

# Risk and Return

## Topic 2

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# Notation

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# Returns

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$$\mathbf{R} = \begin{pmatrix} R_1 \\ R_2 \\ \vdots \\ R_j \\ \vdots \\ R_{n-1} \\ R_n \end{pmatrix}, \quad \bar{\mathbf{R}} = \begin{pmatrix} E(R_1) \\ E(R_2) \\ \vdots \\ E(R_j) \\ \vdots \\ E(R_{n-1}) \\ E(R_n) \end{pmatrix}, \quad R_f$$

# Variance

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$$\sigma^2 = E\left(R - E(R)\right)^2 = E\left(R^2\right) - \left(E(R)\right)^2$$



# Covariance

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$$\sigma_{ij} = E\left[\left(R_i - E(R_i)\right)\left(R_j - E(R_j)\right)\right] = E(R_i R_j) - E(R_i)E(R_j)$$

# Correlation

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$$\rho_{ij} = \frac{\sigma_{ij}}{\sigma_i \sigma_j}$$



# Covariance matrix

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$$\mathbf{V} = \begin{pmatrix} \sigma_{11} & \cdots & \sigma_{1n} \\ \vdots & \ddots & \vdots \\ \sigma_{n1} & \cdots & \sigma_{nn} \end{pmatrix}, \text{ where } \sigma_{ij} = \sigma_{ji}$$

Example:  $\mathbf{V} = \begin{pmatrix} \sigma_{11} & \sigma_{12} & \sigma_{13} \\ \sigma_{21} & \sigma_{22} & \sigma_{23} \\ \sigma_{31} & \sigma_{32} & \sigma_{33} \end{pmatrix}$

# Wealth and investment weights

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$$W_0 = \sum_j S_j = \mathbf{1}' \cdot \mathbf{S} = S_1 + \cdots + S_j + \cdots + S_n \Rightarrow \sum_j x_j = \mathbf{1}' \cdot \mathbf{x} = 1$$

$$\mathbf{S} = \begin{pmatrix} S_1 \\ \vdots \\ S_j \\ \vdots \\ S_n \end{pmatrix}, \quad \mathbf{x} = \frac{\mathbf{S}}{W_0} \begin{pmatrix} x_1 \\ \vdots \\ x_j \\ \vdots \\ x_n \end{pmatrix}, \quad \mathbf{1} = \begin{pmatrix} 1 \\ \vdots \\ 1 \\ \vdots \\ 1 \end{pmatrix}$$



# Income

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$$y = \sum_j R_j \cdot S_j = \mathbf{R}' \cdot \mathbf{S} = \begin{pmatrix} R_1 & \cdots & R_n \end{pmatrix} \cdot \begin{pmatrix} S_1 \\ \vdots \\ S_n \end{pmatrix}$$

# Data

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# The companies

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File: S&P Big Cap 2008-12-31 to 2011-06-30.csv

Stocks: AAPL, DTV, KMB, NWSA, WAG

Dates: 30 Jan 2009 to 30 Jun 2011

$N$ : 30 months

# A Two-Asset Portfolio Frontier

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# The stocks

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AAPL and DTV

$$R = \begin{pmatrix} 0.0489129 \\ 0.0296652 \end{pmatrix}$$

$$\bar{\sigma} = \begin{pmatrix} 0.0684982 \\ 0.0750907 \end{pmatrix}$$

$$V = \begin{pmatrix} 0.00469201 & 0.0024785 \\ 0.0024785 & 0.00563861 \end{pmatrix}$$

# Portfolio expected return

$$E(R_p) = \mathbf{x}' \cdot \mathbf{R} = x_1 E(R_1) + x_2 E(R_2) \quad \text{or} \quad x_1 = \frac{E(R_p) - E(R_2)}{E(R_1) - E(R_2)}$$

$$= x_1 E(R_1) + (1 - x_1) E(R_2)$$

Portfolio	$\sigma_p^2$	$\sigma_p$	$E(R_p)$	$x_1$	$x_2$
1.	0.0101673	0.100833	0.02	-0.502147	1.50215
2. DTV	0.00563861	0.0750907	0.0296652	0	1.
3.	0.00379427	0.0615977	0.04	0.536936	0.463064
4. MVP	0.00378022	0.0614835	0.0409844	0.588079	0.411921
5. AAPL	0.00469201	0.0684982	0.0489129	1.	0
6.	0.00902502	0.0950001	0.06	1.57602	-0.576019



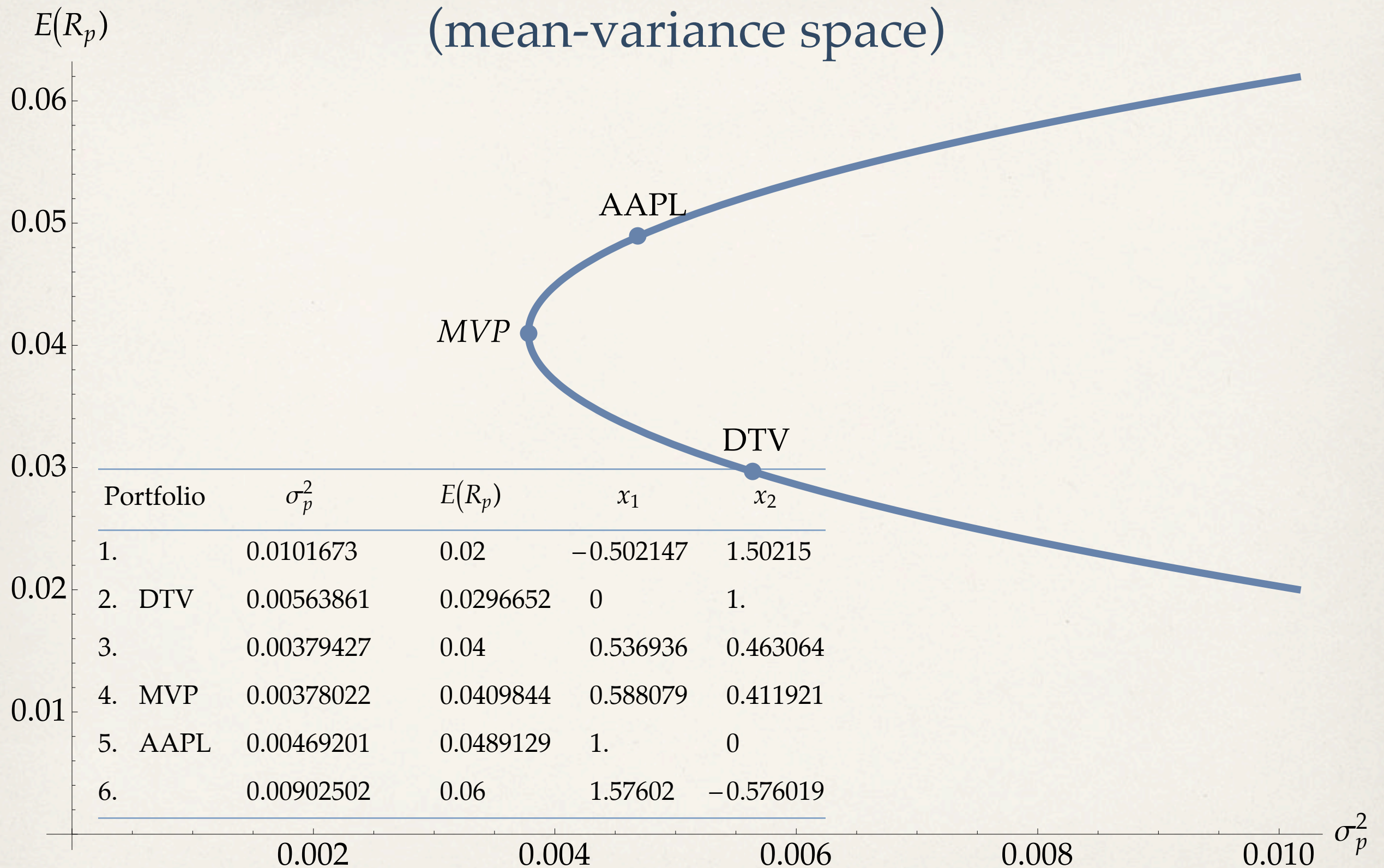
# Portfolio variance

$$\begin{aligned}\sigma_p^2 &= \mathbf{x}' \cdot \mathbf{V} \cdot \mathbf{x} = x_1^2 \sigma_1^2 + x_2^2 \sigma_2^2 + 2x_1 x_2 \sigma_{12} \\ &= x_1^2 \sigma_1^2 + (1 - x_1)^2 \sigma_2^2 + 2x_1 (1 - x_1) \sigma_1 \sigma_2 \rho_{12}\end{aligned}$$

Portfolio	$\sigma_p^2$	$\sigma_p$	$E(R_p)$	$x_1$	$x_2$
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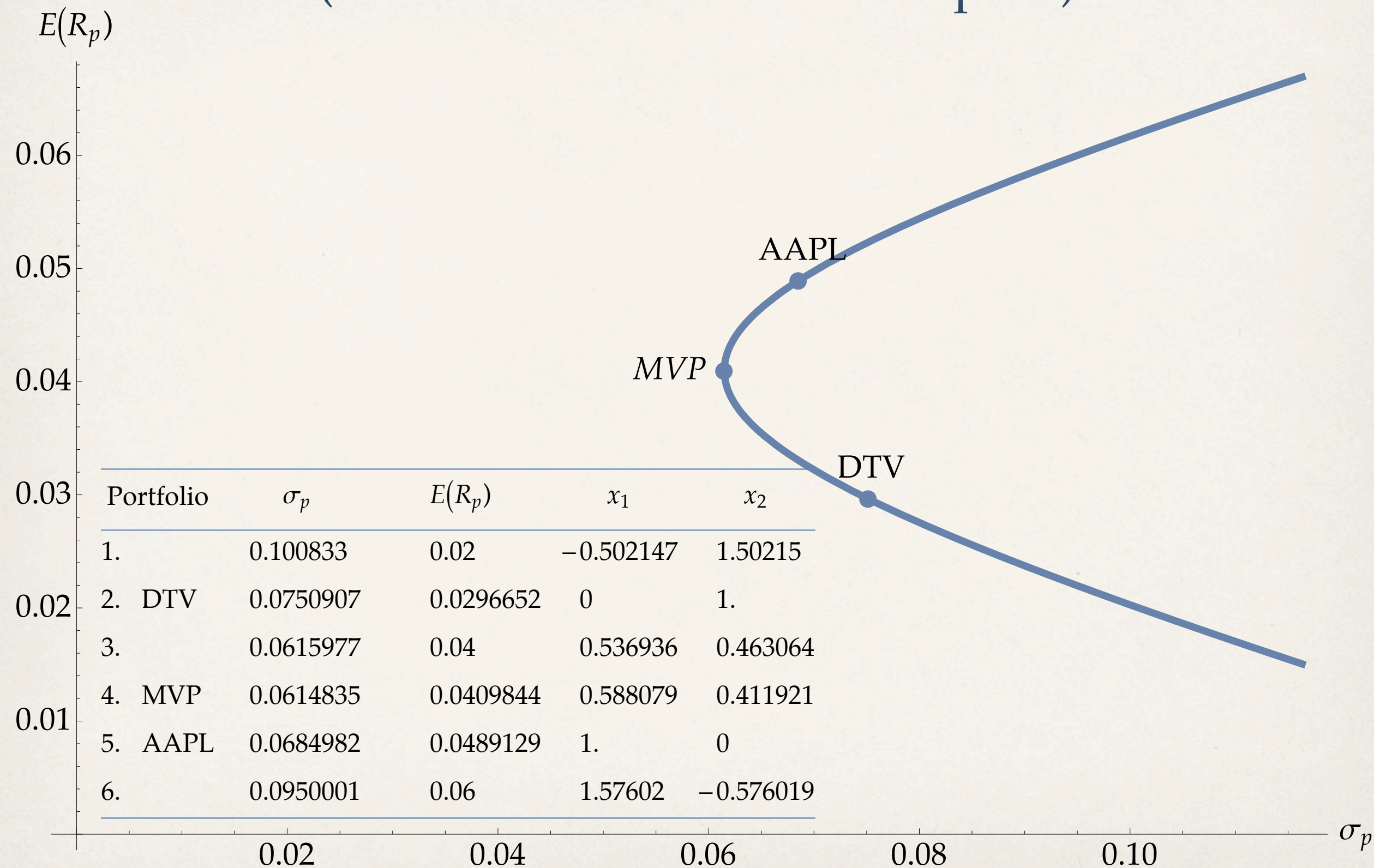
# Minimum-Variance Frontier (mean-variance space)





# Minimum-Variance Frontier

(mean-standard deviation space)



# Special cases

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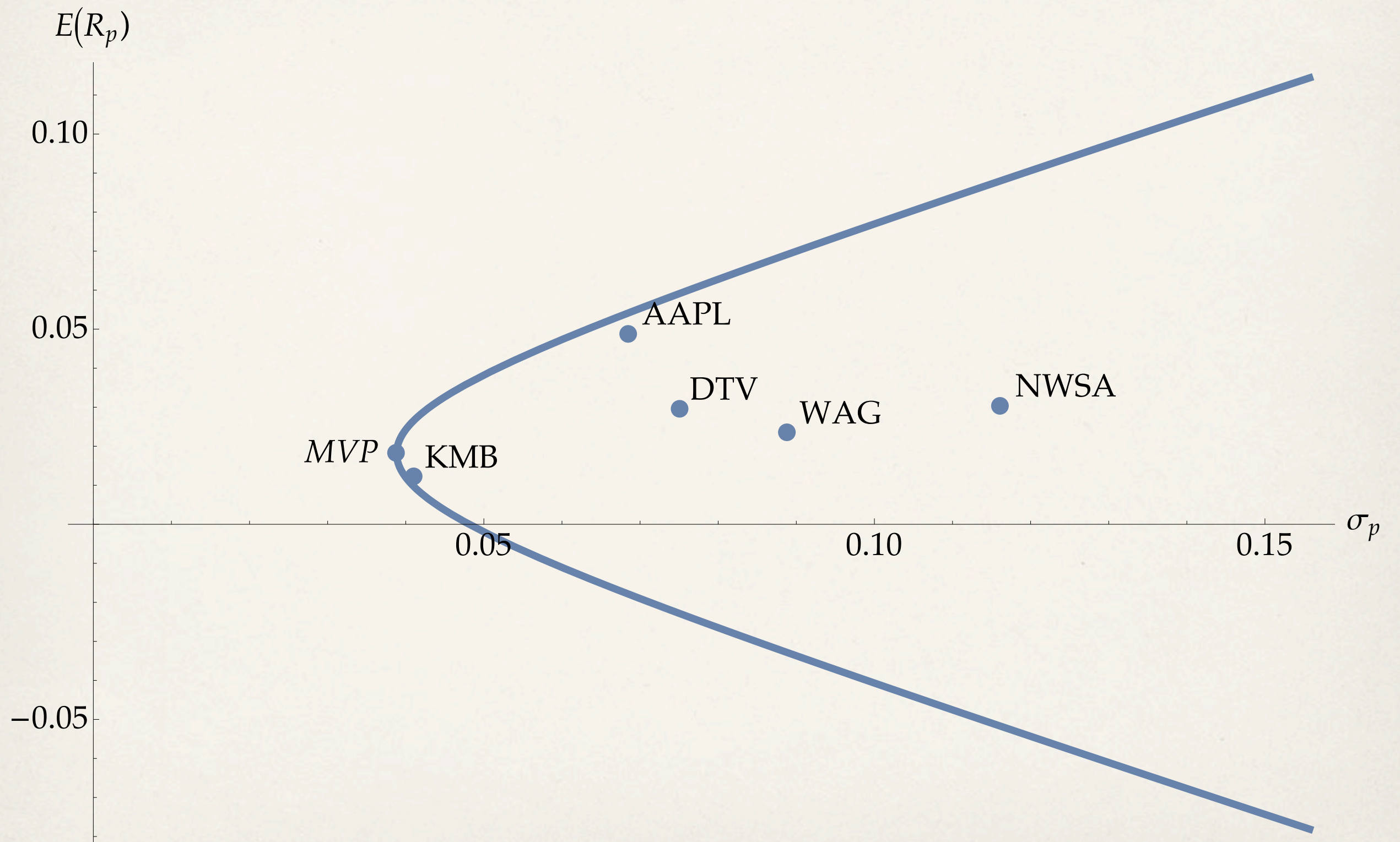
- ✧ Derive the minimum-variance portfolio for a two-asset frontier
- ✧ Derive portfolio standard deviation for the case where the correlation of return between the two assets is
  - ✧  $+1$
  - ✧  $-1$



# Portfolio Mathematics

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# The Goal





# Optimal investment

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$$\underset{\langle S \rangle}{\text{Min}} \sigma_p^2 = \mathbf{S}' \cdot \mathbf{V} \cdot \mathbf{S} + \lambda_1 [\bar{\mathbf{R}}' \cdot \mathbf{S} - \bar{y}] + \lambda_2 [\mathbf{1}' \cdot \mathbf{S} - W_0]$$

$$\mathbf{S} : 2\mathbf{V}\mathbf{S} + \lambda_1 \bar{\mathbf{R}} + \lambda_2 \mathbf{1} = 0$$

$$\lambda_1 : \bar{\mathbf{R}}' \cdot \mathbf{S} - \bar{y}_i = 0$$

$$\lambda_2 : \mathbf{1}' \cdot \mathbf{S} - W_i = 0$$

$$\mathbf{S} = -\mathbf{V}^{-1} \frac{(\lambda_1 \bar{\mathbf{R}} + \lambda_2 \mathbf{1})}{2} = \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \begin{pmatrix} -\lambda_1/2 \\ -\lambda_2/2 \end{pmatrix}$$

# Optimal investment {continued}

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$$\mathbf{S} = -\mathbf{V}^{-1} \frac{(\lambda_1 \bar{\mathbf{R}} + \lambda_2 \mathbf{1})}{2} = \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \begin{pmatrix} -\lambda_1/2 \\ -\lambda_2/2 \end{pmatrix}$$

$$\mathbf{X} = \frac{\mathbf{S}}{W_0} = \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{R}_p \\ 1 \end{pmatrix}$$



$$\begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix}' \mathbf{S} = \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix}' \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \begin{pmatrix} -\lambda_1/2 \\ -\lambda_2/2 \end{pmatrix} = \mathbf{Q} \begin{pmatrix} -\lambda_1/2 \\ -\lambda_2/2 \end{pmatrix}$$

$$\text{where } \mathbf{Q} = \begin{pmatrix} \bar{\mathbf{R}}' \mathbf{V}^{-1} \bar{\mathbf{R}} & \bar{\mathbf{R}}' \mathbf{V}^{-1} \mathbf{1} \\ \mathbf{1}' \mathbf{V}^{-1} \bar{\mathbf{R}} & \mathbf{1}' \mathbf{V}^{-1} \mathbf{1} \end{pmatrix} = \begin{pmatrix} \bar{\mathbf{R}}' \mathbf{V}^{-1} \bar{\mathbf{R}} & \bar{\mathbf{R}}' \mathbf{V}^{-1} \mathbf{1} \\ \mathbf{1}' \mathbf{V}^{-1} \bar{\mathbf{R}} & \mathbf{1}' \mathbf{V}^{-1} \mathbf{1} \end{pmatrix} = \begin{pmatrix} B & A \\ A & C \end{pmatrix}$$

$$\Rightarrow \begin{pmatrix} -\lambda_1/2 \\ -\lambda_2/2 \end{pmatrix} = \mathbf{Q}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix}' \mathbf{S} = \mathbf{Q}^{-1} \begin{pmatrix} \bar{y} \\ W \end{pmatrix}$$



# Optimal investment *{finis}*

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$$\mathbf{x} = \frac{\mathbf{S}}{W_0} = \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{R}_p \\ 1 \end{pmatrix}$$

$$\mathbf{x} = \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \frac{\begin{pmatrix} C & -A \\ -A & B \end{pmatrix}}{D} \begin{pmatrix} \bar{R}_p \\ 1 \end{pmatrix}$$

$$\mathbf{x} = \frac{1}{D} \left[ B(\mathbf{V}^{-1} \mathbf{1}) - A(\mathbf{V}^{-1} \bar{\mathbf{R}}) \right] + \frac{1}{D} \left[ C(\mathbf{V}^{-1} \bar{\mathbf{R}}) - A(\mathbf{V}^{-1} \mathbf{1}) \right] \bar{R}_p$$

$$\mathbf{x} = \mathbf{g} + \mathbf{h} \bar{R}_p \quad \text{Optimal weights are linear in expected portfolio return}$$

# Minimum-variance frontier

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$$\sigma^2 = \mathbf{S}'\mathbf{V}\mathbf{S} = \mathbf{S}'\mathbf{V} \left( \mathbf{V}^{-1} \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{y} \\ W_0 \end{pmatrix} \right)$$

$$\therefore \sigma_p^2 = \left( \mathbf{S}' \begin{pmatrix} \bar{\mathbf{R}} & \mathbf{1} \end{pmatrix} \right) \mathbf{Q}^{-1} \begin{pmatrix} \bar{y} \\ W_0 \end{pmatrix} = \begin{pmatrix} \bar{y} & W_0 \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{y} \\ W_0 \end{pmatrix}$$

$$\therefore \sigma_p^2 = \frac{\sigma^2}{W_0^2} = \begin{pmatrix} \bar{R}_p & 1 \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{R}_p \\ 1 \end{pmatrix}$$

$$\therefore \sigma_p^2 = \frac{1}{D} \left( C\bar{R}_p^2 - 2A\bar{R}_p + B \right) \begin{array}{l} \text{Parabola in mean-variance space} \\ \text{Hyperbola in mean-standard deviation space} \end{array}$$



# Minimum-variance portfolio

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$$\sigma_p^2 = \frac{\sigma^2}{W^2} = \begin{pmatrix} \bar{R}_p & 1 \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{R}_p \\ 1 \end{pmatrix}$$

$$\sigma_p^2 = \frac{1}{D} (C\bar{R}_p^2 - 2A\bar{R}_p + B)$$

$$\frac{d\sigma_p^2}{d\bar{R}_p} = 0 \Rightarrow MVP : (\sigma_p^2, \bar{R}_p) = \left( \frac{1}{C}, \frac{A}{C} \right)$$

# Orthogonal portfolios

## {finding uncorrelated portfolios}

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$$\text{cov}(y_k, y_u) = \mathbf{S}'_k \mathbf{V} \mathbf{S}_u = \begin{pmatrix} \bar{y}_k & W_0 \end{pmatrix} \mathbf{Q}^{-1} \begin{pmatrix} \bar{y}_u \\ W_0 \end{pmatrix}$$

$$\therefore \sigma_{ku} = \text{cov}(\bar{R}_k, \bar{R}_u) = \frac{1}{D} \begin{pmatrix} \bar{R}_k & 1 \end{pmatrix} \begin{pmatrix} C & -A \\ -A & B \end{pmatrix} \begin{pmatrix} \bar{R}_u \\ 1 \end{pmatrix}$$

$$\therefore \sigma_{ku} = \frac{1}{D} C \bar{R}_k \bar{R}_u - A \bar{R}_k - A \bar{R}_u + B$$

$$\therefore \sigma_{ku} = 0 \Rightarrow \bar{R}_u = \frac{A \bar{R}_k - B}{C \bar{R}_k - A} \quad \text{We'll use this later}$$



# A Five-Asset Portfolio Frontier

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# Means and standard deviations

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AAPL	$\bar{R} =$	$\begin{pmatrix} 0.0489129 \\ 0.0296652 \\ 0.0122415 \\ 0.0304791 \\ 0.0235527 \end{pmatrix}$	$\bar{\sigma} =$	$\begin{pmatrix} 0.0684982 \\ 0.0750907 \\ 0.0410235 \\ 0.116067 \\ 0.0888068 \end{pmatrix}$
DTV				
KMB				
NWSA				
WAG				



# Covariances

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$$V = \begin{pmatrix} 0.00469201 & 0.0024785 & 0.00120074 & 0.00474886 & 0.00374848 \\ 0.0024785 & 0.00563861 & 0.00102798 & 0.00369911 & 0.00253769 \\ 0.00120074 & 0.00102798 & 0.00168292 & 0.00199206 & 0.00136948 \\ 0.00474886 & 0.00369911 & 0.00199206 & 0.0134715 & 0.00357577 \\ 0.00374848 & 0.00253769 & 0.00136948 & 0.00357577 & 0.00788664 \end{pmatrix}$$

$$V^{-1} = \begin{pmatrix} 0.00469201 & 0.0024785 & 0.00120074 & 0.00474886 & 0.00374848 \\ 0.0024785 & 0.00563861 & 0.00102798 & 0.00369911 & 0.00253769 \\ 0.00120074 & 0.00102798 & 0.00168292 & 0.00199206 & 0.00136948 \\ 0.00474886 & 0.00369911 & 0.00199206 & 0.0134715 & 0.00357577 \\ 0.00374848 & 0.00253769 & 0.00136948 & 0.00357577 & 0.00788664 \end{pmatrix}$$



# Portfolio math parameters

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$$\mathbf{Q} = \begin{pmatrix} \bar{\mathbf{R}}' \cdot \mathbf{V}^{-1} \cdot \bar{\mathbf{R}} & \bar{\mathbf{R}}' \cdot \mathbf{V}^{-1} \cdot \mathbf{1} \\ \mathbf{1}' \cdot \mathbf{V}^{-1} \cdot \bar{\mathbf{R}} & \mathbf{1}' \cdot \mathbf{V}^{-1} \cdot \mathbf{1} \end{pmatrix}$$

$$= \begin{pmatrix} B & A \\ A & C \end{pmatrix} = \begin{pmatrix} 0.625609 & 12.0295 \\ 12.0295 & 663.12 \end{pmatrix}$$

$$\mathbf{Q}^{-1} = \frac{\begin{pmatrix} C & -A \\ -A & B \end{pmatrix}}{D} = \begin{pmatrix} 2.45467 & -0.0445295 \\ -0.0445295 & 0.00231582 \end{pmatrix}$$

$$D = BC - A^2 = 270.146$$



# Optimal weights

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$$\mathbf{x} = \mathbf{g} + \mathbf{h}\bar{R}_p = \begin{pmatrix} -0.404393 \\ 0.0934881 \\ 1.21151 \\ -0.0238107 \\ 0.123201 \end{pmatrix} + \begin{pmatrix} 31.1317 \\ 1.29537 \\ -20.2935 \\ -3.89382 \\ -8.23981 \end{pmatrix} \bar{R}_p$$



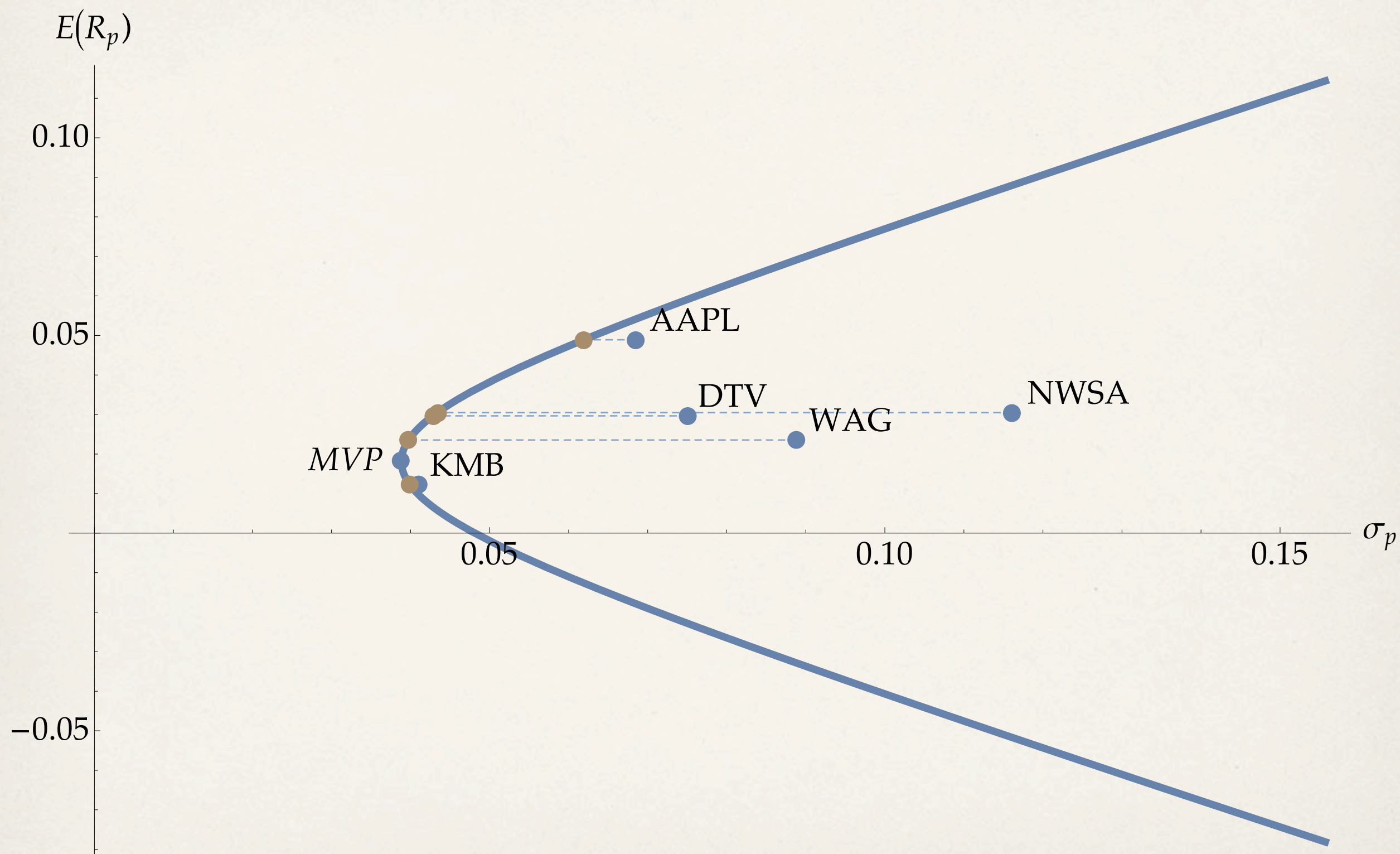
$$\begin{aligned}\sigma_p &= \sqrt{\frac{1}{D}(C\bar{R}_p^2 - 2A\bar{R}_p + B)} \\ &= \sqrt{\frac{1}{270.146}(663.12 \times \bar{R}_p^2 - 2 \times 12.0295 \times \bar{R}_p + 0.625609)} \\ &= \sqrt{\mathbf{x} \cdot \mathbf{V} \cdot \mathbf{x}}\end{aligned}$$

$$\mathbf{x} = \mathbf{g} + \mathbf{h}\bar{R}_p = \begin{pmatrix} -0.404393 \\ 0.0934881 \\ 1.21151 \\ -0.0238107 \\ 0.123201 \end{pmatrix} + \begin{pmatrix} 31.1317 \\ 1.29537 \\ -20.2935 \\ -3.89382 \\ -8.23981 \end{pmatrix} \bar{R}_p$$

Portfolio	$\sigma_p$	$E(R_p)$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
1.	0.0408741	0.01	-0.0930757	0.106442	1.00858	-0.0627488	0.0408028
2. KMB	0.039918	0.0122415	-0.0232947	0.109345	0.963093	-0.0714767	0.0223334
3. MVP	0.0388333	0.0181407	0.160358	0.116987	0.843377	-0.0944472	-0.0262749
4. WAG	0.0397482	0.0235527	0.328843	0.123997	0.733549	-0.115521	-0.0708689
5. DTV	0.0428257	0.0296652	0.519136	0.131915	0.609505	-0.139321	-0.121235
6.	0.0430495	0.03	0.529559	0.132349	0.60271	-0.140625	-0.123993
7. NWSA	0.0433787	0.0304791	0.544476	0.13297	0.592987	-0.142491	-0.127941
8. AAPL	0.0619066	0.0489129	1.11835	0.156848	0.218901	-0.214269	-0.279832
9.	0.0762175	0.06	1.46351	0.17121	-0.00609397	-0.25744	-0.371188

$$MVP: \begin{pmatrix} \sigma & E(R) \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{C}} & \frac{A}{C} \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{663.12}} & \frac{12.0295}{663.12} \end{pmatrix}$$





# Capital Market Line

## {next week}

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