}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) s} \mathscr{L}_{\text {phd }}(\sigma) \mathrm{d} s,
\end{align*}
$$

where

$$
\begin{equation*}
\mathscr{N}(\sigma)=\widehat{A}_{\mathrm{ch}}^{\prime}(\sigma)+\widehat{A}_{\mathrm{ch}}(\sigma) \underline{\widehat{E}}^{\prime}(\sigma) \widehat{F}(\sigma)-\underline{\widehat{E}}^{\prime}(\sigma) \widehat{F}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma) . \tag{3.32}
\end{equation*}
$$

We recall that $\underline{\widehat{E}}(\sigma)$ is defined in (3.4), $\widehat{F}(\sigma)$ in (3.11) and $\widehat{A}_{\text {ch }}(\sigma)$ in (3.9).
For $|\sigma| \leqslant \gamma_{0}$, by Lemma 3.3 and the above equation (3.31), we have

$$
\begin{equation*}
\partial_{\sigma} M(t, \sigma)=\int_{0}^{t} \widehat{\mathscr{T}_{\mathrm{dg}}}(\sigma)^{-1} \widetilde{\mathscr{N}}(\sigma, t, s) \widehat{\mathscr{T}_{\mathrm{dg}}}(\sigma) \mathrm{d} s, \tag{3.33}
\end{equation*}
$$

where

$$
\begin{align*}
\widetilde{\mathscr{N}}(\sigma, t, s) & =\mathrm{e}^{\widehat{A}_{\mathrm{dg}}(\sigma)(t-s)}\binom{S(\sigma)}{P_{q}(\sigma)} \mathscr{N}(\sigma)\left(-\mathscr{F}_{n} \mathbf{e}(\sigma), \mathrm{id}\right) \mathrm{e}^{\widehat{A}_{\mathrm{dg}}(\sigma) s} \\
& =\left(\begin{array}{cc}
-\mathrm{e}^{\lambda(\sigma) t} S(\sigma) \mathscr{N}(\sigma) \mathscr{F}_{n} \mathbf{e}(\sigma) & \mathrm{e}^{\lambda(\sigma)(t-s)} S(\sigma) \mathscr{N}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s} \\
-\mathrm{e}^{\lambda(\sigma) s} \mathrm{e}^{\widehat{A}_{s}(\sigma)(t-s)} P_{q}(\sigma) \mathscr{N}(\sigma) \mathscr{F}_{n} \mathbf{e}(\sigma) & \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma)(t-s)} P_{q}(\sigma) \mathscr{N}(\sigma) \mathrm{e}^{\widehat{A}_{s}(\sigma) s}
\end{array}\right)  \tag{3.34}\\
& =:\left(\begin{array}{cc}
\widetilde{N}_{00}(\sigma, t, s) & \widetilde{N}_{01}(\sigma, t, s) \\
\widetilde{N}_{10}(\sigma, t, s) & \widetilde{N}_{11}(\sigma, t, s)
\end{array}\right) .
\end{align*}
$$

We now evaluate the entries of $\widetilde{\mathscr{N}}$ with expansions combining (3.32) and (3.34). First, recall the definitions of $\widehat{A}_{\mathrm{dg}}(\sigma)$ in (3.9), $S(\sigma)$ in Lemma 3.3, $P_{q}(\sigma)$ in (3.16), $\mathscr{F}_{n}$ in (3.2), and $\mathbf{e}(\sigma)$ in (3.17).
For $\widetilde{N}_{00}$, note that it is smooth with respect to $\sigma$ and

$$
\widetilde{N}_{00}(0, t, s)=-S(0)\left(\widehat{A}_{\mathrm{ch}}^{\prime}(0)+\widehat{A}_{\mathrm{ch}}(0) \underline{\widehat{E}}^{\prime}(0) \widehat{F}(0)-\underline{\underline{E}}^{\prime}(0) \widehat{F}(0) \widehat{A}_{\mathrm{ch}}(0)\right) \mathscr{F}_{n} \mathbf{e}(0) .
$$

We claim that $\widetilde{N}_{00}(0, t, s)=0$. In fact, since $\widehat{A}_{\mathrm{ch}}^{\prime}(0) \mathscr{F}_{n}(\mathbf{e}(0))$ and $\widehat{A}_{\mathrm{ch}}(0) \widehat{\underline{E}}^{\prime}(0)$ are orthogonal to $\mathscr{F}_{n}\left(\mathbf{u}_{\mathrm{ad}}\right)$ in $Y_{2}, S(0) \widehat{A}_{\mathrm{ch}}^{\prime}(0) \mathscr{F}_{n}(\mathbf{e}(0))=0$ and $S(0) \widehat{A}_{\mathrm{ch}}(0) \underline{\widehat{E}}^{\prime}(0)=0$. Moreover, $\widehat{F}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma)=R(\sigma)$, which is defined in (3.11) with $R(0)=0$. Therefore, there exists a positive constant $C$ such that

$$
\left|\widetilde{N}_{00}\right| \leqslant C|\sigma| \mathrm{e}^{-\frac{d}{2} \sigma^{2} t} .
$$

For $\widetilde{N}_{10}$, due to Proposition 6.7 and the fact that $\widetilde{N}_{10}$ is smooth in $\sigma$ with $\widetilde{N}_{10}(0, t, s) \neq 0$, there exists a positive constant $C$ such that

$$
\left\|\tilde{N}_{10}\right\|_{\mathbb{C} \rightarrow Y_{q, \mathrm{~s}}(\sigma)} \leqslant C \mathrm{e}^{-\frac{\gamma_{1}}{2}(t-s)} \mathrm{e}^{-\frac{d}{2} \sigma^{2} s} \leqslant C \mathrm{e}^{-\frac{d}{2} \sigma^{2} t} \mathrm{e}^{-\frac{\gamma_{1}}{4}(t-s)} .
$$

For $\widetilde{N}_{01}$, we have, for any $q \in[1, \infty]$ and $\beta>\frac{1}{2}\left(1-\frac{1}{q}\right)$,

$$
\begin{aligned}
& \left|\left|\left|\tilde{N}_{01}\right| \|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow \mathbb{C}} \leqslant C\right| \mathrm{e}^{\lambda(\sigma)(t-s)}\right|\left(\left\|\mid \widehat{A}_{\mathrm{ch}}^{\prime}(\sigma) \mathrm{e}^{\widehat{\mathrm{A}}_{\mathrm{s}}(\sigma) s}\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}}+\right. \\
& \left|S(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma) \underline{\underline{E}}^{\prime}(\sigma)\right|\left|\left|| \mathrm { e } ^ { \widehat { \mathrm { A } } _ { \mathrm { s } } ( \sigma ) s } | \left\|_{Y_{q, \mathrm{~s}}(\sigma)}+\left|\left|\widehat{F}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\right| \|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow \mathbb{C}}\right)\right.\right.\right. \\
& \leqslant C \mathrm{e}^{-\frac{d}{2} \sigma^{2}(t-s)}\left(\left\|\left|\mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\right|\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}^{\frac{1}{2}}}+|\sigma|| | \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\left|\left\|_{Y_{q, \mathrm{~s}}(\sigma)}+|\sigma|| |\left|\mathrm{e}^{\widehat{\mathrm{A}}_{\mathrm{s}}(\sigma) s}\right|\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{1}}\right)\right. \\
& \leqslant C(\beta) \mathrm{e}^{-\frac{d}{2} \sigma^{2}(t-s)}\left(\left\|\left|\mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\right|\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}^{\frac{1}{2}}}+\left\|\left|\left|\mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\right|\left\|_{Y_{q, \mathrm{~s}}(\sigma)}+|\sigma|| |\left|\mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\right|\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}^{\beta}}\right),\right.\right.
\end{aligned}
$$

where the last inequality results from the fact that, for any $q \in[1, \infty]$ and $\beta>\frac{1}{2}\left(1-\frac{1}{q}\right)$, we have a continuous imbedding

$$
Y_{q}^{\beta} \hookrightarrow Y_{1}
$$

Now, using Proposition 6.7 and 6.5, we can further conclude that

$$
\begin{aligned}
\left\|\widetilde{N}_{01}\right\| \|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow \mathbb{C}} & \leqslant C(q, \beta)\left(s^{-\frac{1}{2}}+1+|\sigma| s^{-\beta}\right) \mathrm{e}^{-\frac{d}{2} \sigma^{2}(t-s)} \mathrm{e}^{-\frac{\gamma_{1}}{2} s} \\
& \leqslant C(q, \beta) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\left(s^{-\frac{1}{2}}+|\sigma|+|\sigma| s^{-\beta}\right) \mathrm{e}^{-\frac{\gamma_{1}}{4} s} .
\end{aligned}
$$

For $\widetilde{N}_{11}$, we have, for any $q \in[1, \infty]$ and $\beta>\frac{1}{2}\left(1-\frac{1}{q}\right)$,

$$
\begin{aligned}
\left\|\left\|\widetilde{N}_{11} \mid\right\|_{Y_{q, \mathrm{~s}}(\sigma)} \leqslant\right. & C\left\|\left|\mathrm{e}^{\widehat{\mathrm{A}}_{\mathrm{s}}(\sigma)(t-s)}\right|\right\|_{Y_{q, \mathrm{~s}}(\sigma)}\left(\left\|\widehat{A}_{\mathrm{ch}}^{\prime}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s} \mid\right\|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}}\right. \\
& +\left|\left\|\mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s}\left|\left\|_{Y_{q, \mathrm{~s}}(\sigma)}+\mid\right\| \widehat{F}(\sigma) \widehat{A}_{\mathrm{ch}}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) s} \|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow \mathbb{C}}\right)\right.\right. \\
\leqslant & C(q, \beta)\left(\mathrm{s}^{-\frac{1}{2}}+1+|\sigma| s^{-\beta}\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t} .
\end{aligned}
$$

Therefore, combining (3.33), (3.26), and the above estimates for entries, we conclude that, for $|\sigma| \in$ $\left[-\gamma_{0}, \gamma_{0}\right], q \in[1, \infty]$ and $\beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$, there exist positive constants $C(q, \beta)$ and $c_{1} \leqslant \frac{d}{2}$ such that

$$
\begin{align*}
& \left(\begin{array}{cc}
\left|\left(\partial_{\sigma} M\right)_{00}(t, \sigma)\right| & \left\|\mid\left(\partial_{\sigma} M\right)_{01}(t, \sigma)\right\|_{Y_{q} \rightarrow \mathbb{C}} \\
\left\|\left|\mid\left(\partial_{\sigma} M\right)_{10}(t, \sigma) \|_{\mathbb{C} \rightarrow Y_{q}}\right.\right. & \left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q}}
\end{array}\right) \\
\leqslant & \left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right) \int_{0}^{t}\left(\begin{array}{cc}
\left|\widetilde{N}_{00}(\sigma, t, s)\right| & \left\|\mid \widetilde{N}_{01}(\sigma, t, s)\right\|_{Y_{q, s}(\sigma) \rightarrow \mathbb{C}} \\
\left\|| | \widetilde{N}_{10}(\sigma, t, s)\right\|_{\mathbb{C} \rightarrow Y_{q, s}(\sigma)} & \left\|| | \widetilde{N}_{11}(\sigma, t, s)\right\|_{Y_{q, \mathrm{~s}}(\sigma)}
\end{array}\right) \mathrm{d} s\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right) \\
\leqslant & C(q, \beta)\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right)\left(\begin{array}{cc}
|\sigma| t & \frac{t^{\frac{1}{2}}}{\sqrt{1+t}}+|\sigma| \frac{t^{1-\beta}}{(1+t)^{1-\beta}} \\
\frac{t^{\frac{1}{2}}}{\sqrt{1+t}} & \frac{t^{\frac{1}{2}}+t^{1-\beta}}{1+t}
\end{array}\right)\left(\begin{array}{cc}
1 & |\sigma| \\
|\sigma| & 1
\end{array}\right) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}  \tag{3.35}\\
\leqslant & C(q, \beta)\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{1+t}} \\
\frac{1}{\sqrt{1+t}} & \frac{1}{1+t}
\end{array}\right)\left(t^{\frac{1}{2}}+t^{1-\beta}\right) \mathrm{e}^{-c_{1} \sigma^{2} t} .
\end{align*}
$$

Here the inequality $(*)$ relies on the fact that for any $\beta \in(0,1)$, there exists a positive constant $C(\beta)$ such that

$$
\int_{0}^{t} \mathrm{e}^{-\frac{\gamma_{1}}{4} s} \mathrm{~d} s \leqslant C(\beta) \frac{t^{\beta}}{(1+t)^{\beta}}, \quad \int_{0}^{t} s^{-\beta} \mathrm{e}^{-\frac{\gamma_{1}}{4} s} \mathrm{~d} s \leqslant C(\beta) \frac{t^{1-\beta}}{(1+t)^{1-\beta}}
$$

On the other hand, for $\gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}, q \in[1, \infty]$ and $\beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$, by the expression (3.31) and Proposition 6.7, there exist positive constants $C(q, \beta)$ and $c_{2}$ such that

$$
\begin{align*}
\left\|\partial_{\sigma} M(t, \sigma)\right\| \|_{\mathbb{C} \times Y_{q}} & \leqslant C(q) \int_{0}^{t}\left\|\mid \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma)(t-s)} \mathscr{N}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) s}\right\|_{q} \mathrm{~d} s \\
& \leqslant C(q, \beta) \mathrm{e}^{-\gamma_{2} t} \int_{0}^{t}\left(s^{-\frac{1}{2}}+1+|\sigma| s^{-\beta}\right) \mathrm{d} s  \tag{3.36}\\
& \leqslant C(q, \beta)\left(t^{\frac{1}{2}}+t+|\sigma| t^{1-\beta}\right) \mathrm{e}^{-\gamma_{2} t} .
\end{align*}
$$

By (3.35) and (3.36), we now conclude that, for any $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty]$ and $\beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$, there exists positive constant $C(q, \beta)$ and $\widetilde{c}$ such that

$$
\left(\begin{array}{cc}
\left|\left(\partial_{\sigma} M\right)_{00}(t, \sigma)\right| & \left\|\left(\partial_{\sigma} M\right)_{01}(t, \sigma)\right\|_{Y_{q} \rightarrow \mathbb{C}} \\
\left\|\left\|\left(\partial_{\sigma} M\right)_{10}(t, \sigma)\right\|_{\mathbb{C} \rightarrow Y_{q}}\right. & \left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q}}
\end{array}\right) \leqslant C(q, \beta)\left(\begin{array}{cc}
1 & \frac{1}{\sqrt{1+t}} \\
\frac{1}{\sqrt{1+t}} & \frac{1}{1+t}
\end{array}\right)\left(t^{\frac{1}{2}}+t^{1-\beta}\right) \mathrm{e}^{-\widetilde{c} \sigma^{2} t} .
$$

We now consider $\left(\partial_{\sigma} M\right)_{11}(t, \sigma)$. For $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right]$, we plug $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$ from (3.20), $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)^{-1}$ from (3.22), and $\widetilde{N}(\sigma, t, s)$ from the last equality in (3.34) into (3.33). We then obtain

$$
\left(\partial_{\sigma} M\right)_{11}(t, \sigma)=\int_{0}^{t}-\widehat{T}_{11}^{-1} \widehat{T}_{10}\left(\widetilde{N}_{00} \widehat{T}_{01}+\widetilde{N}_{01} \widehat{T}_{11}\right)+\widehat{T}_{11}^{-1}\left(\widetilde{N}_{10} \widehat{T}_{01}+\widetilde{N}_{11} \widehat{T}_{11}\right) \mathrm{d} s \cdot(1+\mathrm{O}(\sigma))
$$

More precisely, for $q \in[1, \infty]$ and $\alpha \in(0,1)$, there exists $C$ such that

$$
\begin{aligned}
\left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma) \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}} \leqslant & C \int_{0}^{t}\left[\left\|\widehat{T}_{11}^{-1} \widehat{T}_{10}\right\|_{Y_{q}^{\alpha}}\left(\left|\widetilde{N}_{00}\right|\left|\widehat{T}_{01} \underline{\mathbf{W}}\right|+\left|\widetilde{N}_{01} \widehat{T}_{11} \underline{\mathbf{W}}\right|\right)\right. \\
& \left.+\left\|\widehat{T}_{11}^{-1} \widetilde{N}_{10}\right\|_{Y_{q}^{\alpha}}\left|\widehat{T}_{01} \underline{\mathbf{W}}\right|+\left\|\widehat{T}_{11}^{-1} \widetilde{N}_{11} \widehat{T}_{11} \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}}\right] \mathrm{d} s \\
\stackrel{* *}{\leqslant} & C \int_{0}^{t}\left[\left\|\widehat{T}_{10}\right\|_{Y_{q}^{\alpha}}\left(\left|\widetilde{N}_{00}\right|\left|\widehat{T}_{01} \underline{\mathbf{W}}\right|+\left|\widetilde{N}_{01} \widehat{T}_{11} \underline{\mathbf{W}}\right|\right)\right. \\
& \left.+\left\|\widetilde{N}_{10}\right\|_{Y_{q}^{\alpha}}\left|\widehat{T}_{01} \underline{\mathbf{W}}\right|+\left\|\widetilde{N}_{11} \widehat{T}_{11} \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}}\right] \mathrm{d} s .
\end{aligned}
$$

Here the inequality $(* *)$ relies on the fact that

$$
\begin{array}{rlr}
\widehat{T}_{11}(\sigma): Y_{q}^{\alpha} & \longrightarrow & Y_{q}^{\alpha} \\
\underline{\mathbf{v}} & \longmapsto \underline{\mathbf{v}}-\left\langle\mathbf{e}(\sigma), \mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right\rangle^{-1}\left\langle\underline{\mathbf{v}}, \mathscr{F}_{n}\left(\mathrm{e}^{-\mathrm{i} \sigma x} \mathbf{u}_{\mathrm{ad}}\right)\right\rangle \mathscr{F}_{n} \mathbf{e}(\sigma)
\end{array}
$$

is a uniformly bounded operator for $q \in[1, \infty]$ and $\alpha \in(0,1)$. Using the explicit expressions of the entries of $\widehat{\mathscr{T}}_{\mathrm{dg}}(\sigma)$ in (3.21) and the estimates on the entries of $\widetilde{\mathscr{N}}$ as shown above, we derive the following estimates,

$$
\begin{aligned}
& \int_{0}^{t}\left\|\widehat{T}_{10}\right\|_{Y_{q}^{\alpha}}\left|\widetilde{N}_{00}\left\|\left.\widehat{T}_{01} \underline{\mathbf{W}}|\mathrm{~d} s \leqslant C(q, \alpha)| \sigma\right|^{3} t \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\right\| \underline{\mathbf{W}} \|_{Y_{q}},\right. \\
& \int_{0}^{t}\left\|\widehat{T}_{10}\right\|_{Y_{q}^{\alpha}}\left|\widetilde{N}_{01} \widehat{T}_{11} \underline{\mathbf{W}}\right| \mathrm{d} s \leqslant C(q, \alpha, \beta)|\sigma|\left(\frac{t^{\frac{1}{2}}}{\sqrt{1+t}}+|\sigma| \frac{t^{1-\beta}}{(1+t)^{1-\beta}}\right) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\|\underline{\mathbf{W}}\|_{Y_{q}}, \\
& \int_{0}^{t}\left\|\widetilde{N}_{10}\right\|_{Y_{q}^{\alpha}}\left|\widehat{T}_{01} \underline{\mathbf{W}}\right| \mathrm{d} s \leqslant C(q, \alpha) \frac{t^{\frac{1}{2}}}{\sqrt{1+t}} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}\|\underline{\mathbf{W}}\|_{Y_{q}}, \\
& \int_{0}^{t}\left\|\widetilde{N}_{11} \widehat{T}_{11} \underline{\mathbf{W}}\right\|_{Y_{q}^{\alpha}} \mathrm{d} s \leqslant C(q, \alpha, \beta)\left(t+t^{1-\beta}+\int_{0}^{t}(t-s)^{-\alpha} s^{-1 / 2} \mathrm{~d} s\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}\|\underline{\mathbf{W}}\|_{Y_{q}} .
\end{aligned}
$$

We now conclude that, for $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right], q \in[1, \infty], \beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$ and $\alpha \in(0,1)$, there exists $C(q, \alpha, \beta)>0$ such that

$$
\left\|\mid\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \leqslant C(q, \alpha, \beta)\left[\frac{t^{\frac{1}{2}}+t^{1-\beta}}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}+\left(t^{\frac{1}{2}-\alpha}+t^{1-\beta}\right) \mathrm{e}^{-\frac{\gamma_{1}}{2} t}\right] .
$$

For $\gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}, q \in[1, \infty], \beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$ and $\alpha \in(0,1)$, there exists $C(q, \alpha, \beta)>0$ such that

$$
\begin{aligned}
\left\|\left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}^{\alpha}}\right. & =\| \|(\operatorname{id}-\widehat{\underline{E}}(\sigma) \widehat{F}(\sigma)) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma)(t-s)} \mathscr{N}(\sigma) \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) s}\| \|_{Y_{q} \rightarrow Y_{q}^{\alpha}} \\
& \leqslant C(q, \alpha, \beta) \mathrm{e}^{-\gamma_{2} t} \int_{0}^{t}(t-s)^{-\alpha} s^{-\frac{1}{2}}+s^{-\frac{1}{2}}+1+|\sigma| s^{1-\beta} \mathrm{d} s \\
& \leqslant C(q, \alpha, \beta)\left(t^{\frac{1}{2}-\alpha}+t^{1-\beta}\right) \mathrm{e}^{-\frac{\gamma_{2}}{2} t}
\end{aligned}
$$

Altogether, for $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], q \in[1, \infty], \beta \in\left(\frac{1}{2}\left(1-\frac{1}{q}\right), 1\right)$ and $\alpha \in(0,1)$, there exist $C(q, \alpha, \beta)>0$ and $\widetilde{\gamma}>0$ such that

$$
\left\|\left\|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{q} \rightarrow Y_{q}} \leqslant C(q, \alpha, \beta)\left[\frac{t^{\frac{1}{2}}+t^{1-\beta}}{1+t} \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}+\left(t^{\frac{1}{2}-\alpha}+t^{1-\beta}\right) \mathrm{e}^{-\widetilde{\gamma} t}\right] \text {, for all } t>0 .\right.
$$

## 4 Linear estimates in physical space

According to the outline at the beginning of Section 3, we are now ready to derive the linear estimates for $\mathrm{e}^{A_{\mathrm{nf}} t}$. To be more precise, we first show by Fubini's Theorem that

$$
\mathscr{M}(t)\binom{\underline{\theta}}{\underline{\mathbf{W}}}=\check{M}(t) *\binom{\underline{\theta}}{\underline{\mathbf{W}}},
$$

where $\check{M}(t)$ is the generalized "inverse Fourier transform" of $M(t, \sigma)$. We then employ an argument similar to, but more intricate than, Young's inequality for the case of the scalar heat equation, exploiting the linear Fourier-Bloch estimates in Proposition 3.7 and 3.8 , to obtain the general $L^{p}-L^{q}$ estimate on our linear normal form $\mathrm{e}^{A_{\text {nf }} t}$.
To this end, we first note that $A_{\mathrm{nf}}=\left.\widetilde{A}_{\mathrm{nf}}\right|_{\ell^{1} \times X_{\mathrm{ch}}^{\perp}}$ and thus we have, by (3.8), for any $(\underline{\theta}, \underline{\mathbf{W}}) \in \ell^{1} \times X_{\mathrm{ch}}^{\perp}$,

$$
\mathscr{M}(t)\binom{\underline{\theta}}{\underline{\mathbf{W}}}=\mathrm{e}^{A_{\mathrm{nf}} t}\binom{\underline{\theta}}{\underline{\mathbf{W}}}=\mathrm{e}^{\widetilde{A}_{\mathrm{nf}} t}(\underline{\underline{\theta}} \underline{\underline{\mathbf{W}}})=\mathscr{F}_{\mathrm{nf}}^{-1} \mathrm{e}^{\widehat{A}_{\mathrm{nf}} t} \mathscr{F}_{\mathrm{nf}}(\underline{\underline{\theta}} \underline{\underline{\mathbf{W}}}), \text { for all } t>0
$$

Recall the notation $\mathscr{M}(t)=\mathrm{e}^{A_{\mathrm{nf}} t}$, the definition of $\mathscr{F}_{\mathrm{nf}}$ from (3.2), and the definition of $\widetilde{A}_{\text {nf }}, \widehat{A}_{\text {nf }}$ from (3.6). In addition, by (3.12), we have, for any $(\theta(\sigma), \underline{\widehat{\mathbf{W}}}(\sigma)) \in L^{2}\left(\mathbb{T}_{1}\right) \times L_{\perp}^{2}\left(\mathbb{T}_{1}, \ell^{2}\right)$,

$$
\left(\mathrm{e}^{\widehat{A}_{\mathrm{nf}} t}\binom{\theta}{\widehat{\mathbf{W}}}\right)(\sigma)=\mathrm{e}^{\widehat{A}_{\mathrm{nf}}(\sigma) t}\binom{\theta(\sigma)}{\underline{\widehat{\mathbf{W}}}(\sigma)}=\mathscr{L}_{\text {phd }}(\sigma)^{-1} \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t} \mathscr{L}_{\mathrm{phd}}(\sigma)\binom{\theta(\sigma)}{\underline{\widehat{\mathbf{W}}}(\sigma)} \text {, for a.e. } \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]
$$

To show that $\mathrm{e}^{A_{\mathrm{nf}} t}$ is a generalized convolution, we first define $M(t, \sigma)$ 's "generalized inverse Fourier transform" $\check{M}(t):=\left(\begin{array}{ll}\check{M}_{01} & \check{M}_{01} \\ \check{M}_{10} & \underline{\underline{M}}_{11}\end{array}\right)$, with expressions as follows

$$
\begin{gather*}
\check{\underline{M}}_{00}(t):=\left\{\check{M}_{00}(t, j)\right\}_{j \in \mathbb{Z}}:=\left\{\int_{-\frac{1}{2}}^{\frac{1}{2}} M_{00}(t, \sigma) \mathrm{e}^{\mathrm{i} 2 \pi \sigma j} \mathrm{~d} \sigma\right\}_{j \in \mathbb{Z}}, \\
\check{M}_{01}(t, y):=\left\{\check{M}_{01}(t, y, j)\right\}_{j \in \mathbb{Z}}:=\left\{\int_{-\frac{1}{2}}^{\frac{1}{2}} \sum_{\ell \in \mathbb{Z}}\left(M_{01}\right)_{\ell}(t, \sigma) \mathrm{e}^{-\mathrm{i}(\sigma+\ell) y} \mathrm{e}^{\mathrm{i} 2 \pi j \sigma} \mathrm{~d} \sigma\right\}_{j \in \mathbb{Z}}, \\
\check{M}_{10}(t, x):=\left\{\check{M}_{10}(t, x, j)\right\}_{j \in \mathbb{Z}}:=\left\{\frac{1}{2 \pi} \int_{-\frac{1}{2}}^{\frac{1}{2}} \sum_{\ell \in \mathbb{Z}}\left(M_{10}\right)_{\ell}(t, \sigma) \mathrm{e}^{\mathrm{i}(\sigma+\ell) x} \mathrm{e}^{\mathrm{i} 2 \pi j \sigma} \mathrm{~d} \sigma\right\}_{j \in \mathbb{Z}},  \tag{4.1}\\
\underline{\mathscr{M}}_{11}(t, x, y):=\left\{\check{M}_{11}(t, x, y, j)\right\}_{j \in \mathbb{Z}}:=\left\{\frac{1}{2 \pi} \int_{-\frac{1}{2}}^{\frac{1}{2}} \sum_{\ell, \eta \in \mathbb{Z}}\left(M_{11}\right)_{\ell \eta}(t, \sigma) \mathrm{e}^{\mathrm{i}(\sigma+\ell) x} \mathrm{e}^{-\mathrm{i}(\sigma+\eta) y} \mathrm{e}^{\mathrm{i} 2 \pi j \sigma} \mathrm{~d} \sigma\right\}_{j \in \mathbb{Z}} .
\end{gather*}
$$

We then have the following lemma.
Lemma 4.1 For any $(\underline{\theta}, \underline{\mathbf{W}}) \in \ell^{1} \times X_{\text {ch }}^{\perp}$ and all $t>0$,

$$
\mathscr{M}(t)\binom{\underline{\theta}}{\underline{\mathbf{W}}}=\check{M}(t) *\binom{\underline{\theta}}{\underline{\mathbf{W}}}=\left(\begin{array}{ll}
\check{\mathscr{M}}_{00} * \underline{\theta} & \frac{\check{M}_{01} * \underline{\mathbf{W}}}{\check{\dddot{M}}_{10} * \underline{\theta}}  \tag{4.2}\\
\underline{\mathscr{M}}_{11} * \underline{\mathbf{W}}
\end{array}\right),
$$

where

$$
\begin{aligned}
\check{\underline{M}}_{00} * \underline{\theta} & =\left\{\sum_{k \in \mathbb{Z}} \check{M}_{00}(t, j-k) \theta_{k}\right\}_{j \in \mathbb{Z}}, \\
\check{\underline{M}}_{01} * \underline{\mathbf{W}} & =\left\{\sum_{k \in \mathbb{Z}} \int_{-\pi}^{\pi} \check{M}_{01}(t, y, j-k) \mathbf{W}_{k}(y) \mathrm{d} y\right\}_{j \in \mathbb{Z}}, \\
\check{\underline{M}}_{10} * \underline{\theta} & =\left\{\sum_{k \in \mathbb{Z}} \check{M}_{10}(t, x, j-k) \theta_{k}\right\}_{j \in \mathbb{Z}}, \\
\check{\underline{M}}_{11} * \underline{\mathbf{W}}= & \left\{\sum_{k \in \mathbb{Z}} \int_{-\pi}^{\pi} \check{M}_{11}(t, x, y, j-k) \mathbf{W}_{k}(y) \mathrm{d} y\right\}_{j \in \mathbb{Z}} .
\end{aligned}
$$

Proof. The proof is a straightforward application of Fubini's theorem.

We are now ready to obtain the general $L^{p}-L^{q}$ linear estimates on $\mathscr{M}(t)$. We denote

$$
X_{q}=\left(L^{q}\left(\mathbb{Z}, L^{q}\left(\mathbb{T}_{2 \pi}\right)\right)\right)^{n}, \text { for any } q \in[1, \infty],
$$

and prove the following proposition.
Proposition 4.2 (general $L^{p}-L^{q}$ estimates) For any $1 \leqslant q \leqslant p \leqslant \infty$ and $(\underline{\theta}, \underline{\mathbf{W}}) \in \ell^{1} \times X_{\mathrm{ch}}^{\perp}$, there exists a positive constant $C$ such that, for all $t>0$,

$$
\left(\begin{array}{ll}
\left\|\mathscr{M}_{00}(t) \underline{\theta}\right\|_{\ell^{p}} & \left\|\mathscr{M}_{01}(t) \underline{\mathbf{W}}\right\|_{\ell^{p}}  \tag{4.3}\\
\left\|\mathscr{M}_{10}(t) \underline{\theta}\right\|_{X_{p}} & \left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}}
\end{array}\right) \leqslant C\left(\begin{array}{cc}
(1+t)^{-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}\|\underline{\theta}\|_{\ell_{q}} & (1+t)^{-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)-\frac{1}{2}}\|\underline{\mathbf{W}}\|_{X_{q}} \\
(1+t)^{-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)-\frac{1}{2}}\|\underline{\theta}\|_{\ell^{q}} & t^{-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}(1+t)^{-1}\|\underline{\mathbf{W}}\|_{X_{q}}
\end{array}\right) .
$$

Proof. We illustrate the derivation of the estimates on $\mathscr{M}_{01}$ and sketch the estimates on $\mathscr{M}_{00}$ and $\mathscr{M}_{10}$. Lastly, we show the estimates for $\mathscr{M}_{11}$.
We first notice that, for any $\underline{\mathbf{W}} \in X_{\mathrm{ch}}^{\perp}$ and $1 \leqslant q, r \leqslant p \leqslant \infty$ satisfying $1+\frac{1}{p}=\frac{1}{q}+\frac{1}{r}$, there exists a positive constant $C$ such that

$$
\begin{equation*}
\left\|\mathscr{M}_{01}(t) \underline{\mathbf{W}}\right\|_{\ell^{p}} \leqslant C\left\|\underline{\check{M}}_{01}(t)\right\|_{X_{\infty}}^{\frac{1}{q}-\frac{1}{p}}\left(\sum_{j} \sup _{|y| \leqslant \pi}\left|\check{M}_{01}(t, y, j)\right|\right)^{\frac{1}{r}}\|\underline{\mathbf{W}}\|_{X_{q}} . \tag{4.4}
\end{equation*}
$$

In fact, by Hölder's inequality, we have

$$
\begin{aligned}
\left\|\mathscr{M}_{01}(t) \underline{\mathbf{W}}\right\|_{\ell^{p}}^{p} & =\left\|\underline{\underline{M}}_{01} * \underline{\mathbf{W}}\right\|_{\ell^{p}}^{p}=\sum_{j \in \mathbb{Z}}\left|\sum_{k \in \mathbb{Z}} \int_{-\pi}^{\pi} \check{M}_{01}(t, y, j-k) \mathbf{W}_{k}(y) \mathrm{d} y\right|^{p} \\
& \leqslant \sum_{j \in \mathbb{Z}}\left(\sum_{k \in \mathbb{Z}} \int_{-\pi}^{\pi}\left|\check{M}_{01}(t, y, j-k)\right|^{1-\frac{r}{p}}\left|\mathbf{W}_{k}(y)\right|^{1-\frac{q}{p}}\left(\left|\check{M}_{01}(t, y, j-k)\right|^{r}\left|\mathbf{W}_{k}(y)\right|^{q}\right)^{\frac{1}{p}} \mathrm{~d} y\right)^{p} \\
& \leqslant\left\|\check{M}_{01}(t)\right\|_{X_{r}}^{p-r}\|\underline{\mathbf{W}}\|_{X_{q}}^{p-q} \sum_{j, k \in \mathbb{Z}} \int_{-\pi}^{\pi}\left|\check{M}_{01}(t, y, j-k)\right|^{r}\left|\mathbf{W}_{k}(y)\right|^{q} \mathrm{~d} y \\
& \leqslant\left\|\check{M}_{01}(t)\right\|_{X_{r}}^{p-r}\left(\sup _{|y| \leqslant \pi} \sum_{j \in \mathbb{Z}}\left|\check{M}_{01}(t, y, j)\right|^{r}\right)\|\underline{\mathbf{W}}\|_{X_{q}}^{p} \\
& \leqslant\left[(2 \pi)^{1-\frac{1}{q}}\left\|\check{\mathscr{M}}_{01}(t)\right\|_{X_{\infty}}^{\frac{1}{q}-\frac{1}{p}}\left(\sup _{|y| \leqslant \pi} \sum_{j \in \mathbb{Z}}\left|\check{M}_{01}(t, y, j)\right|\right)^{\frac{1}{r}}\|\underline{\mathbf{W}}\|_{X_{q}}\right]^{p} .
\end{aligned}
$$

Moreover, by (4.1), we have

$$
\begin{equation*}
\left\|\underline{\underline{M}}_{01}(t)\right\|_{X_{\infty}} \leqslant \sup _{|y| \leqslant \pi}\left|\int_{-\frac{1}{2}}^{\frac{1}{2}}\right| \sum_{\ell \in \mathbb{Z}}\left(M_{01}\right)_{\ell}(t, \sigma) \mathrm{e}^{-\mathrm{i}(\sigma+\ell) y}|\mathrm{~d} \sigma| \leqslant \frac{C(\infty)}{\sqrt{1+t}} \int_{-\frac{1}{2}}^{\frac{1}{2}} \mathrm{e}^{-c \sigma^{2} t} \mathrm{~d} \sigma \leqslant \frac{C}{1+t} . \tag{4.5}
\end{equation*}
$$

Here we use the fact that any bounded linear functional on $\ell_{0}^{\infty}$ can be viewed as a bounded linear functional on $\ell^{\infty}$ with the same norm. We now estimate the $X_{1}$ norm of $\left\{\check{M}_{01}(t, y, j)\right\}_{j \in \mathbb{Z}}$. By using Proposition 3.8, there exists $C>0$, independent of the choice of $y \in[-\pi, \pi]$, such that

$$
\begin{align*}
\sum_{j \neq 0}\left|\check{M}_{01}(t, y, j)\right| & =\sum_{j \neq 0}\left(1+\frac{\left(j-\frac{y}{2 \pi}\right)^{2}}{t}\right)^{-\frac{1}{2}}\left(1+\frac{\left(j-\frac{y}{2 \pi}\right)^{2}}{t}\right)^{\frac{1}{2}}\left|\check{M}_{01}(t, y, j)\right| \\
& \leqslant C\left(\int_{\mathbb{R}} \frac{1}{1+\frac{x^{2}}{t}} \mathrm{~d} x\right)^{\frac{1}{2}}\left[\sum_{j}\left(1+\frac{\left(j-\frac{y}{2 \pi}\right)^{2}}{t}\right)\left|\check{M}_{01}(t, y, j)\right|^{2}\right]^{\frac{1}{2}} \\
& \leqslant C t^{\frac{1}{4}}\left(\int_{-\frac{1}{2}}^{\frac{1}{2}} \sum_{\alpha=0}^{1} t^{-\alpha}\left|\sum_{\ell \in \mathbb{Z}}\left(\partial_{\sigma}^{\alpha}\left(M_{01}\right)_{\ell}(t, \sigma)\right) \mathrm{e}^{-\mathrm{i}(\sigma+\ell) y}\right|^{2} \mathrm{~d} \sigma\right)^{\frac{1}{2}}  \tag{4.6}\\
& \stackrel{* * *}{\leqslant} C t^{\frac{1}{4}}\left(\int_{-\frac{1}{2}}^{\frac{1}{2}} \frac{\mathrm{e}^{-2 c \sigma^{2} t}}{1+t} \mathrm{~d} \sigma+\frac{1}{t} \int_{-\frac{1}{2}}^{\frac{1}{2}} \frac{\left(t^{\frac{1}{2}}+t^{1-\frac{3}{4}}\right)^{2} \mathrm{e}^{-2 \widetilde{c} \sigma^{2} t}}{1+t} \mathrm{~d} \sigma\right)^{\frac{1}{2}} \\
& \leqslant C \frac{t^{\frac{1}{4}}+1}{(1+t)^{\frac{3}{4}}} \leqslant \frac{C}{\sqrt{1+t}}, \text { for all } t>0
\end{align*}
$$

Here in the inequality $\left({ }^{(* * *)}\right.$, we applied Proposition 3.8 with $q=\infty$ and $\beta=\frac{3}{4}$ (actually, any fixed $\left.\beta \in\left(\frac{1}{2}, \frac{3}{4}\right]\right)$. Combining (4.4), (4.5), and (4.6), we have that, for all $1 \leqslant q \leqslant p \leqslant \infty$ and $\underline{\mathbf{W}} \in X_{\mathrm{ch}}^{\perp}$, there exists a positive constant $C$ such that

$$
\left\|\sqrt{1+t} \mathscr{M}_{01}(t) \underline{\mathbf{W}}\right\|_{\ell^{p}} \leqslant \frac{C}{(1+t)^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}}\|\underline{\mathbf{W}}\|_{X_{q}} \text {, for all } t \geqslant 0 .
$$

For $\mathscr{M}_{00}$, the steps are the same as above but easier. For $\mathscr{M}_{10}$, we point out two main differences to the above calculation. First, instead of (4.4), we use

$$
\left\|\mathscr{M}_{10}(t) \underline{\theta}_{X_{p}} \leqslant C\right\| \underline{M}_{10}(t)\left\|_{X_{\infty}}^{\frac{1}{q}-\frac{1}{p}}\left(\int_{-\pi}^{\pi}\left(\sum_{j}\left|\check{M}_{10}(t, x, j)\right|\right)^{\frac{p}{r}} \mathrm{~d} x\right)^{\frac{1}{p}}\right\| \underline{\theta} \|_{\ell q} .
$$

Second, to estimate the $Y_{1}$ norm of $\left\{\check{M}_{10}(t, x, j)\right\}_{j \in \mathbb{Z}}$, we use Proposition 3.8 with $q=1$ and $\beta=$ $\frac{1}{2}$ (actually, any fixed $\left.\beta \in\left(0, \frac{3}{4}\right]\right)$, instead of $q=\infty$ and $\beta=\frac{3}{4}$.
The last step of the proof consists of deriving the estimates for $\mathscr{M}_{11}$. We first have

$$
\left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}} \leqslant(2 \pi)^{\frac{1}{r}}\left(\sup _{|x|,|y| \leqslant|\pi|} \sup _{j \in \mathbb{Z}}\left|\check{M}_{11}(t, x, y, j)\right|\right)^{\frac{1}{q}-\frac{1}{p}}\left(\sup _{|x|,|y| \leqslant \pi} \sum_{j \in \mathbb{Z}}\left|\check{M}_{11}(t, x, y, j)\right|\right)^{\frac{1}{r}}\|\underline{\mathbf{W}}\|_{X_{q}} .
$$

On the one hand, we apply Proposition 3.7 with $q=\infty$ and $\alpha>\frac{1}{2}$ and have

$$
\begin{aligned}
\sup _{|x|,|y| \leqslant|\pi| \mid \pi} \sup _{j \in \mathbb{Z}}\left|\check{M}_{11}(t, x, y, j)\right| & \leqslant \sup _{|x|,|y| \leqslant|\pi|} \int_{-\frac{1}{2}}^{\frac{1}{2}}\left|\sum_{\ell, \eta \in \mathbb{Z}}\left(M_{11}\right)_{\ell \eta}(t, \sigma) \mathrm{e}^{\mathrm{i}(\sigma+\ell) x} \mathrm{e}^{-\mathrm{i}(\sigma+\eta) y}\right| \mathrm{d} \sigma \\
& \left.\leqslant C(\alpha) \int_{-\frac{1}{2}}^{\frac{1}{2}}| | \right\rvert\, M_{11}(t, \sigma) \|_{Y_{\infty} \rightarrow Y_{\infty}^{\alpha}} \mathrm{d} \sigma \\
& \leqslant \frac{C(\alpha)}{t^{\alpha}(1+t)^{\frac{3}{2}-\alpha}} .
\end{aligned}
$$

On the other hand, by applying Proposition 3.7 and 3.8 with $q=\infty, \alpha \in\left(\frac{1}{2}, 1\right)$ and $\beta=\frac{3}{4}$, there exists $C(\alpha)$, independent of choices of $x, y \in[-\pi, \pi]$, such that

$$
\begin{aligned}
\sum_{|j|>1}\left|\check{M}_{11}(t, x, y, j)\right| & =\sum_{|j|>1}\left(1+\frac{\left(j+\frac{x-y}{2 \pi}\right)^{2}}{t}\right)^{-\frac{1}{2}}\left(1+\frac{\left(j+\frac{x-y}{2 \pi}\right)^{2}}{t}\right)^{\frac{1}{2}}\left|\check{M}_{11}(t, x, y, j)\right| \\
& \leqslant C t^{\frac{1}{4}}\left(\int_{-\frac{1}{2}}^{\frac{1}{2}} \sum_{\alpha=0}^{1} t^{-\alpha}\left|\sum_{\ell, \eta \in \mathbb{Z}}\left(\partial_{\sigma}^{\alpha}\left(M_{11}\right)_{\ell \eta}(t, \sigma)\right) \mathrm{e}^{\mathrm{i}(\sigma+\ell) x} \mathrm{e}^{-\mathrm{i}(\sigma+\eta) y}\right|^{2} \mathrm{~d} \sigma\right)^{\frac{1}{2}} \\
& \leqslant C(\alpha) t^{\frac{1}{4}}\left(\int_{-\frac{1}{2}}^{\frac{1}{2}}\left\|\left|M_{11}(t, \sigma)\left\|_{Y_{\infty} \rightarrow Y_{\infty}^{\alpha}}^{2} \mathrm{~d} \sigma+\int_{-\frac{1}{2}}^{\frac{1}{2}}\right\|\right|\left(\partial_{\sigma} M\right)_{11}(t, \sigma)\right\|_{Y_{\infty} \rightarrow Y_{\infty}^{\alpha}}^{2} \mathrm{~d} \sigma\right)^{\frac{1}{2}} \\
& \leqslant C(\alpha) t^{\frac{1}{4}}\left(\frac{1}{t^{2 \alpha}(1+t)^{\frac{5}{2}-2 \alpha}}\right)^{\frac{1}{2}} \\
& \leqslant C(\alpha) \frac{1}{t^{\alpha-\frac{1}{4}}(1+t)^{\frac{5}{4}-\alpha}}, \text { for all } t>0 .
\end{aligned}
$$

Moreover, combining the above two estimates, we have that, for given $\alpha \in\left(\frac{1}{2}, 1\right)$, there exists $C(\alpha)>0$ such that

$$
\sup _{|x|,|y| \leqslant \pi} \sum_{j \in \mathbb{Z}}\left|\check{M}_{11}(t, x, y, j)\right| \leqslant \frac{C(\alpha)}{t^{\alpha}(1+t)^{1-\alpha}} .
$$

Therefore, for any $1 \leqslant q \leqslant p \leqslant \infty, \alpha \in\left(\frac{1}{2}, 1\right)$ and $\underline{\mathbf{W}} \in X_{\text {ch }}^{\perp}$, there exists $C(\alpha)>0$ such that

$$
\left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}} \leqslant \frac{C(\alpha)}{(1+t)^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}} \frac{1}{t^{\alpha}(1+t)^{1-\alpha}}\|\underline{\mathbf{W}}\|_{X_{q}} .
$$

Moreover, we can improve the above estimate for $t$ close to zero. Note that for the Laplacian operator, we have the general $L^{p}-L^{q}$ estimate for all $t>0$. As a perturbation of the Laplacian operator, $\mathscr{M}_{11}$ has the same estimate for sufficiently small $t$, which can be seen by using the variation of constant formula as follows.

$$
\left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}}=\left\|(\mathrm{id}-\mathbf{E} * F) \mathrm{e}^{A_{\mathrm{ch}} t} \underline{\mathbf{W}}\right\|_{X_{p}} \leqslant C\left\|\mathrm{e}^{A_{\mathrm{ch}} t} \underline{\mathbf{W}}\right\|_{X_{p}}=C\left\|\mathrm{e}^{A t} \mathbf{W}\right\|_{L^{p}}
$$

where $\underline{\mathbf{W}}=\left\{\mathbf{W}_{j}(x)\right\}_{j \in \mathbb{Z}}$ and $\mathbf{W}(2 \pi j+x)=\mathbf{W}_{j}(x)$ for all $j \in \mathbb{Z}$ and $x \in[-\pi, \pi]$. We now let $\mathbf{V}(t, x)=\mathrm{e}^{A t} \mathbf{W}(x)$ and have

$$
\mathbf{V}(t)=\mathrm{e}^{D \partial_{x x} t} \mathbf{W}+\int_{0}^{t} \mathrm{e}^{D \partial_{x x}(t-s)} \mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{V}(s) \mathrm{d} s
$$

from which we derive

$$
\sup _{0<t \leqslant T} t^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}\|\mathbf{V}(t)\|_{L^{p}} \leqslant\|\mathbf{W}\|_{L^{q}}+C T^{1-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)} \sup _{0<t \leqslant T} t^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}\|\mathbf{V}(t)\|_{L^{p}}
$$

Taking $T$ sufficiently small such that $C T^{1-\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)} \leqslant \frac{1}{2}$, we obtain

$$
\left\|\mathrm{e}^{A t} \mathbf{W}\right\|_{L^{p}} \leqslant \frac{C}{t^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}}\|\mathbf{W}\|_{L^{q}},
$$

which implies that

$$
\left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}} \leqslant \frac{C}{t^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}}\|\underline{\mathbf{W}}\|_{X_{q}}, \text { for all } 0<t \leqslant T .
$$

Therefore, for any $1 \leqslant q \leqslant p \leqslant \infty$, there exists $C>0$ such that

$$
\left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}} \leqslant \frac{C}{t^{\frac{1}{2}\left(\frac{1}{q}-\frac{1}{p}\right)}} \frac{1}{(1+t)}\|\underline{\mathbf{W}}\|_{X_{q}} .
$$

Remark 4.3 By (4.1),(4.2) and a similar argument as in Proposition 4.2, it is not hard to conclude that, for any $j \in \mathbb{Z}^{+}, 1 \leqslant q \leqslant p \leqslant \infty$ and $(\underline{\theta}, \underline{\mathbf{W}}) \in \ell^{1} \times X_{\mathrm{ch}}^{\perp}$, there exists a positive constant $C$ such that, for all $t>0$,

$$
\left(\begin{array}{ll}
\left\|\delta_{+}^{j} \mathscr{M}_{00}(t) \underline{\theta}\right\|_{\ell^{p}} & \left\|\delta_{+}^{j} \mathscr{M}_{01}(t) \underline{\mathbf{W}}\right\|_{\ell^{p}} \\
\left\|\delta_{+}^{j} \mathscr{M}_{10}(t) \underline{\theta}\right\|_{X_{p}} & \left\|\delta_{+}^{j} \mathscr{M}_{11}(t) \underline{\mathbf{W}}\right\|_{X_{p}} \tag{4.7}
\end{array}\right) .
$$

## 5 Maximal regularity and nonlinear stability

In this section, we prove the main theorem-Theorem 1. To achieve this, we first introduce a Banach space that our argument will be based on. We then collect several maximal regularity results since the normal form system is quasilinear. Based on our normal form and the general $L^{p}-L^{q}$ linear estimates, we can apply a fixed point argument to the variation of constant formula, thus obtaining the nonlinear stability result.

We choose $r \in(4,+\infty)$ and define

$$
Z=\left\{(\underline{\theta}, \underline{\mathbf{W}}) \in C\left((0,+\infty), \ell^{1} \times\left(X_{\mathrm{ch}} \cap \mathscr{T}_{\mathrm{ch}}^{-1}\left(H^{2}\right)\right)\right) \mid\|(\underline{\theta}, \underline{\mathbf{W}})\|_{Z}<\infty\right\}
$$

where

$$
\begin{aligned}
\|(\underline{\theta}, \underline{\mathbf{W}})\|_{Z}= & \sup _{t>0}\|\underline{\theta}(t)\|_{\ell^{1}}+\sup _{t>0}(1+t)^{\frac{1}{2}}\|\underline{\theta}(t)\|_{\ell \infty}+\sup _{t>0}(1+t)^{\frac{5}{4}}\left\|\delta^{2} \underline{\theta}\right\|_{\ell^{2}} \\
& +\sup _{t>0}(1+t)^{\frac{1}{2}}\|\underline{\mathbf{W}}\|_{X_{1}}+\sup _{t>0}(1+t)\|\underline{\mathbf{W}}\|_{X_{\infty}}++\sup _{t>0}(1+t)^{\frac{5}{4}}\left\|\delta_{+} \underline{\mathbf{W}}\right\|_{X_{2}} \\
& +\left(\int_{0}^{\infty}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(t)\right\|_{X_{2}}^{r}\right)^{1 / r}
\end{aligned}
$$

Here we have $\delta^{2}:=\delta_{-} \delta_{+}$, where $\delta_{ \pm}$is defined in (2.20).
Lemma 5.1 (maximal regularity) For any given $T>0$ and $r \in(1,+\infty)$, there exists a positive constant $C$ such that the following holds. If $(\underline{\eta}, \underline{\mathbf{v}}) \in L^{r}\left((0, T), \ell^{2} \times X_{2}\right)$ and if $(\underline{\theta}, \underline{\mathbf{w}})$ satisfies

$$
\binom{\underline{\theta}(t)}{\underline{\mathbf{w}}(t)}=\int_{0}^{t} \mathscr{M}(t-s)\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)} \mathrm{d} s, \quad t \in[0, T]
$$

then

$$
\int_{0}^{T}\left\|\partial_{x x} \underline{\mathbf{w}}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \leqslant C \int_{0}^{T}\left(\|\underline{\eta}(t)\|_{\ell^{2}}+\|\underline{\mathbf{v}}(t)\|_{X_{2}}\right)^{r} \mathrm{~d} t
$$

Proof. The result follows from the standard maximal regularity results on the Laplacian operator and the robustness of maximal regularity with respect to lower order perturbations. To see that, we first recall $\mathscr{M}(t)=\mathrm{e}^{A_{\mathrm{nf}} t}$, where $A_{\mathrm{nf}}$ is defined in (2.22). By [13], maximal regularity holds when we replace $A_{\mathrm{nf}}$ by $A_{0}$, which is defined as

$$
A_{0}=\left(\begin{array}{cc}
0 & 0 \\
0 & D \partial_{x x}
\end{array}\right)
$$

Viewing $A_{\mathrm{nf}}$ as a perturbation of $A_{0}$, we have

$$
\binom{\underline{\theta}(t)}{\underline{\mathbf{w}}(t)}=\int_{0}^{t} \mathscr{M}(t-s)\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)} \mathrm{d} s=\int_{0}^{t} \mathrm{e}^{A_{0}(t-s)}\left(\left(A_{\mathrm{nf}}-A_{0}\right)\binom{\underline{\theta}(s)}{\underline{\mathbf{w}}(s)}+\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)}\right) \mathrm{d} s
$$

Then by the maximal regularity property of $A_{0}$, we obtain

$$
\int_{0}^{T}\left\|\partial_{x x \underline{\mathbf{w}}}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \leqslant C \int_{0}^{T}\left(\left\|\left(A_{\mathrm{nf}}-A_{0}\right)\binom{\underline{\theta}(s)}{\underline{\mathbf{w}}(s)}\right\|_{\ell^{2} \times X_{2}}+\left\|\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)}\right\|_{\ell^{2} \times X_{2}}\right)^{r} \mathrm{~d} s
$$

We observe that, for any $\epsilon>0$, there exists $K(\epsilon)>0$ such that

$$
\left\|\left(A_{\mathrm{nf}}-A_{0}\right)\binom{\underline{\theta}}{\underline{\mathbf{w}}}\right\|_{\ell^{2} \times X_{2}} \leqslant \epsilon\left\|A_{0}\binom{\underline{\theta}}{\underline{\mathbf{w}}}\right\|_{\ell^{2} \times X_{2}}+K(\epsilon)\left\|\binom{\underline{\theta}}{\underline{\mathbf{w}}}\right\|_{\ell^{2} \times X_{2}} .
$$

In addition, it is straightforward to see that

$$
\int_{0}^{T}\left\|\binom{\underline{\theta}(t)}{\underline{\mathbf{w}}(t)}\right\|_{\ell^{2} \times X_{2}}^{r} \mathrm{~d} t \leqslant C \int_{0}^{T}\left\|\binom{\underline{\eta}(t)}{\underline{\mathbf{v}}(t)}\right\|_{\ell^{2} \times X_{2}}^{r} \mathrm{~d} t .
$$

The conclusion follows by combing the above three inequalities and taking $\epsilon$ sufficiently small.

We also prove a corollary which will be useful in the proof of nonlinear stability.
Corollary 5.2 For given $\alpha \in \mathbb{R}$ and $r \in(1, \infty)$, there exists a positive constant $C$ such that, if

$$
\binom{\underline{\theta}(t)}{\underline{\mathbf{w}}(t)}=\int_{t-1}^{t} \mathscr{M}(t-s)\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)} \mathrm{d} s, \quad t \geqslant 1,
$$

then

$$
\int_{1}^{\infty}(1+t)^{\alpha}\left\|\partial_{x x} \underline{\mathbf{w}}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \leqslant C \int_{0}^{\infty}(1+t)^{\alpha}\left(\|\underline{\eta}(t)\|_{\ell^{2}}+\|\underline{\mathbf{v}}(t)\|_{X_{2}}\right)^{r} \mathrm{~d} t .
$$

Proof. We first note that, for $t \in[n, n+1), n \in \mathbb{N} \backslash\{0\}$,

$$
\begin{aligned}
\binom{\underline{\theta}(t)}{\underline{\mathbf{w}}(t)} & =\left(\int_{n-1}^{t}-\int_{n-1}^{t-1}\right) \mathscr{M}(t-s)\binom{\underline{\eta}(s)}{\underline{\mathbf{v}}(s)} \mathrm{d} s \\
& =\left(\int_{0}^{t-n+1}-\int_{0}^{t-n}\right) \mathscr{M}(t-n+1-s)\binom{\underline{\eta}(n-1+s)}{\underline{\mathbf{v}}(n-1+s)} \mathrm{d} s .
\end{aligned}
$$

Applying Lemma 5.1 to the above expression, we obtain

$$
\int_{n}^{n+1}\left\|\partial_{x x} \underline{\mathbf{w}}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \leqslant C \int_{n-1}^{n+1}\left(\|\underline{\eta}(t)\|_{\ell^{2}}+\|\underline{\mathbf{v}}(t)\|_{X_{2}}\right)^{r} \mathrm{~d} t .
$$

The conclusion follows from multiplying both sides with $n^{\alpha} \sim(1+t)^{\alpha}$ and summing over $n \in \mathbb{N} \backslash\{0\}$.

Lemma 5.3 If $\left(\underline{\theta}_{0}, \underline{\mathbf{W}}_{0}\right) \in \ell^{1} \times\left(X_{\mathrm{ch}}^{\perp} \cap \mathscr{T}_{\text {ch }}^{-1}\left(H^{2}\right)\right)$, the solution of the linear system

$$
\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}=\mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}
$$

belongs to $Z$ and there exists a positive constant $C_{1}>0$ such that

$$
\begin{equation*}
\left\|\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}\right\|_{Z} \leqslant C_{1}\left\|\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{1} \times\left(X_{\mathrm{ch}} \cap \mathscr{T}_{\mathrm{ch}}^{-1}\left(H^{2}\right)\right)} . \tag{5.1}
\end{equation*}
$$

Proof. By Proposition 4.2, it is straightforward to see that

$$
\begin{array}{rlrl}
\left\|\mathscr{M}_{00}(t) \underline{\theta}_{0}\right\|_{\ell^{1}} & \leqslant C\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}, & \left\|\mathscr{M}_{01}(t) \underline{\mathbf{W}}_{0}\right\|_{\ell^{1}} \leqslant \frac{C}{(1+t)^{1 / 2}}\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{1}}, \\
\left\|\mathscr{M}_{00}(t) \underline{\theta}_{0}\right\|_{\ell \infty} & \leqslant \frac{C}{(1+t)^{1 / 2}}\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}, & \left\|\mathscr{M}_{01}(t) \underline{\mathbf{W}}_{0}\right\|_{\ell \infty} \leqslant \frac{C}{1+t}\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{1}}, \\
\left\|\delta^{2} \mathscr{M}_{00}(t) \underline{\theta}_{0}\right\|_{\ell^{2}} & \leqslant \frac{C}{(1+t)^{5 / 4}}\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}, & \left\|\delta^{2} \mathscr{M}_{01}(t) \underline{\mathbf{W}}_{0}\right\|_{\ell^{2}} \leqslant \frac{C}{(1+t)^{7 / 4}}\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{1}}, \\
\left\|\mathscr{M}_{10}(t) \underline{\theta}_{0}\right\|_{X_{1}} & \leqslant \frac{C}{(1+t)^{1 / 2}}\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}, & \left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}_{0}\right\|_{X_{1}} \leqslant \frac{C}{1+t}\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{1}}, \\
\left\|\mathscr{M}_{10}(t) \underline{\theta}_{0}\right\|_{X_{\infty}} \leqslant \frac{C}{1+t}\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}, & \left\|\mathscr{M}_{11}(t) \underline{\mathbf{W}}_{0}\right\|_{X_{\infty}} \leqslant \frac{C}{1+t}\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{\infty}} .
\end{array}
$$

Moreover, we have

$$
\begin{aligned}
\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(t)\right\|_{X_{2}} & \leqslant C\left\|\delta_{+} A_{0} \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{2} \times X_{2}} \\
& \leqslant C\left(\left\|\delta_{+} A_{\mathrm{nf}} \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{2} \times X_{2}}+\left\|\delta_{+}\left(A_{\mathrm{nf}}-A_{0}\right) \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{2} \times X_{2}}\right) .
\end{aligned}
$$

We need to show that the two terms on the right hand side of the above inequality decay sufficiently fast. On the one hand, we claim that

$$
\left\|\delta_{+} A_{\mathrm{nf}} \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\underline{\mathbf{W}}}_{0}}\right\|_{\ell^{2} \times X_{2}} \leqslant \frac{C}{(1+t)^{\frac{3}{2}}}\left\|\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}}\right\|_{\ell^{2} \times \mathscr{T}_{\mathrm{ch}}^{-1}\left(H^{2}\right)} \text {, for all } t \geqslant 0
$$

Actually, for $t \in[0,1]$, the above inequality is true since $\delta_{+}$is bounded and

$$
\left\|A_{\mathrm{nf}} \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{2} \times X_{2}} \leqslant C\left\|A_{\mathrm{nf}}\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}\right\|_{\ell^{2} \times X_{2}} \leqslant C\left\|\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}}\right\|_{\ell^{2} \times \mathscr{T}_{\mathrm{ch}}^{-1}\left(H^{2}\right)} .
$$

For $t \in[1, \infty]$, we first point out that, to show $\left\|\left\|A_{\mathrm{nf}} \mathscr{M}(t)=A_{\mathrm{nf}} \mathrm{e}^{A_{\mathrm{nf}} t}\right\|_{\ell^{2} \times X_{2}}\right.$ decays with rate $t^{-1}$ as $t$ goes to $\infty$, we only have to show that the supremum norm of its Fourier-Bloch counterpart $\widehat{A}_{\text {nf }}(\sigma) M(t, \sigma)$ decays with rate $t^{-1}$ as $t$ goes to $\infty$, just as in the scalar heat equation case. This is true by applying the steps in Lemma 3.4 and Lemma 3.6 to $\widehat{A}_{\mathrm{nf}}(\sigma) M(t, \sigma)$. Second, it is straightforward to see that the discrete derivative operator $\delta_{+}$gives an extra $t^{-1 / 2}$ decay, which concludes our justification. On the other hand, we have the explicit expression, using that $\delta_{+}$and $A_{\mathrm{nf}}, A_{0}$ commute,

$$
\delta_{+}\left(A_{\mathrm{nf}}-A_{0}\right) \mathscr{M}(t)\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}=\left(\begin{array}{cc}
0 & \delta_{+} \Gamma \\
A_{\mathrm{ch}} \mathbf{E} * & \mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right)-\mathbf{E} * \delta_{+} \Gamma
\end{array}\right)\binom{\delta_{+} \underline{\theta}(t)}{\delta_{+} \underline{\mathbf{W}}(t)} .
$$

We apply Proposition 4.2 again and obtain

$$
\begin{gathered}
\left\|A_{\mathrm{ch}} \mathbf{E} *\left(\delta_{+} \underline{\theta}(t)\right)\right\|_{\ell^{2}} \leqslant C\left\|\delta^{2} \underline{\theta}(t)\right\|_{\ell^{2}} \leqslant \frac{C}{(1+t)^{\frac{5}{4}}}\left(\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}+\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{2}}\right) \\
\left\|\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right)\left(\delta_{+} \underline{\mathbf{W}}(t)\right)\right\|_{X_{2}} \leqslant C\left\|\delta_{+} \underline{\mathbf{W}}(t)\right\|_{X_{2}} \leqslant \frac{C}{(1+t)^{\frac{5}{4}}}\left(\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}+\left\|\underline{\mathbf{W}}_{0}\right\|_{X_{2}}\right)
\end{gathered}
$$

In addition, recalling that $\Gamma$ is defined in (2.20), we conclude that, for any $\epsilon>0$, there exists $K(\epsilon)>0$ such that

$$
\left\|\delta_{+} \Gamma\left(\delta_{+} \underline{\mathbf{W}}(t)\right)\right\|_{X_{2}} \leqslant C\left\|\partial_{x} \delta_{+} \underline{\mathbf{W}}(t)\right\|_{X_{2}} \leqslant \epsilon\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(t)\right\|_{X_{2}}+K(\epsilon)\left\|\delta_{+} \underline{\mathbf{W}}(t)\right\|_{X_{2}}
$$

Therefore, by choosing $\epsilon$ sufficiently small, we conclude that

$$
\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(t)\right\|_{X_{2}} \leqslant \frac{C}{(1+t)^{\frac{5}{4}}}\left(\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}+\left\|\underline{\mathbf{W}}_{0}\right\|_{\mathscr{T h}^{-1}\left(H^{2}\right)}\right)
$$

which shows that

$$
\left(\int_{0}^{\infty}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t\right)^{1 / r} \leqslant C\left(\left\|\underline{\theta}_{0}\right\|_{\ell^{1}}+\left\|\underline{\mathbf{W}}_{0}\right\|_{\mathscr{T}_{\mathrm{ch}}^{-1}\left(H^{2}\right)}\right)
$$

This proves the lemma.
Lemma 5.4 For $\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}<\varepsilon$, where $\varepsilon$ is sufficiently small $\left(0<\varepsilon \leqslant \varepsilon_{0}\right)$, there exists a positive constant $C_{2} \geqslant 1$ such that

$$
\begin{equation*}
\left\|\int_{0}^{t} \mathscr{M}(t-s)\binom{\mathbf{N}^{\theta}(\underline{\theta}(s), \underline{\mathbf{W}}(s))}{\mathbf{N}^{\mathbf{w}}(\underline{\theta}(s), \underline{\mathbf{W}}(s))} \mathrm{d} s\right\|_{Z} \leqslant C_{2}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2} . \tag{5.2}
\end{equation*}
$$

Moreover, for $\left(\underline{\theta}_{1}, \underline{\mathbf{W}}_{1}\right),\left(\underline{\theta}_{2}, \underline{\mathbf{W}}_{2}\right)$ with their norms in $Z$ smaller than $\varepsilon$, we have

$$
\begin{align*}
& \left\|\int_{0}^{t} \mathscr{M}(t-s)\binom{\mathbf{N}^{\theta}\left(\underline{\theta}_{1}(s), \mathbf{W}_{1}(s)\right)-\mathbf{N}^{\theta}\left(\underline{\theta}_{2}(s), \underline{\mathbf{W}}_{2}(s)\right)}{\mathbf{N}^{\mathbf{w}}\left(\underline{\theta}_{1}(s), \underline{\mathbf{W}}_{1}(s)\right)-\mathbf{N}^{\mathbf{w}}\left(\underline{\theta}_{2}(s), \underline{\mathbf{W}}_{2}(s)\right)} \mathrm{d} s\right\|_{Z} \\
\leqslant & C_{2}\left(\sum_{j=1}^{2}\left\|\left(\underline{\theta}_{j}(t), \underline{\mathbf{W}}_{j}(t)\right)\right\|_{Z}\right)\left\|\left(\underline{\theta}_{1}(t)-\underline{\theta}_{2}(t), \underline{\mathbf{W}}_{1}(t)-\underline{\mathbf{W}}_{2}(t)\right)\right\|_{Z} . \tag{5.3}
\end{align*}
$$

Proof. We start with proving the estimate (5.2). The proof is fairly straightforward. The strategy is to use estimates for the linear part $\mathscr{M}(t)$ in Proposition 4.2, the estimates for the nonlinear terms in Lemma 6.2 from the appendix, and the maximal regularity estimates in Lemma 5.1, Corollary 5.2. For simplicity, we denote

$$
\mathbf{N}^{\theta}(s)=\mathbf{N}^{\theta}(\underline{\theta}(s), \underline{\mathbf{W}}(s)), \quad \mathbf{N}^{\mathbf{w}}(s)=\mathbf{N}^{\mathbf{w}}(\underline{\theta}(s), \underline{\mathbf{W}}(s)) .
$$

By Lemma 6.2, we have that

$$
\begin{align*}
& \left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}} \leqslant \frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}+\frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}, \\
& \left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{2}} \leqslant \frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}+\frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}},  \tag{5.4}\\
& \left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}} \leqslant \frac{C}{1+s}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}+\frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}, \\
& \left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{2}} \leqslant \frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}+\frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}} .
\end{align*}
$$

We also exploit the linear estimates from Proposition 4.2 and obtain the following estimates.

$$
\begin{align*}
& N_{1}=\left\|\int_{0}^{t} \mathscr{M}_{00}(t-s) \mathbf{N}^{\theta}(s) \mathrm{d} s\right\|_{\ell^{1}} \leqslant C \int_{0}^{t}\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}} \mathrm{~d} s, \\
& N_{2}=\left\|\int_{0}^{t} \mathscr{M}_{01}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s\right\|_{\ell^{1}} \leqslant C \int_{0}^{t} \frac{\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}}}{(1+t-s)^{\frac{1}{2}}} \mathrm{~d} s, \\
& N_{3}=(1+t)^{\frac{1}{2}}\left\|\int_{0}^{t} \mathscr{M}_{00}(t-s) \mathbf{N}^{\theta}(s) \mathrm{d} s\right\|_{\ell^{\infty}} \leqslant C(1+t)^{\frac{1}{2}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}}}{(1+t-s)^{\frac{1}{2}}} \mathrm{~d} s, \\
& N_{4}=(1+t)^{\frac{1}{2}}\left\|\int_{0}^{t} \mathscr{M}_{01}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s\right\|_{\ell \infty} \leqslant C(1+t)^{\frac{1}{2}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}}}{1+t-s} \mathrm{~d} s, \\
& N_{5}=(1+t)^{\frac{5}{4}}\left\|\delta^{2} \int_{0}^{t} \mathscr{M}_{00}(t-s) \mathbf{N}^{\theta}(s) \mathrm{d} s\right\|_{\ell^{2}} \leqslant C(1+t)^{\frac{5}{4}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}}}{(1+t-s)^{\frac{5}{4}}} \mathrm{~d} s, \\
& N_{6}=(1+t)^{\frac{5}{4}}\left\|\delta^{2} \int_{0}^{t} \mathscr{M}_{01}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s\right\|_{\ell^{2}} \leqslant C(1+t)^{\frac{5}{4}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{2}}}{(1+t-s)^{\frac{3}{2}}} \mathrm{~d} s,  \tag{5.5}\\
& N_{7}=(1+t)^{\frac{1}{2}}\left\|\int_{0}^{t} \mathscr{M}_{10}(t-s) \mathbf{N}^{\theta}(s) \mathrm{d} s\right\|_{X_{1}} \leqslant C(1+t)^{\frac{1}{2}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}}}{(1+t-s)^{\frac{1}{2}}} \mathrm{~d} s, \\
& N_{8}=(1+t)^{\frac{1}{2}}\left\|\int_{0}^{t} \mathscr{M}_{11}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s\right\|_{X_{1}} \leqslant C(1+t)^{\frac{1}{2}} \int_{0}^{t} \frac{\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}}}{1+t-s} \mathrm{~d} s, \\
& N_{12}=(1+t)\left\|\int_{0}^{t} \mathscr{M}_{11}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s\right\|_{X_{\infty}} \leqslant C(1+t) \int_{0}^{t} \frac{\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}}}{(1+t-s)(t-s)^{\frac{1}{2}}} \mathrm{~d} s .
\end{align*}
$$

At this point, we substitute (5.4) into (5.5), estimate the resulting integrals, and find

$$
\begin{equation*}
N_{j} \leqslant C\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}, \text { for all } 1 \leqslant j \leqslant 12 . \tag{5.6}
\end{equation*}
$$

The calculations establishing the estimates for $N_{1}, \ldots, N_{11}$ are based on the following elementary integral estimates.

$$
\int_{0}^{t} \frac{1}{(1+t-s)^{\alpha}} \frac{1}{(1+s)^{\beta}} \mathrm{d} s \leqslant \frac{C}{(1+t)^{\alpha}} \int_{0}^{\frac{t}{2}} \frac{1}{(1+s)^{\beta}} \mathrm{d} s+\frac{C}{(1+t)^{\beta}} \int_{0}^{\frac{t}{2}} \frac{1}{(1+s)^{\alpha}} \mathrm{d} s
$$

For the estimate on $N_{12}$, we just need to show that the following integral expression

$$
h(t)=(1+t)\left(\int_{0}^{t} \frac{1}{(1+t-s)(t-s)^{\frac{1}{2}}} \frac{1}{1+s} \mathrm{~d} s+\left(\int_{0}^{t}\left(\frac{1}{(1+t-s)(t-s)^{\frac{1}{2}}} \frac{1}{(1+s)^{\frac{5}{4}}}\right)^{\frac{r}{r-1}} \mathrm{~d} s\right)^{1-\frac{1}{r}}\right)
$$

has a uniform upper bound for $t \in(0, \infty)$. First, for all $t \in(0,1]$, there exists $C>0$ such that,

$$
h(t) \leqslant 2\left(\int_{0}^{1}(t-s)^{-\frac{1}{2}} \mathrm{~d} s+\left(\int_{0}^{1}(t-s)^{-\frac{r}{2(r-1)}} \mathrm{d} s\right)^{1-\frac{1}{r}}\right) \leqslant C
$$

Second, for $t \in[1, \infty)$, we have

$$
\begin{aligned}
(1+t) \int_{0}^{t} \frac{1}{(1+t-s)(t-s)^{\frac{1}{2}}} \frac{1}{1+s} \mathrm{~d} s & \leqslant \frac{C}{(1+t)^{\frac{1}{2}}} \int_{0}^{\frac{t}{2}} \frac{1}{1+s} \mathrm{~d} s+C \int_{\frac{t}{2}}^{t} \frac{1}{(1+t-s)(t-s)^{\frac{1}{2}}} \mathrm{~d} s \\
& \leqslant C\left(1+\int_{0}^{\infty} \frac{1}{(1+s) s^{\frac{1}{2}}} \mathrm{~d} s\right) \leqslant C
\end{aligned}
$$

Similar arguments show that the second part of $h(t)$ is also uniformly bounded on $[1, \infty)$.
The estimates on $N_{1}, \ldots, N_{12}$ bound the $Z$-norm of the left-hand side of (5.2), except for the maximal regularity component. Thus it remains to show that

$$
\begin{equation*}
\left(\int_{0}^{\infty}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \mathscr{W}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t\right)^{\frac{1}{r}} \leqslant C\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2}, \tag{5.7}
\end{equation*}
$$

where

$$
\mathscr{W}(t)=\int_{0}^{t} \mathscr{M}_{10}(t-s) \mathbf{N}^{\theta}(s)+\mathscr{M}_{11}(t-s) \mathbf{N}^{\mathbf{w}}(s) \mathrm{d} s
$$

For $t \in[0,1]$, by maximal regularity in Lemma 5.1, we have

$$
\begin{align*}
\int_{0}^{1}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \mathscr{W}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t & \leqslant C \int_{0}^{1}\left\|\partial_{x x} \mathscr{W}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \\
& \leqslant C \int_{0}^{1}\left(\left\|\mathbf{N}^{\theta}(t)\right\|_{\ell^{2}}+\left\|\mathbf{N}^{\mathbf{w}}(t)\right\|_{X_{2}}\right)^{r} \mathrm{~d} t  \tag{5.8}\\
& \leqslant C\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2 r}
\end{align*}
$$

For $t \in[1, \infty)$, we split $\mathscr{W}$ into two parts, that is,

$$
\mathscr{W}=\left(\int_{0}^{t-1}+\int_{t-1}^{t}\right) \mathscr{M}_{10}(t-s) \mathbf{N}^{\theta}(s)+\mathscr{M}_{11}(t-s) \mathbf{N}^{\mathrm{w}}(s) \mathrm{d} s=\mathscr{W}_{1}+\mathscr{W}_{2}
$$

By Corollary 5.2, we have

$$
\begin{aligned}
\int_{1}^{\infty}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \mathscr{W}_{2}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t & \leqslant C \int_{1}^{\infty}(1+t)^{r}\left\|\partial_{x x} \mathscr{W}_{2}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \\
& \leqslant C \int_{0}^{\infty}(1+t)^{r}\left(\left\|\mathbf{N}^{\theta}(t)\right\|_{\ell^{2}}+\left\|\mathbf{N}^{\mathbf{w}}(t)\right\|_{X_{2}}\right)^{r} \mathrm{~d} t \\
& \leqslant C\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2 r}
\end{aligned}
$$

By similar arguments as in Lemma 5.3 and with the condition that $t-s>1$, we can show that

$$
\begin{aligned}
& \left\|\delta_{+} \partial_{x x}\left(\mathscr{M}_{10}(t-s) \mathbf{N}^{\theta}(s)+\mathscr{M}_{11}(t-s) \mathbf{N}^{\mathbf{w}}(s)\right)\right\|_{X_{2}} \leqslant \frac{C}{(1+t-s)^{\frac{5}{4}}}\left(\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}}+\left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{2}}\right) \\
\leqslant & \frac{C}{(1+t-s)^{\frac{5}{4}}(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}\left(\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}+(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}\right) .
\end{aligned}
$$

As a result, we obtain

$$
\begin{aligned}
& (1+t)\left\|\delta_{+} \partial_{x x} \mathscr{W}_{1}(t)\right\|_{X_{2}} \leqslant(1+t) \int_{0}^{t-1}\left\|\delta_{+} \partial_{x x}\left(\mathscr{M}_{10}(t-s) \mathbf{N}^{\theta}(s)+\mathscr{M}_{11}(t-s) \mathbf{N}^{\mathbf{w}}(s)\right)\right\|_{X_{2}} \mathrm{~d} s \\
\leqslant & (1+t) \int_{0}^{t-1} \frac{C}{(1+t-s)^{\frac{5}{4}}(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}\left(\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}+(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}\right) \mathrm{d} s \\
\leqslant & \frac{C}{(1+t)^{\frac{1}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2 r}
\end{aligned}
$$

which immediately implies that

$$
\int_{1}^{\infty}(1+t)^{r}\left\|\delta_{+} \partial_{x x} \mathscr{W}_{1}(t)\right\|_{X_{2}}^{r} \mathrm{~d} t \leqslant C\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Z}^{2 r}
$$

Together with (5.8), this establishes (5.7) and concludes the proof.
In a completely analogous fashion, one establishes the Lipshitz estimates.

We now prove our main theorem.
Proof of Theorem 1. The proof is a fixed-point-theorem argument. We first recall the variation of constant formula,

$$
\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}=\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}+\int_{0}^{t} \mathscr{M}(t-s)\binom{\mathbf{N}^{\theta}(\underline{\theta}(s), \underline{\mathbf{W}}(s))}{\mathbf{N}^{\mathbf{w}}(\underline{\theta}(s), \underline{\mathbf{W}}(s))} \mathrm{d} s .
$$

Let $\mathscr{P}$ be the right-hand side of the formula, that is

$$
\mathscr{P}\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}=\binom{\underline{\theta}_{0}}{\underline{\mathbf{W}}_{0}}+\int_{0}^{t} \mathscr{M}(t-s)\binom{\mathbf{N}^{\theta}(\underline{\theta}(s), \underline{\mathbf{W}}(s))}{\mathbf{N}^{\mathbf{w}}(\underline{\theta}(s), \underline{\mathbf{W}}(s))} \mathrm{d} s .
$$

Assume that now the initial value is sufficiently small, that is, for some small $\epsilon>0$,

$$
\left\|\left(\underline{\theta}_{0}, \underline{\mathbf{W}}_{0}\right)\right\|_{\ell^{1} \times\left(Z \cap \mathscr{T}^{-1}\left(H^{2}\right)\right)} \leqslant \epsilon
$$

If $(\underline{\theta}(t), \underline{\mathbf{W}}(t)) \in Z$ with norm smaller than $\varepsilon$, we know that

$$
\mathscr{P}\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)} \in Z .
$$

By Lemma 5.3 and 5.4, we have that

$$
\begin{equation*}
\left\|\mathscr{P}\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}\right\|_{Z} \leqslant C_{1} \epsilon+C_{2}\left\|\binom{\underline{\theta}(t)}{\underline{\mathbf{W}}(t)}\right\|_{Z}^{2} \tag{5.9}
\end{equation*}
$$

Moreover, we have

$$
\begin{equation*}
\left\|\mathscr{P}\binom{\underline{\theta}_{1}(t)}{\underline{\mathbf{W}}_{1}(t)}-\mathscr{P}\binom{\underline{\theta}_{2}(t)}{\underline{\mathbf{W}}_{2}(t)}\right\|_{Z} \leqslant C_{2}\left(\sum_{j=1}^{2}\left\|\left(\underline{\theta}_{j}(t), \underline{\mathbf{W}}_{j}(t)\right)\right\|_{Z}\right)\left\|\left(\underline{\theta}_{1}(t)-\underline{\theta}_{2}(t), \underline{\mathbf{W}}_{1}(t)-\underline{\mathbf{W}}_{2}(t)\right)\right\|_{Z} \tag{5.10}
\end{equation*}
$$

We denote $B=\left\{\underline{\mathbf{V}} \in Z \mid\|\underline{\mathbf{V}}\|_{Z} \leqslant R\right\}$, where $R=\min \left(2 C_{1} \epsilon, \varepsilon\right)$. We now take $\epsilon>0$ small enough so that $2 C_{2} R<1$ and readily conclude, based on (5.9) and (5.10), that $\mathscr{P}(B) \subset B$ and that $\mathscr{P}$ is a strict contraction in $B$. By Banach's fixed point theorem, there is a unique fixed point of $\mathscr{P}$ in $B$, denoted as $(\underline{\theta}(t), \underline{\mathbf{W}}(t))$. Then $(\underline{\theta}(t), \underline{\mathbf{W}}(t))$ is a global solution of $(2.22)$, and if we return to the original variables, we obtain a global solution of (1.2) which satisfies the decay estimate in Theorem 1. This concludes the proof.

## 6 Appendix

### 6.1 Estimates on nonlinear terms

In this section, we derive the estimates on the nonlinear terms $\mathbf{N}^{\theta}$ and $\mathbf{N}^{\mathbf{w}}$ in our normal form (2.21).
Lemma 6.1 For $\|\underline{\mathbf{W}}\|_{X_{\text {ch }}},\|\underline{\theta}\|_{\ell^{1}}<\varepsilon$, where $\varepsilon$ is sufficiently small $\left(0<\varepsilon \leqslant \varepsilon_{0}\right)$, there exists a nondecreasing function $C(\varepsilon)>0$ such that, for all $1 \leqslant p \leqslant \infty$, the nonlinear terms in system (2.21) have the following estimates.

$$
\begin{align*}
\left|\mathbf{N}_{j}^{\theta}\right| \leqslant & C(\varepsilon)\left[\sum_{k=-1}^{0}\left|\left(\delta_{+} \underline{\theta}\right)_{j+k}\right|^{2}+\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|^{3}\right)\left(\sum_{k=-1}^{0}\left|\left(\delta_{+} \underline{\theta}\right)_{j+k}\right|\right)\right. \\
& +\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)\left(\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|+\left|\left(\delta_{+} \partial_{x} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|\right) \\
& \left.+\left\|\mathbf{W}_{j}\right\|_{L^{p}}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|+\left\|\mathbf{W}_{j}^{2}\right\|_{L^{p}}\right],  \tag{6.1}\\
\left\|\mathbf{N}_{j}^{\mathbf{W}}\right\|_{L^{p}} \leqslant & C(\varepsilon)\left[\left(\sum_{k=-1}^{0}\left|\left(\delta_{+} \underline{\theta}\right)_{j+k}\right|\right)\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)+\left|\theta_{j}\right|| | \mathbf{W}_{j} \|_{L^{p}}\right. \\
& +\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)\left(\sum_{k=-1}^{1}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j+k}(-\pi)\right|+\left|\left(\delta_{+} \partial_{x} \underline{\mathbf{W}}\right)_{j+k}(-\pi)\right|\right) \\
& \left.+\left\|\mathbf{W}_{j}\right\|_{L^{p}}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|+\left\|\mathbf{W}_{j}^{2}\right\|_{L^{p}}+\left|\mathbf{N}_{j}^{\theta}\right|+\left|\mathbf{N}_{j+1}^{\theta}\right|+\left|\mathbf{N}_{j-1}^{\theta}\right|\right] .
\end{align*}
$$

Proof. We point out that throughout the proof, we repeatedly exploit the fact that the $L^{2}$ scalar product of an even function and an odd function are zero. We also recall that $\mathbf{u}_{\star}$ is even and $\mathbf{u}_{\mathrm{ad}}$ is odd. By equations $(2.9),(2.10)$ and (2.21), we obtain

$$
\begin{aligned}
\mathbf{N}_{j}^{\theta} & =I_{j}+\left(I I_{j}+I I I_{j}+I V_{j}+V_{j}\right) \mathscr{S}_{j} \text { and } \\
\mathbf{N}_{j}^{\mathrm{w}} & =\left(\mathrm{id}-\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}}\right)^{-1}\left(V I_{j}+V I I_{j}+V I I I_{j}+I X_{j}+\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}} X_{j}\right), \text { where } \\
\mathscr{S}_{j} & =\left(-1+\left\langle\mathbf{W}_{j}(x)+\mathbf{H}_{j}(x), \mathbf{u}_{\mathrm{ad}}^{\prime}\left(x-\theta_{j}\right)\right\rangle\right)^{-1} ; \\
\mathbf{G}_{j} & =\mathbf{G}\left(\theta_{j}, \mathbf{W}_{j}\right)=\left\langle\mathbf{W}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle \psi\left(x-\theta_{j}\right) ; \\
I_{j} & =\left(-\mathscr{S}_{j}-1\right)\left(\delta_{+} \Gamma W\right)_{j} ; \\
I I_{j} & =-\left(\mathbf{W}_{j+1}(-\pi)-\mathbf{W}_{j}(-\pi), D\left(\mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)\right) ; \\
I I I_{j} & =\left(\partial_{x} \mathbf{W}_{j+1}(-\pi)-\partial_{x} \mathbf{W}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)\right) ; \\
I V_{j} & =\left(\partial_{x} \mathbf{H}_{j}(\pi)-\partial_{x} \mathbf{H}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)\right)-\left(\mathbf{H}_{j}(\pi)-\mathbf{H}_{j}(-\pi), D \mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)\right) ; \\
V_{j} & =\left\langle\tilde{g}\left(\theta_{j}, \mathbf{W}_{j}+\mathbf{H}_{j}\right), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)\right\rangle ; \\
V I_{j} & =A\left(\mathbf{H}_{j}-(\underline{\mathbf{E}} * \underline{\theta})_{j}\right) ; \\
V I I_{j} & =-\left(\left(\dot{\mathbf{H}}_{j}-\mathbf{u}_{\star}^{\prime}\left(x-\theta_{j}\right) \dot{\theta}_{j}\right)-(\underline{\mathbf{E}} * \underline{\theta})_{j}+\left\langle\dot{\mathbf{W}}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle \psi\left(x-\theta_{j}\right)\right) ; \\
V I I I_{j} & =\left(\underline{\mathbf{E}} *\left(\delta_{+} \Gamma \underline{\mathbf{W}}-\underline{\dot{\theta}}\right)\right)_{j} ; \\
I X_{j} & =\tilde{g}\left(\theta_{j}, \mathbf{W}_{j}+\mathbf{H}_{j}\right)+\left[\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\left(x-\theta_{j}\right)\right)-\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}(x)\right)\right]\left(\mathbf{W}_{j}+\mathbf{H}_{j}\right) ; \\
X_{j} & =A(\underline{\mathbf{E}} * \underline{\theta})_{j}+A \mathbf{W}_{j}-\left(\underline{\mathbf{E}} * \delta_{+} \Gamma \underline{\mathbf{W}}\right)_{j} .
\end{aligned}
$$

We recall here that $\underline{\mathbf{E}}$ is defined in (2.18) and point out that the term in $V I I_{j}$ involving $\dot{\mathbf{W}}_{j}$ in fact cancels with a contribution from $\dot{\mathbf{H}}_{j}$. We now prove the estimate of $\mathbf{N}_{j}^{\theta}$.
Estimate on $I_{j}:\left|I_{j}\right| \leqslant C(\varepsilon)\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|+\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right)\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|$.
We first recall that $\mathbf{H}_{j}$ is defined in (2.10) and (2.13). We claim that the number $c_{j}$, appearing in the definition of $\mathbf{H}_{j}^{2}$ as in (2.15) and (2.16), can be estimated as

$$
\left|c_{j}\right| \leqslant C(\varepsilon)\left[\left|\left(\delta^{2} \underline{\theta}\right)_{j}\right|+\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right) \sum_{k=-1}^{1} \theta_{j+k}+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right]
$$

where we use notation $\delta^{2}=\delta_{+} \delta_{-}$. In fact, we have

$$
\begin{aligned}
\mid\left\langle\phi ( x ) \left(\mathbf{u}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)\right.\right. & \left.\left.-\mathbf{u}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle \mid
\end{aligned}
$$

We also have $\left|\mathbf{H}_{j}(x)\right| \leqslant C(\varepsilon)\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right)$, from which we obtain the estimate.
Estimate on $I I_{j}:\left|I I_{j}\right| \leqslant C\left|\theta_{j}\right|^{2}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|$.
This is straightforward.
Estimate on $I I I_{j}:\left|I I I_{j}\right| \leqslant C\left|\theta_{j}\right|\left|\left(\delta_{+} \partial_{x} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|$.
This is straightforward.

Estimate on $I V_{j}:\left|I V_{j}\right| \leqslant C\left[\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|^{2}+\mid\left(\left.\delta_{-} \underline{\theta}_{j}\right|^{2}+\left|\left(\delta_{+} \underline{\theta}\right)_{j}+\left(\delta_{-} \underline{\theta}\right)_{j}\right|\left(\left|\theta_{j+1}\right|^{3}+\left|\theta_{j}\right|^{3}+\left|\theta_{j-1}\right|^{3}\right)\right]\right.$.
We first simplify $I V_{j}$ and obtain

$$
I V_{j}=\frac{1}{2}\left(\mathbf{u}_{\star}^{\prime}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}^{\prime}\left(\pi-\theta_{j-1}\right), D \mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)\right)-\frac{1}{2}\left(\mathbf{u}_{\star}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}\left(\pi-\theta_{j-1}\right), D \mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)\right) .
$$

Then, it is not hard to see that

$$
\begin{aligned}
& \left|\frac{1}{2}\left(\mathbf{u}_{\star}^{\prime}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}^{\prime}\left(\pi-\theta_{j-1}\right), D \mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)\right)-\frac{1}{2}\left(\mathbf{u}_{\star, \theta \theta}(\pi)\left(\theta_{j-1}-\theta_{j+1}\right),-D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi) \theta_{j}\right)\right| \\
\leqslant & C\left(\left|\theta_{j}\right|\left|\theta_{j+1}^{3}-\theta_{j-1}^{3}\right|+\left|\theta_{j}\right|^{3}\left|\theta_{j+1}-\theta_{j-1}\right|\right), \\
& \left|\frac{1}{2}\left(\mathbf{u}_{\star}\left(\pi-\theta_{j+1}\right)-\mathbf{u}_{\star}\left(\pi-\theta_{j-1}\right), D \mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)\right)-\frac{1}{2}\left(\frac{1}{2} \mathbf{u}_{\star, \theta \theta}(\pi)\left(\theta_{j+1}^{2}-\theta_{j-1}^{2}\right), D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)\right| \\
\leqslant & C\left(\left|\theta_{j+1}^{4}-\theta_{j-1}^{4}\right|+\left|\theta_{j}\right|^{2}\left|\theta_{j+1}^{2}-\theta_{j-1}^{2}\right|\right), \\
& \frac{1}{2}\left(\mathbf{u}_{\star, \theta \theta}(\pi)\left(\theta_{j-1}-\theta_{j+1}\right),-D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi) \theta_{j}\right)-\frac{1}{2}\left(\frac{1}{2} \mathbf{u}_{\star, \theta \theta}(\pi)\left(\theta_{j+1}^{2}-\theta_{j-1}^{2}\right), D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right) \\
= & \frac{1}{4}\left(\mathbf{u}_{\star, \theta \theta}(\pi), D \mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)\left[\left(\delta_{-} \underline{\theta}\right)_{j}^{2}-\left(\delta_{+} \underline{\theta}\right)_{j}^{2}\right],
\end{aligned}
$$

which establishes the estimate on $I V_{j}$ as claimed.
Estimate on $V_{j}:\left|V_{j}\right| \leqslant C(\varepsilon)\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|^{2}+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|^{2}+\left\|\mathbf{W}_{j}^{2}\right\|_{L^{p}}\right)$.
Noting that $\left|V_{j}\right| \leqslant C(\varepsilon)\left\|\left(\mathbf{W}_{j}+\mathbf{H}_{j}\right)^{2}\right\|_{L^{p}}$ and applying the estimate of $\mathbf{H}_{j}$ to the inequality lead to the above estimate.

Estimate on $\mathscr{S}_{j}:\left|\mathscr{S}_{j}\right| \leqslant C(\varepsilon)$.
This is straightforward.
Combining our estimates of $I_{j}-V_{j}$ and $\mathscr{S}_{j}$, we obtain the first inequality in (6.1).
Now, we have to show that the estimate of $\mathbf{N}_{j}^{\mathbf{w}}$ in (6.1) is true.
Estimate on $V I_{j}$ :

$$
\left|V I_{j}\right| \leqslant C(\varepsilon)\left[\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right) \sum_{k=-1}^{1}\left|\theta_{j+k}\right|+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right] .
$$

First, for $f 2 \pi$-periodic and smooth, we have

$$
\left|f\left(x-\theta_{1}\right)-f\left(x-\theta_{2}\right)-f^{\prime}(x)\left(\theta_{2}-\theta_{1}\right)\right| \leqslant C\left(\left|\theta_{2}-\theta_{1}\right|^{2}+\left|\theta_{2}\right|\left|\theta_{2}-\theta_{1}\right|\right)
$$

If in addition, $f$ is odd, we have

$$
\left|f\left(\theta_{1}\right)-f\left(\theta_{2}\right)-f^{\prime}(0)\left(\theta_{1}-\theta_{2}\right)\right| \leqslant C\left|\theta_{2}^{3}-\theta_{1}^{3}\right| .
$$

The latter implies that

$$
\left|c_{j}-\frac{1}{4}\left(\delta^{2} \underline{\theta}\right)_{j}\right| \leqslant C(\varepsilon)\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|^{2}+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|^{2}+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right) .
$$

Moreover, by the former inequality, we have

$$
\begin{aligned}
\left|V I_{j}\right| & \leqslant C\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|^{2}+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|^{2}+\left|\theta_{j}\right|\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\theta_{j}\right|\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right)+\left|c_{j} A \psi\left(x-\theta_{j}\right)-\frac{1}{4}\left(\delta^{2} \underline{\theta}\right)_{j} A \psi(x)\right| \\
& \leqslant C(\varepsilon)\left[\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right) \sum_{k=-1}^{1}\left|\theta_{j+k}\right|+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right] .
\end{aligned}
$$

## Estimate on $V I I_{j}$ :

$$
\left|V I I_{j}\right| \leqslant C\left[\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)\left(\sum_{k=-1}^{1}\left|\dot{\theta}_{j+k}\right|\right)+\left|\dot{\theta}_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right]
$$

Noting that $(\underline{\mathbf{E}} * \underline{\theta})_{j}$ is the linear part of $\mathbf{H}_{j}+\mathbf{u}_{\star}\left(x-\theta_{j}\right)-\mathbf{u}_{\star}(x)$ and there is no term invorling $\dot{\mathbf{W}}_{j}$ in $V I I_{j}$, we have

$$
\left|V I I_{j}\right| \leqslant C\left(\left|\theta_{j+1}\right|\left|\dot{\theta}_{j+1}\right|+\left|\theta_{j-1}\right|\left|\dot{\theta}_{j-1}\right|+\left|\theta_{j}\right|\left|\dot{\theta}_{j}\right|\right)+\left|c_{j} \psi^{\prime}\left(x-\theta_{j}\right) \dot{\theta}_{j}\right|+\left|\frac{1}{4}\left(\delta^{2} \underline{\underline{\theta}}\right)_{j} \psi(x)-\tilde{\tilde{c}}_{j} \psi\left(x-\theta_{j}\right)\right|,
$$

where $\tilde{c}_{j}=\dot{c}_{j}+\left\langle\dot{\mathbf{W}}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle$.
First, we note that

$$
\left|c_{j} \psi^{\prime}\left(x-\theta_{j}\right) \dot{\theta}_{j}\right| \leqslant C\left|\dot{\theta}_{j}\right|\left[\left|\left(\delta^{2} \underline{\theta}\right)_{j}\right|+\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right) \sum_{k=-1}^{1} \theta_{j+k}+\left|\theta_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right] .
$$

Moreover, we claim that

$$
\begin{aligned}
&\left|\tilde{c}_{j}\right| \leqslant C\left[\left|\left(\delta_{+} \underline{\dot{\theta}}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\dot{\theta}}\right)_{j}\right|+\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)\left(\sum_{k=-1}^{1}\left|\dot{\theta}_{j+k}\right|\right)+\left|\dot{\theta}_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right], \\
&\left|\tilde{c}_{j}-\frac{1}{4}\left(\delta^{2} \dot{\underline{\theta}}\right)_{j}\right| \leqslant C\left[\left(\sum_{k=-1}^{1}\left|\theta_{j+k}\right|\right)\left(\sum_{k=-1}^{1}\left|\dot{\theta}_{j+k}\right|\right)+\left|\dot{\theta}_{j}\right|\left\|\mathbf{W}_{j}\right\|_{L^{p}}\right] .
\end{aligned}
$$

In fact, we have

$$
\begin{aligned}
& \left|\left\langle\phi(x)\left(\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)-\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle\right| \leqslant C\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|\left|\left(\delta_{+} \underline{\dot{\theta}}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\left|\left(\delta_{-} \underline{\dot{\theta}}\right)_{j}\right|\right), \\
& \left|\left\langle\phi^{\prime}\left(x+\theta_{j}\right) \dot{\theta}_{j}\left(\mathbf{u}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)-\mathbf{u}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle\right| \leqslant C\left|\dot{\theta}_{j}\right|\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\right), \\
& \left|\left\langle\left(\phi\left(x+\theta_{j}\right)-\phi(x)\right)\left(\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)-\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle\right| \leqslant C\left|\theta_{j}\right|\left(\left|\left(\delta_{+} \underline{\dot{\theta}}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\dot{\theta}}\right)_{j}\right|\right), \\
& \left|\left\langle\left(\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)+\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle\right| \leqslant C\left(\left|\left(\delta_{+} \underline{\dot{\theta}}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\dot{\theta}}\right)_{j}\right|\right), \\
& \left|\left\langle\left(\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j+1}\right)+\dot{\mathbf{u}}_{\star}\left(x+\theta_{j}-\theta_{j-1}\right)\right), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle+\delta^{2} \underline{\dot{\theta}}_{j}\right| \leqslant C\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|\left|\left(\delta_{+} \underline{\dot{\theta}}\right)_{j}\right|+\left|\left(\delta_{-} \underline{\theta}\right)_{j}\right|\left|\left(\delta_{-} \underline{\dot{\theta}}\right)_{j}\right|\right),
\end{aligned}
$$

which establishes the claim and thus the estimate on $V I I_{j}$.
Estimate on $V I I I_{j}$ :

$$
\left|V I I I_{j}\right| \leqslant C\left(\left|\mathbf{N}_{j}^{\theta}\right|+\left|\mathbf{N}_{j+1}^{\theta}\right|+\left|\mathbf{N}_{j-1}^{\theta}\right|\right)
$$

The calculation is straightforward using the expressions for $\mathbf{K}_{j}$ and $\dot{\theta}_{j}$.
Estimate on $I X_{j}$ :

$$
\left|I X_{j}\right| \leqslant C(\varepsilon)\left[\left(\sum_{k=0}^{1}\left|\left(\delta_{-} \underline{\theta}\right)_{j+k}\right|\right)\left(\sum_{k=-1}^{1} \mid \theta_{j+k}\right)++\left|\theta_{j}\right|\left|\mathbf{W}_{j}\right|+\left|\theta_{j}\right|^{2}\left\|\mathbf{W}_{j}\right\|_{L^{p}}+\left|\mathbf{W}_{j}\right|^{2}\right]
$$

The calculation is straightforward using the estimate on $\mathbf{H}_{j}$.
Estimate on $\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}} X_{j}$ :

$$
\left|\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}} X_{j}\right| \leqslant C(\varepsilon)\left(\left|\theta_{j}\right| \sum_{k=-1}^{1}\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j+k}\right|+\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j+k}(-\pi)\right|\right)+\left|\left\langle A \mathbf{W}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle\right|\right)
$$

Integrating by parts, we have

$$
\begin{aligned}
\left\langle A \mathbf{W}_{j}(x), \mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right\rangle= & \left\langle\mathbf{W}_{j}(x), A^{*}\left(\mathbf{u}_{\mathrm{ad}}\left(x-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(x)\right)\right\rangle+ \\
& \left(\partial_{x} \mathbf{W}_{j+1}(-\pi)-\partial_{x} \mathbf{W}_{j}(-\pi), D\left(\mathbf{u}_{\mathrm{ad}}\left(\pi-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}(\pi)\right)\right)- \\
& \left(\mathbf{W}_{j+1}(-\pi)-\mathbf{W}_{j}(-\pi), D\left(\mathbf{u}_{\mathrm{ad}}^{\prime}\left(\pi-\theta_{j}\right)-\mathbf{u}_{\mathrm{ad}}^{\prime}(\pi)\right)\right) .
\end{aligned}
$$

Thereofore, we have

$$
\left|\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}} X_{j}\right| \leqslant C(\varepsilon)\left|\theta_{j}\right|\left(\sum_{k=-1}^{1}\left(\left|\left(\delta_{+} \underline{\theta}\right)_{j+k}\right|+\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j+k}(-\pi)\right|\right)+\left\|\mathbf{W}_{j}\right\|_{L^{p}}+\left|\left(\delta_{+} \partial_{x} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|\right) .
$$

Estimate on $\left(\mathrm{id}-\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}}\right)^{-1}$ : For any $\underline{\theta} \in \ell^{\infty}$ and $p \in[1, \infty]$, there exists a constant $C>0$ such that

$$
\left\|\left\|\left(\mathrm{id}-\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}}\right)^{-1}\right\|\right\|_{L^{p}} \leqslant C
$$

Combining estimates on $V I_{j}$ to $I X_{j}, \frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}}$ and (id $\left.-\frac{\partial \mathbf{G}_{j}}{\partial \mathbf{W}_{j}}\right)^{-1}$, we obtain the second inequality in (6.1).

Moreover, we have the following lemma.
Lemma 6.2 There exist $C>0$ and $\eta>0$ such that, for all $(\underline{\theta}, \underline{\mathbf{W}}) \in Y$ with its $Y$-norm smaller than $\eta$, we have

$$
\begin{aligned}
& \left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}} \leqslant \frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}, \\
& \left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{2}} \leqslant \frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}, \\
& \left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{1}} \leqslant \frac{C}{1+s}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}, \\
& \left\|\mathbf{N}^{\mathbf{w}}(s)\right\|_{X_{2}} \leqslant \frac{C}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{C}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}} .
\end{aligned}
$$

Proof. The estimates are obtained through a direct calculation from the estimates in Lemma 6.1. We sketch the computation for $\left\|\mathbf{N}^{\theta}(s)\right\|_{\ell^{1}}$, and the others follow similarly.
First, for terms only involving $\underline{\theta}$, we notice that

$$
\begin{aligned}
& \sum_{j \in \mathbb{Z}}\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right|^{2}=-\sum_{j \in \mathbb{Z}} \theta_{j}\left(\delta^{2} \underline{\theta}\right)_{j} \leqslant\|\underline{\theta}\|_{\ell^{2}}\left\|\delta^{2} \underline{\theta}\right\|_{\ell^{2}} \leqslant \frac{1}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}, \\
& \sum_{j \in \mathbb{Z}}\left|\theta_{j}\right|^{3}\left|\left(\delta_{+} \underline{\theta}\right)_{j}\right| \leqslant\|\underline{\theta}\|_{\ell^{\infty}}^{2}\|\underline{\theta}\|_{\ell^{2}}\left\|\delta_{+} \underline{\theta}\right\|_{\ell^{2}} \leqslant\|\underline{\theta}\|_{\ell^{\infty}}^{2}\|\underline{\theta}\|_{\ell^{2}}^{\frac{3}{2}}\left\|\delta^{2} \underline{\theta}\right\|_{\ell^{2}}^{\frac{1}{2}} \leqslant \frac{1}{(1+s)^{2}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{4} .
\end{aligned}
$$

Second, for terms involving $\underline{\mathbf{W}}$, we observe that

$$
\begin{aligned}
\sum_{j \in \mathbb{Z}}\left|\theta_{j} \|\left(\delta_{+} \partial_{x} \underline{\mathbf{W}}\right)_{j}(-\pi)\right| & \leqslant\|\underline{\theta}\|_{\ell^{2}}\left(\sum_{j \in \mathbb{Z}}\left(\int_{-\pi}^{\pi}\left(\partial_{x x} \mathbf{W}_{j+1}(x)-\partial_{x x} \mathbf{W}_{j}(x)\right) \mathrm{d} x\right)^{2}\right)^{\frac{1}{2}} \\
& \leqslant \sqrt{2 \pi}\|\underline{\theta}\|_{\ell^{2}}\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right\|_{X_{2}} \\
& \leqslant \frac{\sqrt{2 \pi}}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}} .
\end{aligned}
$$

Similarly, for $\sum_{j \in \mathbb{Z}}\left|\theta_{j}\right|\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|$, we have

$$
\sum_{j \in \mathbb{Z}}\left|\theta_{j}\left\|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi) \mid \leqslant \sqrt{2 \pi}\right\| \underline{\theta}\left\|_{\ell^{2}}\right\| \delta_{+} \partial_{x} \underline{\mathbf{W}} \|_{X_{2}}\right.
$$

Using the "homogeneous matching boundary conditions" (2.11), we have

$$
\begin{aligned}
\left\|\delta_{+} \partial_{x} \underline{\mathbf{W}}\right\|_{X_{2}} & =\left(-\sum_{j \in \mathbb{Z}} \int_{-\pi}^{\pi}\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(x)\left(\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right)_{j}(x) \mathrm{d} x\right)^{\frac{1}{2}} \\
& \leqslant\left\|\delta_{+} \underline{\mathbf{W}}\right\|_{X_{2}}^{\frac{1}{2}}\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right\|_{X_{2}}^{\frac{1}{2}} \\
& \leqslant\left\|\delta_{+} \underline{\mathbf{W}}\right\|_{X_{2}}+\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right\|_{X_{2}} .
\end{aligned}
$$

We plug the latter estimate into the former one and obtain that

$$
\begin{aligned}
\sum_{j \in \mathbb{Z}}\left|\theta_{j} \|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right| & \leqslant \sqrt{2 \pi}\|\underline{\theta}\|_{\ell^{2}}\left(\left\|\delta_{+} \underline{\mathbf{W}}\right\|_{X_{2}}+\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}\right\|_{X_{2}}\right) \\
& \leqslant \frac{\sqrt{2 \pi}}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{\sqrt{2 \pi}}{(1+s)^{\frac{5}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}} .
\end{aligned}
$$

For $\sum_{j \in \mathbb{Z}}\left\|\mathbf{W}_{j}\right\|_{L^{p}}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right|$, we take $p=2$ and follow steps as above, obtaining the following estimate.
$\sum_{j \in \mathbb{Z}}\left\|\mathbf{W}_{j}\right\|_{L^{2}}\left|\left(\delta_{+} \underline{\mathbf{W}}\right)_{j}(-\pi)\right| \leqslant \frac{\sqrt{2 \pi}}{(1+s)^{2}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}+\frac{\sqrt{2 \pi}}{(1+s)^{\frac{7}{4}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}(1+s)\left\|\delta_{+} \partial_{x x} \underline{\mathbf{W}}(s)\right\|_{X_{2}}$.
For $\sum_{j \in \mathbb{Z}}\left\|\mathbf{W}_{j}^{2}\right\|_{L^{p}} \mid$, we take $p=1$ and obtain that

$$
\sum_{j \in \mathbb{Z}}\left\|\mathbf{W}_{j}^{2}\right\|_{L^{1}} \leqslant\|\underline{\mathbf{W}}\|_{X_{2}}^{2} \leqslant \frac{1}{(1+s)^{\frac{3}{2}}}\|(\underline{\theta}(t), \underline{\mathbf{W}}(t))\|_{Y}^{2}
$$

Combining the above estimate, we establish the first inequality in the lemma.

### 6.2 Bloch wave decomposition

In this section, we present the Bloch wave decomposition of the linear operator $\widetilde{A}$. We first recall that $\widetilde{A}$, as in (3.1), is defined as

$$
\begin{array}{rlc}
\tilde{A}:\left(H^{2}(\mathbb{R})\right)^{n} & \longrightarrow & \left(L^{2}(\mathbb{R})\right)^{n} \\
\mathbf{v} & \longmapsto D \partial_{x x} \mathbf{v}-\mathbf{f}^{\prime}\left(\mathbf{u}_{\star}\right) \mathbf{v} .
\end{array}
$$

We introduce the direct integral [17, XIII.16.]

$$
\begin{align*}
& \mathscr{B}: \quad L^{2}\left(\mathbb{T}_{1},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \longrightarrow \\
& \mathbf{U}(\sigma, x) \longmapsto \int_{\sigma \in \mathbb{T}_{1}} \mathrm{e}^{2}\left(\mathbb{R}^{2}(\mathbb{R})\right)^{n}  \tag{6.2}\\
& \mathbf{U}(\sigma, \cdot) \mathrm{d} \sigma .
\end{align*}
$$

The direct interal is an isometric isomorphism with inverse

$$
\begin{aligned}
& \mathscr{B}^{-1}:\left(L^{2}(\mathbb{R})\right)^{n} \longrightarrow \\
& L^{2}\left(\mathbb{T}_{1},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right) \\
& \mathbf{u}(x) \longmapsto \frac{1}{2 \pi} \sum_{\ell \in \mathbb{Z}^{m}} \mathrm{e}^{\mathrm{i} \ell \cdot x} \widehat{\mathbf{u}}(\sigma+\ell) .
\end{aligned}
$$

The following result from [20, 14] characterizes the Bloch wave decomposition of $\widetilde{A}$.
Theorem 2 (Bloch wave decomposition) The linear operator $\widetilde{A}$ is diagonal in Bloch wave space. To be precise,

$$
\begin{equation*}
\mathscr{B}^{-1} \widetilde{A} \mathscr{B}=\widehat{A}=\int_{-\frac{1}{2}}^{\frac{1}{2}} B(\sigma) \mathrm{d} \sigma, \tag{6.3}
\end{equation*}
$$

where by $\widehat{A}=\int_{-\frac{1}{2}}^{\frac{1}{2}} B(\sigma) \mathrm{d} \sigma$, we mean that, given any $\mathbf{u} \in L^{2}\left(\mathbb{T}_{1},\left(L^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n}\right)$,

$$
(\widehat{A} \mathbf{u})(\sigma)=B(\sigma) \mathbf{u}(\sigma), \text { a.e. } \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right] .
$$

Moreover, we have the following spectral mapping property.

$$
\begin{equation*}
\operatorname{spec}(\widetilde{A})=\operatorname{spec}(\widehat{A})=\bigcup_{\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]} \operatorname{spec}(B(\sigma)) \tag{6.4}
\end{equation*}
$$

### 6.3 Spectral properties of $\left\{\widehat{A}_{\mathrm{ch}}(\sigma)\right\}_{\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]}$

We recall that $\widehat{A}_{\text {ch }}(\sigma)$ is defined in (3.9) as $\widehat{A}_{\mathrm{ch}}(\sigma)=\mathscr{F}_{n} B(\sigma) \mathscr{F}_{n}^{-1}$ and $Y_{q}$ in (3.15) for $1 \leqslant q \leqslant \infty$. We are concerned with their spectral properties as unbounded operators in $Y_{q}$, which is useful for the derivation of the estimates for $M(t, \sigma)$ as defined in (3.23).
We first show the well-definedness of $\widehat{A}_{\mathrm{ch}}(\sigma)$ in $Y_{q}$ in the following lemma.
Lemma 6.3 For any given $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right], \widehat{A}_{\mathrm{ch}}(\sigma)$ is an unbounded closed operator in $Y_{2}$, that is,

$$
\begin{array}{clc}
\widehat{A}_{\mathrm{ch}}(\sigma): \mathscr{D}_{2}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right) \subset Y_{2} & \longrightarrow & Y_{2} \\
\underline{\mathbf{w}} & \longmapsto\left\{-(\sigma+\ell)^{2} D \mathbf{w}_{\ell}+\sum_{k \in \mathbb{Z}} \mathbf{h}_{\ell-k} \mathbf{w}_{k}\right\}_{\ell \in \mathbb{Z}}, \tag{6.5}
\end{array}
$$

where $\mathscr{D}_{2}\left(\widehat{A}_{\text {ch }}(\sigma)\right)=\left\{\mathbf{w} \in Y_{2} \mid\left\{\left(1+m^{2}\right) \mathbf{w}_{m}\right\}_{m \in \mathbb{Z}} \in Y_{2}\right\}$ and $\underline{\mathbf{h}}=\left\{\mathbf{h}_{\ell}\right\}_{\ell \in \mathbb{Z}}=\frac{1}{2 \pi} \int_{-\pi}^{\pi} \mathbf{f}^{\prime}\left(\mathbf{u}_{\star}(x)\right) \mathrm{e}^{-\mathrm{i} k x} \mathrm{~d} x$. Moreover, $\widehat{A}_{\mathrm{ch}}(\sigma)$ can naturally be considered as an unbounded closed operator in $Y_{q}$, with $\mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)=$ $\left\{\mathbf{w} \in Y_{q} \mid\left\{\left(1+m^{2}\right) \mathbf{w}_{m}\right\}_{m \in \mathbb{Z}} \in Y_{q}\right\}$, for all $1 \leqslant q \leqslant \infty$.

Proof. The expression for $\widehat{A}_{\mathrm{ch}}(\sigma)$ in $Y_{2}$ follows from a direct calculation. The extension to $Y_{q}$ follows from the fact that the set $\left\{\underline{\mathbf{w}} \in Y_{\infty} \mid \underline{\mathbf{w}}\right.$ has finitely many nonzero entries $\}$ is dense in $Y_{q}$ and $\mathscr{D}_{q}\left(\widehat{A}_{\text {ch }}(\sigma)\right)$, for all $q \in[1, \infty]$.

We then have the following proposition.
Proposition 6.4 For any fixed $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$ and $p \in[1, \infty], \widehat{A}_{\mathrm{ch}}(\sigma)$ defined in $Y_{q}$ is sectorial and has compact resolvent. In fact, there exist $C>0, \omega \in(\pi / 2, \pi)$ and $\lambda_{0} \in \mathbb{R}$, independent of $\sigma$ and $q$, such that the sector $S\left(\lambda_{0}, \omega\right)=\left\{\lambda \in \mathbb{C}\left|0 \leqslant\left|\arg \left(\lambda-\lambda_{0}\right)\right| \leqslant \omega, \lambda \neq \lambda_{0}\right\} \subseteq \rho\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)\right.$ and

$$
\begin{equation*}
\left\|\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\lambda\right)^{-1}\right\| \|_{Y_{q}} \leqslant C\left|\lambda-\lambda_{0}\right|^{-1}, \text { for all } \lambda \in S\left(\lambda_{0}, \omega\right), \sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right] \text { and } q \in[1, \infty] . \tag{6.6}
\end{equation*}
$$

Moreover, for any fixed $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$, the spectrum of $\widehat{A}_{\mathrm{ch}}(\sigma)$ is independent of the choice of its underlying space $Y_{q}$ and thus denoted as $\operatorname{spec}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)$, for any $q \in[1, \infty]$, with $\operatorname{spec}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right)=\operatorname{spec}(B(\sigma))$, consisting only of isolated eigenvalues with finite multiplicity.

Proof. We view $\widehat{A}_{\text {ch }}(\sigma)$ as a perturbation of the Laplacian in the discrete Fourier space, that is,

$$
\widehat{A}_{\mathrm{ch}}(\sigma)=L(\sigma)+H,
$$

where $L(\sigma) \underline{\mathbf{w}}=\left\{-(\sigma+\ell)^{2} D \mathbf{w}_{\ell}\right\}_{\ell \in \mathbb{Z}}$ and $H \underline{\mathbf{w}}=\left\{\sum_{k \in \mathbb{Z}} \mathbf{h}_{\ell-k} \mathbf{w}_{k}\right\}_{\ell \in \mathbb{Z}}$. It is straightforward to verify that the proposition holds for the Laplacian $L(\sigma)$. We only have to show that the perturbation $H$ is good enough to preserve these properties. Noting that $H \in \mathscr{L}\left(\left(\ell^{p}\right)^{n}\right)$ for any $p \in[1, \infty]$ with its norm uniformly bounded, we have, for $\lambda \in \rho(L(\sigma)),|\lambda|$ sufficiently large,

$$
\begin{equation*}
\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\lambda\right)^{-1}=(L(\sigma)+H-\lambda)^{-1}=(L(\sigma)-\lambda)^{-1}\left(\mathrm{id}+H(L(\sigma)-\lambda)^{-1}\right)^{-1} . \tag{6.7}
\end{equation*}
$$

All assertions in the proposition easily follows from this expression (6.7), except for the fact that the spectrum of $\widehat{A}_{\mathrm{ch}}(\sigma)$ is independent of $q$.
To prove this property, we denote the spectrum of $\widehat{A}_{\text {ch }}(\sigma)$ defined on $Y_{q}$ as $\operatorname{spec}\left(\widehat{A}_{\text {ch }}(\sigma), q\right)$, which consists of eigenvalues with finite multiplicity, accumulating at infinity, only. Given any eigenfunction $\underline{\mathbf{v}}=\left\{\mathbf{v}_{j}\right\}_{j \in \mathbb{Z}}, \underline{\mathbf{v}}$ belongs to $\bigcap_{q \in[1, \infty]} Y_{q}$ since $\underline{\mathbf{v}}$ are smooth, that is, $\mathbf{v}_{j}$ decays algebraically with any rate. This establishes $\operatorname{spec}(\widehat{A}(\sigma), q)=\operatorname{spec}(\widehat{A}(\sigma), p)$, for any $p, q \in[1, \infty]$.

### 6.4 Perturbation results

We apply perturbation theory to the Bloch wave operator $B(\sigma)$ for $\sigma$ near 0 and obtain more detailed spectral information, including the Taylor expansion of $d$ in Hypotheses 1.2.
To this end, we define

$$
\begin{array}{ccc}
F:\left[-\frac{1}{2}, \frac{1}{2}\right] \times \mathbb{C} \times H_{\perp}^{2} & \longrightarrow & L^{2} \\
(\sigma, \lambda, \mathbf{w}) & \longmapsto & (B(\sigma)-\lambda)\left(\mathbf{w}+\mathbf{u}_{\star}^{\prime}\right),
\end{array}
$$

where $H_{\perp}^{2}=\left\{\mathbf{w} \in\left(H^{2}\left(\mathbb{T}_{2 \pi}\right)\right)^{n} \mid\left\langle\mathbf{w}, \mathbf{u}_{\star}^{\prime}\right\rangle=0\right\}$. A standard implicit-function-theorem argument shows that there are a small neighborhood of $\sigma$ at the origin and a smooth function $(\lambda(\sigma), \mathbf{w}(\sigma))$ with $(\lambda(\sigma), \mathbf{w}(\sigma))=0$ on this neighborhood such that $F(\sigma, \lambda(\sigma), \mathbf{w}(\sigma))=0$. We denote $\mathbf{e}(\sigma)=\mathbf{u}_{\star}^{\prime}+\mathbf{w}(\sigma)$.

Similarly, replacing $B(\sigma)$ with its adjoint $B^{*}(\sigma)$, we obtain a smooth continuation of $\mathbf{u}_{\mathrm{ad}}$, denoted as $\mathbf{e}^{*}(\sigma)$. Without loss of generality, we can assume that $\left\langle\mathbf{e}(\sigma), \mathbf{e}^{*}(\sigma)\right\rangle=1$. Moreover, we have the following proposition.

Proposition 6.5 There exist positive numbers $\gamma_{0}$ and $\gamma_{1}$ such that for any $|\sigma| \leqslant \gamma_{0}$ in $\mathbb{R}, B(\sigma)$ has only one simple eigenvalue within the strip $|\operatorname{Re} \lambda| \leqslant \gamma_{1}$ in $\mathbb{C}$, which is exactly the continuation $\lambda(\sigma)$ of the eigenvalue $\lambda(0)=0$. Moreover, $\lambda(\sigma)$ has the Taylor expansion,

$$
\lambda(\sigma)=-d \sigma^{2}+\mathrm{O}\left(|\sigma|^{3}\right),
$$

where $-\gamma_{1} / 4 \leqslant-2 d \sigma^{2}<\operatorname{Re} \lambda(\sigma)<-\frac{d}{2} \sigma^{2}$, for all $\sigma \in\left[-\gamma_{0}, \gamma_{0}\right]$ and

$$
d=-\left\langle 2 \mathrm{i} \frac{\partial^{2} \mathbf{e}(0, x)}{\partial x \partial \sigma}-\mathbf{u}_{\star}^{\prime}(x), D \mathbf{u}_{\mathrm{ad}}(x)\right\rangle
$$

Proof. We first derive the explicit expression of $d$. To do that, taking first and second derivative with respect to $\sigma$ of $F(\sigma, \lambda(\sigma), \mathbf{w}(\sigma))=0$, taking the inner product of the derivatives with $\mathbf{u}_{\mathrm{ad}}$ and letting $\sigma=0$, we have

$$
\begin{aligned}
& \lambda^{\prime}(0)=\left\langle B(0) \partial_{\sigma} \mathbf{e}(0, x)+2 \mathrm{i} D \mathbf{u}_{\star}^{\prime \prime}(x), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle \\
& \lambda^{\prime \prime}(0)=\left\langle B(0) \partial_{\sigma}^{2} \mathbf{e}(0, x)+\left(4 \mathrm{i} D \partial_{x}-2 \lambda^{\prime}(0)\right) \partial_{\sigma} \mathbf{e}(0, x)-2 D \mathbf{u}_{\star}^{\prime}(x), \mathbf{u}_{\mathrm{ad}}(x)\right\rangle .
\end{aligned}
$$

Noting that span $\left\{\mathbf{u}_{\mathrm{ad}}\right\} \perp R g(B(0))$ and the inner product of an even function and an odd function is always 0 , we have

$$
\lambda^{\prime}(0)=0, \quad \lambda^{\prime \prime}(0)=2\left\langle 2 \mathrm{i} \frac{\partial^{2} \mathbf{e}(0, x)}{\partial x \partial \sigma}-\mathbf{u}_{\star}^{\prime}(x), D \mathbf{u}_{\mathrm{ad}}(x)\right\rangle
$$

It remains to prove the uniqueness of the eigenvalue of $B(\sigma)$ in a vertical strip centered at the origin for sufficiently small $\sigma$. First, there is no eigenvalue within the strip far away from the origin due to the fact that, by Proposition $6.4, \operatorname{spec}(B(\sigma))$ is in the same sector for every $\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]$. Secondly, the uniqueness within a small neighborhood of the origin follows from the above perturbation results. For the region inbetween, compactness and the local robustness of resolvent guarantee the absence of eigenvalues within this area.

Remark 6.6 (i) We stress that we may choose $\gamma_{0}$ as small as desired.
(ii) The uniqueness implies that, for $|\sigma|$ sufficiently small, $\lambda(\sigma)$ is a real number since its complex conjugate is also an eigenvalue.

### 6.5 Properties of analytic semigroups $\left\{\mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma) t}\right\}_{\sigma \in\left[-\frac{1}{2}, \frac{1}{2}\right]}$

In this section, we will derive various estimates on $\mathrm{e}^{\widehat{A}_{\text {ch }}(\sigma) t}$. We first note that by $[5,1.4]$ the interpolation space $\mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)^{\alpha}\right)$ is independent of $\sigma$,

$$
\mathscr{D}_{q}\left(\widehat{A}_{\mathrm{ch}}(\sigma)^{\alpha}\right)=\left\{\underline{\mathbf{w}} \in Y_{q} \mid\left\{\left(1+m^{2}\right)^{\alpha} \mathbf{w}_{m}\right\}_{m \in \mathbb{Z}} \in Y_{q}\right\}=: Y_{q}^{\alpha}, \quad\|\underline{\mathbf{w}}\|_{Y_{q}^{\alpha}}=\left\|\left\{\left(1+m^{2}\right)^{\alpha} \mathbf{w}_{m}\right\}_{m \in \mathbb{Z}}\right\|_{Y_{q}} .
$$

We then recall the definitions of $Y_{q, \mathrm{c}}(\sigma), Y_{q, \mathrm{~s}}(\sigma), \widehat{A}_{\mathrm{c}}(\sigma)$ and $\widehat{A}_{\mathrm{s}}(\sigma)$ from (3.18). We now have the following proposition.

Proposition 6.7 For every $q \in[1,+\infty]$ and $\alpha>0$, there exist positive constants $\epsilon \in(0,1), \gamma_{2}, C(q)$, $C(\alpha)$ and $C(\alpha, q)$ such that

$$
\begin{aligned}
\left\|\left|\mathrm{e}^{\widehat{A}_{c}(\sigma) t}\right|\right\|_{Y_{q, \mathrm{c}}(\sigma)} & \leqslant \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}, \text { for all }|\sigma| \leqslant \gamma_{0}, t \geqslant 0, \\
\left\|\mathrm{e}^{\widehat{A}_{c}(\sigma) t}\right\| \|_{Y_{q, \mathrm{c}(\sigma)} \rightarrow Y_{q}^{\alpha}} & \leqslant C(\alpha) \mathrm{e}^{-\frac{d}{2} \sigma^{2} t}, \text { for all }|\sigma| \leqslant \gamma_{0}, t \geqslant 0, \\
\left\|\mid \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) t}\right\|_{Y_{q, \mathrm{~s}}(\sigma)} & \leqslant C(q) e^{-\frac{\gamma_{1}}{2} t}, \text { for all }|\sigma| \leqslant \gamma_{0}, t \geqslant 0, \\
\left\|\mid \mathrm{e}^{\widehat{A}_{\mathrm{s}}(\sigma) t}\right\| \|_{Y_{q, \mathrm{~s}}(\sigma) \rightarrow Y_{q}^{\alpha}} & \leqslant C(\alpha, q) t^{-\alpha} \mathrm{e}^{-\gamma_{1} t / 2}, \text { for all }|\sigma| \leqslant \gamma_{0}, t>0, \\
\left\|\mid \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t}\right\|_{Y_{q}} & \leqslant C(q) \mathrm{e}^{-\epsilon d \sigma^{2} t}, \text { for all }|\sigma| \leqslant \gamma_{0}, t \geqslant 0, \\
\left\|\mid \mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t}\right\|_{Y_{q}} & \leqslant C(q) \mathrm{e}^{-\gamma_{2} t}, \text { for all } \gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}, t \geqslant 0, \\
\left\|\mathrm{e}^{\widehat{A}_{\mathrm{ch}}(\sigma) t}\right\| \|_{Y_{q} \rightarrow Y_{q}^{\alpha}} & \leqslant C(\alpha, p) t^{-\alpha} \mathrm{e}^{-\gamma_{2} t}, \text { for all } \gamma_{0} \leqslant|\sigma| \leqslant \frac{1}{2}, t>0 .
\end{aligned}
$$

Proof. We first derive estimates for the case $|\sigma| \leqslant \gamma_{0}$. For $\widehat{A}_{c}(\sigma)$, we have $\mathrm{e}^{\widehat{A}_{c}(\sigma) t}=\mathrm{e}^{\lambda(\sigma) t}$. The first two inequalities follow directly from the fact that $\operatorname{Re} \lambda(\sigma)<-\frac{d}{2} \sigma^{2}$ and $\mathbf{e}(\sigma)$ is smooth, by Proposition 6.5 , for $|\sigma| \leqslant \gamma_{0}$.

For $\widehat{A}_{\mathrm{s}}(\sigma)$, by Proposition 6.4 and 6.5 , for any $\sigma \in\left(-\gamma_{0}, \gamma_{0}\right)$ and $q \in[1, \infty]$,

$$
\operatorname{spec}\left(\widehat{A}_{\mathrm{s}}(\sigma), q\right) \subset \mathbb{C} \backslash S\left(-\frac{\gamma_{1}}{2}, \tilde{\omega}\right), \text { where } \tilde{\omega} \in\left(\frac{\pi}{2}, \pi\right)
$$

Moreover, for every $q \in[1,+\infty]$, there exists a positive constant $C(q)$ such that

$$
\left\|\left|\left(\widehat{A}_{\mathrm{s}}(\sigma)-\lambda\right)^{-1} \|_{Y_{q, \mathrm{~s}}(\sigma)} \leqslant C(q)\right| \lambda+\left.\frac{\gamma_{1}}{2}\right|^{-1}, \text { for all }|\sigma| \leqslant \gamma_{0} \text { and } \lambda \in S\left(-\frac{\gamma_{1}}{2}, \tilde{\omega}\right) .\right.
$$

Thus, by [5, Thm.1.3.4, 1.4.3], we immediately obtain the two inequalities for $\widehat{A}_{\mathrm{s}}(\sigma)$. The first inequality on $\widehat{A}_{\mathrm{ch}}(\sigma)$ follows directly by combining the first inequality for $\widehat{A}_{\mathrm{c}}(\sigma)$ and the first inequality for $\widehat{A}_{\mathrm{s}}(\sigma)$. We now derive the estimates for the case $\gamma_{0}<|\sigma| \leqslant \frac{1}{2}$. By a similar analysis as in Proposition 6.5, there exists a positive constant $\gamma_{2}$ such that

$$
\operatorname{Re}\left(\operatorname{spec} \widehat{A}_{\mathrm{ch}}(\sigma)\right)<-2 \gamma_{2}, \text { for all } \gamma_{0}<|\sigma| \leqslant \frac{1}{2} .
$$

It is then not hard to conclude that

$$
\operatorname{spec}\left(\widehat{A}_{\mathrm{ch}}(\sigma)\right) \subset \mathbb{C} \backslash S\left(-\gamma_{2}, \tilde{\omega}_{1}\right), \text { where } \tilde{\omega}_{1} \in\left(\frac{\pi}{2}, \pi\right)
$$

Moreover, for every $q \in[1,+\infty]$, there exists a positive constant $C(q)$ such that

$$
\left|\left|\left|\left(\widehat{A}_{\mathrm{ch}}(\sigma)-\lambda\right)^{-1} \|_{Y_{q}} \leqslant C(q)\right| \lambda+\gamma_{2}\right|^{-1}, \text { for all } \gamma_{0}<|\sigma| \leqslant \frac{1}{2} \text { and } \lambda \in S\left(-\gamma_{2}, \tilde{\omega}_{1}\right) .\right.
$$

Therefore, again by [5, Thm.1.3.4, 1.4.3], we immediately obtain the last two inequalities for $\widehat{A}_{\mathrm{ch}}(\sigma)$, which concludes the proof.

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