0 Amendment

Response to
$$25\%$$
; (-1)

This is my corrective action and (my) letter (to you). (0)

1 Overview

1.1 Goal

5/16/24, 8:38 PM

Objective: Build / Compare Two Factor-Based L/S Allocation Models (1)

Beta
$$(\beta)$$
 Constraints (2)

First Strategy
$$(S_{\{1\}})$$
: Target Beta $\beta_T \in [-0.5, 0.5]$ (3)

Second Strategy
$$(S_{\{2\}})$$
: Target Beta $\beta_T \in [-2, 2]$ (4)

$$S_{\{1\}} \cong \text{Value-at-Risk Utility (Robust Optimization)}$$
 (5)

$$S_{\{1\}} \Leftarrow \text{Information Ratio}$$
 (6)

Post optimization, I compare model outcomes while evaluating estimator length se

[covariance matrix $\Sigma \wedge \text{expected returns } \mu$] across market regimes (8)

1.2 Reallocation

Portfolio Allocation
$$\{P_t\} \Leftarrow \text{`03-01-2007'} \sim \text{`03-31-2024'}$$
 (9)

$$P_t \quad orall \, t \in \{t_0, t_1, t_2, \dots, t_n\} \quad ext{where} \quad t_0 = ext{03-01-2007}, \quad t_n = ext{03-31-2024} \quad (10)$$

$$t_i = t_{i-1} + 7 \text{ days} \quad \text{for} \quad i = 1, 2, \dots, n$$
 (11)

1.3 Performance Evaluation

The performance / risk profiles are sensitive to the target Beta and the market en

Low Beta
$$\Rightarrow$$
 Decorrelation; (16)

High Beta
$$\equiv$$
 Antithesis. (17)

• Return:
$$\mu$$
 (19)

• Volatility (Vol):
$$\sigma$$
 (20)

• Skewness (Skew) :
$$\mathbb{E}\left[\left(\frac{x-\mu}{\sigma}\right)^3\right] = \frac{\mu_3}{\sigma^3} = \frac{\kappa_3}{\kappa_2^{3/2}}$$
 (21)

• Sharpe Ratio :
$$\frac{\mathbb{E}[R_a - R_b]}{\sigma_a} = \frac{\mathbb{E}[R_a - R_b]}{\sqrt{\mathbb{V}(R_a - R_b)}}$$
(23)

1.4 Simplification

Look-Back
$$\mu$$
 Estimators: (24)

• Long-Term Estimator (LTE) :
$$LT \Rightarrow LB \in \{180 \text{ Days}\}\$$
 (25)

• Mid-Term Estimator (MTE) :
$$MT \Rightarrow LB \in \{90 \text{ Days}\}\$$
 (26)

• Short-Term Estimator (STE) :
$$ST \Rightarrow LB \in \{40 \text{ Days}, 60 \text{ Days}\}\$$
 (27)

Term-Structure for Covariance
$$\Sigma \wedge \text{Expected Return } \mu$$
. (28)

1.5 Synthesis

$$S_{40}^{90} \equiv \hat{\Sigma} \Rightarrow 40 \text{ Days } \wedge \hat{\boldsymbol{\mu}} \Rightarrow 90 \text{ Days}$$
 (32)

Market Regime Stratification

(36)

2 Strategy

Theory \& Math

2.1 Strategic Formulation

Consider two strategies:

(Strategy I)
$$\begin{cases} \max_{\omega \in \mathbb{R}^n} \ \rho^T \omega - \lambda \sqrt{\omega^T \Sigma \omega} \\ -0.5 \le \sum_{i=1}^n \beta_i^m \omega_i \le 0.5 \\ \sum_{i=1}^n \omega_i = 1, \quad -2 \le \omega_i \le 2, \end{cases}$$
(37)

and

$$\left\{egin{aligned} \max_{\omega \in \mathbb{R}^n} & rac{
ho^T \omega}{ ext{TEV}(\omega)} - \lambda \sqrt{\omega^T \Sigma \omega} \ & -2 \leq \sum_{i=1}^n eta_i^m \omega_i \leq 2 \ & \sum_{i=1}^n \omega_i = 1, \quad -2 \leq \omega_i \leq 2, \end{aligned}
ight.$$

- $\Sigma \equiv \text{covariance matrix between security returns (FF3FM)};$
- $eta_i^m = rac{ ext{Cov}(r_i, r_M)}{\sigma^2(r_M)} \equiv ext{Beta of security } S_i ext{ (CAPM) s.t.}$ $eta_P^m = \sum_{i=1}^n eta_i^m \omega_i \equiv ext{Porfolio Beta;}$
- TEV(ω) = $\sigma(r_P(\omega) r_{SPY}) \equiv$ Tracking Error Volatility; trivial derivation (reader exercise):

$$\sigma(r_P(\omega) - r_{\text{SPY}}) = \sqrt{\omega^{\mathsf{T}} \Sigma \omega - 2\omega^{\mathsf{T}} \text{Cov}(r, r_{\text{SPY}}) + \sigma_{\text{SPY}}^2}.$$
 (39)

2.2 Fama-French Three-Factor Model (FF3FM)

Definition:
$$(40)$$

$$r_i = r_f + eta_i^3 (r_M - r_f) + b_i^s r_{\mathrm{SMB}} + b_i^v r_{\mathrm{HML}} + lpha_i + \epsilon_i$$
 (41)

 $\mathbb{E}[\epsilon_i] = 0;$...

$$\rho_i = r_f + \beta_i^3 (\rho_M - r_f) + b_i^s \rho_{\text{SMB}} + b_i^v \rho_{\text{HML}} + \alpha_i$$
(42)

$$(\hat{\beta}_i^3, \hat{b}_i^s, \hat{b}_i^v)^{\mathsf{T}} \Leftarrow y_i = \rho_i - r_f \tag{44}$$

$$= \hat{\beta}_i^3(\rho_M - r_f) + \hat{\beta}_i^s r_{\text{SMB}} + \hat{b}_i^v \rho_{\text{HML}} + \epsilon_i$$
 (46)

$$\beta_i^m \neq \beta_i^3$$
 | estimated via separate regression / computed directly. (47)

2.3 Executive Summary Formulation

Strategy I
$$(49)$$

1. Objective
$$\equiv$$
 Maximize Returns w/Risk. (50)

- The portfolio's beta must be between -0.5 and 0.5. (52)
- The sum of the weights assigned to each asset in the portfolio must equal 1.
- Each individual weight can range from -2 to 2. (54)

Strategy II
$$(55)$$

1. Objective \equiv Maximize Returns Relative to Tracking Error Volatility (TEV).

- The portfolio's beta must be between -2 and 2. (58)
- The sum of the weights assigned to each asset in the portfolio must equal 1.
- Each individual weight can range from -2 to 2. (60)

3 Assumptions

3.1 Setup

1. Reallocation: '03-01-2007' \sim '03-31-2024' (61)

2. Input Construction: (62)

• LT LB Period :
$$n_{\rm LT} = 120 \mid \Sigma_s \wedge \mu_s \mid {\rm LT} \equiv S_{120}$$
 (63)

• MT LB Period :
$$n_{\rm LT} = 90 \mid \Sigma_s \wedge \mu_s \mid {\rm MT} \equiv S_{90}$$
 (64)

• ST LB Period :
$$n_{\rm LT} = 40 \mid \Sigma_s \wedge \mu_s \mid {
m MT} \equiv S_{40}$$
 (65)

3.
$$\beta_T \in \{0, 1\}$$
 (66)

$$4. \ \lambda \in \{0.10, 0.50\} \tag{67}$$

3.2 Period Analysis

Period Stratification: (68)

• Period
$$1 \equiv \text{Pre-Subprime}$$
 (69)

• Period
$$2 \equiv \text{Subprime}$$
 (70)

• Period
$$3 \equiv \text{Post-Subprime}$$
 (71)

• Period
$$4 \equiv \text{COVID}$$
 (72)

• Period
$$5 \equiv \text{Post-Covid}$$
 (73)

3.3 BackTesting

Definition: Historical Data
$$\Rightarrow$$
 Performance (74)

• BackTest
$$\neq$$
 Forecasts \Rightarrow Snooping Bias / P-Hacking (76)

•
$$\{t_i\}_{i=1}^n$$
: (78)

For the initial date t_1 , use the prior 60 days of historical data to estimate input Store the portfolio weights: ω_{t_1} .

For each subsequent date t_{i+1} , roll the historical data window by 5 days, re-est Store the new portfolio weights: $\omega_{t_{i+1}}$.

Repeat this process until the target date t_n is reached.

4 ToolKit|Arsenal

| | 9 9 9 8 9 33 - | |
|---|---|----------------|
| | $Strat \: I \Rightarrow \mathtt{CVXPY} \mid Strat \: II \Rightarrow \mathtt{Nonlinear} \: \mathtt{Optimizer}$ | (80) |
| | ${ m Data} \ ({ m ETFs}) \ : { m yfinance}$ | (81) |
| • | 1. FXE | (82) |
| • | $2.~{ m EWJ}$ | (83) |
| • | 3. GLD | (84) |
| • | $4. \mathrm{QQQ}$ | (85) |
| • | 5. SPY | (86) |
| • | $6.~\mathrm{SHV}$ | (87) |
| • | 7. DBA | (88) |
| • | 8. USO | (89) |
| • | 9. XBI | (90) |
| • | $10.~\mathrm{ILF}$ | (91) |
| • | 11. EPP | (92) |
| • | $12.~\mathrm{FEZ}$ | (93) |
| | | |
| | To Do: | (94) |
| • | $Task 1: \verb"download'_data(start'_date, end'_date)",$ | 'compute_dail |
| • | Task 2: 'factor_model()'. | (96) |
| • | Task 3: 'optimize_model()' | (97) |

• Task 4: 'backtest(...)' (98)

• Task
$$5$$
: 'analyze(...)' (99)

• Task 6: 'summarize(...)' (100)

5 Performance + Risk Reporting 4 Strats

KPIs: (101)

• Cumulative PnL / Return (102)

• Average Daily Arithmetic / Geometric Return | Daily Min Return (103)

• 10 Day Max Drawdown | Sharpe (104)

• Vol, Skew, (Excess) Kurt, (Modified) VaR, Expected Shortfall (CVar) (10)

Tabular Formulation: (106)

 $S_{40}(eta_T = 0)$ $S_{90}(eta_T = 1)$ $S_{120}(eta_T = 0)$ SPY

Furthermore: (107)

1. Evolution Plot : Cumulative Daily PnL $\mid P_0 = \$100, \text{SPY}_0 = \$100.$ (108)

2. Plot + Analyze (Daily) Return Distribution. (109)