

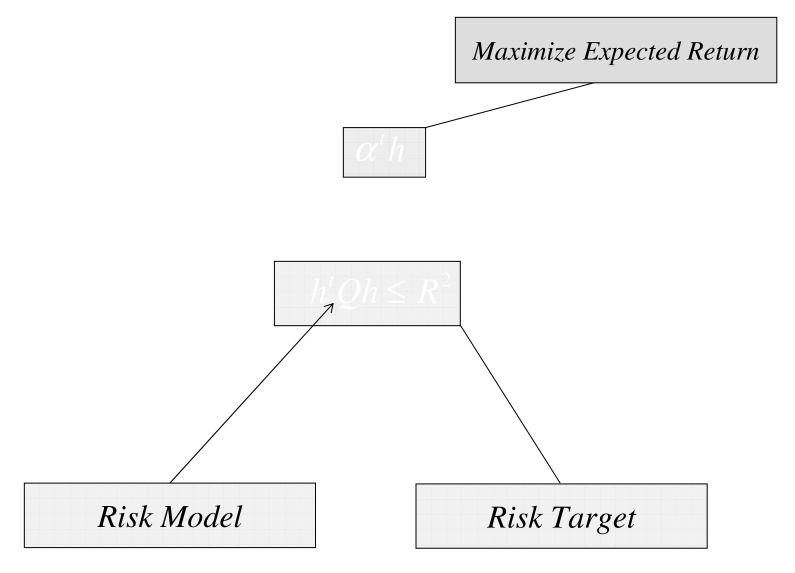
#### Factor Alignment Problems in Optimized Portfolio Construction

October  $4^{th}$  , 2011

Rob Stubbs VP Research Axioma, Inc.

Joint work with Anureet Saxena

#### Mean Variance Optimization Notation



#### How are Q and alpha constructed?

• Factor model of returns

$$r = Xf + \varepsilon$$

• Expected Returns Model

$$E[r] = XE[f]$$

• Factor Risk Model

$$Var[r] = XE[ff^{T}]X^{T} + E[\varepsilon\varepsilon^{T}]$$
$$= X\Omega X^{T} + \Delta$$

## Risk Models and Alpha Interaction

- Are optimal MVO portfolios "biased" with respect to certain risk models?
- How does a risk model used in MVO affect the optimal portfolio?
- Why do risk estimates provided by risk models that were used to construct an MVO portfolio tend to underestimate risk?
- Can an existing risk model be modified to be unbiased when used in optimization?
- And more importantly, can "mean-variance efficiency" be made more efficient?

## Long Only Strategy

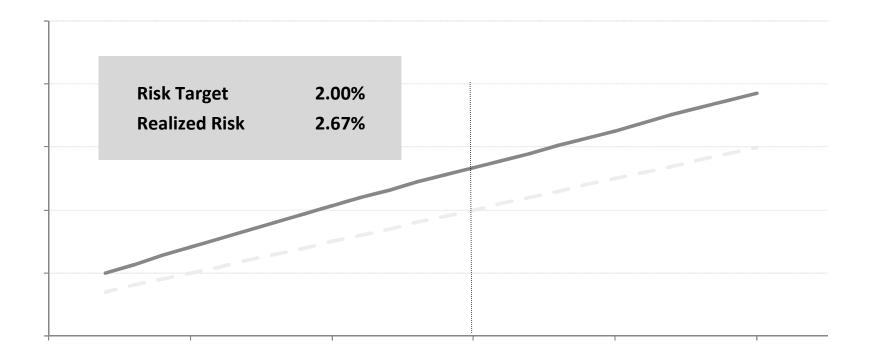
**Maximize Expected Return** 

s.t

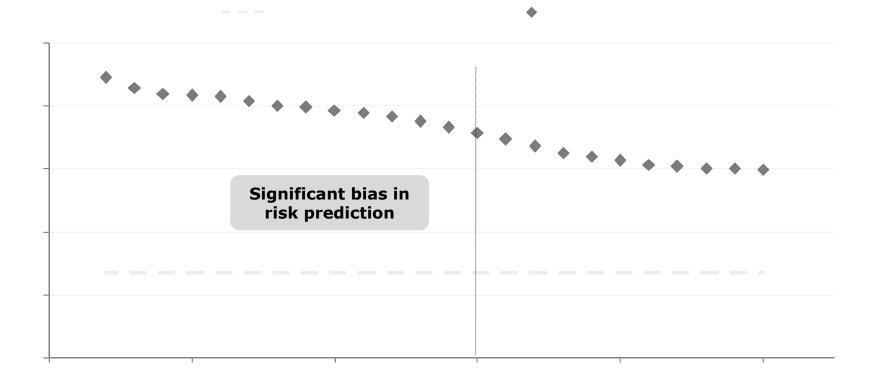
- Fully invested long only portfolio
- GICS Industry exposure constraints
- Active asset bounds constraint
- Active Risk Constraint (2%)

Risk Model = US2AxiomaMH Benchmark = Russell 1000 Growth Monthly backtest, 1999- 2007 time period Alphas are not completely aligned with risk factors

#### **Predicted vs Realized Active Risk**

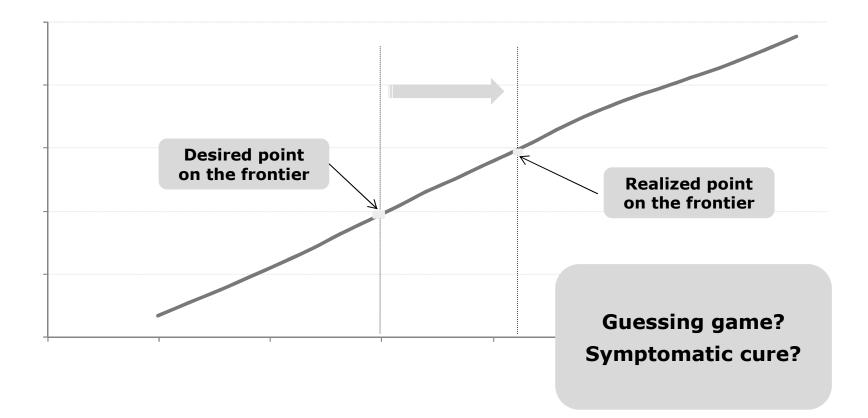


#### **Bias Statistic**

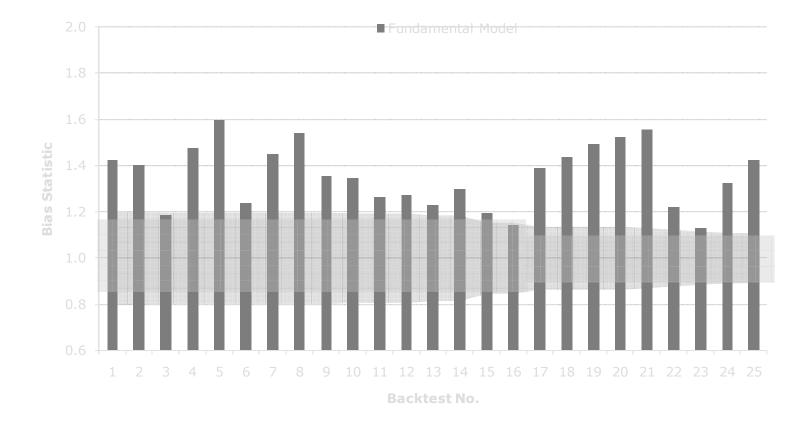


• For our purposes, the bias statistic is (realized risk/predicted risk)

#### **Risk Return Frontier**



#### **Pervasive Problem**

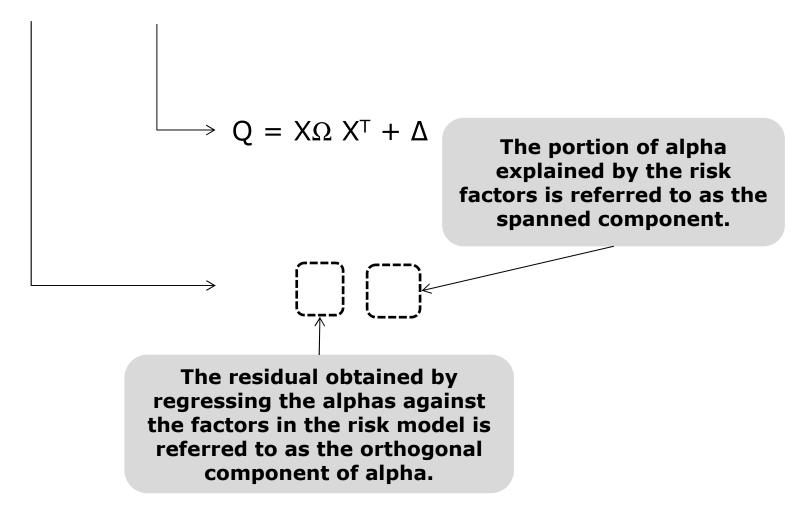


#### **Fact Sheet**

#### **1.** Risk under-estimation of optimized portfolios

How do we explain this phenomenon?

#### **Unconstrained MVO**



#### **Unconstrained MVO**

 $\rightarrow$ 

The optimizer sees no systematic risk in the orthogonal component of alpha and is hence likely to load up on it

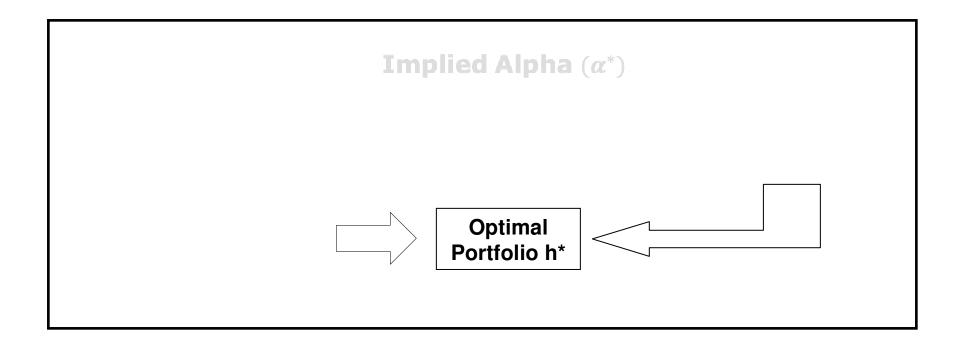
#### Previous Research: The Unconstrained Case

• Consider the Unconstrained Mean-Variance Optimization Problem

- Single Factor Models (*Lee & Stefek '08*)
  Clearly, is overweighted in the optimal solution relative to
- Multi-Factor Models (*Saxena & Stubbs '10*)



## **Constrained MVO**

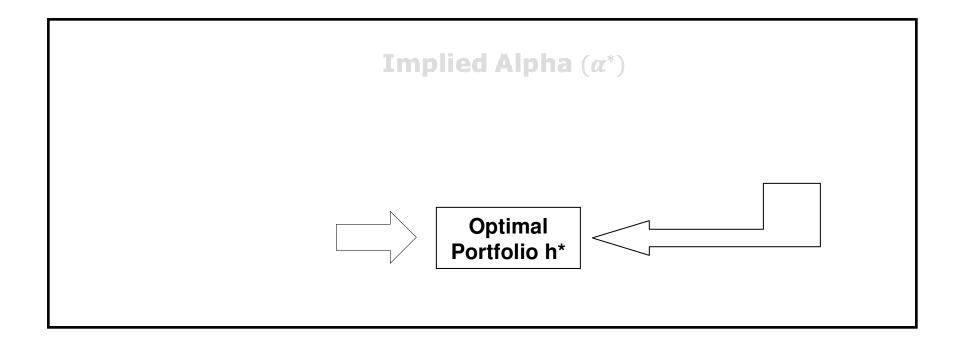


• Implied alpha acts as the *de facto* alpha in the case of constrained MVO problems

• Optimizer sees no systematic risk in the orthogonal component of implied alpha and is hence likely to load up on it.

• Implied alpha is a *dynamic* entity determined by the interaction of alpha, risk factors and constraints via KKT conditions

## **Constrained MVO**



$$AAF = \frac{1}{|h_{\perp}|}h_{\perp} = \frac{1}{|\alpha_{\perp}^*|}\alpha_{\perp}^*$$

An important result that allows us to penalize exposure of the portfolio to the orthogonal component of implied alpha

AAF = Alpha Alignment Factor

## **Extent of overloading?**

Norm-Square Decomposition **Backtest No.** 

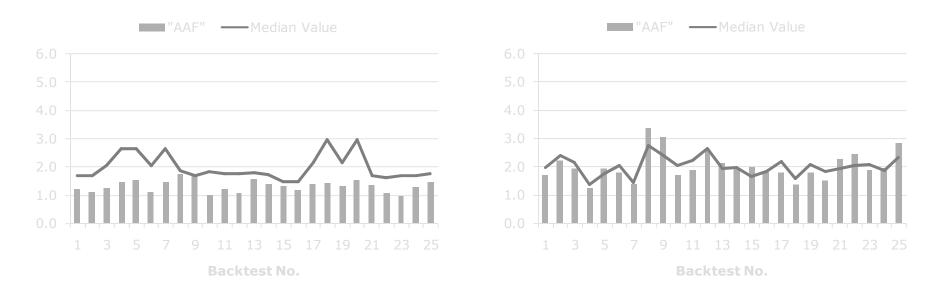
■ Orthogonal ■ Spanned

#### **Fact Sheet**

- **1.** Risk under-estimation of optimized portfolios
- 2. The optimizer assumes that the orthogonal component of the portfolio has no systematic risk and hence overloads on it by a factor of 8!!

Is it really true that the orthogonal component of the portfolio has no systematic risk?

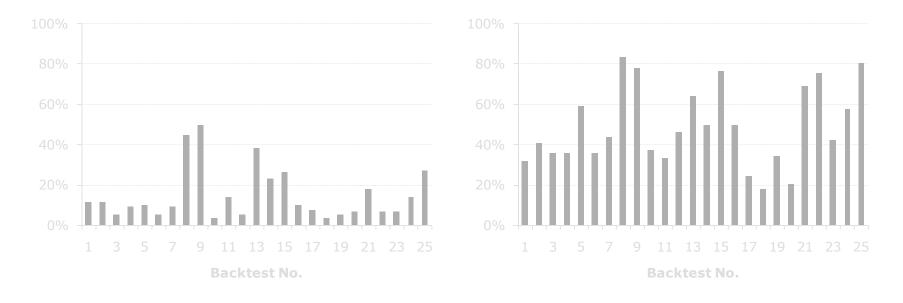
# How does AAF compared with regular risk factors: **RMS t-statistics**



#### Unbiased Setup

#### Portfolio Biased Setup

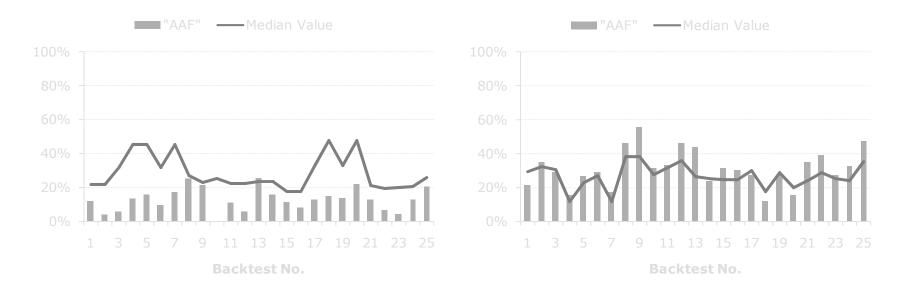
# How does AAF compared with regular risk factors: **AAF Percentile Ranking**



#### Unbiased Setup

#### Portfolio Biased Setup

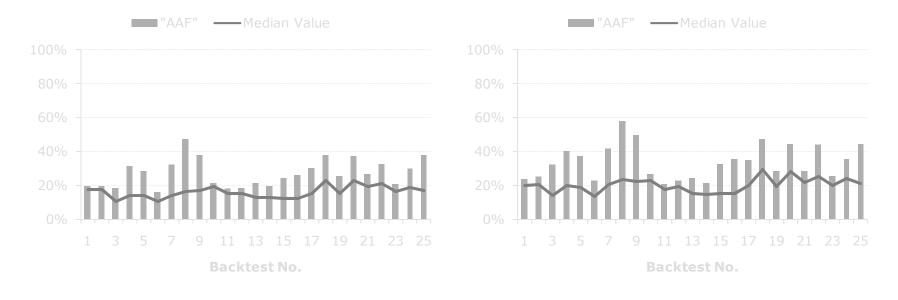
# How does AAF compared with regular risk factors: **Statistical Significance of AAF**



#### Unbiased Setup

#### Portfolio Biased Setup

# How does AAF compared with regular risk factors: **Realized Volatility of AAF**



#### **Unbiased Setup**

#### Portfolio Biased Setup

## **Fact Sheet**

- 1. Risk under-estimation of optimized portfolios
- 2. The optimizer assumes that the orthogonal component of the portfolio has no systematic risk and hence overloads on it.
- 3. The orthogonal component of the portfolio has latent systematic risk.

 $h_{\perp}$  is not an excellent generic risk factor  $h_{\perp}$  can be a useful factor in explaining the volatility of the optimal portfolio

What is the source of this latent systematic risk?

## Sources of Factor Misalignment

Alpha Factors

Proprietary definition of style factors e.g. EBITDA/EV vs E/P

• Constraints

Limiting exposure to custom factors e.g. Exposures to liquidity factor

Proprietary definition of technical factors
 Using different parameters to define common factors
 e.g. Momentum Factors

## Is Misalignment Bad?

- When we work in "theory", we assume that the orthogonal component of alpha is all noise
  - In this case, we do not want exposure to orthogonal alpha (Hence, Misalignment is Bad)
- However, in practice, the orthogonal component of alpha may have positive IC, so, a positive exposure to the orthogonal component of alpha will increase returns (Hence, Misalignment is good)
- But, what if the orthogonal component of alpha also contains systematic risk? (Then, we should be managing the tradeoff given by the misalignment)

## Augmented **Risk Models**

• Assume that the true covariance matrix  $Q_T$  is given by

- Z denotes systematic risk factors which are missing from the user risk model Q.
- Key Idea: Construct an augmented risk model by adding one additional risk factor to the original factor model that can capture the effect of the missing factors.
- Augmented Covariance Matrix
- Goal: Evaluate the effect of augmenting a risk model on the optimal portfolio

## Augmented Risk Models

- denote the optimal portfolio associated with the original covariance matrix Q
- denote the optimal portfolio associated with the augmented covariance matrix
- U(Q) denotes the true utility function of h(Q)
- $U(Q_{y})$  denotes the true utility function of  $h(Q_{y})$

#### **Pushing Frontiers**

#### Theorem

Using an Augmented Risk Model

- Pushes the ex-post frontier upwards
- Eliminates the bias in risk prediction
- Restores Markowitz MVO efficiency

## **Augmented Risk Models**

Theorem

- If  $y^T \alpha_{\perp} = 0$  then the augmented risk model has no impact on the holdings
- Thus if we want to materially affect the optimal portfolio, we need to choose an augmenting factor y which is correlated with  $\alpha_{\perp}$ .
- Using an augmented risk model is tantamount to tilting the portfolio in a direction away from y.

#### **Augmented Risk Models**

Theorem

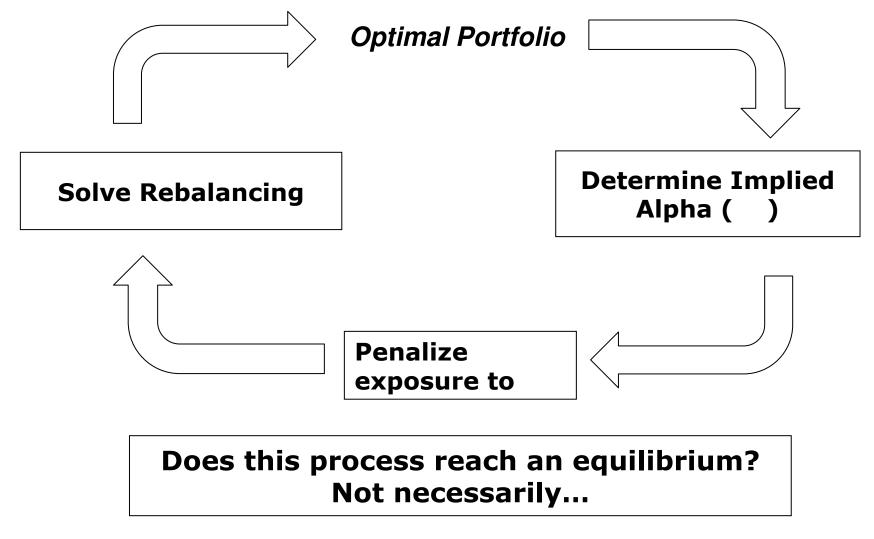
#### • If then

By calibrating the *v* parameter we can control the extent of overloading on the orthogonal component of alpha.

# Manage Exposure to Orthogonal Implied Alpha

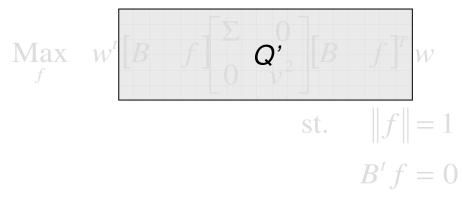
- We really need the implied alpha, not alpha, which means, this process would require us to solve optimization problems to identify the implied alpha
- We would need to re-estimate the whole risk model with the new factor
- We would need to iterate this process, until the orthogonal component of the implied alpha is zero
- Not practical, may not converge -- not recommended
- Alternatively, penalize exposure to the orthogonal implied alpha

#### Penalizing Exposure to the Orthogonal Implied Alpha: Iterative Refinement



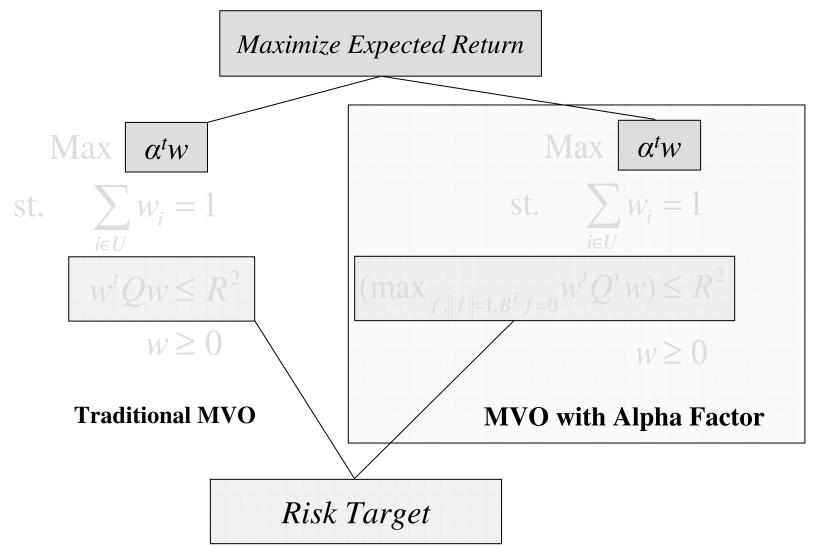
## Finding the Alpha Alignment Factor

• We solve an optimization problem to find an additional factor that increases risk the most, given a fixed portfolio (*w*)



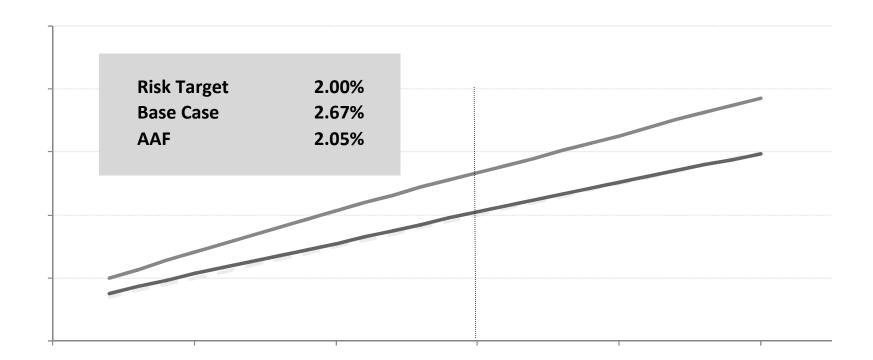
- *v*<sup>2</sup> a fixed constant (new factor volatility)
- $B^T f = 0$  new factor is "orthogonal" to existing factors
- || f || = 1 normalize new factor exposures

#### Integrating the AAF and MVO

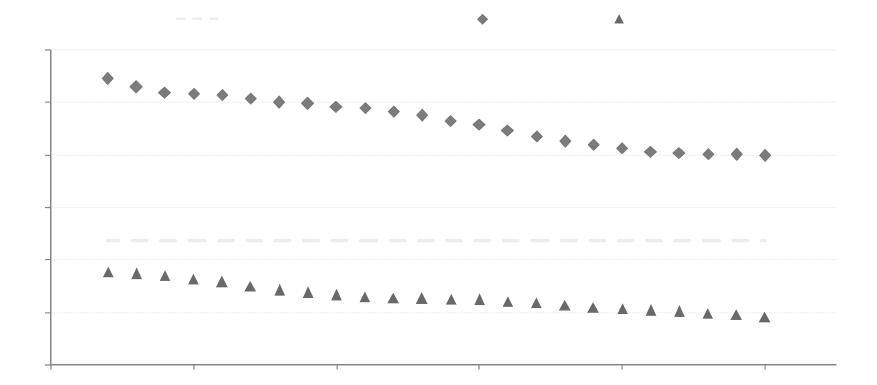


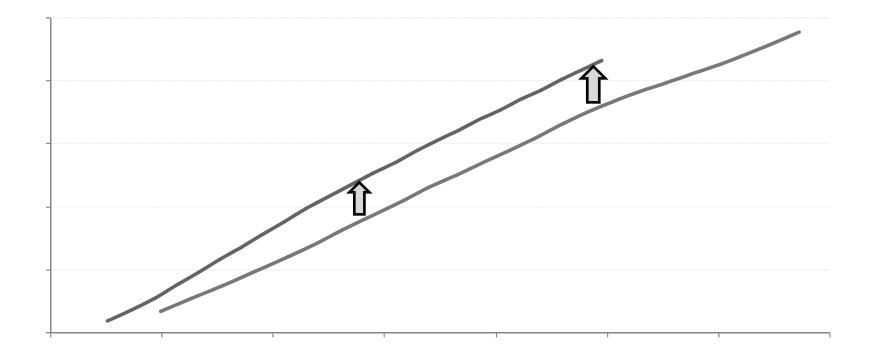
## Robust Optimization provides AAF

- Robust portion has closed form solution:  $f = \frac{w_{\perp}}{\|w_{\perp}\|}$
- "Alpha Alignment Factor" (f) is the orthogonal part of implied alpha
- Adding the Alpha Factor to the optimization problem as we proposed before is, in fact, penalizing the exposure to the orthogonal implied alpha
- Still have to manage the tradeoff; fortunately, it is only one parameter v, the "volatility" of the AAF

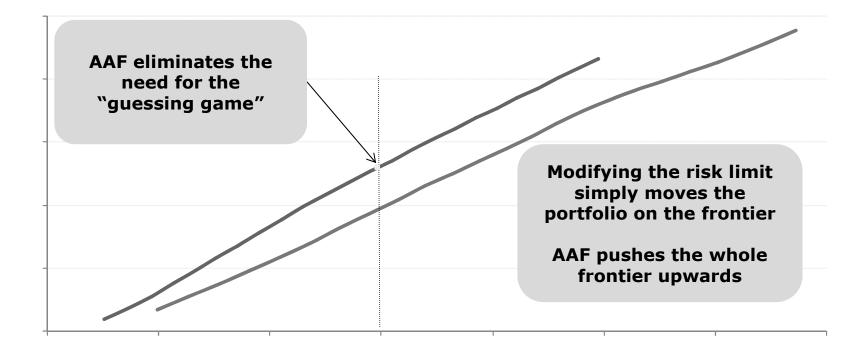


**Unbiased Risk Prediction** 





**Pushing Frontiers, literally!** 

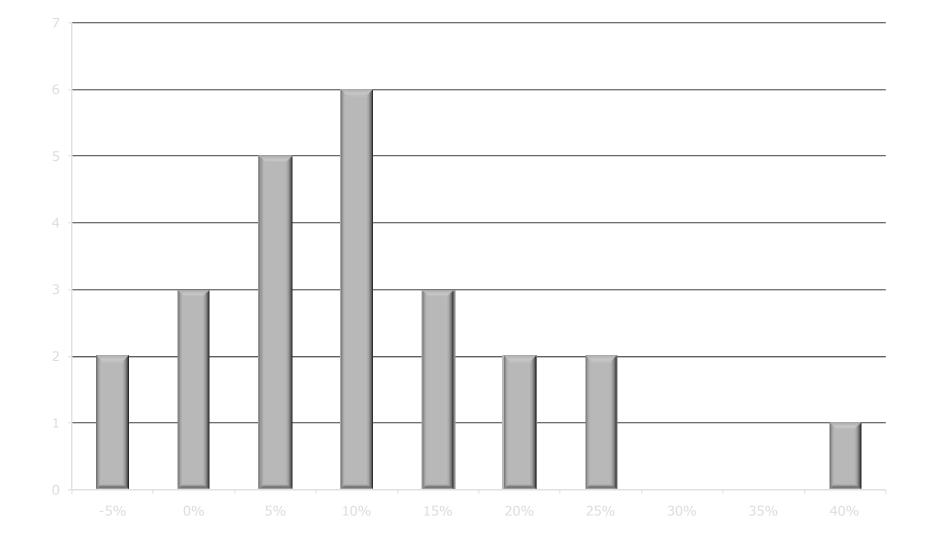


#### **Ex-post performance improvement?**

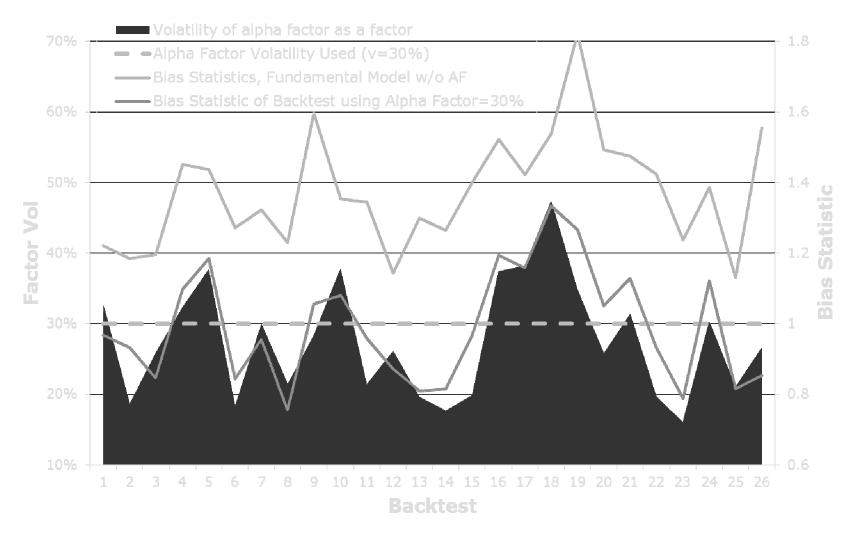
**Q.** How can an approach designed exclusively to remove bias in risk prediction also improve ex-post performance?

- Α.
- Application of risk models in quantitative portfolio construction is not just limited to risk forecasting
- Risk models materially affect the composition of optimal holdings
- Risk models indirectly affect Sharpe/Information ratio, transfer coefficient, risk/budget allocation across various securities, etc.
- Bias in risk models naturally leads to inefficient budget allocation across various securities
- Using AAF not only removes unintended systematic bets but also improves the "efficiency" of the budget allocation process itself

## **Improvements in IR (distribution)**

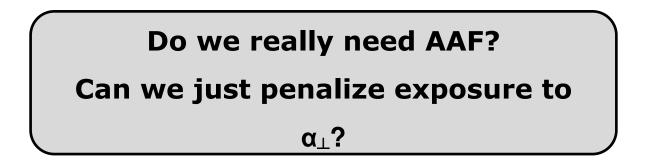


#### Relationship between Bias Statistic and Misestimation of Volatility of Alpha Factor

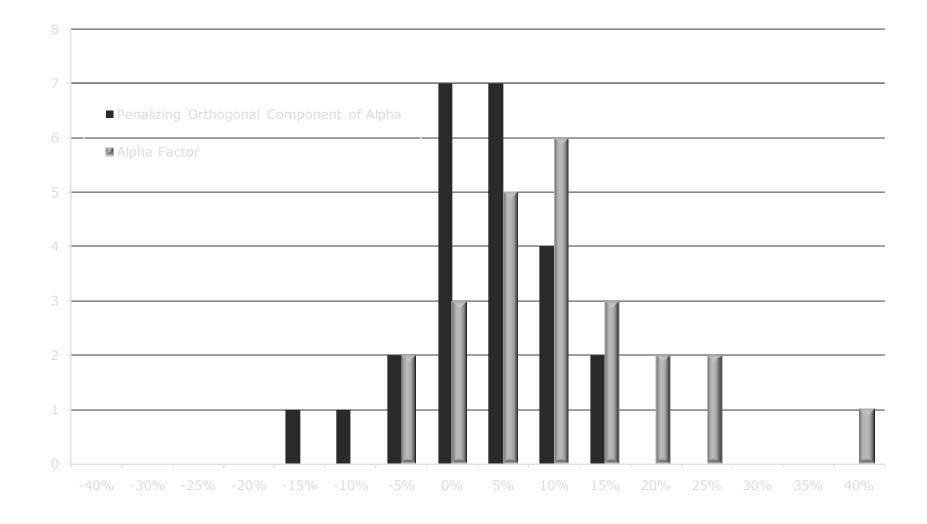


## **Fact Sheet**

- 1. Risk under-estimation of optimized portfolios
- 2. The optimizer assumes that the orthogonal component of the portfolio has no systematic risk and hence overloads on it
- 3. The orthogonal component of the portfolio has latent systematic risk
- 4. Misalignment between alpha factors, risk factors and constraints
- 5. Pushing Frontiers, literally!



#### Improvements in IR (distribution)



## **Fact Sheet**

- 1. Risk under-estimation of optimized portfolios
- 2. The optimizer assumes that the orthogonal component of the portfolio has no systematic risk and hence overloads on it
- 3. The orthogonal component of the portfolio has latent systematic risk
- 4. Misalignment between alpha factors, risk factors and constraints
- 5. Pushing Frontiers, literally!
- 6. Penalizing the exposure to  $\alpha_{\perp}$  is only a partial solution to the alignment problem, and inferior to using the AAF methodology.

#### Theoretical Foundation: Our Contribution

#### • Augmented Risk Models

- Risk model obtained by adding a single auxiliary factor y that is uncorrelated with all the user risk factors.
- Our research is based on detailed analysis of augmented risk models and their capability in capturing the effect of missing systematic risk factors.

#### • Main Theoretical Results

- There is improvement in the ex-post utility function *if and only if* the auxiliary factor *y* is correlated with the orthogonal component of implied alpha.
- Factor alignment problems arise *if and only if* the orthogonal component of implied alpha has systematic risk.
- The effect of all the missing factors can be captured by just one auxiliary factor. The Alpha Alignment Factor is such a factor.
- Using AAF pushes the ex-post risk-return frontier upwards. In other words, it allows the PM to reach a given return target at a lower value of realized risk than obtained without using **Alpha Alignment Factor**.

#### Research Papers (www.axiomainc.com)

- Saxena, A. and Stubbs, R.A., Alpha alignment factor: A solution to the underestimation of risk for optimized active portfolios. Technical report, Axioma, Inc. Research Report #015, February 2010. Submitted to *Journal of Risk*.
- Saxena, A. and Stubbs, R.A., Pushing the Frontier (literally) with the Alpha Alignment Factor. Technical report, Axioma, Inc. Research Report #022, September 2010.
- Ceria, S., Saxena, A. and Stubbs, R.A., Factor Alignment Problems and Quantitative Investing. To appear in *Journal of Portfolio Management, 2012.*
- Saxena, A. and Stubbs, R.A., An empirical case study of factor alignment problems using the United States Expected Returns (USER) model. Technical report, Axioma, Inc. Research Report #036, October 2011. Submitted to *Journal of Investing.*
- Saxena, A. and Stubbs, R.A., Axioma's Alpha alignment factor demystified. Axioma Advisor, September 2010.

## **Next Steps...**

• Custom risk models