

Portfolio Construction with Downside Risk

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Motivation

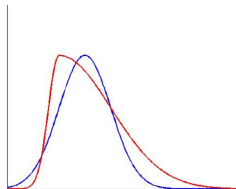
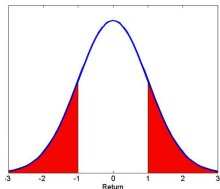
- Asset returns exhibit asymmetric return distributions
- Several asymmetric risk measures have been proposed to handle downside risk
- Portfolio optimization may well benefit from their inclusion
- We discuss several downside risk measures and empirically assess their use in a portfolio optimization study

Motivation

Volatility Issues

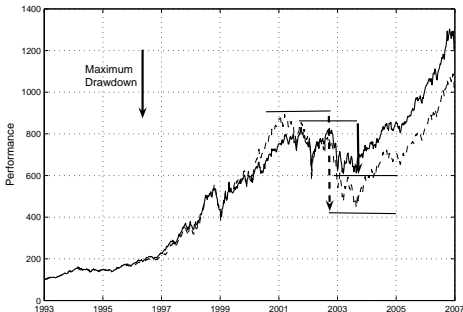
- popular measure for portfolio optimization
- based on the absolute deviation of returns from the mean value
- counts positive deviations as risk—contradicting intuition
- Distributions with equal mean and volatility may entail different downside risks

→ Optimize portfolios with asymmetric risk measures instead



Maximum Drawdown

Maximum Drawdown is the most intuitive measure of downside risk: It is the maximum percentage loss in a given time period



- Independent of return distribution
- Only requires return observations

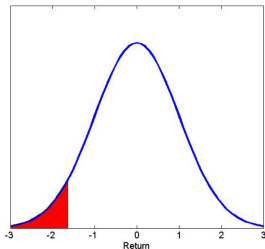
Value at Risk (VaR) and Conditional VaR

Value at Risk

- VaR is a quantile of the return distribution
- Optimization with VaR may lead to fat-tailed return distributions

Conditional VaR:

- CVaR is a conditional expectation: What is the expected return below a given threshold?
- Estimation of CVaR requires certain amount of observations below the threshold



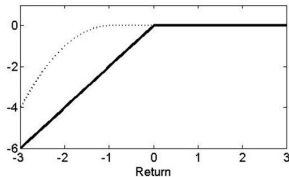
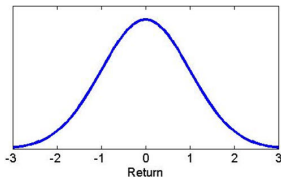
Lower Partial Moments

Semi-Deviation and Semi-Variance

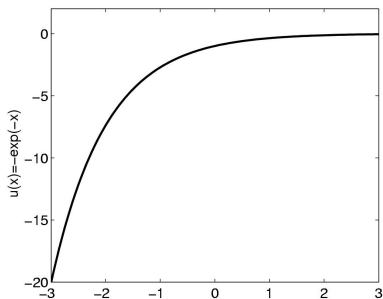
- LPMs are a family of risk measures accounting for downside risk:

$$LPM_{\tau,k}(R) = \mathbb{E} \left((\tau - R)^k \mid R < \tau \right) \cdot P(R < \tau)$$

- Returns are perceived risky when below threshold τ , degree of risk perception dependent of k
- Semideviation: $\tau = \mu$ and $k = 1$
- Semivariance: $\tau = \mu$ and $k = 2$



Loss-oriented Utility



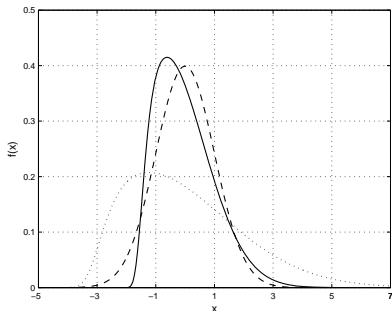
Loss Penalty

- Investor maximizes specific utility function that accounts for downside risk

$$u(x) = -\exp\{-(x - \mathbb{E}(X))\}.$$

- Adjustment for mean return to pursue risk reduction

Skewness



- Definition:

$$\gamma(R) = \frac{\mathbb{E}((R - \mathbb{E}R)^3)}{\sigma(R)^{3/2}}$$

- Right-skewed distributions appeal to downside risk-averse
- Avoid highly volatile portfolios by additionally constraining volatility

$$\hat{\gamma}(X) = \frac{1}{T} \sum_{t=1}^T \left(\frac{X_t - \hat{\mu}}{\hat{\sigma}(X)} \right)^3 \text{ versus } S_B(X) = \frac{Q_3(X) + Q_1(X) - 2Q_2(X)}{Q_3(X) - Q_1(X)}$$

Portfolio Optimization with Downside Risk

- Asymmetric return distributions may severely hamper standard mean-variance optimization
- Including downside risk measures in portfolio optimization may help to construct portfolios that better suits investors' risk perception
- We consider the following portfolio optimization study:
 - A portfolio manager is benchmarked against Euro STOXX 50 with a maximum TE of 5%
 - optimizes his risk exposure according to alternative definitions of downside risk
 - has no additional incentive to exceed the benchmark's return

Portfolio Optimization with Downside Risk

Additionally the manager

- faces constraints:
 - $\boldsymbol{\mu}^T \mathbf{x} \geq \boldsymbol{\mu}^T \mathbf{x}_{BM}$
 - $\sum_{i=1}^N x_i = 1$ and $x_i \geq 0 \quad \forall i$
 - For skewness: $\sigma(\mathbf{x}) = \sqrt{\mathbf{x}^T \mathbf{C} \mathbf{x}} \leq \sigma_{BM}$
- rebalances quarterly using two years' weekly returns

Manager's information set with respect to

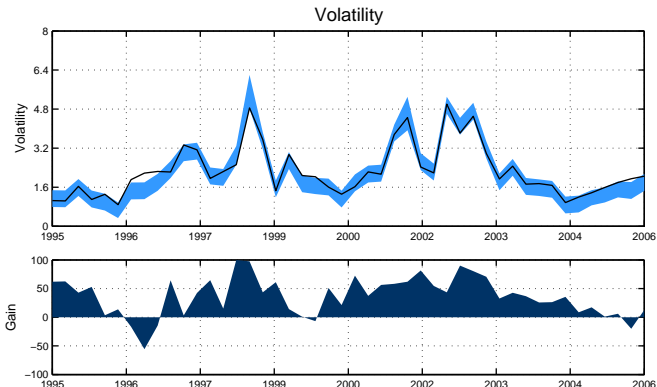
- Risk: has to be estimated from historical data
- Returns: The Manager
 1. has perfect foresight
 2. employs a naïve estimate

Minimizing Downside Risk: Perfect Foresight

Strategy		Risk			Gain average	Hit Rate
		<i>benchmark</i>	<i>optimized</i>	<i>optimal</i>		
1	Volatility	2.538	2.264	1.854	39.96%	89.13%
2	VaR	-0.460	-1.025	-2.165	33.13%	84.78%
3	Conditional VaR	1.702	1.095	0.553	52.83%	91.30%
4	Semideviation	2.106	1.877	1.345	30.01%	73.91%
5	Semivariance	3.679	2.868	2.095	51.16%	84.78%
6	Loss Penalty	33.427	17.908	7.918	60.84%	76.09%
7	Skewness standard	0.149	0.097	-3.437	1.46%	54.35%
8	Skewness robust	0.066	0.019	-0.919	4.74%	50.00%
9	Maximum Drawdown	0.073	0.054	0.032	46.98%	84.78%

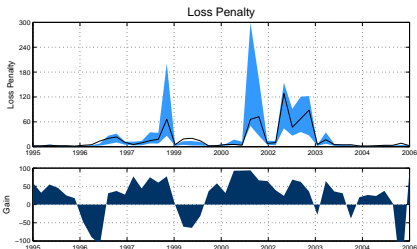
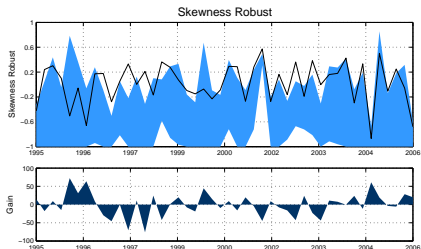
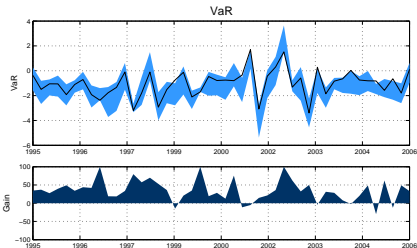
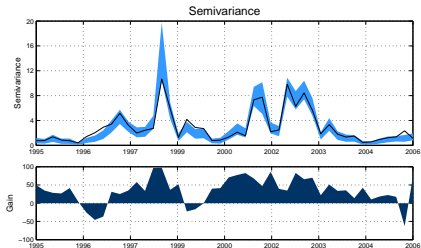
- $Gain = Mean_{\#Optimizations} \left(\frac{risk(optimized) - risk(benchmark)}{risk(optimal) - risk(benchmark)} \right)$
- $HitRate = \frac{\#\{risk(optimized) > risk(benchmark)\}}{\#optimizations}$
- except for skewness the asymmetric measures allow for minimizing downside risk

Minimizing Risk: Evolution over Time

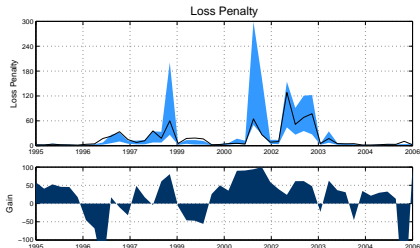
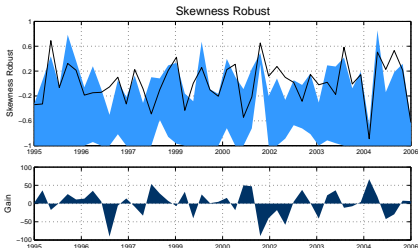
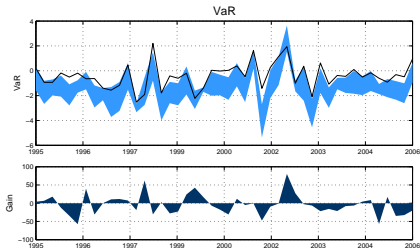
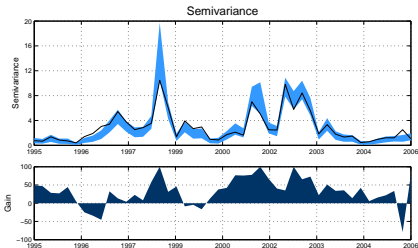


- Upper: Spread between benchmark and optimal portfolio, and risk of optimized portfolio
- Lower: Gain

Some Downside Risk Measures: Perfect Foresight



Some Downside Risk Measures: Forecasted Returns

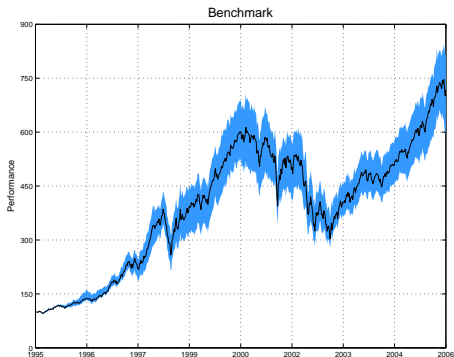


Minimizing Downside Risk: Forecasted Returns

Strategy	Risk			Gain average	Hit Rate
	<i>benchmark</i>	<i>optimized</i>	<i>optimal</i>		
1 Volatility	2.538	2.275	1.854	38.37%	86.97%
2 VaR	-0.460	-0.407	-2.165	-3.09%	39.13%
3 Conditional VaR	1.702	1.635	0.553	5.83%	58.70%
4 Semideviation	2.106	1.908	1.345	26.03%	58.70%
5 Semivariance	3.679	2.862	2.095	51.55%	84.78%
6 Loss Penalty	33.427	17.700	7.918	61.65%	69.57%
7 Skewness standard	0.149	0.029	-3.437	3.35%	58.70%
8 Skewness robust	0.066	0.016	-0.919	5.01%	63.04%
9 Maximum Drawdown	0.073	0.069	0.032	10.67%	67.39%

- The degree of risk reduction is muted as compared to perfect foresight
- Setback most severe for VaR and CVaR
- 4 measures still have significant hit rates: semivariance, semideviation, loss penalty, maximum drawdown

Performance of Downside Risk Strategies



- Spread between the highest and lowest cumulated return of downside strategies
- Solid line represents benchmark performance

- Downside risk control not necessarily at the cost of return
- Downside risk control is not hedging market risk in our setting

Mutual Tracking Errors of Strategies

	BM	Vola	VaR	CVaR	Semi- Dev.	Var.	Loss Penalty	Skewness Std.	Rob.	MDD
BM	0.000	4.972	6.852	7.275	4.742	5.151	5.161	5.133	5.014	6.338
Vola	5.191	2.201	7.889	6.852	2.796	2.109	3.303	7.038	6.062	5.827
VaR	5.200	6.833	8.042	6.102	7.701	8.180	8.141	8.670	7.886	7.240
CVaR	5.743	3.947	6.779	8.127	6.935	6.952	7.101	8.746	7.697	5.603
Semidev.	5.098	2.829	6.764	3.896	3.131	3.278	4.388	6.734	6.042	6.297
Semivar.	5.362	2.180	7.214	4.453	3.310	2.112	2.341	6.418	6.133	5.749
Loss Penalty	5.473	3.318	7.331	5.212	4.422	2.279	1.918	6.209	6.156	5.663
Skewness Std.	5.378	7.124	7.567	7.179	6.895	6.331	6.194	3.866	7.115	7.715
Skewness Rob.	5.018	6.165	7.343	6.587	6.115	6.201	6.510	6.852	6.377	7.349
MDD	5.982	5.540	7.461	5.618	5.919	5.474	5.441	7.104	7.288	6.332

- diagonale: perfect foresight versus forecasted returns
- ex-post TE slightly higher than 5%, CVaR and MDD are most different to the BM
- cluster: volatility, semideviation, and semivariance
- TE between the two skewness strategies amounts to 6.85%

Mutual Downside Risk of Strategies

	BM	Vola	VaR	CVaR	Semi- Dev.	Semi- Var.	Loss Penalty	Skewness Std.	Skewness Rob.	MDD
<i>Panel C: Forecasted</i>										
Vola	2.85	2.54	2.90	2.67	2.59	2.52	2.55	2.85	2.81	2.65
VaR	-0.53	-0.57	-0.41	-0.50	-0.43	-0.55	-0.50	-0.52	-0.51	-0.53
CVaR	1.71	1.53	1.81	1.63	1.57	1.50	1.51	1.73	1.73	1.60
Semidev.	1.04	0.94	1.08	1.00	0.96	0.94	0.95	1.06	1.05	1.00
Semivar.	8.13	6.42	8.38	7.13	6.70	6.32	6.51	8.13	7.90	6.99
Loss Penalty	291.1	70.1	689.7	113.6	107.1	60.8	53.8	196.5	211.3	74.3
Skewness Std.	0.00	0.11	0.08	0.03	0.01	0.11	0.10	0.01	-0.06	0.10
Skewness Rob.	0.07	0.13	0.02	0.09	0.03	0.14	0.07	0.07	0.05	0.03
MDD	50.48	50.03	46.99	46.87	49.80	48.08	45.89	49.63	48.93	49.17
Mean Ranking	6.89	4.56	7.78	5.89	4.89	3.44	3.67	7.00	6.00	4.89

- Semivariance is successful along 4 risk dimensions: Vola, Semideviation, Semivariance, and MDD
- Loss Penalty ranks second
- given the close relationship to semivariance volatility unsurprisingly ranks third

Conclusion

- Asymmetric risk measures capture downside risks
- Instead of volatility one should therefore stick to asymmetric measures when it comes to empirical tests
- Except for skewness all of the tested measures allow for a substantial reduction of portfolio downside risk given perfect foresight of return
- In absence of perfect foresight VaR and CVaR strategies falter while loss penalty, semivariance and maximum drawdown strategies' performance persists
- Two important prerequisites:
 - the asymmetric measure needs to be predictable
 - asset returns need to exhibit significant asymmetry