

portfolio-2.pdf = Modern Portfolio Theory and Investment Analysis

To Do

Compare simple beta estimation results with bayesian analysis results.

Notes

1. Chapter 7 of portfolio-2.pdf for adjusted betas, beta should be adjusted using bayesian technique mentioned.
2. Page 161 of portfolio-2.pdf single index model outperforms multi-index models Burmeister and McElroy (1988).
3. Pg 165(portfolio-2.pdf) 1st para fama french test on forecasting future correlation model(Chan, Karceski, and Lakonishok (1999)). Constant correlation outperforms single index and fama french in forecasting correlations.
4. Measuring Performance using benchmark with modified sharpe portolfio-2.pdf Pg 678
5. Recently Jagannathan and Ma (2003), proved that mean-variance optimizers are already implicitly applying some form of shrinkage to the sample covariance matrix when short sales are ruled out, and that this is generally beneficial in terms of improving weights stability
6. Allen, Lizieri, and Satchell (2019) says,

The theory that high-frequency data can be used to eliminate estimation error in the covariance matrix is constrained in practice because of microstructure issues, thin trading, and departures from normality, all of which place a limit on precision (Hansen and Lunde (2006)). As a result, lower-frequency data tend to be used in conjunction with factor models to estimate the covariance matrix for investment problems involving large numbers of assets.
7. Allen, Lizieri, and Satchell (2019) says

Our results cast doubt on the conclusion of DeMiguel et al. (2009) that mean–variance is unworkable in higher dimensions. Consider that Grinold and Kahn (1999) developed a simple binary model that relates the information coefficient to the number of forecasts that are directionally correct (the “hit rate”). Our IC of 0.07 equates to a hit rate of just 53.5% a month, or an R2 of 0.5%. That such modest levels of forecasting ability can generate meaningful gains in utility is remarkable. The uplift in utility is tangible, and it is a benefit that increases with the size of the asset universe.
8. According to Ehsani and Linnainmaa (2022),

Factor momentum’s ability to span individual stock momentum, but not vice versa, suggests that individual stock momentum is a

manifestation of factor momentum. An investor who trades individual stock momentum indirectly times factors, and an investor who directly times factors performs better. The indirect method loses out because it also takes positions based on noise. The other possible sources of momentum profits do not contribute to these profits, and so their inclusion renders the strategy unnecessarily volatile.

Basically saying that stock momentum is a result of factor momentum.

References

- Allen, David, Colin Lizieri, and Stephen Satchell. 2019. “In Defense of Portfolio Optimization: What If We Can Forecast?” *Financial Analysts Journal* 75 (3): 20–38.
- Burmeister, Edwin, and Marjorie B McElroy. 1988. “Joint Estimation of Factor Sensitivities and Risk Premia for the Arbitrage Pricing Theory.” *The Journal of Finance* 43 (3): 721–33.
- Chan, Louis KC, Jason Karceski, and Josef Lakonishok. 1999. “On Portfolio Optimization: Forecasting Covariances and Choosing the Risk Model.” *The Review of Financial Studies* 12 (5): 937–74.
- Ehsani, Sina, and Juhani T Linnainmaa. 2022. “Factor Momentum and the Momentum Factor.” *The Journal of Finance* 77 (3): 1877–1919.
- Hansen, Peter R, and Asger Lunde. 2006. “Realized Variance and Market Microstructure Noise.” *Journal of Business & Economic Statistics* 24 (2): 127–61.
- Jagannathan, Ravi, and Tongshu Ma. 2003. “Risk Reduction in Large Portfolios: Why Imposing the Wrong Constraints Helps.” *The Journal of Finance* 58 (4): 1651–83.