



# Drivers in the Case of Two Assets





### Future Returns Are Random In Nature

#### Optimizing Portfolio requires expectations:

- about average portfolio return (mean)
- about how far off it may be (variance)

Why?



Portfolio Return Is A Random Variable



### Past Performance to Predictions

	Mean Portfolio Return
Computed on a sample of T Historical Returns	$\hat{\mu} = \frac{R_1 + R_2 + \ldots + R_T}{T}$
When the return is a random variable	$\mu = E[R]$

	Portfolio Return Variance
Computed on a sample of T Historical Returns	$\hat{\sigma}^2 = \frac{(R_1 - \hat{\mu})^2 + (R_2 - \hat{\mu})^2 + \dots + (R_T - \hat{\mu})^2}{T - 1}$
When the return is a random variable	$\sigma^2 = E[(R - \mu)^2]$





### Drivers of Mean & Variance

Assume two assets:

Asset 1	Asset 2
Weight: w₁	Weight: w₂
Return: R <sub>1</sub>	Return: R <sub>2</sub>

- Portfolio Return  $P = w_1 * R_1 + w_2 * R_2$
- Thus:  $E[P] = w_1^* E[R_1] + w_2^* E[R_2]$



### Portfolio Return Variance

Again, for a portfolio with 2 assets

#### Variance of Portfolio Return

$$var(P) = E[(P - E[P])^{2}] = w_{1}^{2} * var(R_{1})$$

$$+w_{2}^{2} * var(R_{1})$$

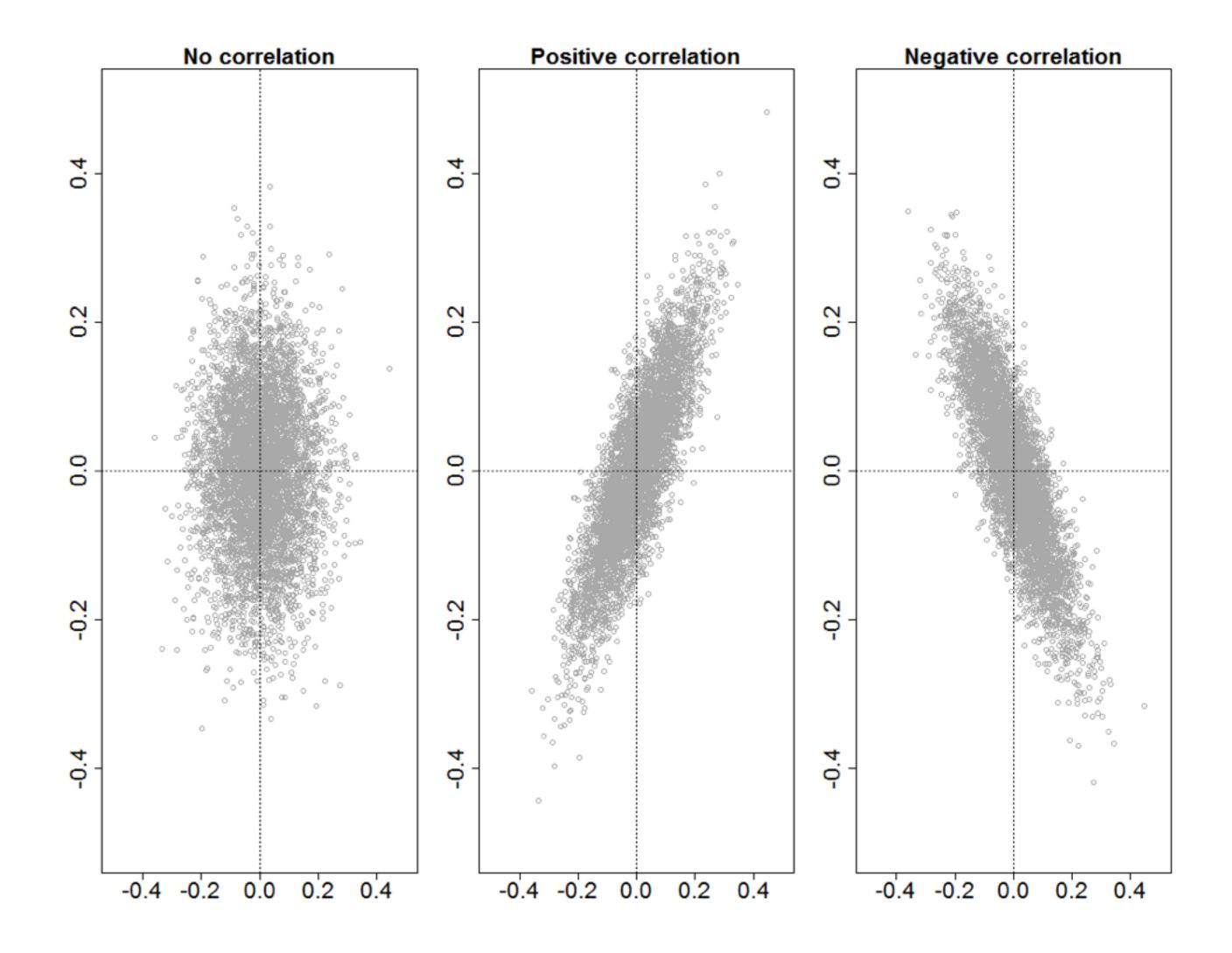
$$+2 * w_{1} * w_{2} * cov(R_{1}, R_{2})$$

#### Covariance between return 1 and 2

$$Cov(R_1, R_2) = E[(R_1 - E[R_1])(R_2 - E(R_2))]$$
  
=  $StdDev(R_1) * StdDev(R_1) * corr(R_1, R_2)$ 



## Correlations





# Take Away Formulas

- E[Portfolio Return] =  $E(P) = w_1 * E[R_1] + w_2 * E[R_2]$
- var(Portfolio Return) =  $var(P) = w_1^2 * var(R_1)$   $+w_2^2 * var(R_1)$  $+2*w_1*w_2*cov(R_1,R_2)$





# Let's practice!





# Using Matrix Notation



### Variables at Stake for N Assets

• w: the N x 1 column-matrix of portfolio weights

• R: the N x 1 column-matrix of asset returns

 μ: the N x 1 column-matrix of expected returns

$$w = egin{bmatrix} w_1 \ w_2 \ dots \ w_N \end{bmatrix}$$

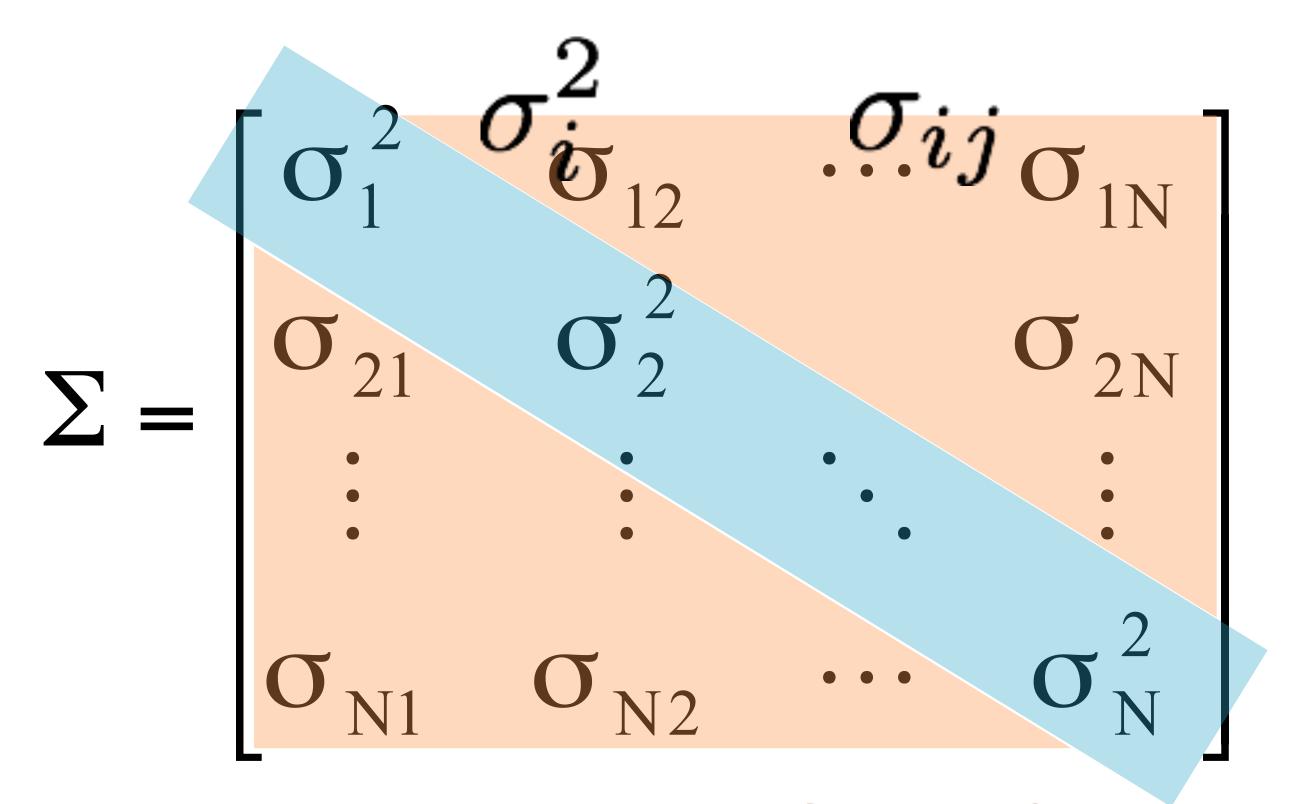
$$R = \begin{bmatrix} R_1 \\ R_2 \\ \vdots \\ R_N \end{bmatrix}$$

$$\mu = egin{bmatrix} \mu_1 \ \mu_2 \ dots \ \mu_N \end{bmatrix}$$



### Variables at Stake for N Assets

 $\bullet$   $\Sigma$ : The N x N covariance matrix of the N asset returns:



Covariance: Outside Diagonal

Variance: On Diagonal

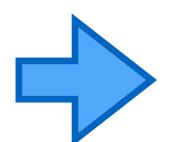




# Generalizing from 2 to N Assets

#### Portfolio Return

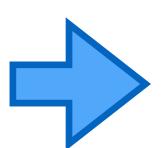
$$w_1 * R_1 + w_2 * R_2$$



$$w_1 * R_1 + \ldots + w_N * R_N$$

#### Portfolio Expected Return

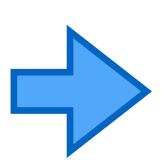
$$w_1 * \mu_1 + w_2 * \mu_2$$



$$w_1*\mu_1+\ldots+w_N*\mu_N$$

#### Portfolio Variance

$$w_1^2 * var(R_1) + w_2^2 * var(R_1)$$
  
  $+2 * w_1 * w_2 * cov(R_1, R_2)$ 



$$w_1^2 + var(R_1) + \ldots + w_N^2 * var(R_N)$$
  
  $+2 * w_1 * w_2 * cov(R_1, R_2) + \ldots$   
  $+2 * w_{N-1} * w_N * cov(R_{N-1}, R_N)$ 



# Matrices Simplify the Notation

- Avoid large number of terms by using matrix notation
- We have 4 matrices:
  - weights (w), returns (R), expected returns ( $\mu$ ), and covariance matrix ( $\Sigma$ )

$$w = egin{bmatrix} w_1 \ w_2 \ dots \ w_N \end{bmatrix} \qquad w' = egin{bmatrix} w_1 \ w_2 \ \cdots \ w_N \end{bmatrix}$$

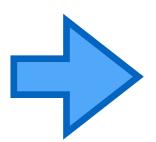


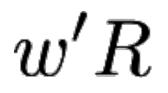


# Simplifying the Notation

#### Portfolio Return

$$w_1 * R_1 + \ldots + w_N * R_N$$





#### Portfolio Expected Return

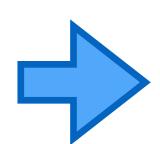
$$w_1*\mu_1+\ldots+w_N*\mu_N$$



$$\overline{w'\mu}$$

#### Portfolio Variance

$$w_1^2 + var(R_1) + \dots + w_N^2 * var(R_N)$$
  
  $+2 * w_1 * w_2 * cov(R_1, R_2) + \dots$   
  $+2 * w_{N-1} * w_N * cov(R_{N-1}, R_N)$ 



 $w'\Sigma w$ 





# Let's practice!





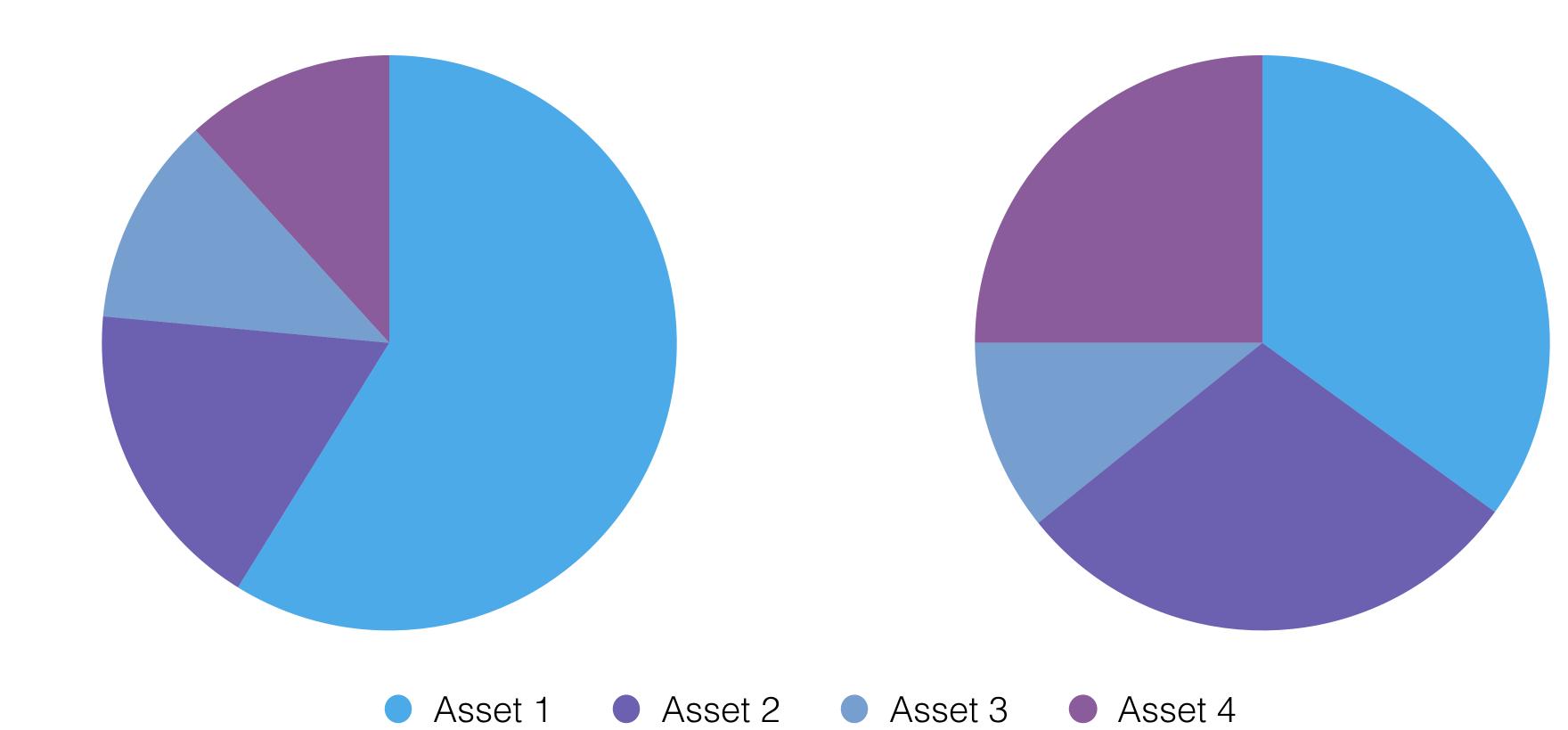
# Portfolio Risk Budget



### Who Did It?

Capital Allocation Budget

Portfolio Volatility Risk





### Portfolio Volatility In Risk Contribution

• Portfolio Volatility = 
$$\sum_{i=1}^{N} RC_i$$

• Where: 
$$RC_i = \frac{w_i(\sum w)_i}{\sqrt{w'\sum w}}$$

- risk contribution of asset i depends on
  - 1. the complete matrix of weights w
  - 2. the full covariance matrix  $\sum$



### Percent Risk Contribution

$$\%RC_i = \frac{RC_i}{\text{portfolio volatility}}$$

where 
$$\sum_{i=1}^{N} \% RC_i = 1$$

Relatively more risky assets:  $\%RC_i>w_i$ 

Relatively less risky assets:  $\%RC_i < w_i$ 





# Let's practice!