

# Introduction to Portfolio Theory

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**Session goals:**

- Understand risk, return, and diversification concepts.
- Construct and interpret efficient frontier portfolios.
- Apply CAPM to evaluate systematic risk.

# Session Agenda

- Risk & return
- Two-asset & multi-asset portfolios
- Diversification
- Efficient frontier
- CAPM and Beta

# Motivation

- Investors rarely hold a single security
- Portfolio outcomes depend on:
  - Individual asset behavior
  - How assets move together
- Central idea: **Risk can be reduced without sacrificing expected return**
- This insight earned Markowitz the Nobel Prize.



# Return as a Random Variable

- For a single asset with price  $S(0)$ ,  $S(1)$ , the return is:

$$K = \frac{S(1) - S(0)}{S(0)}$$

- Return is random because  $S(1)$  is uncertain.
- Expected return measures central tendency:

$$\mu = \mathbb{E}(K)$$

- Risk is captured by standard deviation:

$$\sigma = \sqrt{\text{Var}(K)}$$

# Why Variance as Risk

- Captures spread of possible outcomes.
- Penalizes deviations both above and below the mean (risk-averse stance).
- Works naturally with linear combinations of assets.
- Integrates directly into optimization problems (quadratic structure).

# Two Securities: Portfolio Return

For weights  $w_1, w_2$  with  $w_1 + w_2 = 1$ :

- Portfolio return:

$$K_V = w_1 K_1 + w_2 K_2$$

- Portfolio expected return:

$$\mu_V = w_1 \mu_1 + w_2 \mu_2$$

- This is a linear, intuitive relationship.

# Two Securities: Portfolio Risk

- Portfolio variance:
$$\sigma_V^2 = w_1^2 \sigma_1^2 + w_2^2 \sigma_2^2 + 2w_1 w_2 \text{Cov}(K_1, K_2)$$
- Using correlation  $\rho_{12}$ :
$$\sigma_V^2 = w_1^2 \sigma_1^2 + w_2^2 \sigma_2^2 + 2w_1 w_2 \rho_{12} \sigma_1 \sigma_2$$
- **Key insight:** Risk depends on *co-movement*, not individual volatilities alone.



# Diversification Mechanism

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

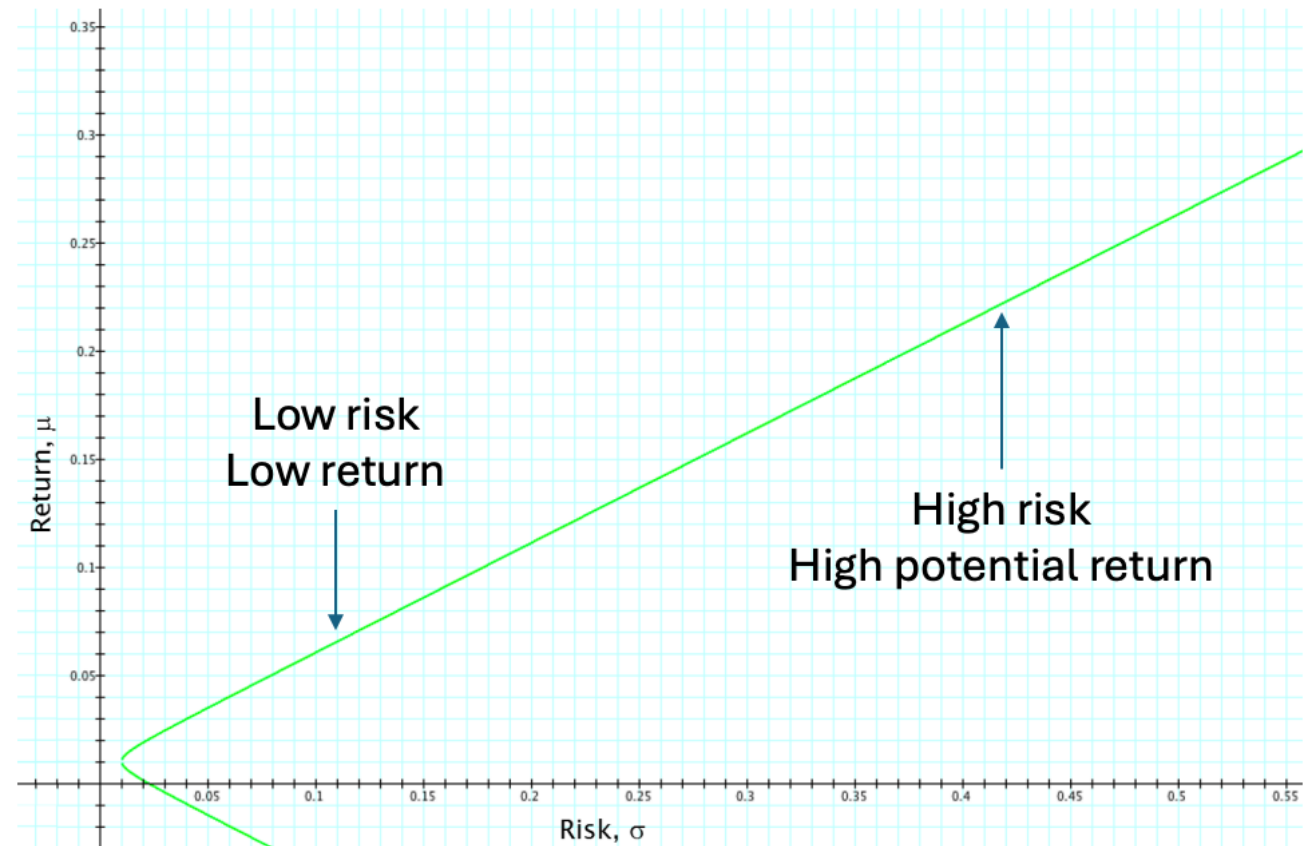
- If  $\rho_{12} < 1$ , diversification reduces portfolio variance.
- Greatest benefit when  $\rho_{12}$  is *negative*.
- Even when both assets are individually risky, the combination may be *less risky than either*.
- Diversification cannot eliminate *systematic* (market-wide) risk.



# Two-asset Theory: Core Idea

Two-Asset Theory states:

- Any efficient portfolio composed of two risky assets must lie on a smooth curve in  $(\sigma, \mu)$  space.
- The curve is parabolic when variance is plotted against weight.
- Every investor choosing between only two risky assets will pick a point on this curve based on risk preference.
- All efficient portfolios of two risky assets are convex combinations of the two assets.



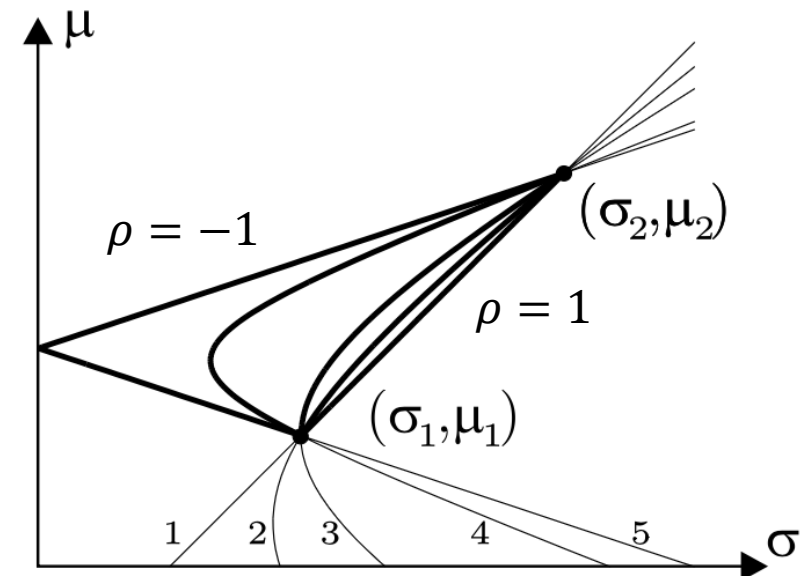
# Two-asset Efficient Frontier

For assets 1 and 2, the efficient frontier is the **upper branch** of the curve generated by:

$$\mu(w) = w\mu_1 + (1 - w)\mu_2$$

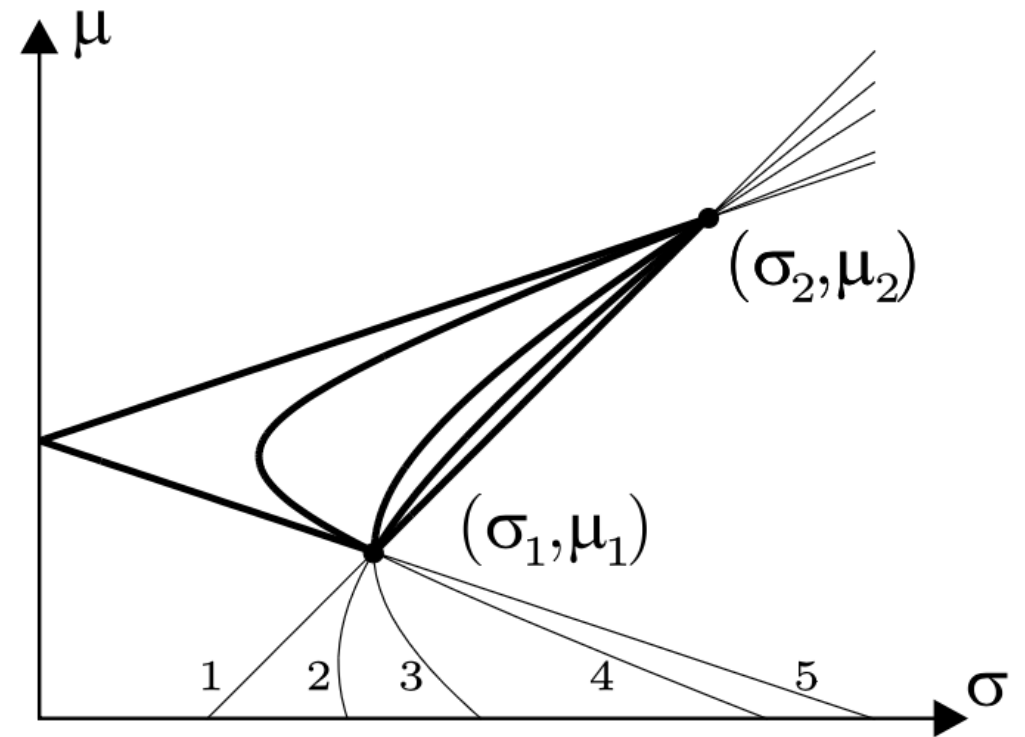
$$\sigma^2(w) = w^2\sigma_1^2 + (1 - w)^2\sigma_2^2 + 2w(1 - w)\rho_{12}\sigma_1\sigma_2$$

- Each point corresponds to a  $w$ .
- Upper part of the curve is *efficient*.
- Lower part of the curve is dominated.
- Endpoints correspond to holding 100% of one asset.
- Diversification improves risk-return trade-offs.



# Conditions for the Efficient Set

- If  $\frac{\sigma_1}{\sigma_2} < \rho_{12} \leq 1$ , then there is a portfolio with short selling such that  $\sigma_V < \sigma_1$ , but for each portfolio without short selling  $\sigma_V \geq \sigma_1$  (lines 1 and 2).
- If  $\rho_{12} = \frac{\sigma_1}{\sigma_2}$ , then  $\sigma_V \geq \sigma_1$  for each portfolio (line 3).
- If  $-1 \leq \rho_{12} < \frac{\sigma_1}{\sigma_2}$ , then there is a portfolio without short selling such that  $\sigma_V < \sigma_1$  (lines 4 and 5).



# Minimum Variance Portfolio

For  $-1 < \rho_{12} < 1$  the portfolio with minimum variance is attained at

$$w_0 = \frac{\sigma_2^2 - \rho_{12}\sigma_1\sigma_2}{\sigma_1^2 + \sigma_2^2 - 2\rho_{12}\sigma_1\sigma_2}$$

**How?**

$$\sigma_V^2 = w^2\sigma_1^2 + (1-w)^2\sigma_2^2 + 2w(1-w)\rho_{12}\sigma_1\sigma_2$$

- Step-1: Compute the derivative and equate it to 0.

$$\frac{d\sigma_V^2}{dw} = 2w\sigma_1^2 - 2(1-w)\sigma_2^2 + 2(1-w)\rho_{12}\sigma_1\sigma_2 - 2w\rho_{12}\sigma_1\sigma_2 = 0$$

- Step-2: Check if the second derivative is positive.

$$2\sigma_1^2 + 2\sigma_2^2 - 4\rho_{12}\sigma_1\sigma_2 > 2\sigma_1^2 + 2\sigma_2^2 - 4\sigma_1\sigma_2 = 2(\sigma_1 - \sigma_2)^2 \geq 0$$

# Portfolio with several securities

- A portfolio with  $n$  securities can be described in terms of its weights

$$w_i = \frac{x_i S_i(0)}{V(0)}, \quad i = 1, \dots, n$$

where  $x_i$  is the number of shares of type  $i$  in the portfolio.

## Matrix representation

$$\mathbf{w} = [w_1 \ w_2 \ \dots \ w_n]$$

Their weights adding up to 1 can be written as:

$$1 = \mathbf{u}\mathbf{w}^T$$

where  $\mathbf{u} = [1 \ 1 \ \dots \ 1]$

The expected returns of the portfolio are represented as

$$\mathbf{m} = [\mu_1 \ \mu_2 \ \dots \ \mu_n]$$

# Portfolio with several securities

The covariance matrix is written as:

$$\mathbf{C} = \begin{bmatrix} c_{11} & \cdots & c_{1n} \\ \vdots & \ddots & \vdots \\ c_{n1} & \cdots & c_{nn} \end{bmatrix}$$

Hence, the expected return and variance of the portfolio can be given as:

$$\mu_V = \mathbf{m}\mathbf{w}^T; \quad \sigma_V^2 = \mathbf{w}\mathbf{C}\mathbf{w}^T$$

- The diagonal elements are simply the variance of returns:  $c_{ii} = \text{Var}(K_i)$ .
- $\mathbf{C}$  is a symmetric and positive definite matrix, where the inverse  $\mathbf{C}^{-1}$  is possible.

# Minimum Variance Portfolio

The portfolio with the smallest variance has weights:

$$\mathbf{w}_0 = \frac{\mathbf{u}\mathbf{C}^{-1}}{\mathbf{u}\mathbf{C}^{-1}\mathbf{u}^T}$$

**How?**

$$F(\mathbf{w}, \lambda) = \mathbf{w}\mathbf{C}\mathbf{w}^T - \lambda(\mathbf{u}\mathbf{w}^T - 1)$$

Equating the partial derivative of  $F$  with  $w_i$ , we get

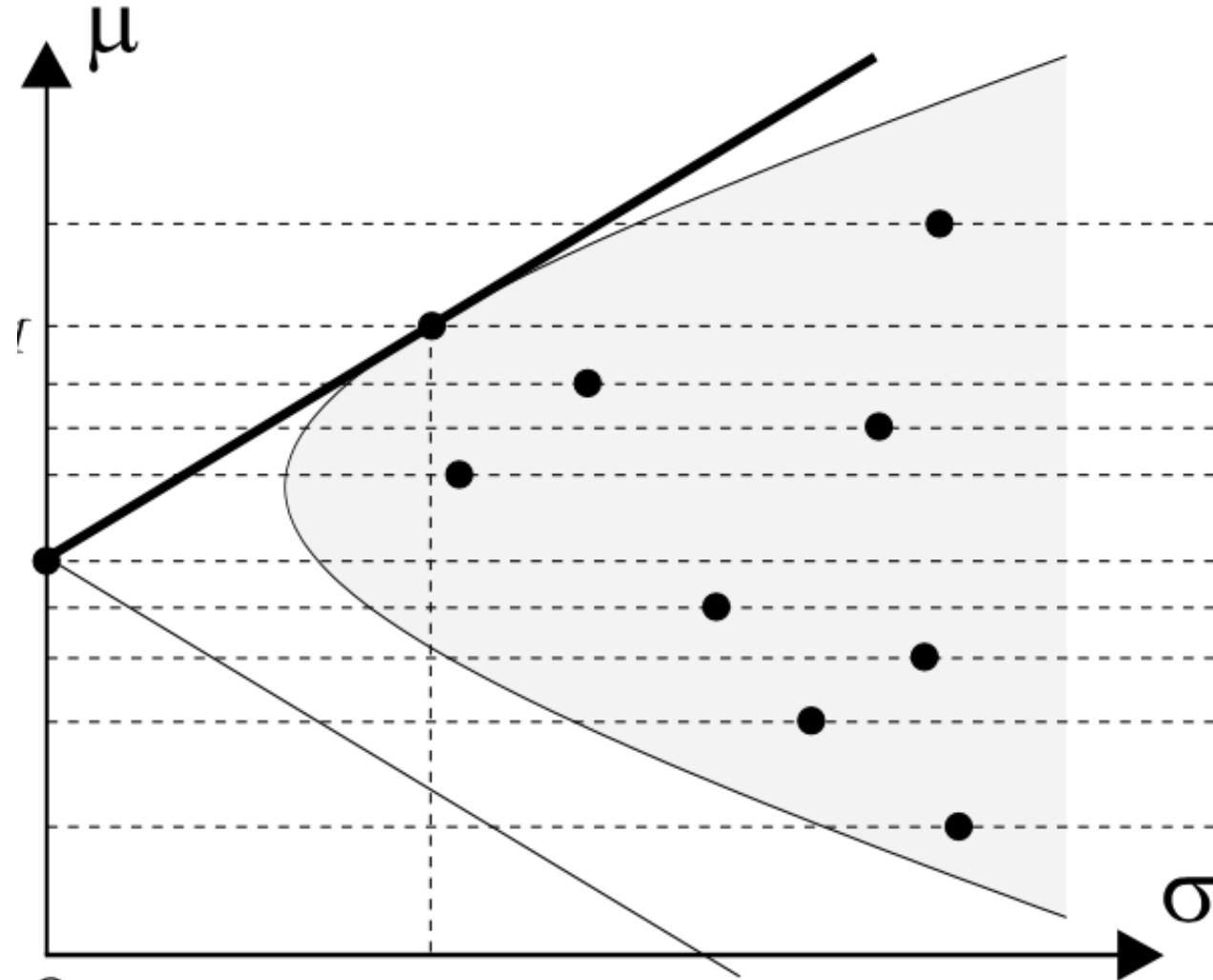
$$2\mathbf{w}\mathbf{C} - \lambda\mathbf{u} = 0 \Rightarrow \mathbf{w} = \frac{\lambda}{2}\mathbf{u}\mathbf{C}^{-1}$$

Substituting this in the constraint  $1 = \mathbf{u}\mathbf{w}^T$ , we get:

$$1 = \frac{\lambda}{2}\mathbf{u}\mathbf{C}^{-1}\mathbf{u}^T$$

Solving the above equations gives us  $\mathbf{w}_0$ .

# Minimum Variance for an Expected Return





# Minimum Variance for an Expected Return

Solve:  $\min \mathbf{w} \mathbf{C} \mathbf{w}^T$ , subject to: (a)  $\mathbf{w} \mathbf{m}^T = \mu$ , and (b)  $\mathbf{w} \mathbf{u}^T = 1$

$$G(\mathbf{w}, \lambda_1, \lambda_2) = \mathbf{w} \mathbf{C} \mathbf{w}^T - \lambda_1 (\mathbf{w} \mathbf{m}^T - \mu) - \lambda_2 (\mathbf{w} \mathbf{u}^T - 1)$$

Taking partial derivatives with respect to  $\mathbf{w}$ ,  $\lambda_1$  and  $\lambda_2$ , and solving, we get.

$$2\mathbf{w} = \lambda_1 \mathbf{m} \mathbf{C}^{-1} + \lambda_2 \mathbf{u} \mathbf{C}^{-1}$$

$$\begin{bmatrix} \lambda_1 \\ \lambda_2 \end{bmatrix} = 2M^{-1} \begin{bmatrix} \mu \\ 1 \end{bmatrix}$$

where,

$$M = \begin{bmatrix} \mathbf{m} \mathbf{C}^{-1} \mathbf{m}^T & \mathbf{u} \mathbf{C}^{-1} \mathbf{m}^T \\ \mathbf{m} \mathbf{C}^{-1} \mathbf{u}^T & \mathbf{u} \mathbf{C}^{-1} \mathbf{u}^T \end{bmatrix}$$

The final solution is of the form:  $\mathbf{w} = \mu \mathbf{a} + \mathbf{b}$

# Two-fund Theorem

Let  $w_1, w_2$  be the weights of any two portfolios  $V_1, V_2$  on the minimum variance line with expected returns  $\mu_{V_1} \neq \mu_{V_2}$ .

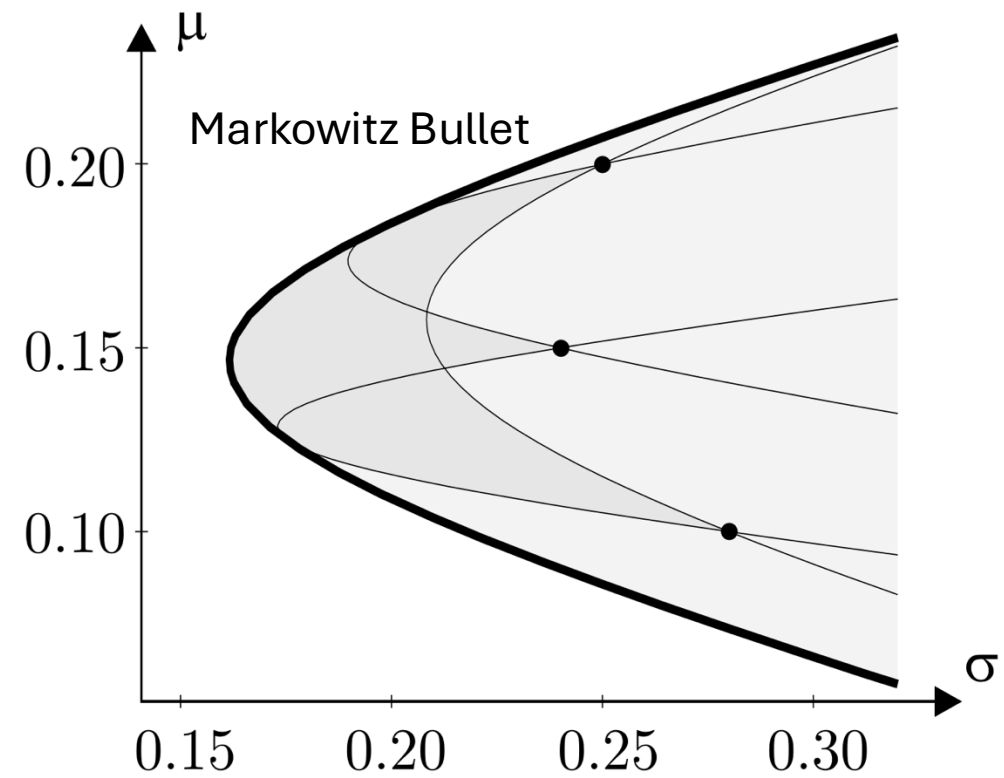
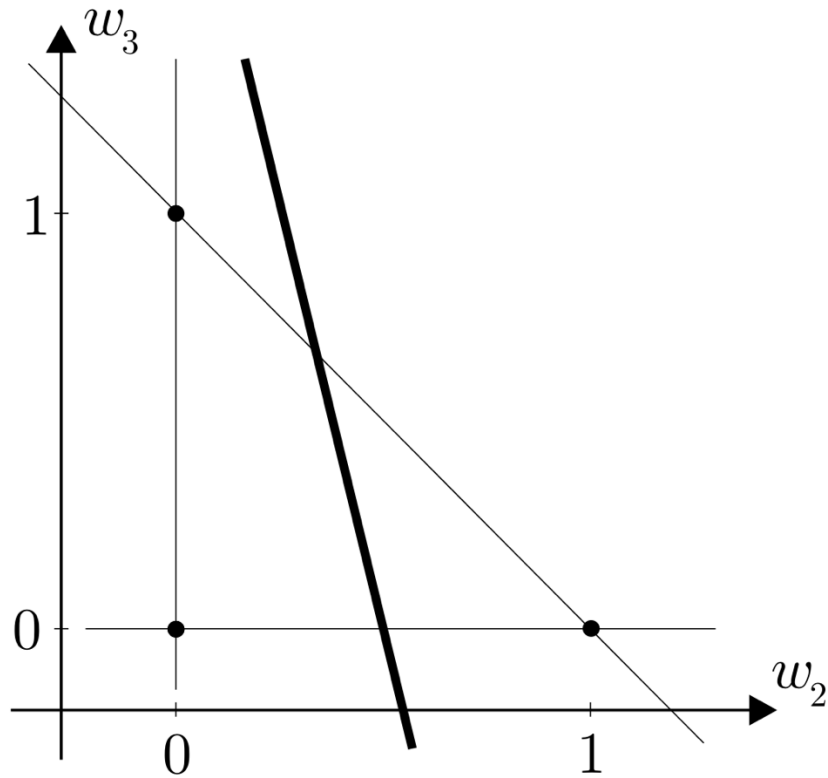
Each portfolio  $V$  on the minimum variance line can be expressed as:

$$\mathbf{w} = \alpha \mathbf{w}_1 + (1 - \alpha) \mathbf{w}_2$$

## Practical significance

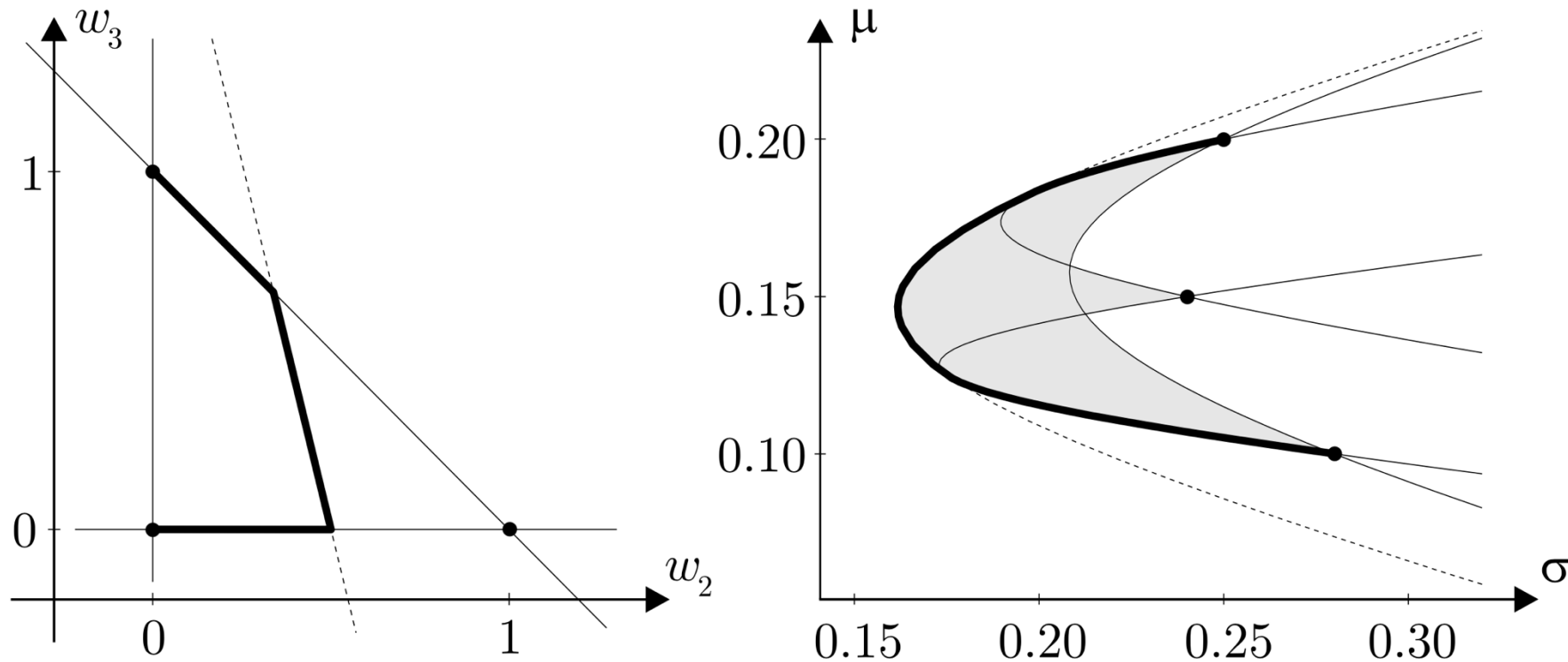
- Any portfolio on the minimum variance line can be realized by splitting the wealth between just two investment funds.
- Trading the units of only two investment funds can significantly reduce the costs and simplify the procedure as compared to simultaneous transactions in  $n$  risky assets.
- Ofcourse the target expected returns can only be between  $\mu_{V_1}$  and  $\mu_{V_2}$ .

# Feasible portfolio with 3 securities



# Feasible portfolio with 3 securities

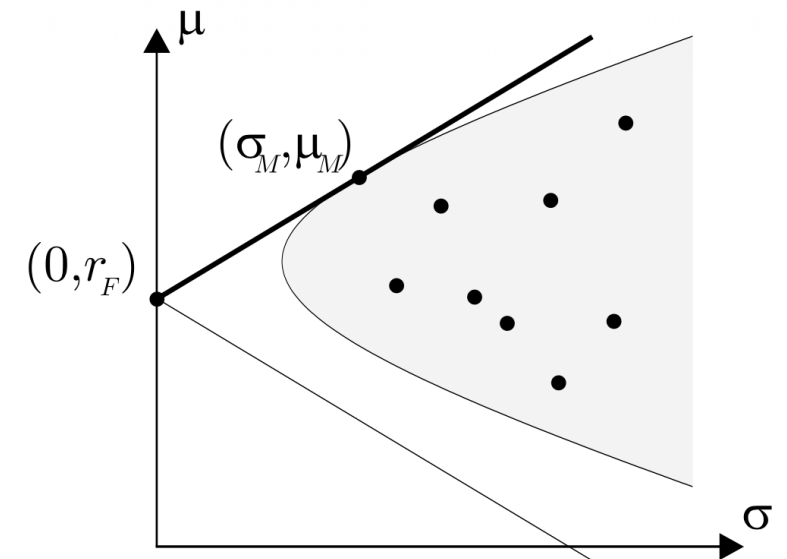
*without short selling*



# Market Portfolio

- Let's add a risk-free security with return  $r_F$ .
- Every rational investor expecting a dominating portfolio will select a portfolio on this half-line, called the **Capital Market Line (CML)**.
- The argument works until  $r_F$  is not too high, i.e., it is less than the expected return of the min. variance portfolio.
- $(\sigma_M, \mu_M)$  is referred to as the **Market Portfolio**.
- This is the steepest (highest gradient) line passing through  $(0, r_F)$ , given as:

$$\frac{\mu_V - r_F}{\sigma_V} = \frac{\mathbf{w}\mathbf{m}^T - r_F}{\sqrt{\mathbf{w}\mathbf{C}\mathbf{w}^T}}$$



# Market Portfolio

- To calculate the market portfolio, we need to maximize the slope:

$$\max \frac{\mathbf{w}\mathbf{m}^T - r_F}{\sqrt{\mathbf{w}\mathbf{C}\mathbf{w}^T}}, \quad \text{s. t. } \mathbf{w}\mathbf{u}^T = 1$$

- Solving it through the Lagrange function, leads to

$$\mathbf{w}_M = \frac{(\mathbf{m} - r_F\mathbf{u})\mathbf{C}^{-1}}{(\mathbf{m} - r_F\mathbf{u})\mathbf{C}^{-1}\mathbf{u}^T}$$

- Any portfolio  $(\sigma, \mu)$  on the CML satisfies:  $\mu = r_F + \frac{\mu_M - r_F}{\sigma_M} \sigma$ 
  - The term  $\frac{\mu_M - r_F}{\sigma_M} \sigma$  is called the *risk premium*.
  - It is the additional return over and above the risk-free return  $r_F$ , which compensates for exposure to risk.

# CAPM: Capital Asset Pricing Model

- We know that CML is tangent to the efficient frontier at  $(\sigma_M, \mu_M)$ .
- Assume another portfolio  $V$  characterized by  $(\sigma_V, \mu_V)$ .
- We can construct a range of portfolios from  $M$  and  $V$ .
- These portfolios should form a hyperbola which should technically be a tangent to the CML at  $(\sigma_M, \mu_M)$ .
- Assume the weight of any such portfolio to be  $x$  for  $V$  and  $(1 - x)$  for  $M$ .

$$\sigma_P = \left( x^2 \sigma_V^2 + (1 - x)^2 \sigma_M^2 + 2x(1 - x) \text{Cov}(K_V, K_M) \right)^{\frac{1}{2}}$$

$$\mu_P = x\mu_V + (1 - x)\mu_M$$

# CAPM: Capital Asset Pricing Model

- Taking derivatives with respect to  $x$  at  $x = 0$ , we get

$$\left. \frac{\partial \sigma_P}{\partial x} \right|_{x=0} = \frac{\text{Cov}(K_V, K_M) - \sigma_M^2}{\sigma_M}; \left. \frac{\partial \mu_P}{\partial x} \right|_{x=0} = \mu_V - \mu_M$$

- The slope of CML should equate the slope of the curve at  $x = 0$ .

$$\frac{\frac{\mu_V - \mu_M}{\text{Cov}(K_V, K_M) - \sigma_M^2}}{\sigma_M} = \frac{\mu_M - r_F}{\sigma_M}$$

- Solving this gets us:

$$\mu_V = r_F + \frac{\text{Cov}(K_V, K_M)}{\sigma_M^2} (\mu_M - r_F)$$

- This is referred to as the *beta factor*, given as:  $\beta_V = \frac{\text{Cov}(K_V, K_M)}{\sigma_M^2}$ .



# CAPM: Capital Asset Pricing Model

## CAPM theorem

- Suppose that the risk-free rate  $r_F$  is lower than the expected return  $\mu_{MVP}$  of the minimum variance portfolio (so that the market portfolio  $M$  exists).
- Then, the expected return  $\mu_V$  on any feasible portfolio  $V$  is given by
$$\mu_V = r_F + \beta_V(\mu_M - r_F)$$
- The term  $\beta_V(\mu_M - r_F)$  is called risk premium.
- This becomes same as the risk premium of the market portfolio if  $V$  is on the capital market line.

$$\beta_V(\mu_M - r_F) = \frac{\mu_M - r_F}{\sigma_M} \sigma_V$$

- Suppose we want to approximate  $K_V$  by a linear function  $\beta K_M + \alpha$ , the error of approximation is given as

$$\epsilon = K_V - (\beta K_M + \alpha)$$

# Thank you!