

Cap M, 2 securities.

Ω = sample spaces.

$S_1(0)$, $S_2(0)$ given.

$S_1(0)$, $S_2(0) : \Omega \rightarrow (0, \infty)$

If $\Omega = \{1, \dots, m\}$ \leftarrow m possible outcomes,

write: $S_i^j(1) \equiv$ value of i^{th} stock under outcome j .

Return variables for each stock, are defined as:

$$K_i = \frac{S_i(1) - S_i(0)}{S_i(0)} ; \quad K = \begin{pmatrix} K_1 \\ K_2 \end{pmatrix}$$

We need only concern ourselves with the parameters

$$\mu_i = E K_i \quad i=1,2.$$

$$\sigma_i^2 = C_i = \text{var } K_i$$

$$C_{12} = \text{cov}(K_1, K_2)$$

$$\mu_K = \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} ; \quad \Sigma_K = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

PARAMETERIZE the portfolios by the set 2

$$W = \left\{ \begin{pmatrix} w_1 \\ w_2 \end{pmatrix} \in \mathbb{R}^2 : w_1 + w_2 = 1 \right\}.$$

* let us use notation $\mathbf{1} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ so $w^\top \mathbf{1} = w_1 + w_2 = 1$.

For given portfolio $w \in W$

$$K_w = (w_1 \ w_2) \begin{pmatrix} K_1 \\ K_2 \end{pmatrix} = w_1 K_1 + w_2 K_2 = w^\top K.$$

The expected return of the portfolio is then:

~~$$\mu_w = E(K_w) = E(w_1 K_1 + w_2 K_2) =$$~~

$$= w_1 \mu_1 + w_2 \mu_2 = w^\top \mu_K.$$

The variance (risk)² is

~~$$\sigma_w^2 = \text{var}(K_w) = \text{var}(w_1 K_1 + w_2 K_2)$$~~
~~$$= w_1^2 \sigma_1^2 + 2 w_1 w_2 C_{12} + w_2^2 \sigma_2^2$$~~

~~$$\sigma_w^2 = \text{var}(K_w) = \text{var}(w_1 K_1 + w_2 K_2)$$~~

$$= w_1^2 \sigma_1^2 + 2 w_1 w_2 C_{12} + w_2^2 \sigma_2^2$$

$$= w^\top \sum_K w$$

It is helpful to write the

portfolio weight as $w_1 = s$
 $w_2 = 1-s$.

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$$\mu_w = \mu_s = s\mu_1 + (1-s)\mu_2 = \mu_2 + s(\mu_1 - \mu_2).$$

$$\sigma_w^2 = \sigma_s^2 = s^2 \sigma_1^2 + (1-s)^2 \sigma_2^2 + 2s(1-s)\sigma_{12}.$$

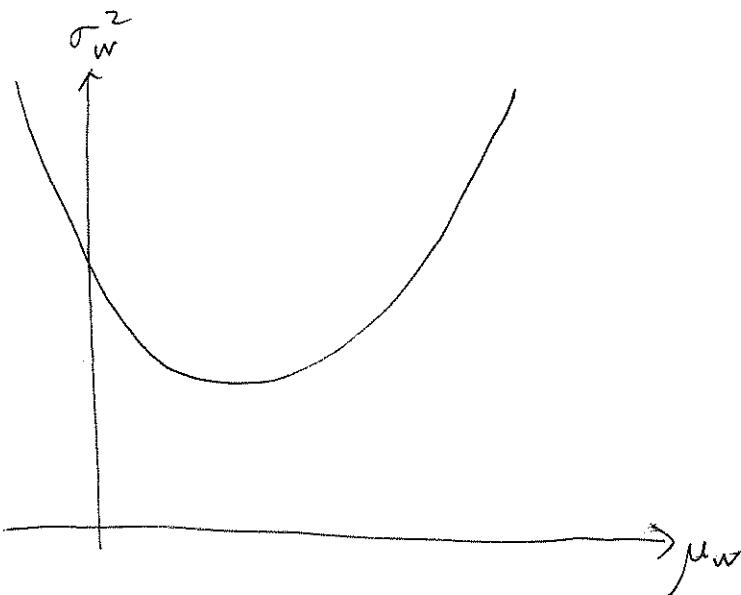
Notice μ_w is linear in $s \rightarrow s$ is linear in μ_w .

σ_w^2 is quadratic in s

$$s = \frac{\mu_w - \mu_2}{\mu_1 - \mu_2}.$$

$\therefore \sigma_w^2$ is quadratic in μ_w -

$$\sigma^2(s) = \sigma^2 \left(\frac{\mu_w - \mu_2}{\mu_1 - \mu_2} \right).$$



~~Horizontal axis~~ σ_w^2 .

MINIMIZE $\sigma_w^2 \dots$

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$$\frac{d\sigma_w^2}{ds} = 2s\sigma_1^2 + [2(1-s) - 2s]C_{12} - 2(1-s)\sigma_2^2 = 0$$

(1) s minimizing σ_w^2 is (provided $\begin{cases} \sigma_1 \neq \sigma_2 \\ -\sigma_1^2 \leq C_{12} < \sigma_1^2 \text{ if } \sigma_1 = \sigma_2 \end{cases}$)

$$s = \frac{\sigma_2^2 - C_{12}}{\sigma_1^2 - 2C_{12} + \sigma_2^2}$$

\therefore portfolio minimizing σ_w^2 is

$$(4) \quad W = \begin{pmatrix} s \\ 1-s \end{pmatrix} = \begin{pmatrix} \frac{\sigma_2^2 - C_{12}}{\sigma_1^2 - 2C_{12} + \sigma_2^2} \\ \frac{\sigma_1^2 - C_{12}}{\sigma_1^2 - 2C_{12} + \sigma_2^2} \end{pmatrix} = \cancel{\begin{pmatrix} \sigma_2^2 \\ \sigma_1^2 \end{pmatrix} + \cancel{\begin{pmatrix} C_{12} \\ C_{12} \end{pmatrix}} - \cancel{\begin{pmatrix} \sigma_2^2 - C_{12} \\ \sigma_1^2 - C_{12} \end{pmatrix}}} \\ = \frac{\Sigma^{-1} \mathbf{1}}{\mathbf{1}^\top \Sigma^{-1} \mathbf{1}}$$

(2) if $\sigma_1 = \sigma_2$ + ~~$C_{12} = \sigma_1^2 = \sigma_2^2$~~ $C_{12} = \sigma_1^2 = \sigma_2^2$.

$$\begin{aligned} \sigma_w^2 &= s^2\sigma_1^2 + (1-s)^2\sigma_2^2 + 2s(1-s)C_{12} \\ &= (s^2 + (1-s)^2 + 2s(1-s))\sigma_1^2 \\ &= (s + (1-s))^2\sigma_1^2 = \sigma_1^2 \end{aligned}$$

$\therefore \sigma_w^2$ is constant.

Notice the formula (4) may imply $s < 0$ or $1-s < 0$
this would require short selling to find
minimal risk.

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Find σ_w as a function of μ_w

$$s = \frac{\mu_w - \mu_2}{\mu_1 - \mu_2}$$

$$\sigma_w^2 = \left(\frac{\mu_w - \mu_2}{\mu_1 - \mu_2} \right)^2 \sigma_1^2 + \left(\frac{\mu_1 - \mu_w}{\mu_1 - \mu_2} \right) \sigma_2^2 - 2c_{12} \frac{(\mu_w - \mu_2)(\mu_w - \mu_1)}{(\mu_1 - \mu_2)^2}$$

write: $\mu_m = \mu_w(w_m)$; $\sigma_m^2 = \sigma_w^2(w_m)$

$$\sigma_w^2 - A^2 (\mu_w - \mu_m)^2 = \sigma_m^2$$

$$\sigma_w = \left(A^2 (\mu_w - \mu_m)^2 + \sigma_m^2 \right)^{1/2}$$

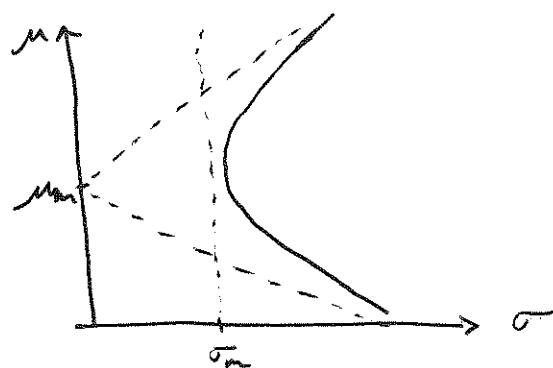
Asymptotes:

$$\text{as } \mu_w \rightarrow \pm\infty, \quad \sigma_w \sim A(\mu_w - \mu_m)$$

hyperbolz is centered @ $(\mu_m, 0)$.

Asym. lines

$$\mu = \mu_m \pm A\sigma \quad ; \quad A = \frac{\sigma_1^2 + \sigma_2^2 - 2c_{12}}{(\mu_1 - \mu_2)^2} > 0$$



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Let us consider simplest case first,

that $|C_{12}| = \sigma_1 \sigma_2 \Leftrightarrow |P_{12}| = \left| \frac{C_{12}}{\sigma_1 \sigma_2} \right| = 1$.

In this case,

$$K_1 = t K_2 + P.$$

$$\text{So } \sigma_1^2 = t^2 \sigma_2^2.$$

Assume $t \neq 1$ (in this case σ_w^2 is constant).

Find w for minimum variance.

$$W = \begin{pmatrix} S_o \\ 1-S_o \end{pmatrix} = \dots$$

$$S_o = \frac{\sigma_2^2 - t \sigma_2^2}{t^2 \sigma_2^2 + \sigma_2^2 - 2t \sigma_2^2} = \frac{1-t}{(1-t)^2} = \frac{1}{1-t}.$$

$$W = \begin{pmatrix} \frac{1}{1-t} \\ \frac{-t}{1-t} \end{pmatrix}$$

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(i) $P_{12} = 1$

$$t > 0 \Rightarrow$$

$$S_0 = \frac{1}{1-t} < 0 \quad \text{if } t > 1$$

$$1 - S_0 = \frac{-1}{1-t} < 0 \quad \text{if } 0 < t < 1.$$

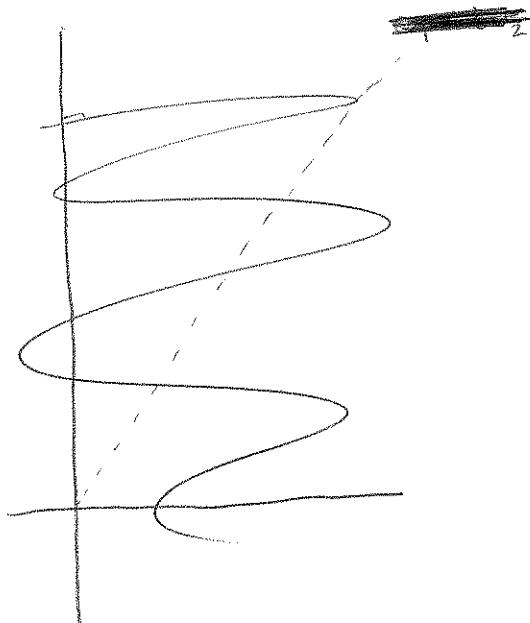
∴ finding min variance requires short selling

(ii) $P_{12} = -1$

$$t < 0$$

$$\left. \begin{array}{l} S_0 = \frac{1}{1-t} \\ 1 - S_0 = \frac{t}{1-t} \end{array} \right\} \text{both belong to } (0, 1)$$

∴ minimal variance does not require
short selling.



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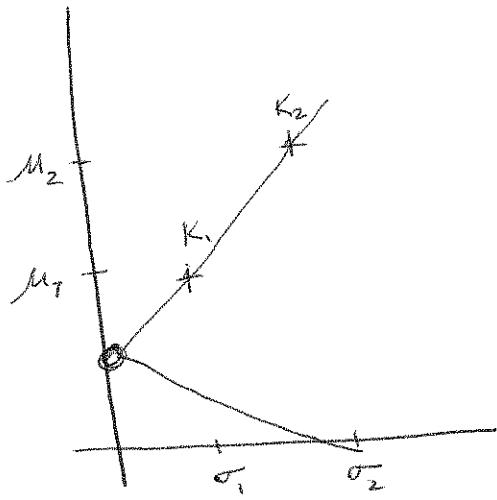
Notice

$$K_1 = t K_2 + P$$

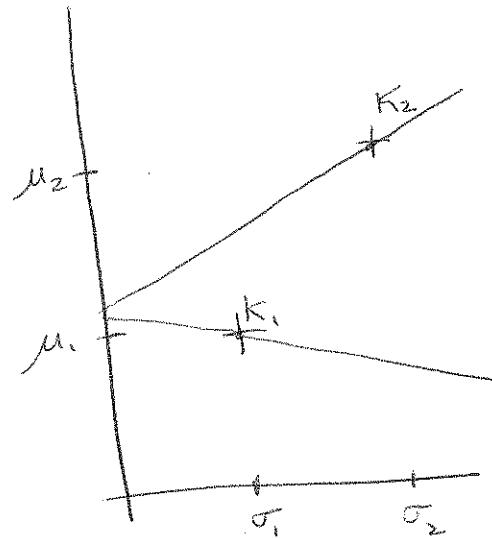
$$\Leftrightarrow M_1 = t \mu_2 + P$$

$$\sigma_1 = |t| \sigma_2$$

$$t > 0$$



$$t < 0$$



min variance, $\sigma_w^2 = 0$.

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~~with short sell between~~

$\rho_{12} \sim 1$ mvp req short sell

$\rho_{12} \sim -1$ mvp does not req short sell

where is bdry?

$0 < S_0 < 1 \leftarrow$ mvp w/o short sell.

$$S_0 = \frac{\sigma_2^2 - c_{12}}{\sigma_1^2 + \sigma_2^2 - 2c_{12}}$$

$$S_0 < 1$$

$$S_0 \geq 0$$

$$\sigma_2^2 - c_{12} < \sigma_1^2 + \sigma_2^2 - 2c_{12}$$

$$\sigma_2^2 - c_{12} > 0$$

$$c_{12} < \sigma_1^2$$

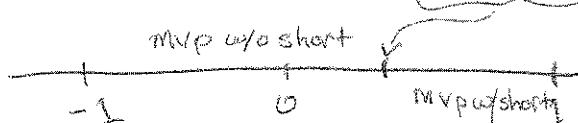
$$c_{12} < \sigma_2^2$$

$$\rho_{12} < \frac{\sigma_1}{\sigma_2}$$

$$\rho_{12} < \frac{\sigma_2}{\sigma_1}$$

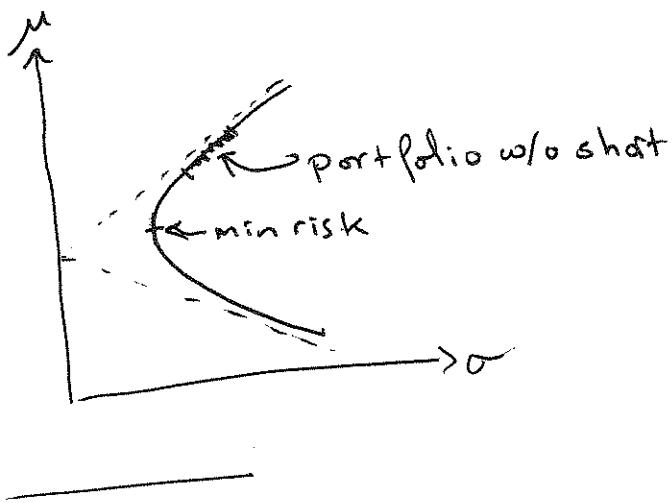
mvp w/ short sell is possible iff

$$\rho_{12} < \underbrace{\frac{\sigma_2}{\sigma_1} \wedge \frac{\sigma_1}{\sigma_2}}_{\rho^*} = \rho^*$$



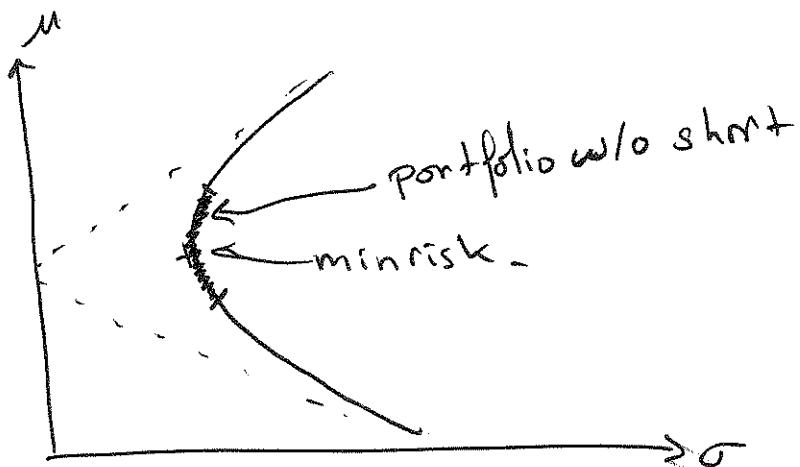
$$1 > \rho > \frac{\sigma_2}{\sigma_1} \wedge \frac{\sigma_1}{\sigma_2} = \min \left\{ \frac{\sigma_2}{\sigma_1}, \frac{\sigma_1}{\sigma_2} \right\}.$$

min variance portfolio requires short.



$$-1 < \rho < \frac{\sigma_2}{\sigma_1} \wedge \frac{\sigma_1}{\sigma_2}$$

min variance portfolio w/o short



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Suppose

$$\sigma_1^2 = \gamma_2 \quad . \quad \sigma_2^2 = \gamma_4 \quad . \quad c_{12} = \gamma_8 \quad .$$

$$p_{12} = \frac{\gamma_8}{\gamma_2} \gamma_2 = \frac{\gamma_8}{4}$$

$$\therefore p^c = \frac{\gamma_4}{\gamma_2} = \gamma_2$$

$\because p_{12} < p^c \Rightarrow$ does not require short sell

$$S_0 = \frac{\gamma_2 - \gamma_8}{\gamma_2 + \gamma_4 - 2\gamma_8} = \frac{3/8}{7/8} = 3/7 \in (0, 1).$$

$$\sigma_1^2 = \frac{1}{3} \quad . \quad \sigma_2^2 = \frac{1}{9} \quad . \quad c_{12} = \frac{1}{6}$$

$$p_{12} = \frac{\gamma_6}{\gamma_3 \gamma_8} = \frac{\sqrt{3}}{2}$$

$$p^c = \frac{\gamma_9}{\gamma_3} = \gamma_3$$

$p_{12} > p^c \Rightarrow$ require short sell.

Indeed

$$S_0 = \frac{\gamma_9 - \gamma_6}{\gamma_9 + \gamma_3 - 2\gamma_6} = \frac{2-3}{2+6-6} = -\frac{1}{2} < 0$$