

ECON 337901

FINANCIAL ECONOMICS

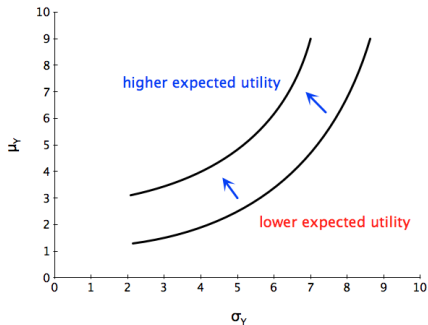
Peter Ireland

Boston College

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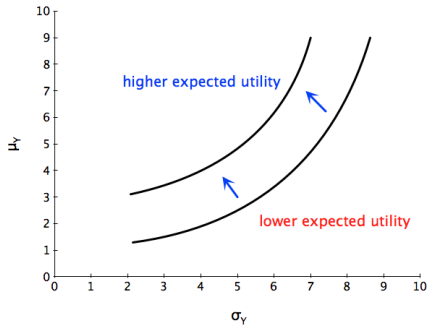
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Justifying Mean-Variance Utility



Since μ_Y is a “good” and σ_Y is a “bad,” indifference curves slope up. But the indifference curves are still convex.

Justifying Mean-Variance Utility



But what does the “budget constraint” look like in this diagram? To see, we need to consider the gains from diversification.

The Gains From Diversification

The position and shape of the constraint is determined in large part by the gains from diversification.

To see where these gains come from, consider forming a portfolio from two risky assets:

$\tilde{r}_1, \tilde{r}_2 =$ random returns

$\mu_1, \mu_2 =$ expected returns

$\sigma_1, \sigma_2 =$ standard deviations

Assume $\mu_1 > \mu_2$ and $\sigma_1 > \sigma_2$ to create a trade-off between expected return and risk **if** the investor must choose between one or the other.

The Gains From Diversification

The expected portfolio return

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

is a weighted average of the expected returns on the individual asset returns, but the standard deviation of the portfolio return

$$\sigma_P = [w^2\sigma_1^2 + (1 - w)^2\sigma_2^2 + 2w(1 - w)\sigma_1\sigma_2\rho_{12}]^{1/2}$$

is **not** a weighted average of the standard deviations of the returns on the individual assets and can be reduced by choosing a mix of assets ($0 < w < 1$) when ρ_{12} is less than one and, especially, when ρ_{12} is negative.

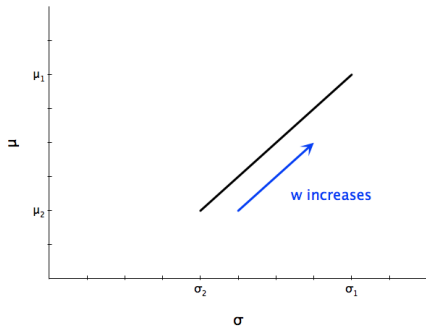
The Gains From Diversification

With $\rho_{12} = 1$,

$$\begin{aligned}\sigma_P &= [w^2\sigma_1^2 + (1-w)^2\sigma_2^2 + 2w(1-w)\sigma_1\sigma_2\rho_{12}]^{1/2} \\ &= [w^2\sigma_1^2 + (1-w)^2\sigma_2^2 + 2w(1-w)\sigma_1\sigma_2]^{1/2} \\ &= \{[w\sigma_1 + (1-w)\sigma_2]^2\}^{1/2} \\ &= |w\sigma_1 + (1-w)\sigma_2|.\end{aligned}$$

In this special case, the standard deviation of the return on the portfolio **is** a weighted average of the standard deviations of the returns on the individual assets.

The Gains From Diversification



When $\rho_{12} = 1$, so that individual asset returns are perfectly correlated, there are no gains from diversification.

The Gains From Diversification

With $\rho_{12} = -1$,

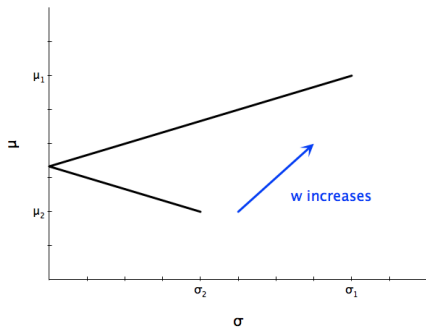
$$\begin{aligned}\sigma_P &= [w^2\sigma_1^2 + (1-w)^2\sigma_2^2 + 2w(1-w)\sigma_1\sigma_2\rho_{12}]^{1/2} \\ &= [w^2\sigma_1^2 + (1-w)^2\sigma_2^2 - 2w(1-w)\sigma_1\sigma_2]^{1/2} \\ &= \{[w\sigma_1 - (1-w)\sigma_2]^2\}^{1/2} \\ &= |w\sigma_1 - (1-w)\sigma_2|.\end{aligned}$$

In this special case, the setting

$$w = \frac{\sigma_2}{\sigma_1 + \sigma_2}$$

creates a “synthetic” risk free portfolio!

The Gains From Diversification



When $\rho_{12} = -1$, so that individual asset returns are perfectly, but negatively correlated, risk can be eliminated via diversification.

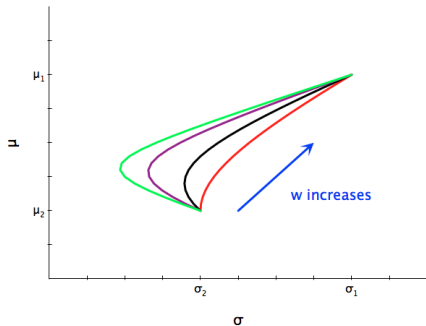
The Gains From Diversification

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

$$\sigma_P = [w^2\sigma_1^2 + (1 - w)^2\sigma_2^2 + 2w(1 - w)\sigma_1\sigma_2\rho_{12}]^{1/2}$$

In all intermediate cases, there will still be gains from diversification. These gains will become stronger as ρ_{12} declines from 1 to -1 .

The Gains From Diversification



As ρ_{12} decreases from 0.5 to 0 to -0.5 to -0.75, the gains from diversification strengthen.

The Gains from Diversification

$$\tilde{r}_p = w\tilde{r}_1 + (1 - w)\tilde{r}_2$$

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

$$\sigma_P = [w^2\sigma_1^2 + (1 - w)^2\sigma_2^2 + 2w(1 - w)\sigma_1\sigma_2\rho_{12}]^{1/2}$$

PS11: Suppose $\mu_1 = 8$, $\mu_2 = 4$, $\sigma_1 = 8$, and $\sigma_2 = 4$.

Calculate μ_P and σ_P for various values of w when $\rho_{12} = 0$ and $\rho_{12} = -0.5$. The gains from diversification are strongest when ρ_{12} is negative, but still present whenever $\rho_{12} < 1$.

The Gains from Diversification

From *Barron's*, 16 November 2015, pp.S3-4:

Richard Bernstein, chief investment officer at Richard Bernstein Advisors, was a top-rated investment strategist at Merrill Lynch before founding his own firm in 2009. The New York-based firm uses ETFs to execute its macro strategy. Bernstein is also behind three mutual funds, including the \$498 million [Eaton Vance Richard Bernstein All Asset Strategy](#) (ticker: EARAX), an asset-allocation mutual fund.

Is there still value in high-yield municipals?

Bernstein: You don't find that kind of pricing in high-yield munis anymore, but we still hold them. Treasuries we hold because they are, right now, negatively correlated to every other asset class. If you're looking for basic diversification, like we are, then you're always looking for negatively correlated assets to mute the volatility of your portfolio.

The Efficient Frontier

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

$$\sigma_P = [w^2\sigma_1^2 + (1 - w)^2\sigma_2^2 + 2w(1 - w)\sigma_1\sigma_2\rho_{12}]^{1/2}$$

In the case with two risky assets, the choice of w simultaneously determines μ_P and σ_P . But with more than two risky assets, the portfolio problem takes on an added dimension, since then we can ask: how can we select w_1, w_2, \dots, w_N to minimize σ_P for any given choice of μ_P ?

The Efficient Frontier

Consider two portfolios, A and B , with expected returns μ_A and μ_B and standard deviations σ_A and σ_B .

Portfolio A is said to exhibit **mean-variance dominance** over portfolio B if either

$$\mu_A > \mu_B \text{ and } \sigma_A \leq \sigma_B$$

or

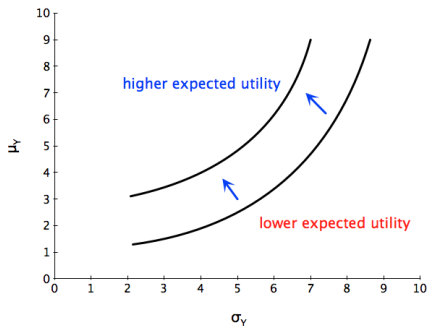
$$\mu_A \geq \mu_B \text{ and } \sigma_A < \sigma_B$$

The Efficient Frontier

Hence, choosing portfolio shares to minimize variance for a given mean will allow us to characterize the **efficient frontier**:

1. The set of all portfolios that are **not** mean-variance dominated by any other portfolio.
2. The set of all portfolios that are of potential interest to investors with mean variance utility.
3. The “budget constraint” in Markowitz’s diagram.

The Efficient Frontier



Here are the indifference curves in Markowitz's diagram. Now we want to find out what the constraint looks like when there are more than two risky assets.

The Efficient Frontier

With three assets, for example, an investor can choose

w_1 = share of initial wealth allocated to asset 1

w_2 = share of initial wealth allocated to asset 2

$1 - w_1 - w_2$ = share of wealth allocated to asset 3

The Efficient Frontier

Given the choices of w_1 and w_2 :

$$\tilde{r}_P = w_1 \tilde{r}_1 + w_2 \tilde{r}_2 + (1 - w_1 - w_2) \tilde{r}_3$$

$$\mu_P = w_1 \mu_1 + w_2 \mu_2 + (1 - w_1 - w_2) \mu_3$$

$$\begin{aligned} \sigma_P^2 &= w_1^2 \sigma_1^2 + w_2^2 \sigma_2^2 + (1 - w_1 - w_2)^2 \sigma_3^2 \\ &\quad + 2w_1 w_2 \sigma_1 \sigma_2 \rho_{12} \\ &\quad + 2w_1 (1 - w_1 - w_2) \sigma_1 \sigma_3 \rho_{13} \\ &\quad + 2w_2 (1 - w_1 - w_2) \sigma_2 \sigma_3 \rho_{23} \end{aligned}$$

The Efficient Frontier

Our problem is to solve

$$\min_{w_1, w_2} \sigma_P^2 \text{ subject to } \mu_P = \bar{\mu}$$

for a given value of $\bar{\mu}$.

But since we are more used to solving constrained **maximization** problems, consider the reformulated, but equivalent, problem:

$$\max_{w_1, w_2} -\sigma_P^2 \text{ subject to } \mu_P = \bar{\mu}$$

The Efficient Frontier

Set up the Lagrangian, using the expressions for σ_P and μ_P derived previously:

$$\begin{aligned}L(w_1, w_2, \lambda) = & -w_1^2\sigma_1^2 - w_2^2\sigma_2^2 - (1 - w_1 - w_2)^2\sigma_3^2 \\ & - 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ & - 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ & - 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23} \\ & + \lambda[w_1\mu_1 + w_2\mu_2 + (1 - w_1 - w_2)\mu_3 - \bar{\mu}]\end{aligned}$$

The Efficient Frontier

$$\begin{aligned}L(w_1, w_2, \lambda) = & -w_1^2\sigma_1^2 - w_2^2\sigma_2^2 - (1 - w_1 - w_2)^2\sigma_3^2 \\ & - 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ & - 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ & - 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23} \\ & + \lambda[w_1\mu_1 + w_2\mu_2 + (1 - w_1 - w_2)\mu_3 - \bar{\mu}]\end{aligned}$$

Most of these objects are **data**:

$$\mu_1, \mu_2, \mu_3, \sigma_1, \sigma_2, \sigma_3, \rho_{12}, \rho_{13}, \rho_{23}$$

And the target $\bar{\mu}$ is given as well.

The Efficient Frontier

PS12: As in PS11, suppose $\mu_1 = 8$, $\mu_2 = 4$, $\sigma_1 = 8$, and $\sigma_2 = 4$.

Now introduce a third asset, with $\mu_3 = 6$ and $\sigma_3 = 6$.

Assume for simplicity that $\rho_{12} = \rho_{13} = \rho_{23} = 0$.

We can achieve a target expected return $\bar{\mu} = 6$ by investing only in asset 3. The portfolio will then have $\sigma_P = \sigma_3 = 6$.

How much better can we do by choosing portfolio weights optimally? A lot better, even with only 3 assets and zero correlations.

The Efficient Frontier

$$\begin{aligned}L(w_1, w_2, \lambda) = & -w_1^2\sigma_1^2 - w_2^2\sigma_2^2 - (1 - w_1 - w_2)^2\sigma_3^2 \\ & - 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ & - 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ & - 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23} \\ & + \lambda[w_1\mu_1 + w_2\mu_2 + (1 - w_1 - w_2)\mu_3 - \bar{\mu}]\end{aligned}$$

With $\bar{\mu} = 6$, $\mu_1 = 8$, $\mu_2 = 4$, $\mu_3 = 6$, $\sigma_1 = 8$, $\sigma_2 = 4$, $\sigma_3 = 6$,
and $\rho_{12} = \rho_{13} = \rho_{23} = 0$:

$$\begin{aligned}L(w_1, w_2, \lambda) = & -64w_1^2 - 16w_2^2 - 36(1 - w_1 - w_2)^2 \\ & + \lambda[8w_1 + 4w_2 + 6(1 - w_1 - w_2) - 6]\end{aligned}$$

The Efficient Frontier

$$L(w_1, w_2, \lambda) = -64w_1^2 - 16w_2^2 - 36(1 - w_1 - w_2)^2 \\ + \lambda[8w_1 + 4w_2 + 6(1 - w_1 - w_2) - 6]$$

FOC for w_1 :

$$-128w_1^* + 72(1 - w_1^* - w_2^*) + \lambda^*(8 - 6) = 0$$

FOC for w_2 :

$$-32w_2^* + 72(1 - w_1^* - w_2^*) + \lambda^*(4 - 6) = 0$$

Constraint:

$$8w_1^* + 4w_2^* + 6(1 - w_1^* - w_2^*) = 6$$

The Efficient Frontier

FOC for w_1 :

$$-128w_1^* + 72(1 - w_1^* - w_2^*) + \lambda^*(8 - 6) = 0$$

FOC for w_2 :

$$-32w_2^* + 72(1 - w_1^* - w_2^*) + \lambda^*(4 - 6) = 0$$

Constraint:

$$8w_1^* + 4w_2^* + 6(1 - w_1^* - w_2^*) = 6$$

Three **linear** equations in three unknowns: w_1^* , w_2^* , and λ^* .

The Efficient Frontier

Start with the constraint:

$$8w_1^* + 4w_2^* + 6(1 - w_1^* - w_2^*) = 6$$

$$2w_1^* - 2w_2^* = 0$$

$$w_1^* = w_2^*$$

Since $\mu_1 = 8$ and $\mu_2 = 4$, maintaining the target expected return $\mu_P = 6$ requires allocating equal shares to assets 1 and 2.

The Efficient Frontier

Substitute

$$w^* = w_1^* = w_2^*$$

Into the FOC for w_1 :

$$-128w_1^* + 72(1 - w_1^* - w_2^*) + \lambda^*(8 - 6) = 0$$

$$-128w^* + 72(1 - 2w^*) + 2\lambda^* = 0$$

and the FOC for w_2 :

$$-32w_2^* + 72(1 - w_1^* - w_2^*) + \lambda^*(4 - 6) = 0$$

$$-32w^* + 72(1 - 2w^*) - 2\lambda^* = 0$$

The Efficient Frontier

Solve for w^* by elimination:

$$-128w^* + 72(1 - 2w^*) + 2\lambda^* = 0$$

$$-32w^* + 72(1 - 2w^*) - 2\lambda^* = 0$$

$$-160w^* + 144(1 - 2w^*) = 0$$

The Efficient Frontier

$$-160w^* + 144(1 - 2w^*) = 0$$

After you find the numerical values of $w_1^* = w^*$, $w_2^* = w^*$, and $w_3^* = 1 - w_1^* - w_2^* = 1 - 2w^*$, compute

$$\sigma_P = (64w_1^{*2} + 16w_2^{*2} + 36w_3^{*2})^{1/2}$$

It will be **much** smaller than 6. Optimal portfolio allocation yields a substantial reduction in risk while still maintaining the expected return of $\bar{\mu} = 6$ percent.

The Efficient Frontier

$$\begin{aligned}L(w_1, w_2, \lambda) = & -w_1^2\sigma_1^2 - w_2^2\sigma_2^2 - (1 - w_1 - w_2)^2\sigma_3^2 \\ & - 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ & - 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ & - 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23} \\ & + \lambda[w_1\mu_1 + w_2\mu_2 + (1 - w_1 - w_2)\mu_3 - \bar{\mu}]\end{aligned}$$

First-order condition for w_1 :

$$\begin{aligned}0 = & -2w_1^*\sigma_1^2 + 2(1 - w_1^* - w_2^*)\sigma_3^2 - 2w_2^*\sigma_1\sigma_2\rho_{12} \\ & - 2(1 - w_1^* - w_2^*)\sigma_1\sigma_3\rho_{13} + 2w_1^*\sigma_1\sigma_3\rho_{13} \\ & + 2w_2^*\sigma_2\sigma_3\rho_{23} + \lambda^*\mu_1 - \lambda^*\mu_3\end{aligned}$$

The Efficient Frontier

$$\begin{aligned}L(w_1, w_2, \lambda) = & -w_1^2\sigma_1^2 - w_2^2\sigma_2^2 - (1 - w_1 - w_2)^2\sigma_3^2 \\ & - 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ & - 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ & - 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23} \\ & + \lambda[w_1\mu_1 + w_2\mu_2 + (1 - w_1 - w_2)\mu_3 - \bar{\mu}]\end{aligned}$$

First-order condition for w_2 :

$$\begin{aligned}0 = & -2w_2^*\sigma_2^2 + 2(1 - w_1^* - w_2^*)\sigma_3^2 - 2w_1^*\sigma_1\sigma_2\rho_{12} \\ & + 2w_1^*\sigma_1\sigma_3\rho_{13} - 2(1 - w_1^* - w_2^*)\sigma_2\sigma_3\rho_{23} \\ & + 2w_2^*\sigma_2\sigma_3\rho_{23} + \lambda^*\mu_2 - \lambda^*\mu_3\end{aligned}$$

The Efficient Frontier

The two first-order conditions and the constraint

$$\begin{aligned}0 &= -2w_1^* \sigma_1^2 + 2(1 - w_1^* - w_2^*) \sigma_3^2 - 2w_2^* \sigma_1 \sigma_2 \rho_{12} \\ &\quad - 2(1 - w_1^* - w_2^*) \sigma_1 \sigma_3 \rho_{13} + 2w_1^* \sigma_1 \sigma_3 \rho_{13} \\ &\quad + 2w_2^* \sigma_2 \sigma_3 \rho_{23} + \lambda^* \mu_1 - \lambda^* \mu_3\end{aligned}$$

$$\begin{aligned}0 &= -2w_2^* \sigma_2^2 + 2(1 - w_1^* - w_2^*) \sigma_3^2 - 2w_1^* \sigma_1 \sigma_2 \rho_{12} \\ &\quad + 2w_1^* \sigma_1 \sigma_3 \rho_{13} - 2(1 - w_1^* - w_2^*) \sigma_2 \sigma_3 \rho_{23} \\ &\quad + 2w_2^* \sigma_2 \sigma_3 \rho_{23} + \lambda^* \mu_2 - \lambda^* \mu_3\end{aligned}$$

$$w_1^* \mu_1 + w_2^* \mu_2 + (1 - w_1^* - w_2^*) \mu_3 = \bar{\mu}$$

form a system of three equations in the three unknowns: w_1^* , w_2^* , and λ^* .

The Efficient Frontier

Moreover, the equations are **linear** in the unknowns w_1^* , w_2^* , and λ^* :

$$\begin{aligned} 0 = & -2w_1^*\sigma_1^2 + 2(1 - w_1^* - w_2^*)\sigma_3^2 - 2w_2^*\sigma_1\sigma_2\rho_{12} \\ & - 2(1 - w_1^* - w_2^*)\sigma_1\sigma_3\rho_{13} + 2w_1^*\sigma_1\sigma_3\rho_{13} \\ & + 2w_2^*\sigma_2\sigma_3\rho_{23} + \lambda^*\mu_1 - \lambda^*\mu_3 \end{aligned}$$

$$\begin{aligned} 0 = & -2w_2^*\sigma_2^2 + 2(1 - w_1^* - w_2^*)\sigma_3^2 - 2w_1^*\sigma_1\sigma_2\rho_{12} \\ & + 2w_1^*\sigma_1\sigma_3\rho_{13} - 2(1 - w_1^* - w_2^*)\sigma_2\sigma_3\rho_{23} \\ & + 2w_2^*\sigma_2\sigma_3\rho_{23} + \lambda^*\mu_2 - \lambda^*\mu_3 \end{aligned}$$

$$w_1^*\mu_1 + w_2^*\mu_2 + (1 - w_1^* - w_2^*)\mu_3 = \bar{\mu}$$

Given specific values for μ_1 , μ_2 , μ_3 , σ_1 , σ_2 , σ_3 , ρ_{12} , ρ_{13} , ρ_{23} , and $\bar{\mu}$ they can be solved quite easily.

The Efficient Frontier

In linear algebra, a **vector** is just a column of numbers. With $N \geq 3$ assets, you can organize the portfolio shares and expected returns into a vectors:

$$w = \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_N \end{bmatrix} \quad \text{and} \quad \mu = \begin{bmatrix} \mu_1 \\ \mu_2 \\ \vdots \\ \mu_N \end{bmatrix}$$

where

$$w_1 + w_2 + \dots + w_N = 1$$

Also in linear algebra, the **transpose** of a vector just reorganizes the column as a row; for example:

$$w' = [w_1 \quad w_2 \quad \dots \quad w_N]$$

The Efficient Frontier

Meanwhile, the variances and covariances can be organized into a **matrix** – a collection of rows and columns:

$$\Sigma = \begin{bmatrix} \sigma_1^2 & \sigma_1\sigma_2\rho_{12} & \dots & \sigma_1\sigma_N\rho_{1N} \\ \sigma_1\sigma_2\rho_{12} & \sigma_2^2 & \dots & \sigma_2\sigma_N\rho_{2N} \\ \vdots & \vdots & \dots & \vdots \\ \sigma_1\sigma_N\rho_{1N} & \sigma_2\sigma_N\rho_{2N} & \dots & \sigma_N^2 \end{bmatrix}$$

The Efficient Frontier

Using the rules from linear algebra for multiplying vectors and matrices, the expected return on any portfolio with shares in the vector w is

$$\mu'w$$

and the variance of the random return on the portfolio is

$$w'\Sigma w.$$

Hence, the problem of minimizing the variance for a given mean can be written compactly as

$$\max_w -w'\Sigma w \text{ subject to } \mu'w = \bar{\mu} \text{ and } \ell'w = 1$$

where ℓ is a vector of N ones.

The Efficient Frontier

$$\max_w -w'\Sigma w \text{ subject to } \mu'w = \bar{\mu} \text{ and } \ell'w = 1$$

Problems of this form are called **quadratic programming problems** and can be solved very quickly on a computer even when the number of assets N is large.

We can also add more constraints, such as $w_i \geq 0$, ruling out short sales.

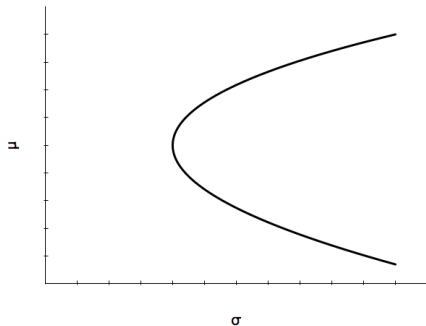
The Efficient Frontier

Going back to the case with three assets, once the optimal shares w_1^* and w_2^* have been found, the minimized standard deviation can be computed using the general formula

$$\begin{aligned}\sigma_P^2 &= w_1^2\sigma_1^2 + w_2^2\sigma_2^2 + (1 - w_1 - w_2)^2\sigma_3^2 \\ &\quad + 2w_1w_2\sigma_1\sigma_2\rho_{12} \\ &\quad + 2w_1(1 - w_1 - w_2)\sigma_1\sigma_3\rho_{13} \\ &\quad + 2w_2(1 - w_1 - w_2)\sigma_2\sigma_3\rho_{23}\end{aligned}$$

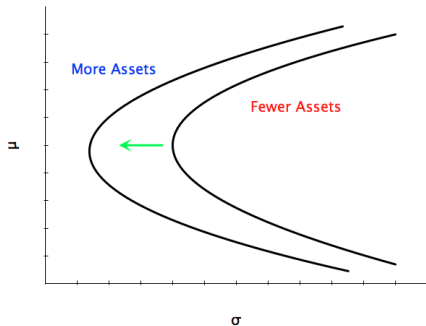
Doing this for various values of $\bar{\mu}$ allows us to trace out the **minimum variance frontier**.

The Efficient Frontier



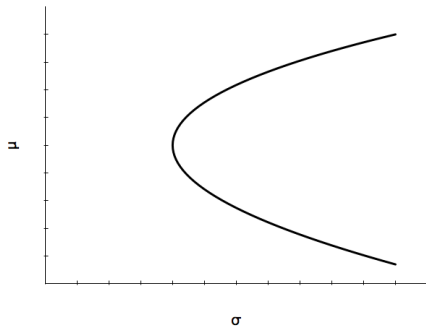
Tracing out the minimized σ_p for each value of $\mu_p = \bar{\mu}$ produces the **minimum variance frontier**.

The Efficient Frontier



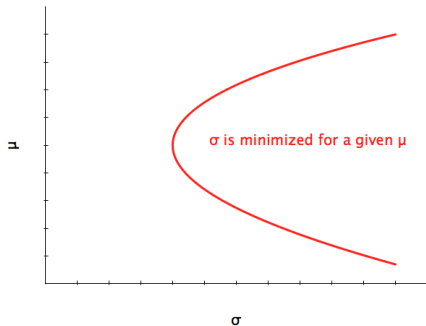
Adding assets shifts the minimum variance frontier to the left, as opportunities for diversification are enhanced.

The Efficient Frontier



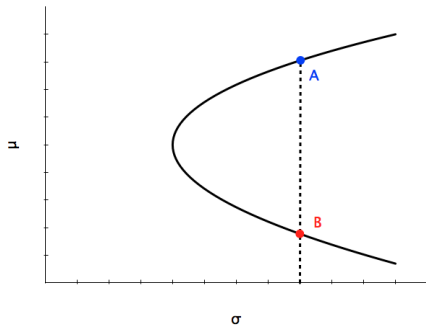
However, the minimum variance frontier retains its sideways parabolic shape.

The Efficient Frontier



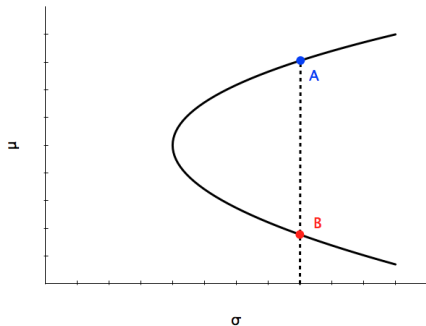
The **minimum variance frontier** traces out the minimized variance or standard deviation for each required mean return.

The Efficient Frontier



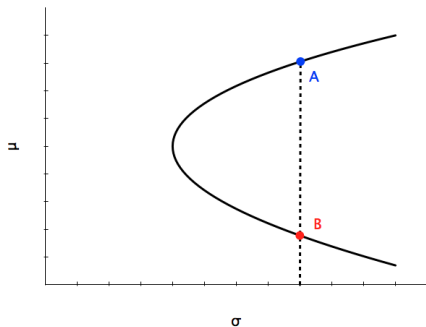
Both **portfolio A** and **portfolio B** minimize σ_p for a given value of μ_p .

The Efficient Frontier



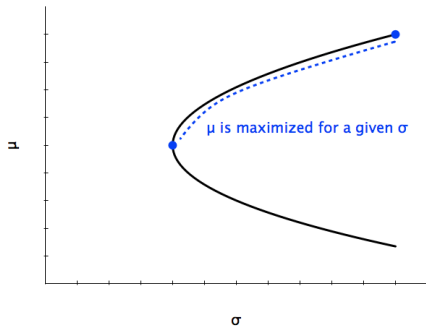
But **portfolio A** has a higher expected return that with the same standard deviation as **portfolio B**.

The Efficient Frontier



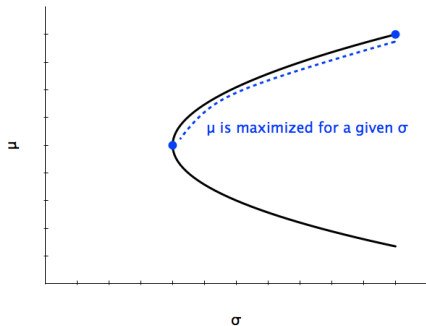
That is, **portfolio A** exhibits mean-variance dominance over **portfolio B**, since it offers a higher expected return with the same standard deviation.

The Efficient Frontier



Hence, the **efficient frontier** extends only along the top arm of the minimum variance frontier.

The Efficient Frontier



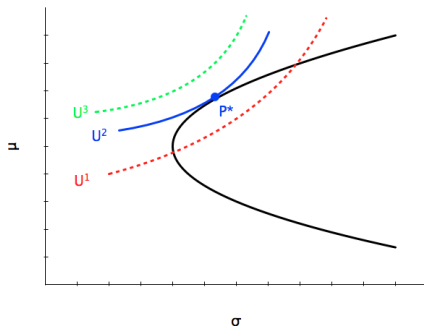
All other feasible portfolios are dominated in mean-variance by those on the **efficient frontier**.

The Efficient Frontier

Recall that any of the following assumptions imply that indifference curves in this $\sigma - \mu$ diagram slope upward and are convex:

1. Risks are small enough to justify a second-order Taylor approximation to any increasing and concave Bernoulli utility function within the vN-M expected utility framework
2. Investors have vN-M expected utility with quadratic Bernoulli utility functions
3. Asset returns are normally distributed and investors have vN-M expected utility with increasing and concave Bernoulli utility functions

The Efficient Frontier



Portfolios along U^1 are suboptimal. Portfolios along U^3 are infeasible. Portfolio P^* , located where U^2 is tangent to the efficient frontier, is optimal.