

# Multi-Factor Models Performance Measurement Black–Litterman Model

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# Pricing Anomalies

- Empirical research on CAPM has uncovered various systematic pricing “anomalies”
- Size effect: “small-cap” stocks (with low market value) tend to outperform “big-cap” stocks (with high market value)
- Value effect: “value” stocks (with high book-to-market ratio) outperform “growth” stocks (with low book-to-market ratio)
- Momentum effect: “winner” stocks (with recent high returns) tend to outperform “loser” stocks (with recent low returns)
- Pricing anomalies persist even after adjusting for differences in expected return due to differences in exposure to market risk

# Multi-Factor Linear Pricing Models

- Since APT provides no guidance on nature of systematic risk factors, popular approach is to augment market model with additional risk factors to account for pricing anomalies
- Similar to market model, use returns on **factor-mimicking portfolios** to represent effect of risk factors
- In principle, factor-mimicking portfolio has sensitivity of one to specific risk factor and sensitivity of zero to other risk factors
- In practice, factor-mimicking portfolios will not be completely insensitive to other risk factors
- Factor-mimicking portfolios are usually zero-investment portfolios, which reduces correlation among risk factors

# Fama–French 3-Factor Model – Part 1

- Fama and French (1993) use three-factor model for realised returns, with additional risk factors for size and value effects:

$$R_i - R_f = \alpha_i + \beta_i (R_m - R_f) + \gamma_i (R_s - R_b) + \delta_i (R_h - R_l) + \tilde{\epsilon}_i$$

- Risk factor for size effect (“SMB”) is return on portfolio that is long on small-cap stocks and short on big-cap stocks
- Risk factor for value effect (“HML”) is return on portfolio that is long on value stocks and short on growth stocks

# Fama–French 3-Factor Model – Part 2

- Economic intuition is that “size risk” and “value risk” are separate types of systematic risk that cannot be combined with each other or market risk
- If Fama–French three-factor model provides correct description of realised returns, then  $\alpha_i$  will not be significantly different from zero for any asset or (passive) portfolio
- Hence risk premium for any asset is given by:

$$E[R_i - R_f] = \beta_i E[R_m - R_f] + \gamma_i E[R_s - R_b] + \delta_i E[R_h - R_l]$$

# Carhart 4-Factor Model – Part 1

- Carhart (1994) augments Fama–French three-factor model with additional risk factor for momentum effect:

$$R_i - R_f = \alpha_i + \beta_i (R_m - R_f) + \gamma_i (R_s - R_b) + \delta_i (R_h - R_l) + \zeta_i (R_u - R_d) + \epsilon_i$$

- Risk factor for momentum effect (“UMD”) is return on portfolio that is long on winners and short on losers
- Performs better than Fama–French three-factor model for actively-managed portfolios, since fund managers tend to buy winners and sell losers

# Carhart 4-Factor Model – Part 2

- Economic intuition is that “momentum risk” cannot be combined with other types of systematic risk
- If Carhart four-factor model provides correct description of realised returns, then  $\alpha_i$  will not be significantly different from zero for any asset or portfolio
- Hence risk premium for any asset is given by:

$$E[R_i - R_f] = \beta_i E[R_m - R_f] + \gamma_i E[R_s - R_b] + \delta_i E[R_h - R_l] + \zeta_i E[R_u - R_d]$$

# Sharpe Ratio

- **Sharpe ratio** is risk premium per unit of standard deviation:

$$SR_i = \frac{E[R_i - R_f]}{\sqrt{\text{Var}[R_i - R_f]}}$$

- Denominator is designed to capture total risk, i.e., both systematic and idiosyncratic risk
- However, denominator ignores higher moments such as skewness and kurtosis, so may not fully reflect risk of investment if return distribution is not normal



# Treynor Ratio

- **Treynor ratio** is risk premium per unit of market risk:

$$TR_i = \frac{E[R_i - R_f]}{\beta_i}$$

- Denominator is designed to capture systematic risk, but cannot easily be extended to account for possibility of multiple sources of systematic risk
- Measures “reward-to-risk” ratio for asset as part of well-diversified portfolio

# Jensen's Alpha

- **Jensen's alpha** is intercept coefficient from market model regression using excess returns:

$$\alpha_i = E[R_i - R_f] - \beta_i E[R_m - R_f]$$

- For passive portfolio, represents pricing error for CAPM
- For actively-managed portfolio, represents “abnormal” or “value added” premium due to managerial skill
- Can extend to allow for multiple sources of systematic risk by using intercept coefficient from regression using Fama–French three-factor model or Carhart four-factor model

# Information Ratio

- **Information ratio** (or **appraisal ratio**) is mean deviation from target (or benchmark) return, per unit of **tracking error**:

$$IR_i = \frac{E[R_i - R_t]}{\sqrt{\text{Var}[R_i - R_t]}}$$

- Same as Sharp ratio when target return is risk-free rate
- Measures ability of fund manager to exceed target return, relative to amount of additional risk exposure

# Downside Risk

- **Downside risk** of investment is risk that realised return on investment will fall below target return
- One measure of downside risk is **below-target semi-variance**:

$$SV[R_i; R_t] = E \left[ \min \left\{ R_i - R_t, 0 \right\}^2 \right]$$

- Can distinguish between asymmetric return distributions with same variance but different skewness
- Intuition is that investors are only concerned with probability and magnitude of underperformance, not overperformance

# Sortino Ratio

- **Sortino ratio** is mean deviation from target (or benchmark) return, per unit of below-target semi-deviation:

$$StR_i = \frac{E[R_i - R_t]}{\sqrt{SV[R_i; R_t]}}$$

- Will produce rankings similar to information ratio when return distribution is close to symmetric, and mean asset return is close to mean target return
- May deviate from rankings produced by information ratio when return distribution is highly skewed, or when mean asset return deviates substantially from mean target return

# Issues with Efficient Frontier

- Biggest econometric issue with mean-variance-efficient frontier is difficulty of estimating mean return
- Standard error of sample mean is  $\sigma/\sqrt{n}$ , where  $\sigma$  is standard deviation of return and  $n$  is number of observations
- Returns are volatile and data is limited, so estimate of mean return tends to have large standard error
- Related issue is instability of efficient frontier: small change in vector of mean returns can induce large change in composition of frontier portfolios
- Fischer Black and Robert Litterman developed model in 1990 to overcome issues with efficient frontier

# Excess Returns – Part 1

- Excess returns for tradable assets follow normal distribution:

$$\tilde{\mathbf{R}} \sim N(\tilde{\boldsymbol{\mu}}, \boldsymbol{\Sigma})$$

- Vector of (unknown) risk premiums is also random variable:

$$\tilde{\boldsymbol{\mu}} = \boldsymbol{\pi} + \tilde{\boldsymbol{\epsilon}}$$

- Here  $\tilde{\boldsymbol{\epsilon}}$  is error term with (independent) normal distribution:

$$\tilde{\boldsymbol{\epsilon}} \sim N(\mathbf{0}, \boldsymbol{\Sigma}_{\mu})$$

# Excess Returns – Part 2

- For simplicity, assume that  $\Sigma_\mu$  is proportional of  $\Sigma$ :

$$\Sigma_\mu = \tau \Sigma$$

- In practice, often set  $\tau = 1/n$ , where  $n$  is number of data points used to estimate  $\Sigma$
- Hence distribution of excess returns around  $\pi$ :

$$\tilde{R} \sim N(\pi, (1 + \tau) \Sigma)$$



# Implied Risk Premiums

- Use observed weights of individual assets in market portfolio to determine implied risk premiums:

$$\pi = \lambda \Sigma \mathbf{w}_m$$

- Here  $\lambda$  is coefficient of relative risk aversion (based on initial wealth), for investor with constant absolute risk aversion
- Calibrate  $\lambda$  using market risk premium or Sharpe ratio:

$$\lambda = \frac{\mathbf{w}_m' \pi}{\mathbf{w}_m' \Sigma \mathbf{w}_m} = \frac{R_m - R_f}{\sigma_m^2} = \frac{S_m}{\sigma_m}$$

# Investor Views

- Black-Litterman model also incorporates investor's "views" on (absolute or relative) expected returns of risky assets
- Suppose that investor has  $k \geq 1$  views on expected returns
- Let  $\mathbf{P}$  be  $k \times n$  vector of asset weights corresponding to investor's views, and let  $\mathbf{Q}$  be  $k \times 1$  vector of expected returns corresponding to investor's views
- Also let  $\mathbf{\Omega}$  be  $k \times k$  covariance matrix of error terms derived from confidence of investor's views
- For simplicity, require that investor's views be uncorrelated with one another, so that  $\mathbf{\Omega}$  is diagonal

# Example: Investor Views

- Suppose that investible universe consists of three risky assets (or asset classes, or mutual funds)
- Investor expects first risky asset to earn return of 5% per year
- Investor expects that second risky asset to outperform third risky asset by 100 basis points (i.e., 1%) per year
- Then  $\mathbf{P} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & -1 \end{bmatrix}$  and  $\mathbf{Q} = \begin{bmatrix} 0.05 \\ 0.01 \end{bmatrix}$
- Notice that first row of  $\mathbf{P}$  corresponds to absolute view (where weights sum to one), while second row of  $\mathbf{P}$  corresponds to relative view (where weights sum to zero)

# Bayes' Theorem

- Bayes' theorem is used to update probability of given hypothesis  $H$  when new evidence  $E$  is observed:

$$\Pr(H|E) = \frac{\Pr(E|H)}{\Pr(E)}\Pr(H)$$

- $\Pr(H)$  is **prior distribution**, or probability that  $H$  is true before any new evidence is observed
- $\Pr(E|H)$  is conditional probability of observing  $E$ , assuming that  $H$  is true, while  $\Pr(E)$  is unconditional probability of observing  $E$ , which is independent of  $H$
- $\Pr(H|E)$  is **posterior distribution**, or probability that  $H$  is true given that we have observed  $E$

# Posterior Distribution of Asset Returns

- Black-Litterman model uses Bayes' theorem to update prior distribution of excess returns using investor's views
- Conditional on investor's views, excess returns have (posterior) normal distribution of  $N(\hat{\pi}, \mathbf{M})$ , where:

$$\hat{\pi} = \pi + \tau \mathbf{\Sigma} \mathbf{P}' (\tau \mathbf{P} \mathbf{\Sigma} \mathbf{P}' + \mathbf{\Omega})^{-1} (\mathbf{Q} - \mathbf{P} \pi),$$

$$\mathbf{M} = \left( \frac{1}{\tau} \mathbf{\Sigma}^{-1} + \mathbf{P}' \mathbf{\Omega}^{-1} \mathbf{P} \right)^{-1}$$