

Financial Econometrics

Factor Models of Returns

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Factor model of returns

- Simple equation for return on asset i as a function of m factors

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- f_{jt} is factor j in period t and there are m factors
 - The factors are common to all returns, e.g. risk factors
 - α_i is the constant term and the coefficients β_{i1} are called “factor loadings”
- One convenient matrix representation for each period is

$$\mathbf{r}_t = \boldsymbol{\alpha} + \boldsymbol{\beta}\mathbf{f}_t + \boldsymbol{\varepsilon}_t$$

where

- \mathbf{r}_t is a k by one vector of returns at t
- $\boldsymbol{\alpha}$ is a k by one vector of constant terms
- $\boldsymbol{\beta}$ is a k by m vector of factor loadings
- \mathbf{f}_t is a k by one vector of factors
- $\boldsymbol{\varepsilon}_t$ is a k by one vector of error terms

Statistical properties

- In

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

we generally assume the factors are stationary with

$$\begin{aligned} E \mathbf{f}_t &= \boldsymbol{\mu}_f \\ \text{Cov}[\mathbf{f}_t] &= \boldsymbol{\Sigma}_f \end{aligned}$$

where $\boldsymbol{\mu}_f$ is an n by 1 vector and $\boldsymbol{\Sigma}_f$ is an m by m matrix

- $\boldsymbol{\Sigma}_f$ is not bold because it does not render correctly

- The innovations are stationary with

$$\begin{aligned} E \varepsilon_{it} &= 0 \\ \text{Cov}[\varepsilon_{it}, \varepsilon_{js}] &= \begin{cases} \sigma_i^2 & \text{if } i = j \text{ and } t = s \\ 0 & \text{if } i \neq j \text{ or } t \neq s \end{cases} \\ \text{Cov}[\varepsilon_{it}, f_{js}] &= 0 \text{ for all } i \text{ and } j \text{ and } t \text{ and } s \end{aligned}$$

- The ε_{it} are idiosyncratic factors

- Generally $T \gg$ number of assets k and number of factors m

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- Suppose that the factors are observed macroeconomic variables such as GDP, the return on the stock market, the return on bonds, etc.
- Then this is a general “macroeconometric factor model”

Single-factor stock market models

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- The market model

- Suppose that the assets are k stocks
- Let there be one factor f_{1t} , where f_{1t} is the return on the value-weighted market portfolio
- Then the factor model is the “market model”

- CAPM

- Let r_{it} be the return on a stock less the risk-free rate
- Let f_{1t} be the return on the value-weighted portfolio of stocks less the risk-free rate
- Then the factor model is the Capital Asset Pricing Model

Multi-Factor macroeconomic factor models

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- The multi-factor model due to Chen, Ross and Roll
 - Let the returns be excess returns on stocks
 - Let the factors be estimated macroeconomic variables such as GDP minus expected GDP, inflation minus expected inflation and similar “unexpected” factors
 - First estimate regressions of variables that help to predict the variables
 - A VAR would be one way to do this
 - Or possibly an ECM
 - Then get estimated factors from actual values minus predicted values
 - Then estimate regressions for each of the stocks to estimate their sensitivity to the factors

Fundamental Factor models

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- Fundamental factor models use firm-specific variables as the factors
 - Let the returns be excess returns on stocks
 - Let the factors be estimated firm-specific variables such as industry, book to market value, size and similar variables
 - Two approaches
 - Barra
 - Fama-French

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- The Barra approach uses mean-adjusted returns, so the vector representation is

$$\tilde{\mathbf{r}}_t = \boldsymbol{\beta}\mathbf{f}_t + \boldsymbol{\varepsilon}_t$$

where $\tilde{\mathbf{r}}_t$ are mean-adjusted returns

- Let the variables be estimated firm-specific variables such as industry, book to market value, size and similar variables
- Treat the variables as the factor loadings and interpret them as firm-specific variables and factor loadings β_{ij}
- Object is to estimate \mathbf{f}_t
- Can't estimate by OLS because the errors are heteroskedastic
 - **Estimate a cross-section regression for each period**
 - Then calculate firm-specific standard deviation of residuals across periods
 - Then use this in cross-section Generalized Least Squares Regression (GLS) for each period

Fama-French factor model

- The general representation of factor models is

$$r_{it} = \alpha_i + \beta_{i1}f_{1t} + \dots + \beta_{im}f_{mt} + \varepsilon_{it}, \quad i = 1, \dots, k, \quad t = 1, \dots, T$$

- The Fama-French approach uses three factors
 - Market excess return
 - Book to market (value)
 - Size (small firms relative to large firms)
- Factor variables for regression for latter two variables are included as returns to the factor
 - Take a firm-specific variable such as the ratio of book to market
 - Sort the assets by book to market
 - Form a portfolio that is long in the top fifth of the sorted returns and short in the bottom fifth of the sorted returns
 - The observed return on this portfolio is the factor realization f_{jt}
- **Estimate a time-series regression for each firm to obtain factor sensitivities, β_{ij}**