

Risk Model in details

- Statistical Models
- Fundamental Models (BARRA)
- Macro Models

Estimation example: CAPM model

$$E(R_i) = R_f + B_i * (E(R_m) - R_f) + E(\text{Alpha}), \quad R_f = 0$$

$$B_i = \text{Cov}(R_i, R_m) / \text{Var}(R_m)$$

Intuition:

$R(i,t)$ - stock return for stock “i”

$$R(i,t) = (\text{Price}(i,t+1) - \text{Price}(i,t)) / \text{Price}(i,t)$$

Question: How to define market?

- Market cap weighted S&P500
- Equal weighted index
- Statistical methodology

Estimation example (1): CAPM

Construct index equal weighted index

$$\text{MarketReturn}(t) = 1/N \sum(R(i,t), i=1\dots N)$$

Run regression to estimate betas:

$$R(i,t) = B_i * \text{MarketReturn} + e(i,t)$$

$$e(i,t) \sim N(0, \text{Sigma}(i))$$

Use OLS to estimate B_i

Other choices for market proxy are possible

Determining market structure from the data

PCA (Principal Component Analysis) approach

Consider time window: $t=0, \dots, T$ for a universe of N stocks

$R(i,t)$ ----> $N \times T$ matrix

Average return $A(i) = \langle R(i,t) \rangle = 1/T \text{ Sum}(R(i,t), t=1 \dots T)$

Volatility $S(i) = 1/(T-1) \text{ Sum}((R(i,t)-A(i))^2, t = 1 \dots T)$

Define standardized return $Y(i,t) = R(i,t) / S(i)$

Correlation Matrix

$$G(i,j) = 1/(T-1) \sum(Y(i,t)*Y(j,t) , t=1...T)$$

$$\text{Rank}(G) \leq \min(N, T)$$

$$\text{Eigenvalues: } L_1 > L_2 \geq L_3 \geq \dots L_N > 0$$

$$\text{Eigenvectors: } F_1, F_2, \dots$$

$$G F_1 = L_1 F_1 \text{ etc } \dots$$

$$\text{Variance explained by "M" eigenvectors: } 1/N \sum(L_i, i=1...M)$$

Spectral Representation of Correlation Matrix

$G \rightarrow N \times N$ matrix

$F(i,k) \rightarrow$ eigenvector corresponding to eigenvalue $L(k)$

$$G(i,j) = \text{Sum}(F(i,k) * F(j,k) * L(k), k=1 \dots \text{Rank}(G))$$

$A(i, k) \rightarrow$ factor loading of risk model “i” - stock index, “k” - factor index

$$A(i, k) = F(i, k) * \text{sqrt}(L(k)) / S(i)$$

Example

50 largest eigenvalues using the 1400 US
stocks with cap >1BB cap (Jan 2007)



Exercise

Derive 1-factor model using PCA

Compare it with simple minded approach described above

Problems with PCA approach

- Curse of dimensionality
- Low dimensional risk factors (eg. industries) can't be determined
- Latent factors

Basic structure of Statarb models

$R(i,t)$ = Return Attributed to Risk (aka systematic returns) + Alpha + Unexplained Residual

Return Attributed to Risk Model = $\text{Sum}(F(i,k) * m(k,t), k = 1 \dots \text{NRISKFACTORS})$

$m(k,t)$ - unpredictable, for example IID variables drawn from $N(0, S_m(k))$

$S_m(k)$ - volatility of risk factor “k”

Linear Alpha Models

Features (aka alpha factors) $X(i,t,p)$

“i” - stock label $i = 1 \dots N$

“t” - time label

“p” - feature index

$\text{Alpha}(i,t) = \text{Sum}(X(i,t,p) * w(p), \quad p = 1 \dots \text{NumberOfFeatures})$

Example of Alpha Factors/Features

Philosophy (Predictability in the markets exists for 2 reasons)::

- Information is processed at different speed by different players (fundamental factors)
- Providing (liquidity) service to other players in the market

Examples of Alpha Factors (2)

Fundamental factors:

- Earning surprise: $(\text{ActualEarning} - \text{EpsConsensus}) / \text{EpsConsensus}$
- Changes in analyst recommendations: $\text{RecValue} - \text{PrevRecValue}$

$\text{RecValue} = 2$ (Strong Buy), 1 (Buy), 0 (Neutral) -1 (Sell) -2 (Strong Sell)

- Changes in Eps estimates: $(\text{EpsConsensus} - \text{EpsConsensusPrev}) / (\text{EpsConsensus} + \text{EpsConsensusPrev})$

Examples of Alpha Factors (3)

Technical factors:

- Mean reversion factor: $-R(i,t-1)$
- Volume spikes: $V(i,t) / \text{MovingAverage}(V(i,t))$

Improving Alpha Factors

- Control outliers: ranking, winsorization
- Orthogonal projection relative to risk factors

Combining different Alpha Factors

- Use Ordinary Linear Regression to find optimal weights for each factor
- Advanced machine learning methods

Risk Factors versus Alpha Factors

The distinction is ephemeral, it depends on your ability to forecast future value of the feature/risk factor return

The most known example is “momentum”

Constructing momentum factor:

$$\text{Mmnt}(i,t) = \text{Rank}((P(i,t) - P(i,t-252))/P(i,t-252))$$

$$\text{Mmnt}(i,t) \text{ ---> } \text{Mmnt}(i,t) - \text{SectorAverage}(\text{Mmnt}(i,t))$$

Optimal Trading (Optimization)

Use inputs: Risk Model/Alpha Model/Trading Cost Model

- Maximize return for a given level of risk
- Impose common sense constraints: a) Position limits b) Trading limits c) Sector/Industry exposures d) Other constraints

Optimization Example

- $\text{TradingCost} = |n(i,t+1) - n(i,t)| * P(i,t) * \text{RelativeSpread}(i)$
- $\text{Alpha1}(i, t) = 1\text{d mean reversion}$
- $\text{Alpha2}(i,t) = \text{changes in analyst recommendations}$
- Risk Model: 1 factor CAPM

$$\text{CombinedAlpha}(i,t) = \text{Alpha1}(i, t) + \text{Alpha2}(i,t)$$

$$R(i,t) = \text{Beta}(i) * \text{market}(i,t) + \text{CombinedAlpha}(i, t) + \text{Residual}(i, t)$$

Optimization (2)

Given previous weights $w_0(i)$, find new weights $w(i)$ such that

$$\text{Max}(w(i) * \text{CombinedAlpha}(i,t) - \text{TradingCost}(|w(i) - w_0(i)|))$$

$$\text{Sum}(w(i) * \text{Beta}(i)) = 0$$

Extra constraints: $\text{Sum}(w(i)) = 0$ market neutrality

$$w(i) < \text{Position Limit}$$

$$|w(i) - w_0(i)| < \text{Trading Constraint}$$