Multi risk factor model Minimum variance optimization under constraints Market neutrality 130/30 long short portfolio

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Plan

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1. Factor model and stochastic processes

We have $N \in \mathbb{N}$ stocks. Their returns $R_i, \forall i \in \{1, \dots, N\}$ are modeled by N random processes χ_i corresponding to idiosyncratic risk together with K random processes f_A :

$$R_i = \chi_i + \sum_{A=1}^K \widetilde{\Omega}_{iA} f_A \tag{1}$$

$$\langle \chi_i, \chi_j \rangle = \xi_{ij}^2 \delta_{ij} \tag{2}$$

$$\langle \chi_i, f_A \rangle = 0 \tag{3}$$

$$\langle f_A, f_B \rangle = \Phi_{AB} \tag{4}$$

$$\langle R_i, R_j \rangle = \Theta_{ij} \tag{5}$$

2. Covariance matrix

The R_i covariance matrix can be written as:

$$\Theta = \Xi + \widetilde{\Omega}\Phi\widetilde{\Omega}^{\mathsf{T}} \tag{6}$$

where $\Xi_{ij} = \xi_{ij}^2 \delta_{ij}$ is the diagonal idiosyncratic part of the variance and Φ the $K \times K$ factor covariance matrix, $\widetilde{\Omega}$ is the $K \times K$ factor loading matrix. Note that K << N gives more out of samples stability. Through Cholesky:

$$\Theta = \Xi + \Omega \Omega^{\mathsf{T}} \tag{7}$$

where $\Omega = \widetilde{\Omega}\widetilde{\Phi}$ and $\widetilde{\Phi}\widetilde{\Phi}^{\intercal} = \Phi$

Note that the standard variance for a portfolio with weights ω decomposes as :

$$\omega^{\mathsf{T}}\Omega\omega = \sum_{i=1}^{N} \xi_{ij}^2 + f^{\mathsf{T}}\Phi f \tag{8}$$

where $f=\Omega\omega$ is the factor values for our portfolio with weights ω

3. Regressing returns over factor loading/exposure

At each observation time we have to regress our stock returns to our stock factors:

$$R(t) = \sum_{k=1}^{K} \beta_k(t) \times F_k(t) + \epsilon(t)$$
 (9)

$$R(t) = \sum_{k=1}^{K} f_k(t) \times \Omega_k(t) + \epsilon(t)$$
 (10)

 Ω_i , F_i is the factor exposure or loading for the stock i and β_i , f_i is the factor return or factor itself.

The variance of $\epsilon(t)$ is the idiosyncratic risk for stock i

The covariance of $(f_k(t))_{k \in \{1,\dots,K\}}$ is the factor covariance matrix.

Here a proprietary strategy is to be defined to compute the factor matrix: a rolling standard deviation or garch volatility estimation for idiosyncratic variance and a rolling Ledoit-Wolf covariance matrix estimation

4. Minimum variance optimization rebalancement at each date

$$\begin{cases} \omega \in \mathbb{R}^{N}, \omega_{i} > 0, \omega_{i} < 1 \\ \sum_{i=1}^{N} \omega_{i} = 1 \\ \sum_{i=1}^{n} \omega_{i} \beta_{i} = \sum_{i=1}^{n} \omega_{i} \frac{\operatorname{Cov}(R_{i}, R_{market})}{\operatorname{Var}(R_{i})} = 1 \end{cases}$$

$$\begin{cases} x = (f, \omega) \in \mathbb{R}^{(N+K)}, \omega_{i} > 0, \omega_{i} < 1 \\ \sum_{i=1}^{N} \omega_{i} = 1 \\ \sum_{i=1}^{n} \omega_{i} \beta_{i} = \sum_{i=1}^{n} \omega_{i} \frac{\operatorname{Cov}(R_{i}, R_{market})}{\operatorname{Var}(R_{i})} = 1 \\ \Omega \omega = f \end{cases}$$

$$\begin{cases} \omega \Sigma \omega^{\mathsf{T}} - q \times \mu^{\mathsf{T}} \omega \right\}, \text{ becomes}$$

$$\{xQx^{\mathsf{T}} - q \times \widetilde{\mu}^{\mathsf{T}} x\} \text{ (12)}$$

where:

$$Q = \begin{pmatrix} \Phi & 0 \cdots 0 \\ \hline \xi_{11} & \mathbf{0} \\ 0 & \ddots \\ \vdots & \mathbf{0} & \\ \vdots & \mathbf{0} & \\ 0 & & & \\ \xi_{nn} \end{pmatrix}, \widetilde{\mu} = \begin{pmatrix} 0 \\ \mu \end{pmatrix}$$
(13)

5. Long/Short 130/30 Market Neutral portfolio

$$\operatorname{argmin} \qquad \{xQx^{\mathsf{T}} - q \times \widetilde{\mu}^{\mathsf{T}}x\} \quad (14)$$

$$\begin{cases}
x = (f, \omega) \in \mathbb{R}^{(N+K)}, \omega_{i} > 0, \omega_{i} < 1 \\
\sum_{i=1}^{N} \omega_{i} = 1 \\
\sum_{i=1}^{n} \omega_{i}\beta_{i} = \sum_{i=1}^{n} \omega_{i} \frac{\operatorname{Cov}(R_{i}, R_{market})}{\operatorname{Var}(R_{i})} = 1 \\
\Omega\omega = f
\end{cases}$$

where:

$$Q = \begin{pmatrix} \Phi & 0 \cdots 0 \\ \xi_{11} & \mathbf{0} \\ 0 & \ddots \\ \vdots & \mathbf{0} & \vdots \\ 0 & 0 & \xi_{nn} \end{pmatrix}, \widetilde{\mu} = \begin{pmatrix} 0 \\ \mu \end{pmatrix}$$
 (15)

6. 130/30 long/short portfolio