## **Drivers in the case** of two assets

#### INTRODUCTION TO PORTFOLIO ANALYSIS IN R

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Optimizing Portfolio requires expectations: about average portfolio return (mean)

about how far off it may be (variance)





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### Why?





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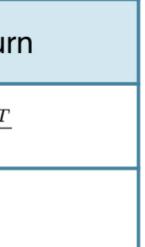
Portfolio Return Is A Random Variable





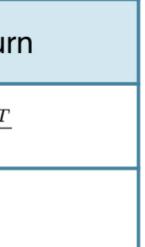
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Computed on a sample of T Historical Returns	$\hat{\mu} = \frac{R_1 + R_2 + \ldots + R_T}{T}$
When the return is a random variable	$\mu = E[R]$





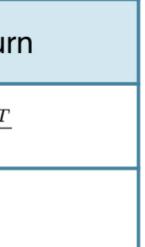
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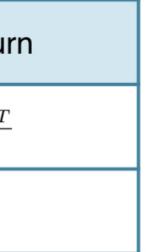


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Computed on a sample of T Historical Returns	$\hat{\sigma}^2 = \frac{(R_1 - \hat{\mu})^2 + (R_2 - \hat{\mu})^2 + \ldots + (R_2 - \hat{\mu})^2}{T - 1}$
When the return is a random variable	$\sigma^2 = E[(R-\mu)^2]$



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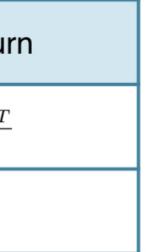
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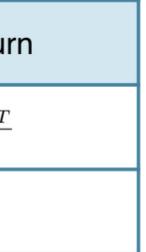
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$$(R_T - \hat{\mu})^2$$

### **Drivers of mean & variance**

• Assume two assets:

Asset 1	Asset 2
Weight: $w_1$	Weight: $w_2$
Return: $R_1$	Return: $R_2$

- Portfolio Return,  $P = w1 \cdot R1 + w2 \cdot R2$
- Thus:  $E[P] = w_1 \cdot E[R_1] + w_2 \cdot E[R_2]$



### **Portfolio return variance**

Again, for a portfolio with 2 assets

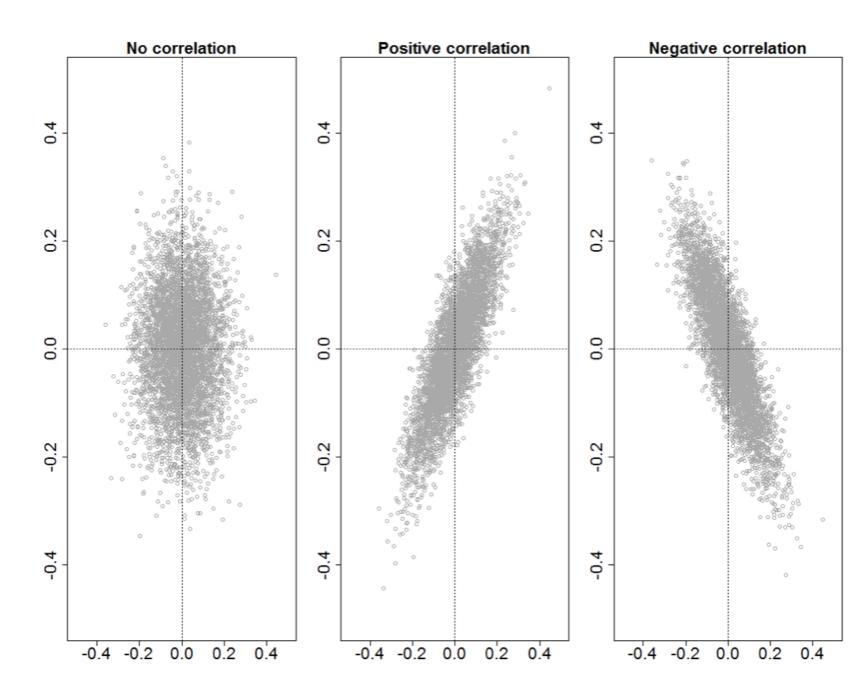
•  $var(P) = w_1^2 \cdot var(R_1) + w_2^2 \cdot var(R_2) + 2 \cdot w_1 \cdot w_2 \cdot cov(R_1, R_2)$ 

**Covariance between return 1 and 2** 

- $Cov(R_1, R_2)$  $\circ = E[(R_1 - E[R_1])(R_2 - E(R_2))]$ 
  - $\circ = StdDev(R_1) \cdot StdDev(R_2) \cdot corr(R_1, R_2)$



### Correlations



### R datacamp

### Take away formulas

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



# Let's practice!



# Using matrix notation

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### Variables at stake for n assets

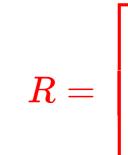
w: the N x 1 column-matrix of portfolio weights:

$$w = \left[egin{array}{cc} w_1 \ w_2 \ \ldots \ w_N \end{array}
ight]$$

$$\mu$$
: the  $N$  x 1 column-matrix of expected returns:

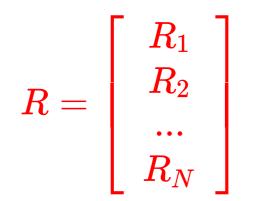
$$\mu = \left[ egin{array}{c} \mu_1 \ \mu_2 \ \ldots \ \mu_N \end{array} 
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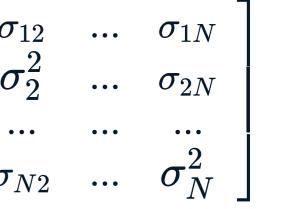
R: the  $N \ge 1$  column-matrix of asset returns:



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

$$w = \left[egin{array}{ccc} \sigma_1^2 & \sigma \ \sigma_{21} & \sigma \ \dots & \ddots \ \sigma_{N1} & \sigma \end{array}
ight.$$





### Generalizing from 2 to n assets

### Portfolio Return

$$w_1 * R_1 + w_2 * R_2$$
  $w_1 * R_1 +$ 



### $\ldots + w_N * R_N$

### Generalizing from 2 to n assets



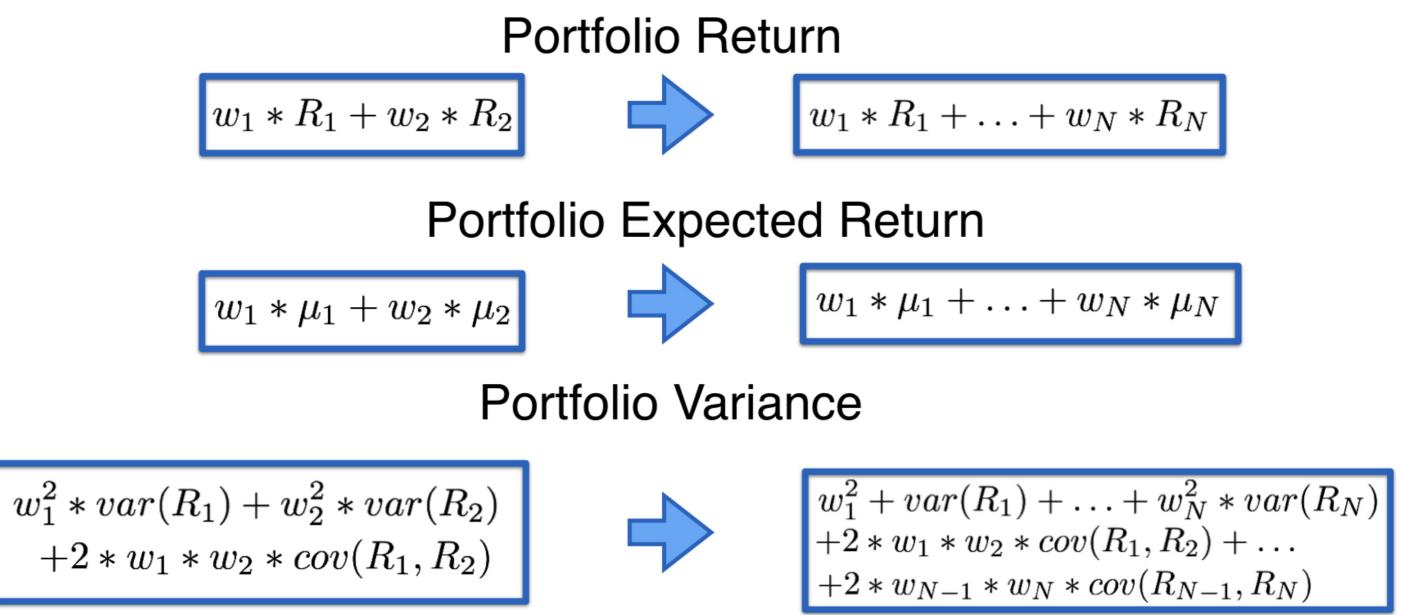
### Portfolio Expected Return

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



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

### Generalizing from 2 to n assets



### 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 = \left[egin{array}{c} w_1 \ w_2 \ ... \ w_N \end{array}
ight] w' = \left[egin{array}{c} w_1 \ w_2 \ ... \ w_N \end{array}
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### Simplifying the notation

### Portfolio Return

$$w_1 * R_1 + \ldots + w_N * R_N$$
  $w'R$ 



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### Portfolio Expected Return

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





### **Portfolio Variance**

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

tacamp

$$w'\Sigma w$$

# Let's practice!



### Portfolio risk budget INTRODUCTION TO PORTFOLIO ANALYSIS IN R



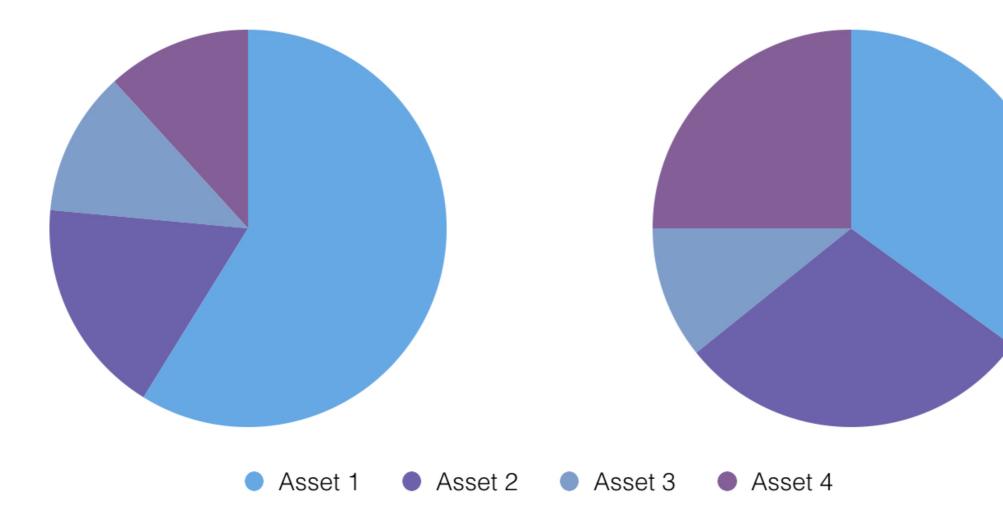
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### Who did it?

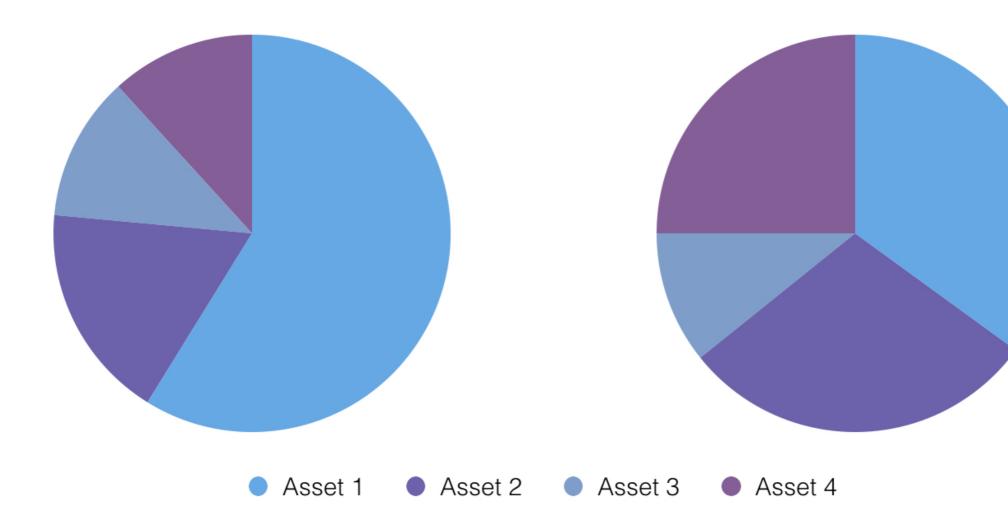
R datacamp



### Who did it?

#### **Capital Allocation Budget**

#### Portfolio Volatility Risk



#### V datacamp

### Portfolio volatility in risk contribution

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

where: 
$$RC_i = rac{w_i(\Sigma w)_i}{\sqrt{w'\Sigma w}}$$

- Risk contribution of asset i depends on ullet
  - 1. the complete matrix of weights w
  - 2. the full covariance matrix  $\Sigma$



### **Percent risk contribution**

$$\% RC_i = rac{RC_i}{ ext{Portfolio volatility}}$$

• where 
$$\sum_{i=1}^N \% RC_i = 1$$

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

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



# Let's practice!

