



Università Commerciale  
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# Lecture 5 – Introduction to Mean-Variance Analysis: the Opportunity Set and the Efficient Frontier

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Prep Course in Quant Methods for Finance

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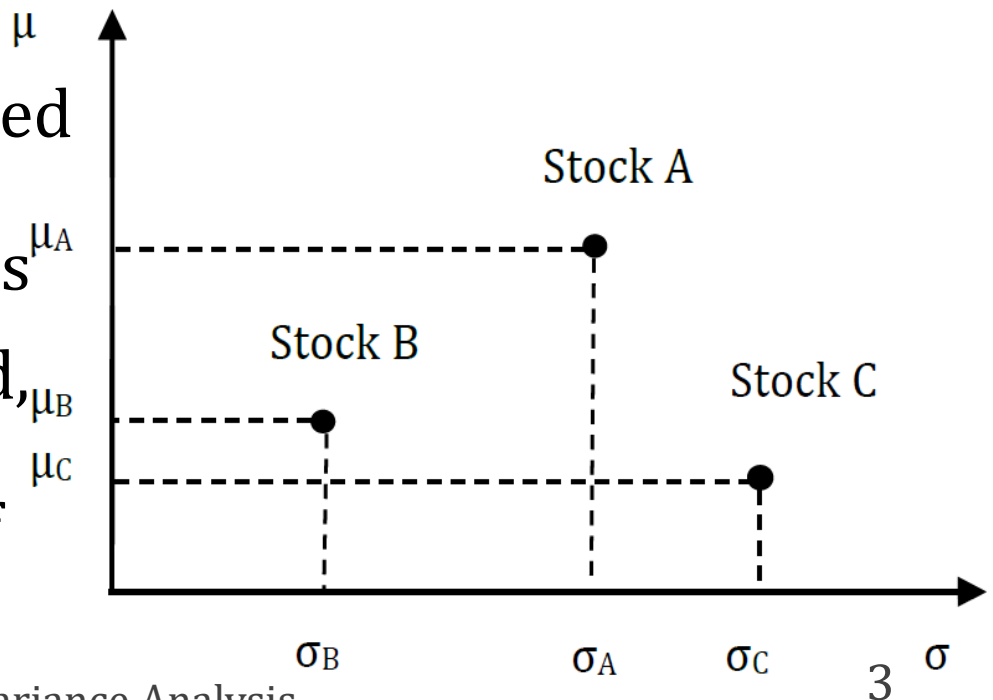
# Outline and objectives

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- Mean-variance and efficient frontiers: logical meaning
  - Guidolin-Pedio, chapter 3, sec. 1
- The case of no borrowing and lending and two risky assets
  - Guidolin-Pedio, chapter 3, sec. 1.1
- Generalizations to the case of  $N$  risky assets
  - Guidolin-Pedio, chapter 3, sec. 1.2
- Two-fund separation result
  - Guidolin-Pedio, chapter 3, sec. 1.2
- Extension to unlimited borrowing and lending
  - Guidolin-Pedio, chapter 3, sec. 2
- Limited borrowing and lending
  - Guidolin-Pedio, chapter 3, sec. 2
- Short-sale constraints
  - Guidolin-Pedio, chapter 3, sec. 3

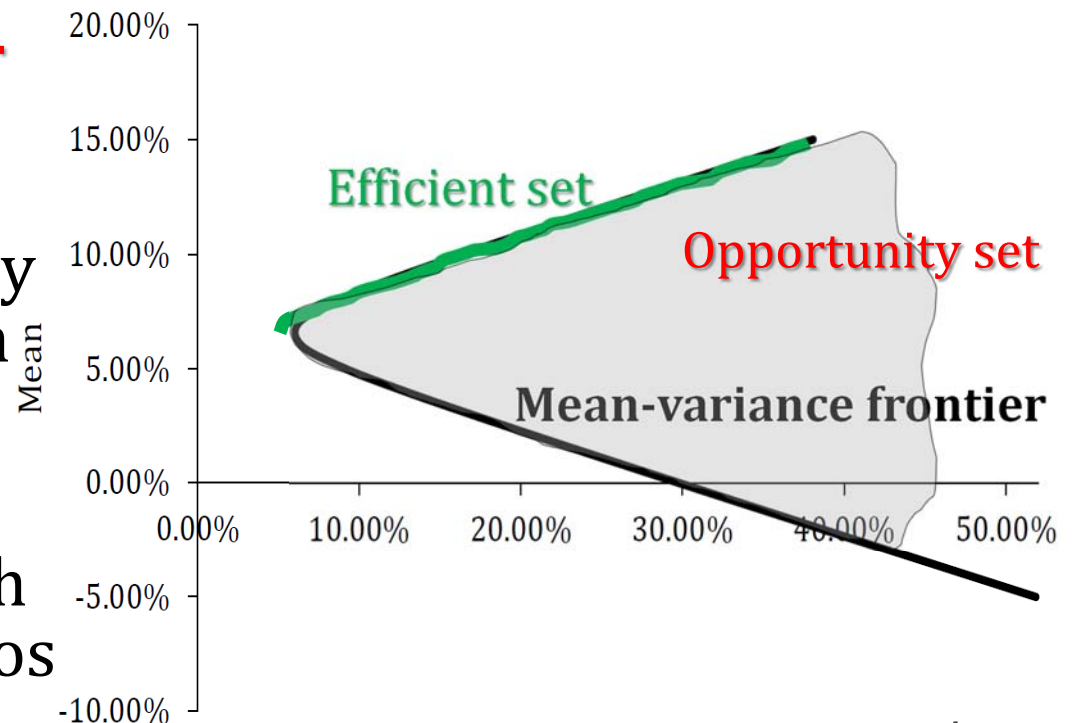
# Mean Variance Framework: Key Concepts

- We review the development of the celebrated mean-variance framework introduced by Markowitz in the 1950s
- Initially at least, risky assets only, no borrowing or lending
- Assume that for some reasons, the joint distribution of asset returns is completely characterized by their means, variances, and covariances
- We represent each asset (and portfolio of assets) in a two-dimensional diagram, where expected portfolio return is plotted on the vertical axis and standard deviation is on the horizontal axis
- Not all securities may be selected, e.g., stock C is dominated by the remaining two stocks in terms of MV dominance



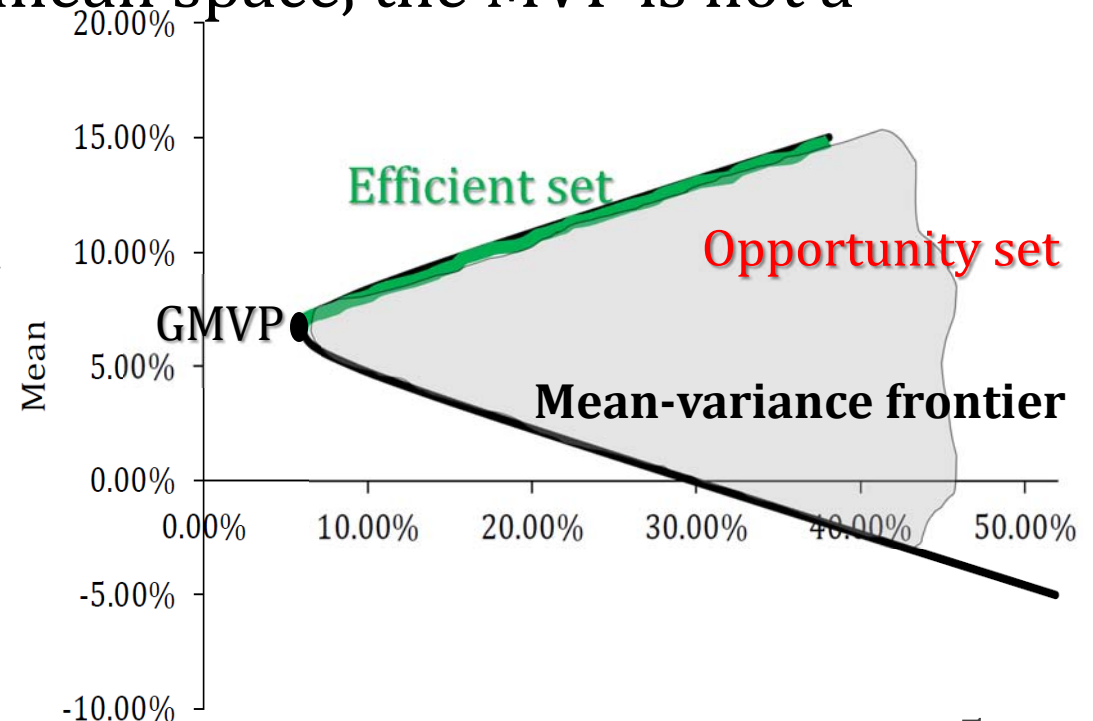
# Mean Variance Framework: Key Concepts

- According to **MV criterion** a portfolio is efficient if and only if there is no other portfolio that allows the investor to achieve the same expected return with a lower level of risk or a higher level of expected return with the same level of risk
- Three key notions: i) the **opportunity set** (feasible region), which includes all the portfolios (both efficient and inefficient) that the investor is able to build given securities in the asset menu
- (ii) the **mean-variance frontier** (aka minimum variance frontier, MVF) subset of the opportunity set containing only the portfolio(s) with minimum variance for any target level of expected returns
- (iii) the **efficient frontier**, which only includes efficient portfolios



# Mean Variance Framework: Key Concepts

- Because it is possible that a portfolio exists which has a higher return than another portfolio with the same level of risk, only portfolios that have a higher expected return than the global minimum variance portfolio (GMVP) are efficient
- The preferences of the investor(s) for risk are not relevant to the determination of the efficient frontier
- In the classical standard dev.-mean space, the MVF is not a function, but a «right-rotated hyperbola»
- The GMVP is Global Minimum Variance Ptf. and it is of high interest because «separates» the efficient set from the MVF
- The structure GMVP does not depend on expected returns



# The Efficient Frontier with Two Risky Assets

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- Assume no borrowing or lending at the risk-free rate
- Re-cap of a few basic algebraic relationships that exploit the fact that with two risky assets,  $\omega_B = 1 - \omega_A$ 
  - See textbook for detailed derivations  $\mu_P = \omega_A \mu_A + (1 - \omega_A) \mu_B$
  - Portfolio mean & variance:  $\sigma_P^2 = \omega_A^2 \sigma_A^2 + (1 - \omega_A)^2 \sigma_B^2 + 2\omega_A(1 - \omega_A)\sigma_{AB}$
  - Using the definitions of correlation and of standard deviation:
$$\sigma_P = \sqrt{\omega_A^2 \sigma_A^2 + (1 - \omega_A)^2 \sigma_B^2 + 2\omega_A(1 - \omega_A)\rho_{AB}\sigma_A\sigma_B}$$
  - Solve std. dev equation for  $\omega_A$  and plug the result into mean equation  $\Rightarrow$  a system of 2 equations in 2 unknowns
  - The system has in general a unique solution  $\Rightarrow$  the opportunity set is a curve and it coincides with the mean-variance frontier (there is only one possible level of risk for a given level of return)
  - The shape of set depends on the correlation between the 2 securities
- Three possible cases: (i)  $\rho_{AB} = +1$ ; (ii)  $\rho_{AB} = -1$ ; (iii)  $\rho_{AB} \in (-1, 1)$
- Case (i):  $\rho_{AB} = +1$ : the expression for  $\sigma_P^2$  becomes a perfect square sum and this simplifies the algebra

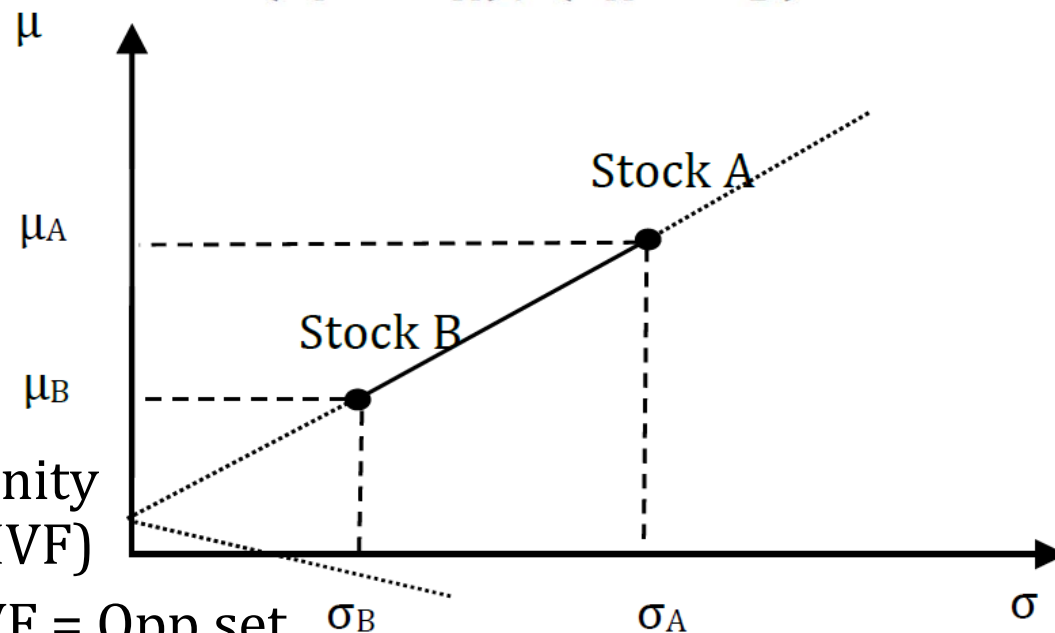
# The Efficient Frontier with Two Risky Assets

- After algebra (see textbook), we have:

$$\mu_P = \frac{\sigma_P - \sigma_B}{\sigma_A - \sigma_B} \mu_A + \left(1 - \frac{\sigma_P - \sigma_B}{\sigma_A - \sigma_B}\right) \mu_B = \mu_B + \frac{\sigma_P - \sigma_B}{\sigma_A - \sigma_B} (\mu_A - \mu_B)$$

the equation of a straight line, with slope  $(\sigma_P - \sigma_A)/(\sigma_A - \sigma_B)$

- In the picture, dashed lines == portfolios require short selling
- Without short sales, the least risky stock == GMVP
- With short sales, the GMVP has zero risk
- In this special case, the opportunity set = mean-variance frontier (MVF)
- With no short sales, EffSet = MVF = Opp set



- Case (ii):  $\rho_{AB} = -1$ : the expression for  $\sigma_P^2$  becomes a perfect square difference and this simplifies the algebra (see textbook) to yield:

$$\sigma_P = \omega_A \sigma_A - (1 - \omega_A) \sigma_B \quad \text{or to} \quad \sigma_P = -\omega_A \sigma_A + (1 - \omega_A) \sigma_B$$

- Yet, each of the equations only holds when the RHS is positive

# The Efficient Frontier with Two Risky Assets

- The opportunity set is a straight line, but its slope depends on which of the equations above holds
- If the first equation applies, the opportunity set is equal to:

$$\mu_P = \frac{\sigma_P + \sigma_B}{\sigma_A + \sigma_B} \mu_A + \left(1 - \frac{\sigma_P + \sigma_B}{\sigma_A + \sigma_B}\right) \mu_B = \mu_B + \frac{\sigma_P + \sigma_B}{\sigma_A + \sigma_B} (\mu_A - \mu_B)$$

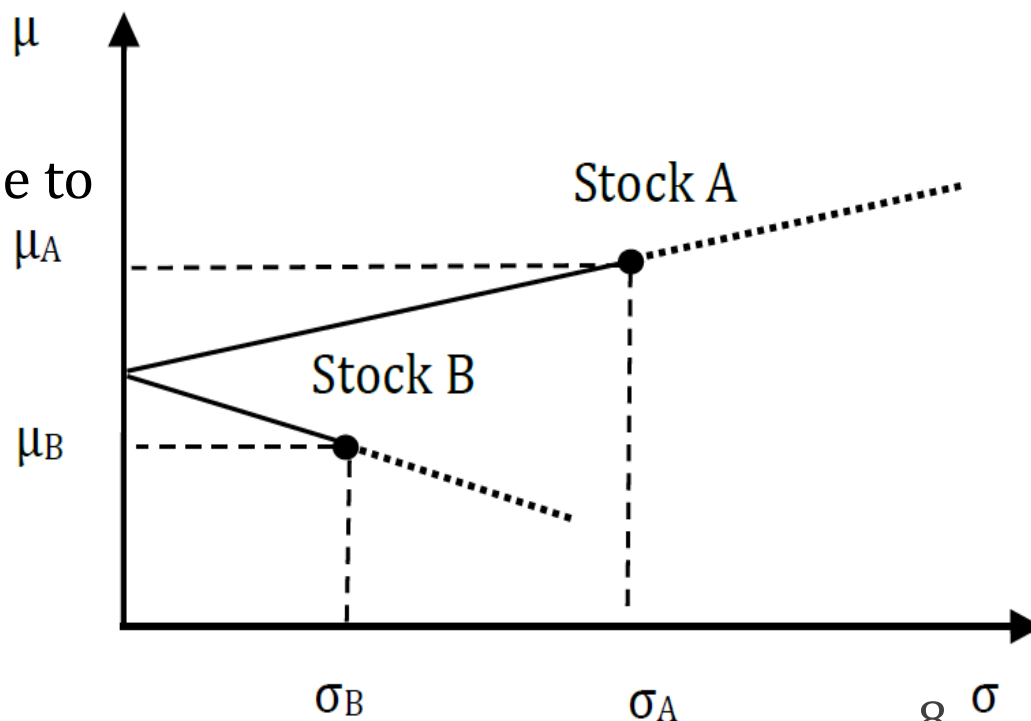
while if the second equation holds, the opportunity set is equal to:

$$\mu_P = \frac{\sigma_B - \sigma_P}{\sigma_A + \sigma_B} \mu_A + \left(1 - \frac{\sigma_B - \sigma_P}{\sigma_A + \sigma_B}\right) \mu_B = \mu_B + \frac{\sigma_B - \sigma_P}{\sigma_A + \sigma_B} (\mu_A - \mu_B)$$

- In the picture, dashed lines ==  $\mu$  portfolios require short selling
- Even without short sales, possible to find a combination that has zero  $\mu_A$  variance, i.e., it is risk-free
- Such a riskless portfolio is GMVP
- The expression for such a ptf. is:

$$\omega_A^{GMVP} \sigma_A - (1 - \omega_A^{GMVP}) \sigma_B = 0 \text{ or}$$

$$-\omega_A^{GMVP} \sigma_A + (1 - \omega_A^{GMVP}) \sigma_B = 0$$



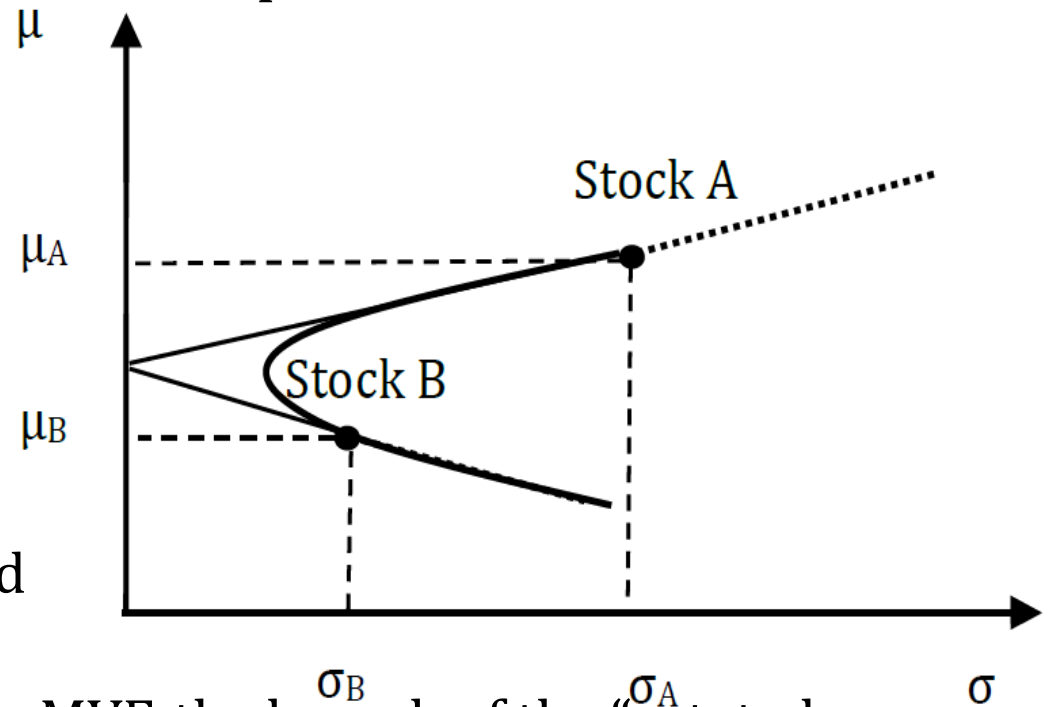


# The Efficient Frontier with Two Risky Assets

$$\implies \omega_A^{GMVP} = \frac{\sigma_B}{\sigma_A + \sigma_B}$$

- Case (iii):  $\rho_{AB} \in (-1, 1)$ : In this case, although tricks exist to trace it out, the MVF does not have a closed-form expression

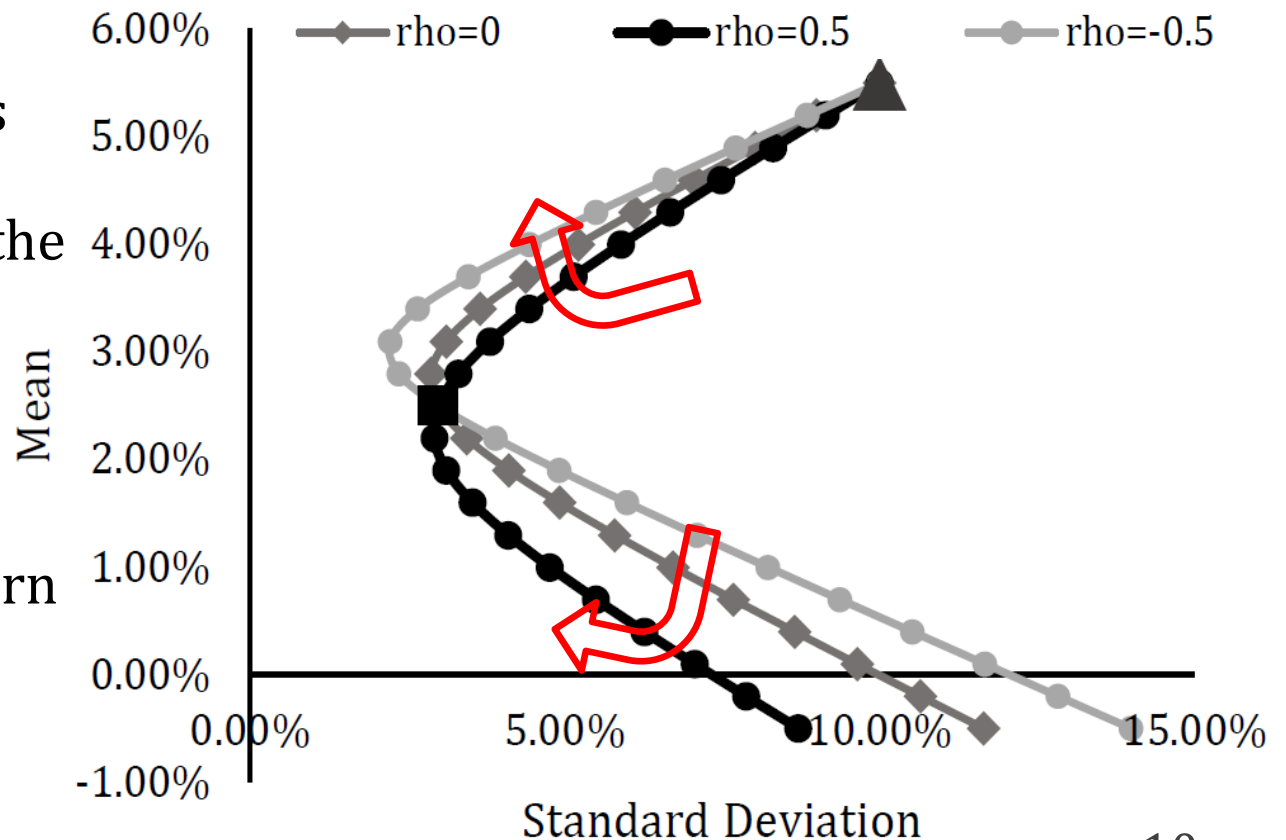
- The MVF is non-linear, a parabola (i.e., a quadratic function) in the variance-mean space
- Or a (branch of) hyperbola in standard deviation-mean space
- In such a space, the MVF is not a function, it is just a «correspondence», a “right-rotated hyperbola”
- The efficient set == a portion of the MVF, the branch of the “rotated hyperbola” that lies above (and includes) the GMVP
- To distinguish the efficient set from the MVF we have to find the GMVP:



$$\frac{\partial \sigma_P^2}{\partial \omega_A} = 2\omega_A \sigma_A^2 - 2(1 - \omega_A) \sigma_B^2 + 2(1 - 2\omega_A) \rho_{AB} \sigma_A \sigma_B \implies \omega_A^{GMVP} = \frac{\sigma_B^2 - \rho_{A,B} \sigma_A \sigma_B}{\sigma_A^2 + \sigma_B^2 - 2\rho_{A,B} \sigma_A \sigma_B}$$

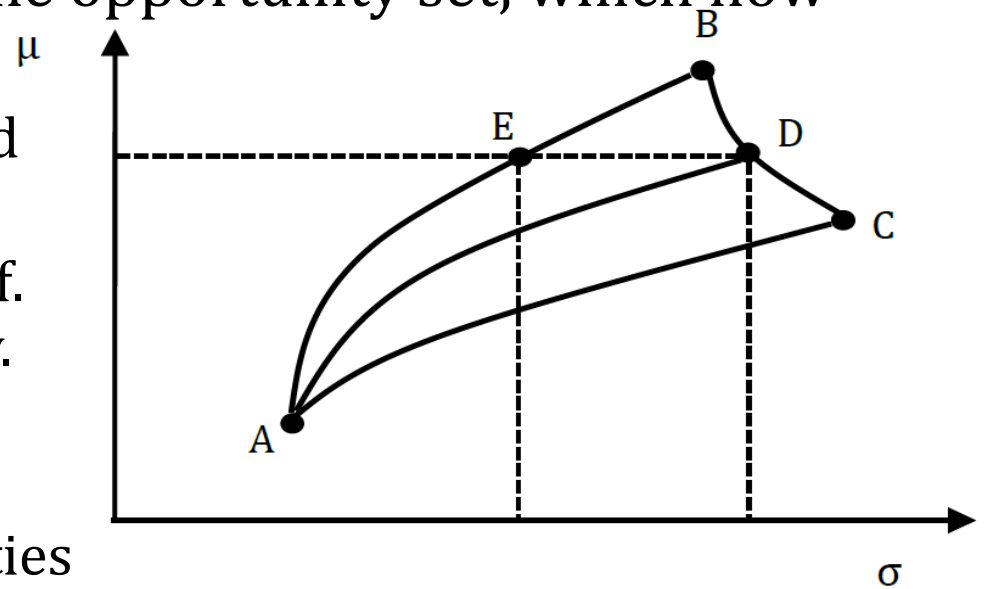
# One Numerical Example

- For instance, consider stock A with  $\mu_A = 5.5\%$  and  $\sigma_A = 10\%$ , and stock B with  $\mu_B = 2.5\%$  and  $\sigma_B = 3\%$
- Draw MVF in Excel for  $\rho_{A,B} = 0$ ,  $\rho_{A,B} = 0.5$ , and  $\rho_{A,B} = -0.5$
- See textbook for calculations and details and book's website for exercises in Excel related to this case
  - When the  $\rho_{A,B} < 0$ , it is possible to form ptf. that have a lower risk than each of the 2 assets
  - Clearly, as  $\rho_{A,B}$  declines risk characterizing the GMVP moves towards the left, inward
  - The entire MVF rotates upward, less risk may be borne for identical expected portfolio return
  - Note that the GMVP often needs to include short positions



# The Case of N Risky Assets

- Usually investors choose among a large number of risky securities
  - E.g., allocation among the 500 stocks in the S&P 500
- Extend our framework to the general case, with  $N$  risky assets
- The MVF no longer coincides with the opportunity set, which now becomes **a region** and not a line



- Ptf. D, a combination of assets B and C, is not MV efficient
- It gives the same mean return as ptf. E but implies a higher standard dev. and a risk-averse investor would never hold portfolio D

- To exclude all the inefficient securities and ptf., as first step the investor needs to trace out the MVF, i.e., select ptf. with minimum variance (std. dev.) for each level of  $\mu$

- Only interested in the upper bound of the feasible region

- We solve the following **quadratic programming problem**:

$$\min_{\{\omega\}} \frac{1}{2} \omega' \Sigma \omega \text{ Subject to}$$

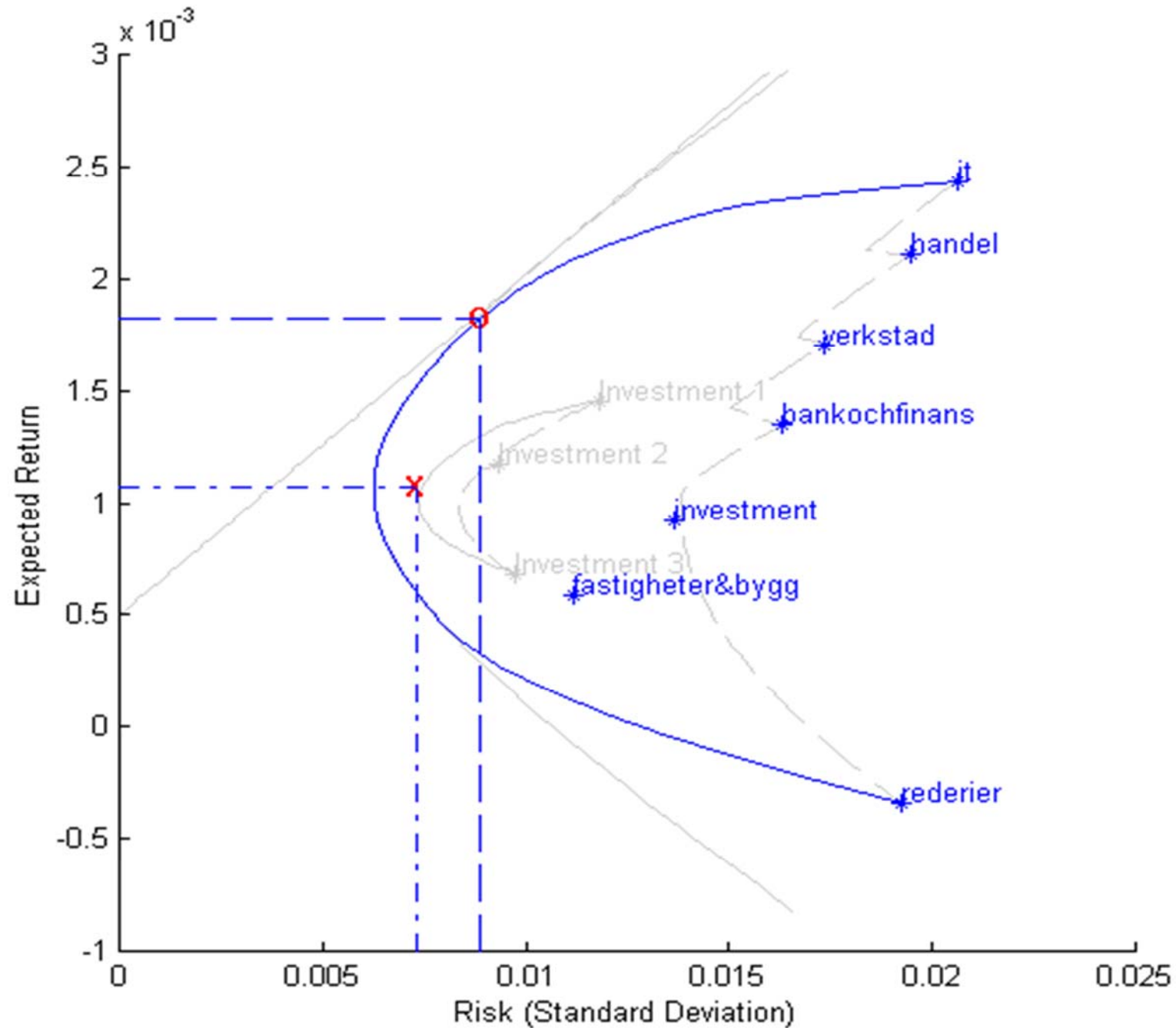
$N \times 1$  vector of 1s

$$\mathbf{1}' \omega = 1$$

$$\mu' \omega = \bar{\mu}$$

Target mean

# The Case of N Risky Assets



# The Case of N Risky Assets

- For the time being, no short-sale restrictions have been imposed
  - To solve the program, assume that no pair or general combination of asset returns are linearly dependent
  - $\Rightarrow \Sigma$  is nonsingular and invertible; in fact,  $\Sigma$  is (semi-)positive definite
- Under these conditions, it is a constrained minimization problem that can be solved through the use of Lagrangian multiplier method
- See the textbook for algebra and details
- If one defines  $A \equiv \boldsymbol{\mu}' \boldsymbol{\Sigma}^{-1} \boldsymbol{\iota}$ ,  $B \equiv \boldsymbol{\mu}' \boldsymbol{\Sigma}^{-1} \boldsymbol{\mu}$ ,  $C \equiv (\boldsymbol{\iota}' \boldsymbol{\Sigma}^{-1} \boldsymbol{\iota})$ , and  $D = BC - A^2$  then the unique solution to the problem,  $\boldsymbol{\omega}^*$ , is:

$$\boxed{\boldsymbol{\omega}^* = \mathbf{g} + \mathbf{h} \bar{\mu}} \quad \mathbf{g} = \frac{1}{D} [B(\boldsymbol{\Sigma}^{-1} \boldsymbol{\iota}) - A(\boldsymbol{\Sigma}^{-1} \boldsymbol{\mu})], \quad \mathbf{h} = \frac{1}{D} [C(\boldsymbol{\Sigma}^{-1} \boldsymbol{\mu}) - A(\boldsymbol{\Sigma}^{-1} \boldsymbol{\iota})]$$

i.e., any combination of MVF ptf. weights gives another MVF ptf.

- Consider two MVF ptf.  $P_1$  and  $P_2$  with mean  $\mu_{P_1}$  and  $\mu_{P_2}$ , and assume that  $P_3$  is a generic portfolio on the MVF: always possible to find a quantity  $x$  such that  $\mu_{P_3} = x\mu_{P_1} + (1-x)\mu_{P_2}$
- Other MVF ptf:  $\boldsymbol{\omega}_{P_3} = x\boldsymbol{\omega}_{P_1} + (1-x)\boldsymbol{\omega}_{P_2} = x(\mathbf{g} + \mathbf{h}\mu_{P_1}) + (1-x)(\mathbf{g} + \mathbf{h}\mu_{P_2})$   
 $= \mathbf{g} + \mathbf{h}(x\mu_{P_1} + (1-x)\mu_{P_2}) = \mathbf{g} + \mathbf{h}\mu_{P_3},$

# Recap: The Efficient Frontier, in General

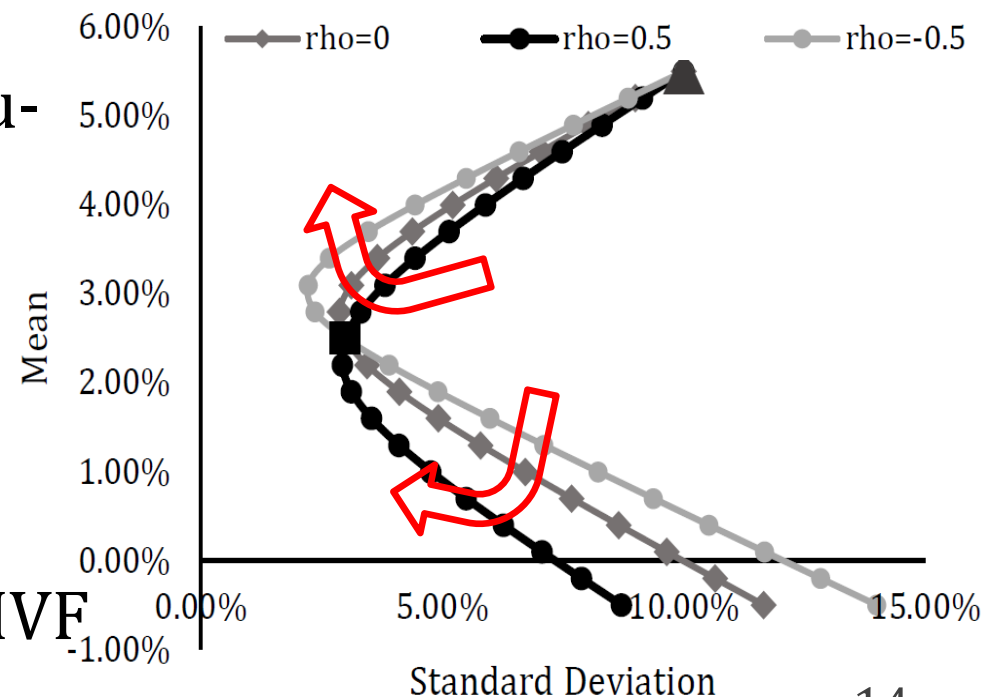
- The GMVP and the entire MVF depend strongly on the correlation structure of security returns: the lower are the correlations (on average), the more the efficient set moves up and to the left, improving the risk-expected return trade-off
- The position and shape of the MVF reflects the diversification opportunities that a given asset menu offers
- Even though, MVF ptf's are solutions of a complex quadratic programming problem, in the absence of constraints, their structure is relatively simple:

$$\omega^* = \mathbf{g} + \mathbf{h} \bar{\mu}$$

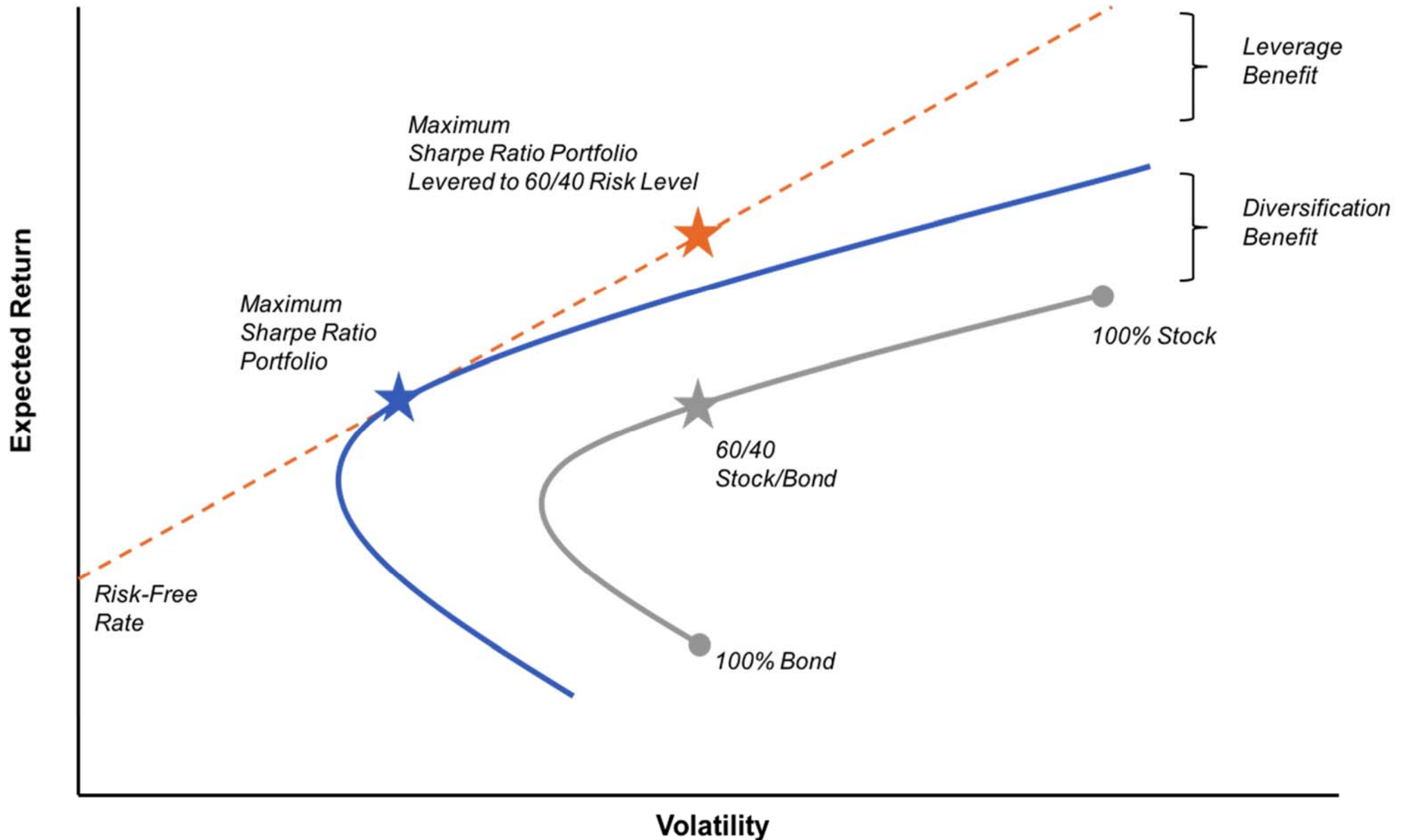
$$\mathbf{g} = \frac{1}{D} [B(\Sigma^{-1} \boldsymbol{\iota}) - A(\Sigma^{-1} \boldsymbol{\mu})]$$

$$\mathbf{h} = \frac{1}{D} [C(\Sigma^{-1} \boldsymbol{\mu}) - A(\Sigma^{-1} \boldsymbol{\iota})]$$

- Combinations of MVF ptf's. are MVF



# Recap: The Efficient Frontier, in General



# Two-Fund Separation

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It is sufficient to know two points (portfolios) on the mean-variance frontier to generate all the others

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- All MV-optimizers are satisfied by holding a combination of **two mutual funds** (provided they are MV efficient), regardless of preferences
- Their heterogeneous preferences will only impact the way in which they combine the two funds that they choose to hold
- In equilibrium, if all investors are rational MV optimizers, the market portfolio, being a convex combination of the optimal portfolios of all the investors, has to be an efficient set portfolio
- As for the shape of MVF when  $N$  assets are available, this is a rotated hyperbola as in case of 2 assets:
$$\sigma_P^2 = \frac{1}{D} [C(\mu_P)^2 - 2A\mu_P + B]$$
  - Equation of a parabola with vertex in  $((1/C)^{1/2}, A/C)$ , which also represents the global minimum variance portfolio
  - The textbook shows that GMV weights are:  $\mathbf{w}_{GMVP} = \frac{\Sigma^{-1} \mathbf{1}}{C} = \frac{\Sigma^{-1} \mathbf{1}}{\mathbf{1}'\Sigma^{-1}\mathbf{1}}$



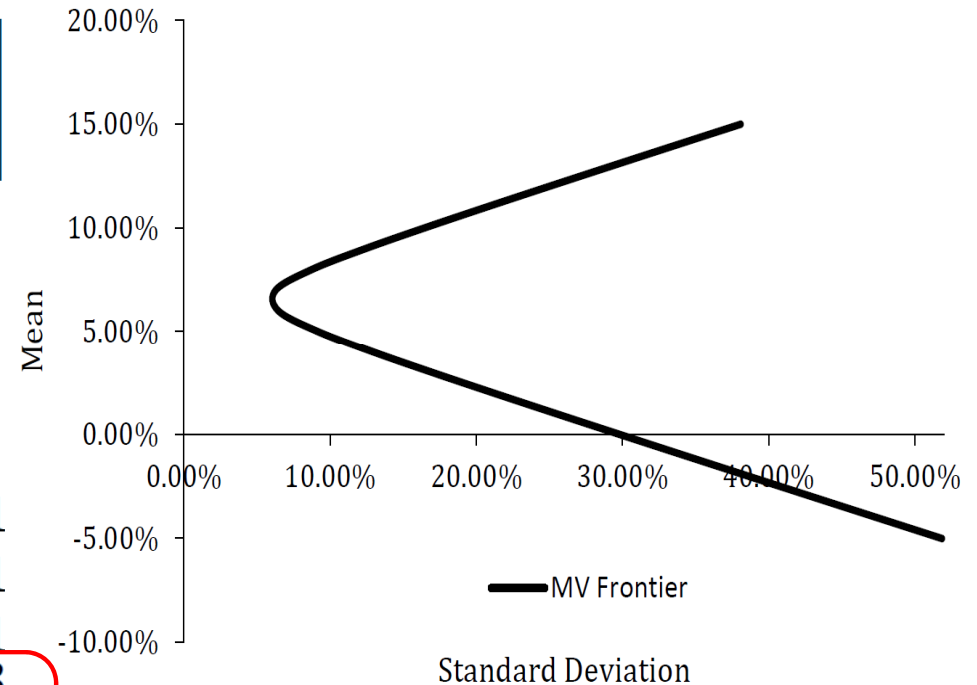
# One Strategic Asset Allocation Example

- Consider three assets – U.S. Treasury, corporate bonds, and equity – characterized by the mean vector and the variance-covariance matrix:

$$\mu = \begin{bmatrix} 6.00 \\ 7.50 \\ 9.00 \end{bmatrix} \quad \Sigma = \begin{bmatrix} 0.005 & 0.004 & -0.002 \\ 0.004 & 0.008 & 0.003 \\ -0.002 & 0.003 & 0.025 \end{bmatrix}$$

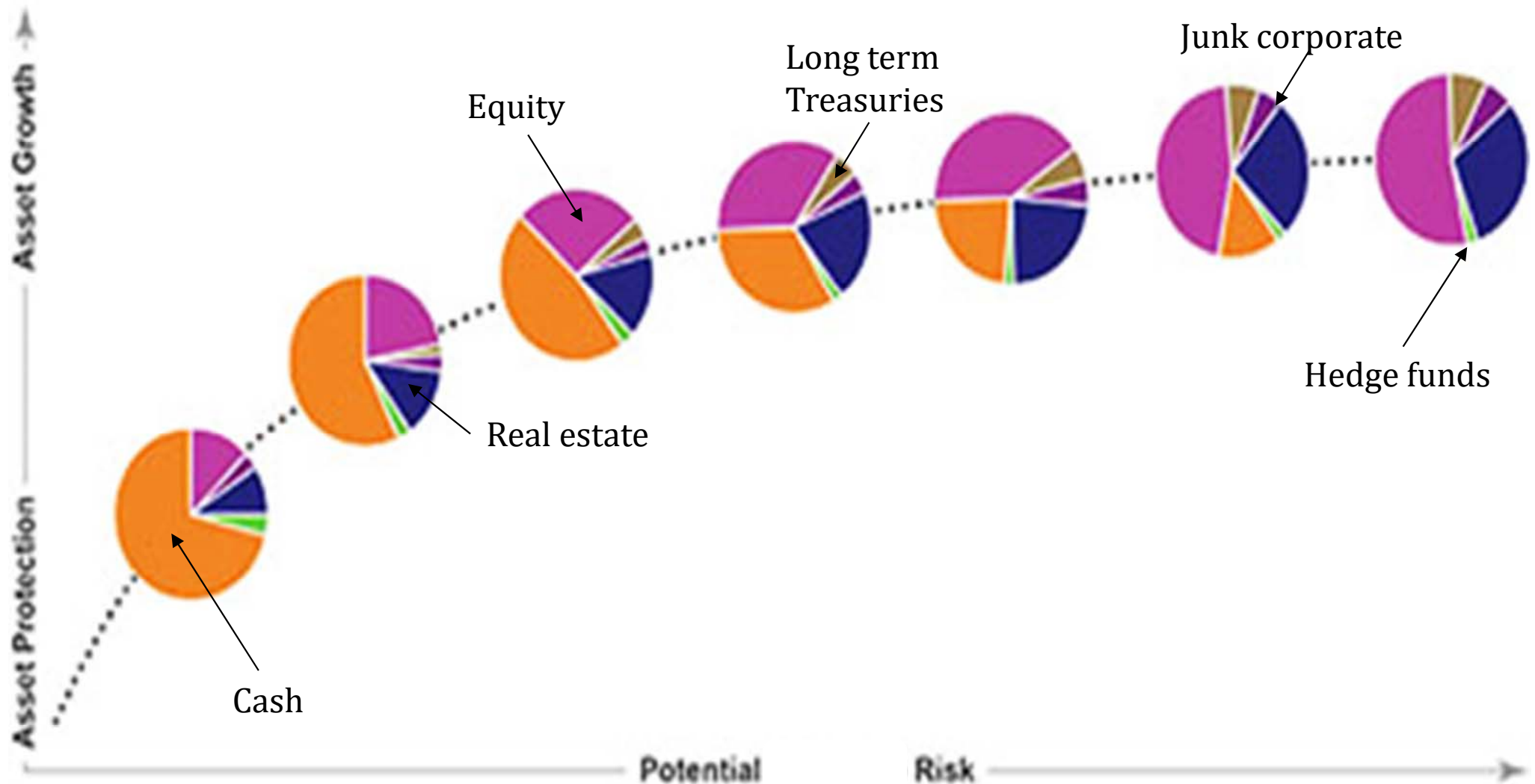
- The textbook guides you to perform calculations of A, B, C, D **using Excel**:

$$\begin{aligned} \mathbf{g} &= \frac{1}{D} [B(\Sigma^{-1} \mathbf{1}) - A(\Sigma^{-1} \boldsymbol{\mu})] \\ &= \frac{1}{14.21} \left\{ 1.26 \cdot \begin{bmatrix} 402.11 & -223.16 & 58.95 \\ -223.16 & 254.74 & -48.42 \\ 58.95 & -48.42 & 50.53 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right. \\ &\quad \left. \cdot \begin{bmatrix} 402.11 & -223.16 & 58.95 \\ -223.16 & 254.74 & -48.42 \\ 58.95 & -48.42 & 50.53 \end{bmatrix} \cdot \begin{bmatrix} 6.00 \\ 7.50 \\ 9.00 \end{bmatrix} \right\} = \begin{bmatrix} 4.63 \\ -3.27 \\ -0.37 \end{bmatrix} \end{aligned}$$



$$\begin{aligned} \mathbf{h} &= \frac{1}{D} [C(\Sigma^{-1} \boldsymbol{\mu}) - A(\Sigma^{-1} \mathbf{1})] = \frac{1}{14.21} \left\{ 282.10 \cdot \begin{bmatrix} 402.11 & -223.16 & 58.95 \\ -223.16 & 254.74 & -48.42 \\ 58.95 & -48.42 & 50.53 \end{bmatrix} \cdot \begin{bmatrix} 6.00 \\ 7.50 \\ 9.00 \end{bmatrix} \right. \\ &\quad \left. - 18.5 \cdot \begin{bmatrix} 402.11 & -223.16 & 58.95 \\ -223.16 & 254.74 & -48.42 \\ 58.95 & -48.42 & 50.53 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right\} = \begin{bmatrix} -57.78 \\ 48.89 \\ 8.89 \end{bmatrix} \end{aligned}$$

# Unlimited, Riskless Borrowing and Lending



- Always recall that as you move on the MVF frontier, ptf. structure is affected, often in ways that are hard to guess (see the evolution of the green slice)

# Unlimited, Riskless Borrowing and Lending

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- So far, we have ignored the existence of a **risk-free asset** == a security with return  $R^f$  known with certainty and zero variance and zero covariance with all risky assets
  - Buying such a riskless asset == lending at a risk-free rate to issuer
  - Assume investor is able to leverage at riskless rate
  - There is no limit to the amount that the investor can borrow or lend at the riskless rate (we shall remove this assumption later)
- Fictional experiment in which the possibility to borrow and lend at  $R^f$  is offered to investor who already allocated among N risky assets
- X is the fraction of wealth in an efficient frontier, risky portfolio (A) characterized  $\mu_A$  and  $\sigma_A$ , respectively; a share  $1 - X$  is invested in the riskless asset, to obtain mean and standard deviation:

$$\mu_P = X\mu_A + (1 - X)R^f = R^f + X(\mu_A - R^f)$$

$$\sigma_P = \sqrt{\sigma_P^2} = \sqrt{X^2\sigma_A^2} = X\sigma_A$$

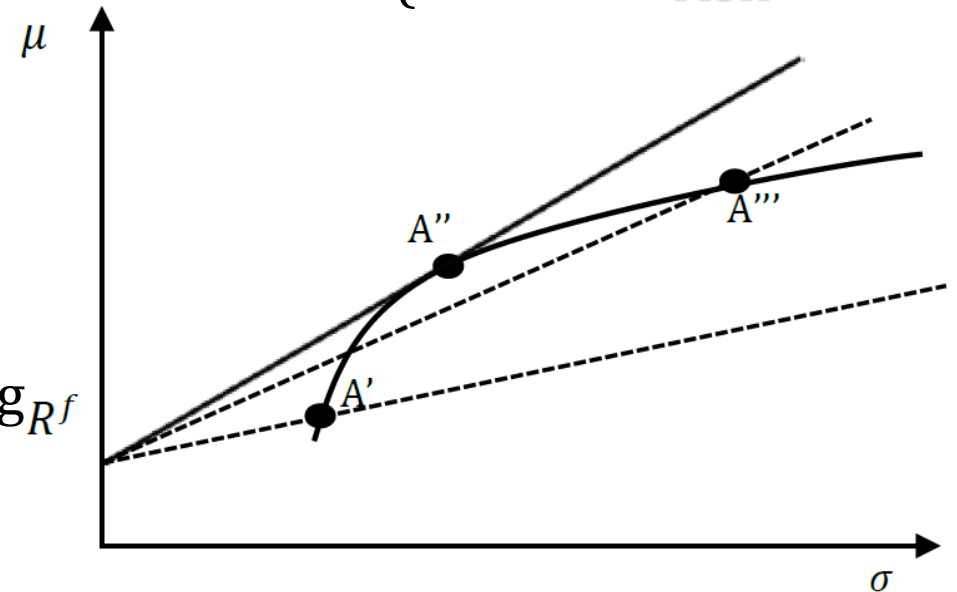
- Solving from X in the first equation and plugging into the second:

$$\mu_P = R^f + \frac{\sigma_P}{\sigma_A}(\mu_A - R^f) = R^f + \frac{(\mu_A - R^f)}{\sigma_A}\sigma_P$$

# Unlimited, Riskless Borrowing and Lending

The capital transformation line measures at what rate unit risk (st. dev.) can be transformed into average excess return (risk premium)

- The equation of a straight line with intercept  $R^f$  and slope  $(\mu_A - R^f)/\sigma_A$
- This line is sometimes referred to as **capital transformation line**
- The term  $(\mu_A - R^f)/\sigma_A$  is called **Sharpe ratio** (SR), the total reward for taking a certain amount of risk, represented by the st. dev.
  - SR is the mean return in excess of the risk-free rate (called the **risk premium**) per unit of volatility
  - The plot shows 3 transformation lines for 3 choices of the risky benchmark A ( $A'$ ,  $A''$ , and  $A'''$ ) on the efficient frontier
  - Points to the left of A involve lending  $R^f$  at the risk-free rate while the ones to the right involve borrowing
  - As investors prefer more to less, they will welcome a “rotation” of the straight line passing through  $R^f$  as far as possible in a counterclockwise direction, **until tangency**



# The Tangency Portfolio and the Capital Market Line

Under no frictions and homogeneous beliefs, there exists a tangency ptf. that maximizes the slope of the transformation line

- Assuming beliefs are homogeneous, there are no frictions or taxes, and that individuals face the same  $R^f$  and identical asset menus, all rational, non-satiated investors hold the same **tangency portfolio**
- It is combined with a certain share of risk-free lending or borrowing
- While the share of wealth an investor lends or borrows at  $R^f$  depends on the investor's preference for risk, **the risky portfolio should be the same for all the investors**

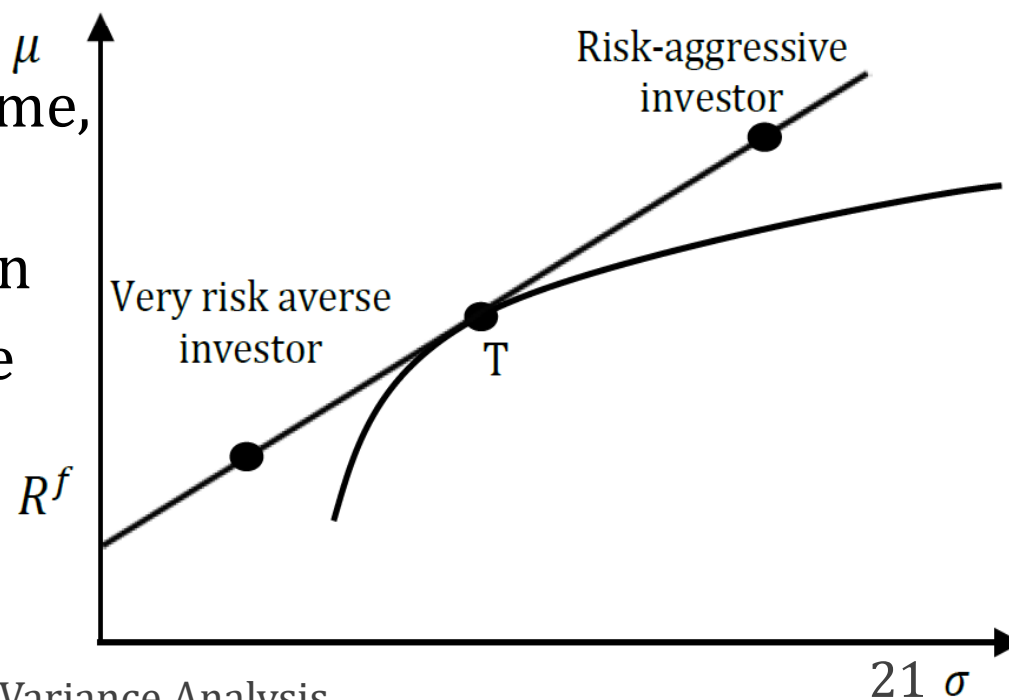
- The steepest CTL gets a special name, the **Capital Market Line** (CML)

- Special case of two-fund separation

- To determine the tangency ptf. one needs to solve:

$$\max_{\{\omega\}} \frac{(\omega' \mu - R^f)}{(\omega' \Sigma \omega)^{\frac{1}{2}}}$$

subject to  $\omega' \mathbf{1} = 1$



# The Tangency Portfolio and the Capital Market Line

- The textbook explains how the problem may be written as a simple unconstrained max problem that we can solve by solving the FOCs:

$$\max_{\{\omega\}} \frac{\omega'(\mu - R^f \mathbf{1})}{(\omega' \Sigma \omega)^{1/2}}$$

- The resulting vector of optimal ptf. weights is:  $\omega_T = \frac{\Sigma^{-1}(\mu - R^f \mathbf{1})}{A - CR^f}$
- Using the same data as in the strategic asset allocation example on three assets – U.S. Treasury, corporate bonds, and equity – we have:

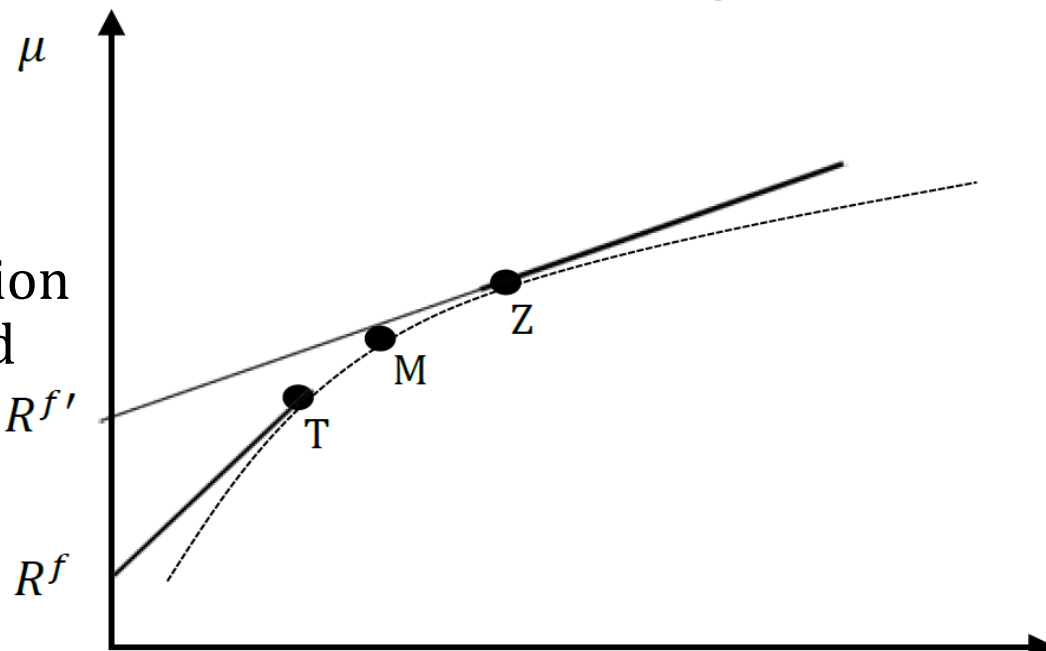
$$\omega_T = \frac{1}{18.5 - 282.11 \cdot 2.5\%} \begin{bmatrix} 402.11 & -223.16 & 58.95 \\ -223.16 & 254.74 & -48.42 \\ 58.95 & -48.42 & 50.53 \end{bmatrix} \begin{bmatrix} 3.50\% \\ 5.00\% \\ 6.50\% \end{bmatrix} = \begin{bmatrix} 0.59 \\ 0.16 \\ 0.25 \end{bmatrix}$$

- Textbook gives indications on how to use Microsoft Excel's Solver<sup>®</sup>
  - The Solver will iteratively change the values of the cells that contain the weights until the value of the Sharpe ratio is maximized
- Up to this point, we have assumed that the investor can borrow money at the same riskless rate at which she can lend
- More reasonable assumption: the investor is able to borrow money, but at a higher rate than the one of the risk free (long) investment

# Unlimited, Riskless Borrowing and Lending

When lending and borrowing is possible at different rates, it is no longer possible to determine a single tangency portfolio

- The figure shows how the CML is modified when borrowing is only possible at a rate  $R^{f'} > R^f$
- There are now two CTLs, both tangent to the efficient frontier
  - All the points falling on the portion of the efficient frontier delimited by T (below) and Z (above) will be efficient even though these do not fall on the straight, CML-type line
- While constructing the efficient frontier, we have assumed “equality” constraints (e.g., portfolio weights summing to one), but no “inequality” constraints (e.g., positive portfolio weights)
- Inequality constraints complicate the solution techniques
- However, unlimited short-selling assumption is often unrealistic (see margin accounts)



# Short-Selling Constraints

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- When short-selling is not allowed, portfolio weights should be positive, i.e. the constraint  $\omega \geq 0$  (to be interpreted in an element-by-element basis) has to be imposed
  - When  $\omega$  has to be positive, the unconstrained maximum may be at a value of that is not feasible
  - Therefore, it is necessary to impose the Kuhn-Tucker conditions
  - The textbook gives a heuristic introduction to what these are
  - Fortunately, Microsoft's Excel Solver<sup>®</sup> offers the possibility to solve the problem numerically, by-passing these complex analytical details
- Consider again our earlier strategic asset allocation example and let's set  $\bar{\mu} = 9\%$ 
  - In the absence of constraints, the solution is  $\omega_T = \begin{bmatrix} -56.66\% \\ 113.33\% \\ 43.33\% \end{bmatrix}$
  - This makes sense because the second asset is characterized by a large Sharpe ratio and hence must be exploited to yield a high mean return by leveraging the first security
  - Selling -57% of the first security is a major hurdle
  - Under nonnegativity constraints we obtain:  $\omega_T^{constrain} = \begin{bmatrix} 0\% \\ 0\% \\ 100\% \end{bmatrix}$



# Short-Selling Constraints

