

Realized Semibetas:
Disentangling “good” and “bad” downside risks

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Motivation

- ▶ Do expected returns vary significantly with the degree of diversification offered during market downturns/upturns?
 - ▶ We measure this using **semibetas**, defined below.
- ▶ Can we better explain the **cross-section of expected returns** by drawing on the information in semibetas? Compare with:
 - ▶ Mean-semivariance framework of Hogan and Warren (1972, 1974, *JFQA*)
 - ▶ Downside beta framework of Ang Chen and Xing (2006, *RFS*)
 - ▶ Co-skewness and co-kurtosis pricing, see Kraus and Litzenberger (1976, *JF*), Harvey and Siddique (2000, *JF*), Langlois (2019, *JFE*) and others.

Semicovariances and Semibetas

- ▶ Ignoring the mean, the covariance between a stock (r) and the market (f) is:

$$\begin{aligned} \text{Cov}(r, f) &= E[r \cdot f] \\ &= E[r \cdot f \cdot \mathbf{1}(r < 0, f < 0)] + E[r \cdot f \cdot \mathbf{1}(r > 0, f > 0)] \\ &\quad + E[r \cdot f \cdot \mathbf{1}(r < 0, f > 0)] + E[r \cdot f \cdot \mathbf{1}(r > 0, f < 0)] \end{aligned}$$

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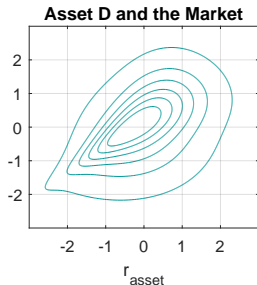
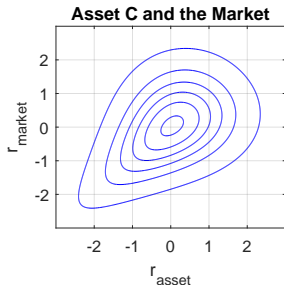
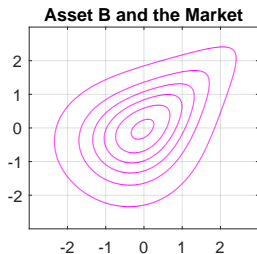
- ▶ Scaling by $\text{Var}(f)$, we obtain an exact decomposition of beta into **semibetas**:

$$\begin{aligned} \beta &= \frac{\text{Cov}(r, f)}{\text{Var}(f)} \\ &= \frac{\mathcal{N} + \mathcal{P} + \mathcal{M}^+ + \mathcal{M}^-}{\text{Var}(f)} \\ &\equiv \beta^{\mathcal{N}} + \beta^{\mathcal{P}} - \beta^{\mathcal{M}^+} - \beta^{\mathcal{M}^-} \end{aligned}$$

- ▶ As $\mathcal{M} < 0$, we switch the sign when defining $\beta^{\mathcal{M}}$ to ease interpretation.

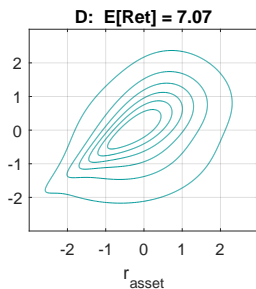
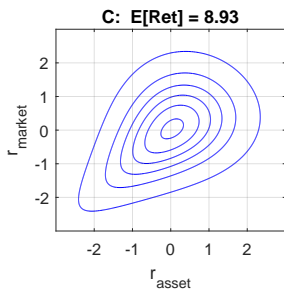
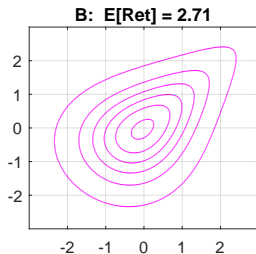
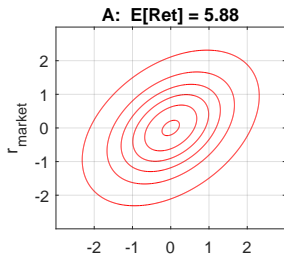
Which asset would you prefer?

A: Gaussian, B: Correlated booms, C: Correlated crashes, D: Less correlated crashes



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A: Gaussian, B: Correlated booms, C: Correlated crashes, D: Less correlated crashes



Outline

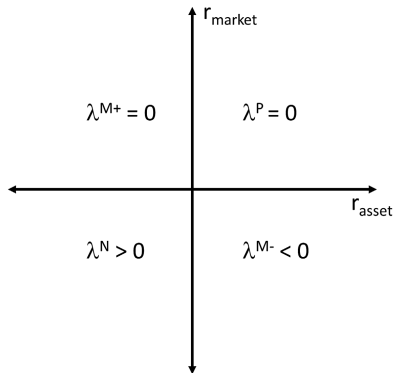
- ▶ Mean-semivariance pricing
- ▶ Realized semibetas
- ▶ Cross-sectional pricing
- ▶ Betting on semibetas

Mean-Semivariance Pricing

- ▶ Investors only care about **downside return variation**
 - ▶ Only the covariation with negative aggregate market returns should be priced
 - ▶ Standard CAPM and single beta security market line too simplistic
 - ▶ Traces back to early work by Roy (1952, *Ecta*), Markowitz (1959, *book*), Mao (1970, *JF*), Hogan and Warren (1972, 1974, *JFQA*), Bawa and Lindenberg (1977, *JFE*), Fishburn (1977, *AER*)
 - ▶ May also be justified by prospect theory and the notion of loss aversions proposed by Kahneman and Tversky (1979, *Ecta*)

Mean-semivariance pricing

- ▶ Hogan and Warren, 1974, *JFQA*:



- ▶ Naturally suggests measuring risk using semibetas:

$$\beta \equiv \frac{\text{Cov}(r, f)}{\text{Var}(f)} = \beta^N + \beta^P - \beta^{M^+} - \beta^{M^-}$$

Realized semibetas

- ▶ True semibetas are **latent**:

$$\beta_{t,i}^{\mathcal{N}} \equiv \frac{\mathcal{N}_{t,i}}{\mathcal{V}_t} \quad \beta_{t,i}^{\mathcal{P}} \equiv \frac{\mathcal{P}_{t,i}}{\mathcal{V}_t} \quad \beta_{t,i}^{\mathcal{M}^+} \equiv \frac{-\mathcal{M}_{t,i}^+}{\mathcal{V}_t} \quad \beta_{t,i}^{\mathcal{M}^-} \equiv \frac{-\mathcal{M}_{t,i}^-}{\mathcal{V}_t}$$

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- ▶ But using results in Bollerslev, Li, Patton and Quaedvlieg (2020, *Econometrica*), these can be consistently ($m \rightarrow \infty$) estimated by **realized semibetas**:

$$\begin{aligned} \widehat{\beta}_{t,i}^{\mathcal{N}} &\equiv \frac{\sum_{k=1}^m r_{t,k,i}^- f_{t,k}^-}{\sum_{k=1}^m f_{t,k}^2} & \widehat{\beta}_{t,i}^{\mathcal{P}} &\equiv \frac{\sum_{k=1}^m r_{t,k,i}^+ f_{t,k}^+}{\sum_{k=1}^m f_{t,k}^2} \\ \widehat{\beta}_{t,i}^{\mathcal{M}^-} &\equiv \frac{-\sum_{k=1}^m r_{t,k,i}^+ f_{t,k}^-}{\sum_{k=1}^m f_{t,k}^2} & \widehat{\beta}_{t,i}^{\mathcal{M}^+} &\equiv \frac{-\sum_{k=1}^m r_{t,k,i}^- f_{t,k}^+}{\sum_{k=1}^m f_{t,k}^2} \end{aligned}$$

where $r_{t,k,i}^+ \equiv \max(r_{t,k,i}, 0)$ and $r_{t,k,i}^- \equiv \min(r_{t,k,i}, 0)$

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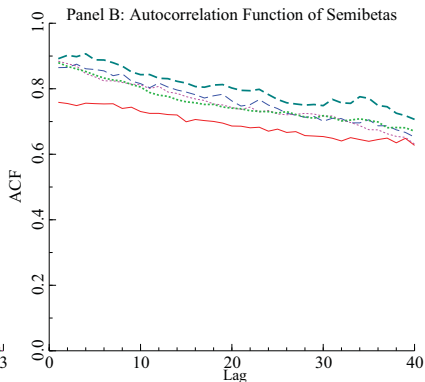
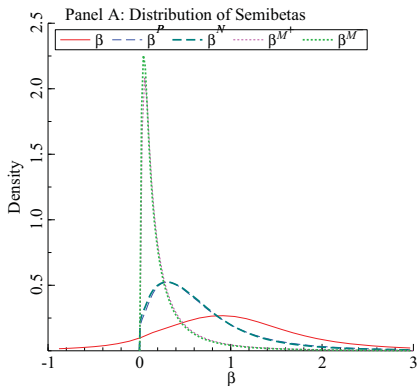
- ▶ By construction $\widehat{\beta}_{t,i} = \widehat{\beta}_{t,i}^{\mathcal{N}} + \widehat{\beta}_{t,i}^{\mathcal{P}} - \widehat{\beta}_{t,i}^{\mathcal{M}^+} - \widehat{\beta}_{t,i}^{\mathcal{M}^-}$, for all m

Realized semibetas

- ▶ High-frequency intraday data from TAQ
- ▶ S&P 500 constituent stocks, 1993–2014
 - ▶ $T = 5,541$ trading days
 - ▶ 1,049 unique securities ($\bar{N} = 722$)
 - ▶ 15-minute returns excluding overnight (so $m = 26$)
- ▶ Total daily returns based on CRSP to account for overnight and dividends

Realized semibetas

- ▶ Unconditional distributions and autocorrelations:

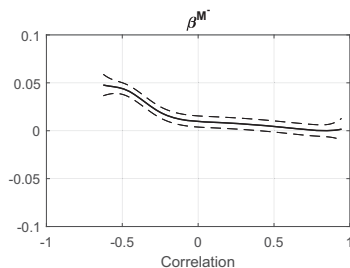
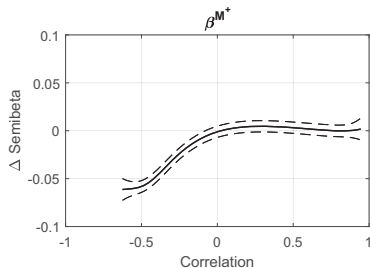
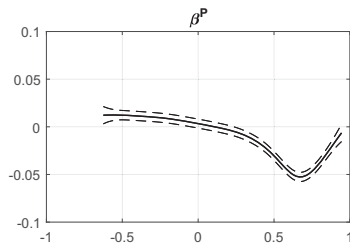
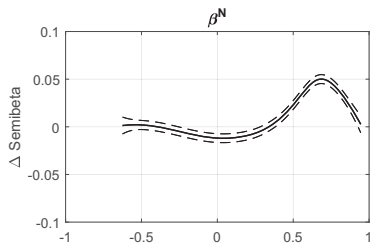


- ▶ Distributions of two (dis)concordant semibetas almost indistinguishable
- ▶ Semibetas more strongly autocorrelated than conventional betas

Realized semibetas

► Average deviations from implied Gaussian values as a function of ρ

► β^N and β^{M^-} typically higher; β^P and β^{M^+} typically lower



Cross-sectional pricing

- ▶ Standard Fama-MacBeth approach, with cross-sectional regressions:

$$r_{t+1,i} = \lambda_{0,t+1} + \lambda_{t+1}^{\mathcal{N}} \hat{\beta}_{t,i}^{\mathcal{N}} + \lambda_{t+1}^{\mathcal{P}} \hat{\beta}_{t,i}^{\mathcal{P}} + \lambda_{t+1}^{\mathcal{M}^+} \hat{\beta}_{t,i}^{\mathcal{M}^+} + \lambda_{t+1}^{\mathcal{M}^-} \hat{\beta}_{t,i}^{\mathcal{M}^-} + \epsilon_{t+1,i}$$

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- ▶ Risk premium estimates, $\hat{\lambda}^j = \frac{1}{T-1} \sum_{t=2}^T \hat{\lambda}_t^j$:

β	$\beta^{\mathcal{N}}$	$\beta^{\mathcal{P}}$	$\beta^{\mathcal{M}^+}$	$\beta^{\mathcal{M}^-}$	R^2
4.58					2.70
3.04					
	22.54	-1.58	-4.29	-8.48	5.43
	5.62	-0.52	-0.86	-2.02	

- ▶ Stocks with higher $\beta^{\mathcal{N}}$ earn significantly **higher** risk premium
- ▶ Stocks with higher $\beta^{\mathcal{M}^-}$ earn significantly **lower** risk premium

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- ▶ Stocks with higher $\beta^{\mathcal{M}^-}$ earn significantly **lower** risk premium
- ▶ H_t^{CAPM} : $\lambda_t^{\mathcal{N}} = \lambda_t^{\mathcal{P}} = -\lambda_t^{\mathcal{M}^+} = -\lambda_t^{\mathcal{M}^-}$
 - ▶ Rejected at 5% level for 68% of 5,541 days in sample

Short-sales constraints

- ▶ In a frictionless market without any short-sales constraints:

$$H_t^{SYM} : \lambda_t^{\mathcal{N}} = -\lambda_t^{\mathcal{M}^-} \cap \lambda_t^{\mathcal{P}} = -\lambda_t^{\mathcal{M}^+}$$

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- ▶ Rejected at 5% level for 58% of 5,541 days in sample

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- ▶ Rejected at 5% level for 58% of 5,541 days in sample
- ▶ D'Avolio (2002, *JFE*) and Henderson, Jostova and Philipov (2019, *wp*) indicate that almost *all* of the S&P 500 stocks can easily and cheaply be borrowed
 - ▶ Diffs in risk premia cannot be attributed to “hard” short-selling constraints

Short-sales constraints

- ▶ In a frictionless market without any short-sales constraints:

$$H_t^{SYM} : \lambda_t^N = -\lambda_t^{M^-} \cap \lambda_t^P = -\lambda_t^{M^+}$$

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- ▶ D'Avolio (2002, *JFE*) and Henderson, Jostova and Philipov (2019, *wp*) indicate that almost *all* of the S&P 500 stocks can easily and cheaply be borrowed
 - ▶ Diffs in risk premia cannot be attributed to “hard” short-selling constraints
- ▶ Legal and institutional constraints impede many individual and institutional investors from short-selling
 - ▶ Creates limits to arbitrage as in Pontiff (1996, *QJE*) and Schleifer and Vishny (1997, *JF*), and related **arbitrage risks** (Hong and Sraer, 16, *JF*)

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 - ▶ Turnover (Harris and Raviv, 93, *RFS*; Blume, Easley and O'Hara, 94, *JF*)

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 - ▶ Turnover (Harris and Raviv, 93, *RFS*; Blume, Easley and O'Hara, 94, *JF*)
- ▶ We apply a linear rotation to our model to make it easier to observe whether there is a violation of the “symmetric” pricing restrictions:

$$r_i = \lambda_0 + \lambda^{\mathcal{N}} (\beta_i^{\mathcal{N}} - \beta_i^{\mathcal{M}^-}) + \lambda^{\mathcal{P}} (\beta_i^{\mathcal{P}} - \beta_i^{\mathcal{M}^+}) + \delta^{\mathcal{M}^+} \beta_i^{\mathcal{M}^+} + \delta^{\mathcal{M}^-} \beta_i^{\mathcal{M}^-} + \epsilon_{t+1}$$

- ▶ If the symmetric pricing restrictions hold, then $\delta^{\mathcal{M}^+} = \delta^{\mathcal{M}^-} = 0$

Arbitrage risk and semibeta pricing

	β^N	β^P	δ^{M-}	δ^{M+}	β^N	β^P	δ^{M-}	δ^{M+}	R^2
Panel A: Full-Sample Estimates									
	22.54	-1.58	12.16	-5.81					5.43
	5.62	-0.52	2.22	-0.93					
Panel B: Sorting on Arbitrage Risk									
	Below Median				Above Median				
IVOL	22.29	-0.10	3.99	-3.41	17.43	-4.63	34.81	10.19	6.62
	5.45	-0.03	0.72	-0.52	4.03	-1.52	4.73	1.33	
TO	22.23	-12.64	-2.10	-9.54	19.57	1.69	8.06	-5.13	6.93
	5.42	-4.40	-0.34	-1.37	4.66	0.50	2.27	-0.76	

- ▶ Differences in semibeta risk premia are driven by stocks that are more difficult to short (higher IVOL) or are more difficult to value (higher turnover).

Including control variables

- ▶ Including some “standard,” lower frequency, control variables:

β^N	β^P	β^{M^+}	β^{M^-}	ME	BM	MOM	REV	IVOL	ILLIQ	R^2
22.54	-1.58	-4.29	-8.48							5.43
5.62	-0.52	-0.86	-2.02							
22.47	-5.67	-2.90	-12.20	-2.23	-1.77	0.11				8.23
5.75	-2.02	-0.65	-3.14	-3.83	-1.95	3.47				
20.36	-2.91	1.68	-6.15	-7.42	-1.65	0.09	-0.55	-3.07	-4.88	10.32
5.44	-1.08	0.41	-1.68	-7.71	-1.87	2.55	-5.82	-3.56	-6.42	

- ▶ Average R^2 s increase
- ▶ Still the case that:
 - ▶ Stocks with higher β^N earn **higher** returns
 - ▶ Stocks with higher β^{M^-} earn **lower** returns

Comparison with up- and down-side betas

- Up- and down-side betas (Ang, Chen and Xing, 2006, *RFS*):

β^N	β^P	β^{M^+}	β^{M^-}	β^+	β^-	R^2
22.54	-1.58	-4.29	-8.48			5.43
5.62	-0.52	-0.86	-2.02			
				-1.17	6.88	3.70
				-1.11	5.54	
17.31	-8.10	-12.66	-3.86	-2.40	7.90	6.61
3.86	-0.13	0.03	-1.79	-1.67	0.82	

where $\hat{\beta}_{t,i}^+ \equiv \frac{\sum_{k=1}^m r_{t,k,i} f_{t,k}^+}{\sum_{k=1}^m (f_{t,k}^+)^2}$ and $\hat{\beta}_{t,i}^- \equiv \frac{\sum_{k=1}^m r_{t,k,i} f_{t,k}^-}{\sum_{k=1}^m (f_{t,k}^-)^2}$

- β^N and β^{M^-} remain significant in combined regression

Comparison with up- and down-side betas

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$\beta^{\mathcal{N}}$	$\beta^{\mathcal{P}}$	$\beta^{\mathcal{M}^+}$	$\beta^{\mathcal{M}^-}$	β^+	β^-	R^2
22.54	-1.58	-4.29	-8.48			5.43
5.62	-0.52	-0.86	-2.02			
				-1.17	6.88	3.70
				-1.11	5.54	
17.31	-8.10	-12.66	-3.86	-2.40	7.90	6.61
3.86	-0.13	0.03	-1.79	-1.67	0.82	

Note: $\hat{\beta}_{t,i}^+ = (\hat{\beta}_{t,i}^{\mathcal{P}} - \hat{\beta}_{t,i}^{\mathcal{M}^+}) \frac{\sum_{k=1}^m f_{t,k}^2}{\sum_{k=1}^m (f_{t,k}^+)^2}$ and $\hat{\beta}_{t,i}^- = (\hat{\beta}_{t,i}^{\mathcal{N}} - \hat{\beta}_{t,i}^{\mathcal{M}^-}) \frac{\sum_{k=1}^m f_{t,k}^2}{\sum_{k=1}^m (f_{t,k}^-)^2}$

- Pricing implications coincide if $H_t^{SYM} : \lambda_t^{\mathcal{P}} = -\lambda_t^{\mathcal{M}^+} \cap \lambda_t^{\mathcal{N}} = -\lambda_t^{\mathcal{M}^-}$
 - Rejected at 5% level for 58% of 5,541 days in sample

Comparison with co-skewness and co-kurtosis

- ▶ Coskewness and Cokurtosis (Harvey and Siddique, 2000, *JF*):

$\beta^{\mathcal{N}}$	$\beta^{\mathcal{P}}$	$\beta^{\mathcal{M}^+}$	$\beta^{\mathcal{M}^-}$	CSK	CKT	R^2
22.54	-1.58	-4.29	-8.48			5.43
5.62	-0.52	-0.86	-2.02			
				-4.40	0.81	1.52
				-1.55	0.76	
30.92	-3.79	-3.89	-16.33	10.09	-3.59	6.26
6.20	-1.12	-0.76	-3.69	2.66	-3.22	

$$CSK_{t,i} \propto \frac{1}{m} \sum_{k=1}^m (r_{t,k,i} - \bar{r}_{t,i})(f_{t,k} - \bar{f}_t)^2$$

$$CKT_{t,i} \propto \frac{1}{m} \sum_{k=1}^m (r_{t,k,i} - \bar{r}_{t,i})(f_{t,k} - \bar{f}_t)^3$$

- ▶ $\beta^{\mathcal{N}}$ and $\beta^{\mathcal{M}^-}$ remain significant; additional information about the tails captured by CSK and CKT

High frequency data; longer investment horizons

β	$\beta^{\mathcal{N}}$	$\beta^{\mathcal{P}}$	$\beta^{\mathcal{M}^+}$	$\beta^{\mathcal{M}^-}$	ME	BM	MOM	REV	IVOL	ILLIQ	R^2
Weekly:											
4.69											2.37
3.72											
	14.58	0.96	5.20	-13.80							5.07
	5.90	0.50	1.87	-3.71							
	10.85	-0.52	3.67	-13.58	-6.36	-1.94	0.08	-0.35	-1.06	-3.63	10.83
	5.92	-0.38	1.76	-5.03	-7.50	-2.30	2.54	-4.39	-1.40	-5.77	
Monthly:											
2.93											1.90
2.71											
	8.70	1.67	3.51	-3.36							4.45
	4.39	1.15	1.55	-1.30							
	4.42	-0.57	-0.65	-6.41	-4.89	-1.86	0.08	-0.25	0.98	-2.27	10.79
	3.55	-0.58	-0.44	-3.34	-7.06	-2.58	2.64	-4.15	1.56	-4.60	

- ▶ $\beta^{\mathcal{N}}$ and $\beta^{\mathcal{M}^-}$ remain significant at **weekly** and **monthly** horizons; estimated premiums decrease with horizon as predictability diminishes

Lower frequency data

- ▶ Now we consider **monthly semibetas** based on **daily** data
 - ▶ All CRSP common stocks (with codes 10 and 11 and price > \$5)
 - ▶ $\bar{N} = 390$ stocks per period (unbalanced panel)
 - ▶ Longer January 1963 to December 2017 sample
 - ▶ $T = 660$ months
 - ▶ Fewer “high frequency” observations per period
 - ▶ $m \approx 21$ observations

Lower frequency data

- ▶ Consider **monthly semibetas** and **monthly** Fama-MacBeth regressions:

β	$\beta^{\mathcal{N}}$	$\beta^{\mathcal{P}}$	$\beta^{\mathcal{M}^+}$	$\beta^{\mathcal{M}^-}$	ME	BM	MOM	REV	IVOL	ILLIQ	R^2
4.10											2.36
3.77											
	10.43	1.40	4.15	-6.42							5.22
	4.46	0.87	1.15	-2.03							
	8.66	-0.66	5.60	-14.09	-2.55	-0.47	0.06				10.70
	3.56	-0.43	1.42	-3.72	-4.93	-0.40	2.14				
	6.59	-1.90	6.33	-15.59	-2.08	-0.75	0.07	-0.12	-1.60	2.40	13.38
	2.85	-1.06	1.50	-3.82	-4.22	-0.66	2.61	-1.97	-1.48	2.44	

- ▶ Premiums for $\beta^{\mathcal{N}}$ and $\beta^{\mathcal{M}^-}$ remain significant

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	4.46	0.87	1.15	-2.03							
	8.66	-0.66	5.60	-14.09	-2.55	-0.47	0.06				10.70
	3.56	-0.43	1.42	-3.72	-4.93	-0.40	2.14				
	6.59	-1.90	6.33	-15.59	-2.08	-0.75	0.07	-0.12	-1.60	2.40	13.38
	2.85	-1.06	1.50	-3.82	-4.22	-0.66	2.61	-1.97	-1.48	2.44	

- ▶ Premiums for $\beta^{\mathcal{N}}$ and $\beta^{\mathcal{M}^-}$ remain significant

- ▶ $H_t^{CAPM} : \lambda_t^{\mathcal{N}} = \lambda_t^{\mathcal{P}} = -\lambda_t^{\mathcal{M}^+} = -\lambda_t^{\mathcal{M}^-}$

- ▶ Rejected at 5% level for 45% of 659 months in sample

Comparison with other measures

- ▶ **Monthly** semibetas and other measures:

β^N	β^P	β^{M^+}	β^{M^-}	β^+	β^-	CSK	CKT	R^2
10.43	1.40	4.15	-6.42					5.22
4.46	0.87	1.15	-2.03					
				1.06	3.16			3.42
				1.61	3.74			
12.37	2.90	2.41	-7.56	-6.64	-0.90			5.57
4.97	1.03	1.20	-2.48	-0.95	-0.28			
						5.00	1.98	1.69
						2.81	2.57	
18.11	-2.27	2.87	-12.09			12.10	-2.80	6.49
4.98	-1.04	0.81	-3.40			4.26	-3.43	

- ▶ Premiums for β^N and β^{M^-} always significant

Comparison with other measures

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β^N	β^P	β^{M^+}	β^{M^-}	β^+	β^-	CSK	CKT	R^2
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12.37	2.90	2.41	-7.56	-6.64	-0.90			5.57
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4.98	-1.04	0.81	-3.40			4.26	-3.43	

- ▶ Premiums for β^N and β^{M^-} always significant
- ▶ $H_t^{SYM} : \lambda_t^N = -\lambda_t^{M^-} \cap \lambda_t^P = -\lambda_t^{M^+}$
 - ▶ Rejected at 5% level for 34% of 659 months in sample

Betting on semibetas

- ▶ We next consider long/short portfolios based on **semibetas**
- ▶ Conventional CAPM security market line “too flat”
 - ▶ Friend and Blume (1970, *AER*), Black, Jensen and Scholes (1972, *Book*), Kandel (1984, *JFE*), Shanken (1985, *JFE*), Fama and French (1992, *JF*), Fama and French (2006, *JF*)
- ▶ Recent literature on betting against beta (BAB)
 - ▶ Frazzini and Pedersen (2014, *JFE*), Cederburg and O’Doherty (2016, *JF*), Bali, Brown, Murray and Tang (2017, *JFQA*), Novy-Marx and Velikov (2018, *wp*), Schneider, Wagner and Zechner (2020, *JF*)

Betting on semibetas

- ▶ We will look at:
 - ▶ Betting on β (standard case)
 - ▶ Betting on $\beta^{\mathcal{N}}$
 - ▶ Betting against $\beta^{\mathcal{M}^-}$
 - ▶ Betting **on** $\beta^{\mathcal{N}}$ and **against** $\beta^{\mathcal{M}^-}$ (“Semi β ”)

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- ▶ Value-weighted long/short positions in high/low quintile of S&P 500 stocks
 - ▶ Avoids small and difficult to short micro-cap stocks, and the use of rank-weighted portfolios
- ▶ Sharpe ratios, alphas and factor loadings:
 - ▶ Four-factor Fama-French-Carhart model: MKT, SML, HML, MOM
 - ▶ Five-factor Fama-French model: MKT, SML, HML, RMW (profitability: robust minus weak), CMA (investment: conservative minus aggressive)

Betting on semibetas

► **Daily** rebalancing:

	β		Semi β		$\beta^{\mathcal{N}}$		$\beta^{\mathcal{M}^-}$	
Avg ret	4.98		9.76		10.98		7.66	
Std dev	16.57		9.30		16.89		8.00	
Sharpe	0.30		1.05		0.65		0.96	
α	2.19	4.17	8.35	9.68	8.05	10.62	7.76	7.88
	0.87	1.77	6.32	7.38	3.36	4.58	4.17	4.19
β_{MKT}	0.59	0.51	0.30	0.25	0.61	0.52	-0.02	-0.03
	67.61	55.78	64.70	47.90	73.15	56.89	-2.34	-3.44
β_{SMB}	0.30	0.12	0.30	0.21	0.40	0.22	0.20	0.20
	18.10	7.36	33.91	22.61	25.00	13.49	16.08	14.98
β_{HML}	-0.02	0.18	-0.01	0.11	-0.08	0.13	0.05	0.08
	-1.24	10.58	-1.61	11.10	-4.75	7.57	3.82	6.08
β_{MOM}	-0.24		-0.14		-0.22		-0.07	
	-19.53		-22.46		-19.01		-7.31	
β_{RMW}		-0.50		-0.28		-0.53		-0.04
		-22.15		-22.56		-23.70		-2.28
β_{CMA}		-0.35		-0.28		-0.44		-0.13
		-13.21		-19.09		-16.74		-5.95
R^2	58.15	60.26	55.92	59.55	60.59	64.38	6.72	7.42

Betting on semibetas

► **Weekly** rebalancing:

	β		Semi β		β^N		β^{M^-}	
Avg ret	1.18		5.47		7.76		2.40	
Std dev	15.30		8.64		15.50		7.52	
Sharpe	0.08		0.63		0.50		0.32	
α	-1.11	0.71	4.37	5.66	5.54	7.71	2.40	2.84
	-0.46	0.31	3.45	4.35	2.52	3.46	1.46	1.72
β_{MKT}	0.52	0.44	0.25	0.20	0.51	0.43	-0.02	-0.04
	60.65	48.93	55.60	38.70	66.56	49.31	-3.40	-5.54
β_{SMB}	0.30	0.13	0.30	0.22	0.39	0.23	0.21	0.21
	18.43	7.83	35.58	23.57	26.64	14.41	19.12	17.76
β_{HML}	-0.08	0.08	-0.06	0.02	-0.12	0.05	0.00	-0.01
	-4.68	4.54	-7.02	2.26	-8.05	2.96	-0.10	-0.55
β_{MOM}	-0.20		-0.11		-0.20		-0.01	
	-17.14		-17.93		-19.12		-1.85	
β_{RMW}		-0.47		-0.26		-0.48		-0.05
		-21.38		-21.18		-22.57		-3.05
β_{CMA}		-0.26		-0.22		-0.35		-0.08
		-9.80		-14.89		-14.07		-4.48
R^2	51.93	54.10	47.31	51.03	53.19	56.27	7.21	8.44

Betting on semibetas

► **Monthly** rebalancing:

	β		Semi β		β^N		β^{M^-}	
Avg ret	-0.64		4.04		3.10		4.33	
Std dev	14.43		8.39		14.72		7.18	
Sharpe	-0.04		0.48		0.21		0.60	
α	-2.48	-1.34	2.96	4.22	1.17	2.85	4.09	4.94
	-1.17	-0.64	2.30	3.24	0.55	1.34	2.83	3.42
β_{MKT}	0.46	0.41	0.23	0.19	0.46	0.39	0.00	-0.02
	62.38	50.99	51.68	36.54	62.55	47.54	0.90	-4.01
β_{SMB}	0.26	0.11	0.31	0.23	0.38	0.23	0.24	0.23
	18.67	7.27	36.10	24.59	26.93	14.85	25.18	22.62
β_{HML}	-0.08	0.05	-0.06	0.01	-0.08	0.08	-0.04	-0.06
	-5.79	3.22	-6.46	1.06	-5.48	5.15	-3.64	-5.76
β_{MOM}	-0.21		-0.10		-0.22		0.02	
	-20.65		-16.43		-21.44		2.22	
β_{RMW}		-0.42		-0.26		-0.45		-0.08
		-21.18		-21.10		-21.89		-5.86
β_{CMA}		-0.17		-0.19		-0.30		-0.08
		-7.03		-12.88		-12.57		-4.91
R^2	49.21	49.52	46.76	50.29	50.83	52.42	11.71	13.32

Betting on semibetas

- ▶ What about **transaction costs**?
- ▶ Many practical procedures to help mitigate the cost of trading
 - ▶ Bertsiman and Lo (1998, *JFinMkts*), Engle and Ferstenberg (2007, *JPorMgmt*), Obizhaeva and Wang (2013, *JFinMkts*)
 - ▶ Dependent on specific trading strategies and settings

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 - ▶ Dependent on specific trading strategies and settings
- ▶ Trading only partially towards the “target” (Garleanu and Pedersen, 2013, *JF*):

$$\omega_t = \lambda\omega_{t-1} + (1 - \lambda)\omega_t^{Target}, \quad \lambda = 0.95$$

- ▶ We will assume transaction costs proportional to the turnover of the long and short positions
- ▶ We will consider round-trip costs of 0.5% and 1.0% (Novy-Marx and Velikov, 2016, *RFS*)

Betting on semibetas: T-costs and partial adjustment

- **Alphas** remain economically large and statistically significant:

T-cost	0%		0%		0.5%		1.0%	
Adjustment	Full		Partial		Partial		Partial	
Avg ret	4.04		4.62		4.32		4.02	
Std dev	8.39		7.77		7.77		7.77	
Sharpe	0.48		0.59		0.56		0.52	
α	2.96	4.22	3.09	5.31	2.79	5.01	2.49	4.71
	2.30	3.24	3.05	5.33	2.76	5.03	2.46	4.72
β_{MKT}	0.23	0.19	0.25	0.17	0.25	0.17	0.25	0.17
	51.68	36.54	69.25	44.87	69.23	44.85	69.18	44.81
β_{SMB}	0.31	0.23	0.27	0.20	0.27	0.20	0.27	0.20
	36.10	24.59	40.61	27.95	40.58	27.93	40.54	27.89
β_{HML}	-0.06	0.01	-0.13	-0.11	-0.13	-0.11	-0.13	-0.11
	-6.46	1.06	-18.72	-14.51	-18.69	-14.49	-18.66	-14.46
β_{MOM}	-0.10		0.01		0.01		0.01	
	-16.43		2.59		2.59		2.59	
β_{RMW}		-0.26		-0.26		-0.26		-0.26
		-21.10		-27.01		-27.01		-27.00
β_{CMA}		-0.19		-0.20		-0.20		-0.20
		-12.88		-18.16		-18.15		-18.13
R^2	46.76	50.29	52.20	58.90	52.19	58.90	52.16	58.87

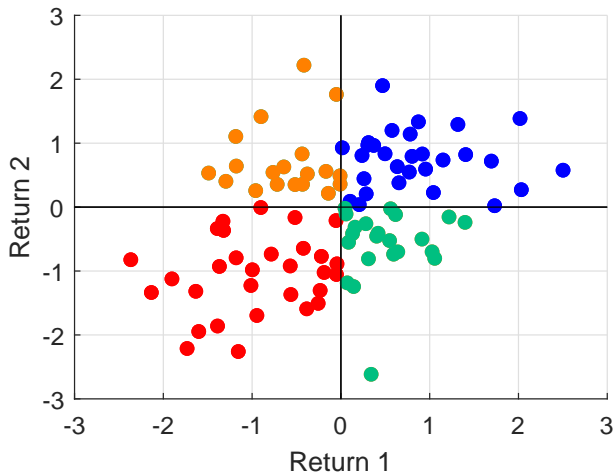
Summary

- ▶ We propose the use of **semibetas** to gain richer information about the diversification benefits offered by various assets
 - ▶ Semibetas offer an exact decomposition: $\beta = \beta^{\mathcal{N}} + \beta^{\mathcal{P}} - \beta^{\mathcal{M}+} - \beta^{\mathcal{M}-}$
 - ▶ We estimate **realized semibetas** using high frequency data
- ▶ We show that semibetas are better able to explain the cross-section of expected returns than existing alternatives
 - ▶ As expected, much better than the CAPM, and also significantly better than up- and down-side betas
 - ▶ We find that the risk premium associated with $\beta^{\mathcal{N}}$ is around 23%, with $\beta^{\mathcal{M}-}$ around -9%, and is not different from zero for the other semibetas.
- ▶ We find that a long-short portfolio based on semibetas generates large and significant alphas
 - ▶ Sharpe ratios are double that of the market; alphas are 8-9%
 - ▶ Don't bet on or against beta, bet **on and against** the "right" **semibetas**

Appendix

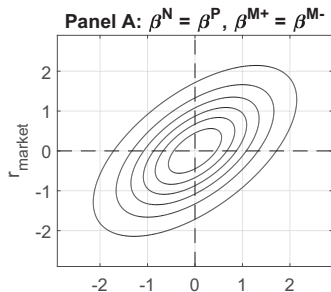
Computing semibetas

There's nothing to it (once you've got the data ready)



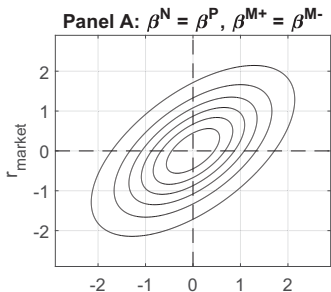
Mean-semivariance pricing

- ▶ Jointly normally distributed market and asset return:



Mean-semivariance pricing

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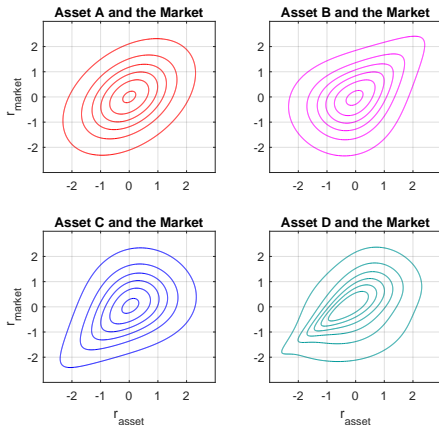
- ▶ No new information over traditional CAPM β :

$$\beta^{\mathcal{N}} = \beta^{\mathcal{P}} = \frac{1}{2\pi} \left(\sqrt{\frac{\sigma_r^2}{\sigma_f^2} - \beta^2} + \beta \arccos \left(-\frac{\sigma_f}{\sigma_r} \beta \right) \right)$$

$$\beta^{\mathcal{M}^-} = \beta^{\mathcal{M}^+} = \frac{1}{2\pi} \left(\sqrt{\frac{\sigma_r^2}{\sigma_f^2} - \beta^2} - \beta \arccos \left(\frac{\sigma_f}{\sigma_r} \beta \right) \right)$$

Mean-semivariance pricing

- ▶ Jointly non-normally distributed market and asset returns:



- ▶ Identical CAPM betas, but very different semibetas and expected returns:

- ▶ $E(r^B) < E(r^A) < E(r^D) < E(r^C)$

Betting on the competition

	$\beta^- - \beta^+$		β^-		β^+		CKT - CSK		CSK		CKT	
Avg ret	2.96		7.53		-3.61		-0.16		-1.21		0.54	
Std dev	5.45		15.56		14.67		6.30		7.70		9.55	
Sharpe	0.54		0.48		-0.25		-0.03		-0.16		0.06	
α	2.56	2.93	4.78	6.80	-1.67	-2.94	-0.60	-0.03	-1.44	-1.27	-0.11	0.86
	2.10	2.42	1.90	2.76	-0.69	-1.28	-0.49	-0.02	-0.86	-0.76	-0.06	0.47
β_{MKT}	0.04	0.03	0.55	0.47	-0.47	-0.41	0.13	0.11	0.02	0.01	0.24	0.21
	9.32	6.38	61.82	48.63	-54.89	-45.37	30.25	23.49	2.60	1.94	36.63	28.96
β_{SMB}	0.01	0.01	0.27	0.12	-0.25	-0.10	-0.02	-0.06	0.02	0.03	-0.06	-0.15
	1.01	1.43	16.31	7.10	-15.85	-6.02	-2.58	-7.10	1.87	2.11	-4.96	-11.23
β_{HML}	-0.02	-0.04	-0.04	0.13	-0.01	-0.20	-0.09	-0.06	-0.03	-0.06	-0.16	-0.06
	-2.53	-4.09	-2.07	7.05	-0.38	-11.98	-11.15	-6.98	-2.45	-4.85	-12.20	-4.72
β_{MOM}	0.04		-0.17		0.25		-0.01		0.04		-0.06	
	6.29		-14.09		21.19		-2.51		4.43		-7.07	
β_{RMW}	0.00		-0.43		0.42		-0.10		0.00		-0.21	
	-0.30		-18.06		19.01		-9.15		0.26		-12.22	
β_{CMA}	-0.01		-0.32		0.30		-0.04		0.03		-0.10	
	-1.04		-11.57		11.42		-2.70		1.69		-5.05	
R^2	2.67	1.93	52.93	55.64	50.17	50.32	17.43	18.91	0.89	0.58	27.91	30.36

► Lower Sharpe ratios, and smaller, less significant alphas

Betting on semibetas: Conditional alphas

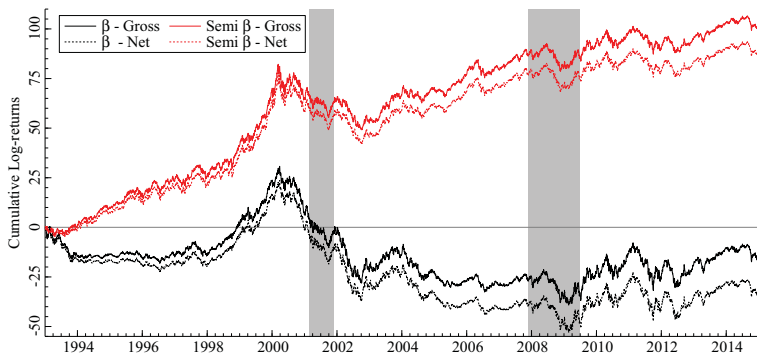
- ▶ Time varying conditional betas may bias unconditional alphas
 - ▶ Jagannathan and Wang (1996, *JF*), Lewellen and Nagel (2006, *JFE*)
- ▶ Conditional alphas following Cederburgh and O'Doherty (2016, *JF*):

	FFC4				FF5			
	β	Semi β	β^N	β^{M^-}	β	Semi β	β^N	β^{M^-}
I	3.01	6.35	7.48	2.00	3.25	8.07	9.61	4.83
	1.02	2.65	2.42	1.05	0.85	3.08	2.60	1.95
II	3.87	7.72	10.09	4.25	3.18	7.59	9.32	4.91
	1.27	3.56	3.38	1.86	1.01	4.15	2.90	2.36
III	3.58	6.75	8.57	4.21	2.87	7.57	8.88	4.91
	1.24	3.34	3.28	2.18	0.92	3.77	3.23	2.13

- ▶ Same qualitative findings for all sets of instruments I, II and III
- ▶ Conditional alphas for semi β strategy remain large and significant

Betting on Semibetas

- ▶ **Timing** of returns:



- ▶ Semibeta strategy generally performs well
- ▶ Consistent with “betting against beta” (Frazzini and Pedersen, 2014, *JFE*), conventional beta strategy generally performs poorly