



INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

The Capital Asset Pricing Model

Dakota Wixom

Quantitative Analyst | QuantCourse.com



The Founding Father of Asset Pricing Models

CAPM

The Capital Asset Pricing Model is the fundamental building block for many other asset pricing models and factor models in finance.



Excess Returns

To calculate excess returns, simply subtract the risk free rate of return from your total return:

$$\text{Excess Return} = \text{Return} - \text{Risk Free Return}$$

Example:

Investing in Brazil:

10% Portfolio Return - 15% Risk Free Rate = -5% Excess Return

Investing in the US:

10% Portfolio Return - 3% Risk Free Rate = 7% Excess Return



The Capital Asset Pricing Model

$$E(R_P) - RF = \beta_P(E(R_M) - RF)$$

- $E(R_P) - RF$: The excess expected return of a stock or portfolio P
- $E(R_M) - RF$: The excess expected return of the broad market portfolio B
- RF : The regional risk free-rate
- β_P : Portfolio beta, or exposure, to the broad market portfolio B



Calculating Beta Using Co-Variance

To calculate historical beta using co-variance:

$$\beta_P = \frac{Cov(R_P, R_B)}{Var(R_B)}$$

- β_P : Portfolio beta
- $Cov(R_P, R_B)$: The co-variance between the portfolio (P) and the benchmark market index (B)
- $Var(R_B)$: The variance of the benchmark market index



Calculating Beta Using Co-Variance in Python

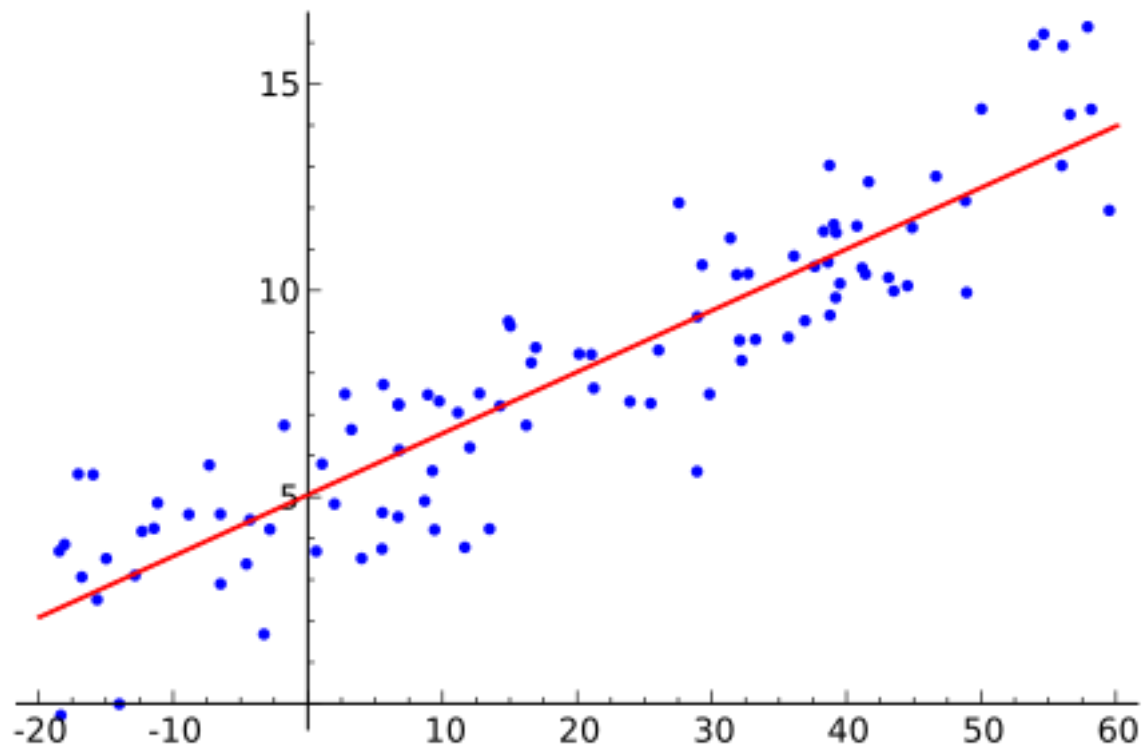
Assuming you already have excess portfolio and market returns in the object `Data`:

```
In [1]: covariance_matrix = Data[["Port_Excess", "Mkt_Excess"]].cov()
In [2]: covariance_coefficient = covariance_matrix.iloc[0,1]
In [3]: benchmark_variance = Data["Mkt_Excess"].var()
In [4]: portfolio_beta = covariance_coefficient / benchmark_variance
In [5]: portfolio_beta
Out [5]: 0.93
```



Linear Regressions

Example of a linear regression:



Regression formula in matrix notation:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\epsilon},$$

$$\mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix},$$

$$\mathbf{X} = \begin{pmatrix} \mathbf{x}_1^\top \\ \mathbf{x}_2^\top \\ \vdots \\ \mathbf{x}_n^\top \end{pmatrix} = \begin{pmatrix} 1 & x_{11} & \cdots & x_{1p} \\ 1 & x_{21} & \cdots & x_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{n1} & \cdots & x_{np} \end{pmatrix},$$

$$\boldsymbol{\beta} = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \vdots \\ \beta_p \end{pmatrix}, \quad \boldsymbol{\epsilon} = \begin{pmatrix} \epsilon_1 \\ \epsilon_2 \\ \vdots \\ \epsilon_n \end{pmatrix}.$$



Calculating Beta Using Linear Regression

Assuming you already have excess portfolio and market returns in the object `Data`:

```
In [1]: import statsmodels.formula.api as smf
In [2]: model = smf.ols(formula='Port_Excess ~ Mkt_Excess', data=Data)
In [3]: fit = model.fit()
In [4]: beta = fit.params["Mkt_Excess"]
In [5]: beta
Out [5]: 0.93
```




R-Squared vs Adjusted R-Squared

To extract the adjusted r-squared and r-squared values:

```
In [1]: import statsmodels.formula.api as smf
In [2]: model = smf.ols(formula='Port_Excess ~ Mkt_Excess', data=Data)
In [3]: fit = model.fit()
In [4]: r_squared = fit.rsquared
In [5]: r_squared
Out [5]: 0.70
In [6]: adjusted_r_squared = fit.rsquared_adj
Out [6]: 0.65
```



INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

Let's practice!



INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

Alpha and Multi-Factor Models

Dakota Wixom

Quantitative Analyst | QuantCourse.com



The Fama-French 3 Factor Model

The Fama-French 3- factor model:

$$R_P = RF + \beta_M(R_M - RF) + b_{SMB} \cdot SMB + b_{HML} \cdot HML + \alpha$$

- SMB: The small minus big factor
- b_{SMB} : Exposure to the SMB factor
- HML: The high minus low factor
- b_{HML} : Exposure to the HML factor
- α : Performance which is unexplained by any other factors
- β_M : Beta to the broad market portfolio B

The Fama-French 3 Factor Model

	Portfolio	Market_Excess	SMB	HML	RF
Date					
2013-01-03	-0.005066	-0.0014	0.0014	0.0004	0.0
2013-01-04	0.004024	0.0055	0.0019	0.0043	0.0
2013-01-07	0.004421	-0.0031	-0.0009	-0.0037	0.0
2013-01-08	-0.004659	-0.0027	0.0004	-0.0007	0.0
2013-01-09	0.004636	0.0034	0.0024	-0.0041	0.0



The Fama-French 3 Factor Model in Python

Assuming you already have excess portfolio and market returns in the object `Data`:

```
In [1]: import statsmodels.formula.api as smf
In [2]: model = smf.ols(formula='Port_Excess ~ Mkt_Excess + SMB + HML', data=Data)
In [3]: fit = model.fit()
In [4]: adjusted_r_squared = fit.rsquared_adj
In [5]: adjusted_r_squared
Out [5]: 0.90
```



P-Values and Statistical Significance

To extract the `HML` p-value, assuming you have a fitted regression model object in your workspace as `fit`:

```
In [1]: fit.pvalues["HML"]  
Out [1]: 0.0063
```

To test if it is statistically significant, simply examine whether or not it is less than a given threshold, normally 0.05:

```
In [1]: fit.pvalues["HML"] < 0.05  
Out [2]: True
```



Extracting Coefficients

To extract the `HML` coefficient, assuming you have a fitted regression model object in your workspace as `fit`:

```
In [1]: fit.params["HML"]  
Out [1]: 0.502  
In [2]: fit.params["SMB"]  
Out [2]: -0.243
```




Alpha and the Efficient Market Hypothesis

Assuming you already have a fitted regression analysis in the object `fit`:

```
In [1]: portfolio_alpha = fit.params["Intercept"]
In [2]: portfolio_alpha_annualized = ((1+portfolio_alpha)**252)-1
In [3]: portfolio_alpha_annualized
Out [3]: 0.045
```



INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

Let's practice!



INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

Expanding the 3-Factor Model

Dakota Wixom

Quantitative Analyst | QuantCourse.com



Fama French 1993

The original paper that started it all:



Journal of Financial Economics
Volume 33, Issue 1, February 1993, Pages 3-56



Common risk factors in the returns on stocks and bonds ☆

Eugene F. Fama , Kenneth R. French



Cliff Asness on Momentum

A paper published later by Cliff Asness from AQR:

PERSPECTIVE

Fama on Momentum

February 5, 2016 - [Cliff Asness](#)

Topics - [Factor/Style Investing](#), [Momentum](#)



The Fama-French 5 Factor Model

In 2015, Fama and French extended their previous 3-factor model, adding two additional factors:

- **RMW:** Profitability
- **CMA:** Investment

The RMW factor represents the returns of companies with high operating profitability versus those with low operating profitability.

The CMA factor represents the returns of companies with aggressive investments versus those who are more conservative.



The Fama-French 5 Factor Model

	Portfolio	Market_Excess	SMB	HML	RMW	CMA	RF
Date							
2013-01-03	-0.005066	-0.0014	0.0014	0.0004	0.0020	0.0023	0.0
2013-01-04	0.004024	0.0055	0.0019	0.0043	-0.0037	0.0027	0.0
2013-01-07	0.004421	-0.0031	-0.0009	-0.0037	-0.0013	-0.0012	0.0
2013-01-08	-0.004659	-0.0027	0.0004	-0.0007	-0.0012	0.0009	0.0
2013-01-09	0.004636	0.0034	0.0024	-0.0041	-0.0007	-0.0015	0.0



The Fama-French 5 Factor Model in Python

Assuming you already have excess portfolio and market returns in the object `Data`:

```
In [1]: import statsmodels.formula.api as smf
In [2]: model = smf.ols(formula='Port_Excess ~ Mkt_Excess + SMB + HML + RMW + CM
In [3]: fit = model.fit()
In [4]: adjusted_r_squared = fit.rsquared_adj
In [5]: adjusted_r_squared
Out [5]: 0.92
```




INTRODUCTION TO PORTFOLIO RISK MANAGEMENT IN PYTHON

Let's practice!