

Discussion of:
A Unified Model of Distress Risk Puzzles
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Outline

- Paper's Goal: lay out theory that explains the “puzzles” raised in:
 - Campbell, Hilscher, and Szilagyi (2008, 2011)
 - Friewald, Wagner, and Zechner (2014)
- Discussion Outline:
 - Review CHS evidence.
 - I'll omit any discussion of FWZ.
 - This paper's model and model implications
 - Empirical tests.

Campbell, Hilscher, and Szilagyi (2008, 2011)

- Response to Fama and French (1996), who argued that value premium might be a distress premium.
- Found that firms with **high** distress risk deliver abnormally **low** returns.
 - Calculate an EDF based on Shumway (2001) bankruptcy model.
 - Builds on Dichev (1998), who found a negative relation between the Altman (1968) and Ohlson (1980) measures and future returns.
 - Also consistent with Garlappi, Shu, and Yan (2008), who use the Moodys-KMV measure of distance-to-default, with similar results (see also Garlappi and Yan, 2011).
- These results have continued to hold in a 1981–2010 sample (Anginer and Yıldızhan, 2017)

Campbell, Hilscher, and Szilagyi (2008, 2011)

- CHS (2011) abstract states:

... We find that distressed stocks have high stock return volatility and high market betas and that they tend to underperform safe stocks by more at times of high market volatility and risk aversion. However, investors in distressed stocks have not been rewarded for bearing these risks. Instead, distressed stocks have had very low returns, both relative to the market and after adjusting for their high risk.

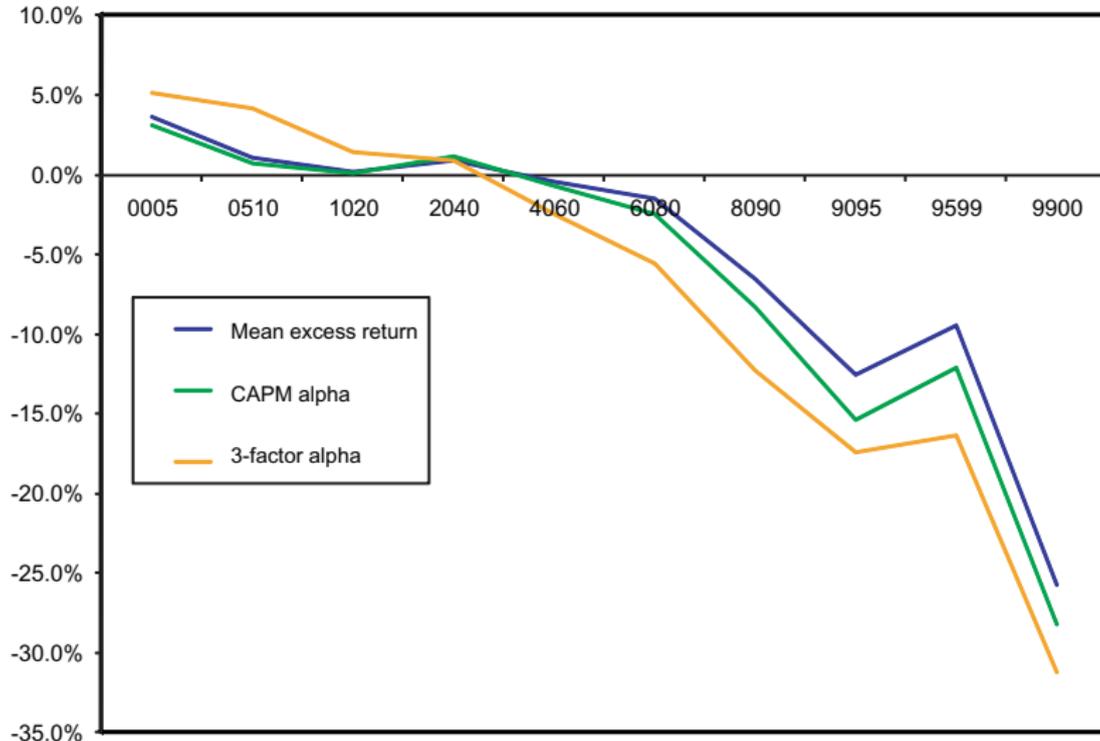
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Campbell, Hilscher, and Szilagyi (2011)

Mean excess returns (%/year) (net of the market return):



Campbell, Hilscher, and Szilagyi (2008)

PD_t^{IP}-sorted portfolio results:

	Excess ret.	CAPM alpha	Three-factor alpha	Four-factor alpha	MKT	SMB	HML
Low	0.608** (2.01)	0.166 (0.99)	0.433*** (2.86)	0.096 (0.72)	0.879*** (23.63)	0.109** (2.17)	-0.462*** (8.05)
2	0.569** (2.55)	0.095 (1.51)	0.090 (1.42)	0.022 (0.36)	0.898*** (54.84)	0.110*** (4.66)	-0.141*** (-2.72)
3	0.534** (2.51)	0.092 (1.48)	0.034 (0.55)	0.043 (0.69)	1.033*** (60.30)	0.116*** (4.66)	-0.071*** (-2.75)
4	0.553* (1.92)	-0.059 (-0.70)	-0.168** (-2.06)	-0.075 (-0.96)	1.170*** (54.03)	0.249*** (7.90)	0.069** (2.13)
5	0.496 (1.54)	-0.175 (-1.64)	-0.279*** (-2.73)	-0.167* (-1.69)	1.252*** (45.85)	0.367*** (9.24)	-0.021 (-0.84)
6	0.385* (1.70)	-0.112 (-0.79)	-0.157 (-1.19)	0.056 (0.47)	1.254*** (36.52)	0.389*** (9.37)	0.013 (0.32)
7	0.408* (1.68)	-0.089 (-0.65)	-0.224* (-1.77)	-0.043 (-0.37)	1.245*** (41.65)	0.458*** (10.52)	0.031 (0.68)
8	0.308 (0.92)	-0.371*** (-2.73)	-0.476*** (-3.99)	-0.280*** (-2.61)	1.171*** (36.10)	0.358*** (7.57)	0.027 (0.55)
9	0.200 (0.44)	-0.596** (-2.17)	-0.653*** (-2.67)	-0.375*** (-2.85)	1.425*** (23.64)	0.920*** (12.44)	0.053 (0.58)
High	-0.576 (1.19)	-1.216*** (3.87)	-1.509*** (5.29)	-0.736*** (3.24)	1.511*** (21.63)	0.923*** (9.82)	0.430*** (3.99)
High-Low	-1.184** (2.34)	-1.382*** (2.96)	-1.942*** (4.68)	-0.832*** (2.64)	0.632*** (5.69)	0.814*** (10.96)	0.892*** (6.25)

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Friewald, Wagner, and Zechner (2014)

- Stocks with a **high** credit risk premium, defined as $\mathbb{E}_t^{\mathbb{Q}}[S_{t+\tau}^T] - \mathbb{E}_t^{\mathbb{P}}[S_{t+\tau}^T]$,
 - Here, the expected spreads under the \mathbb{P} - and \mathbb{Q} -measures are based on the CDS term structure.
 - earn **high** average returns and CAPM α s.

What is going on?

There are (at least) two possibilities:

- ① The market is completely wrong about which firms are going to default (inattention?)
 - The market is consistently (negatively) surprised when the “distressed” (high $PD_t^{\mathbb{P}}$) firms default, and positively surprised when the “healthy” (low $PD_t^{\mathbb{P}}$) firms don't!
 - This leads to lower realized (not expected) returns for the high $PD_t^{\mathbb{P}}$ firms.
- ② The market's expectations are correct, but previous empirical tests haven't properly measured risk.
 - “Distressed” (high unconditional $PD^{\mathbb{P}}$) are actually “safe”, in that they earn comparatively good returns in bad (high marginal utility) times.
 - The “healthy” (low $PD^{\mathbb{P}}$) firms earn low returns (default) in the worst (highest marginal utility) economic times.

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 - The “healthy” (low $PD^{\mathbb{P}}$) firms earn low returns (default) in the worst (highest marginal utility) economic times.

This paper's model

This paper presents a very nice dynamic model consistent with (2):

- “Healthy” (unconditionally low $PD^{\mathbb{P}}$ firms) issue debt in expansions.
 - When the economy weakens, and $\mathbb{E}_t[R_{m,t+1}]$ increases, the high debt levels of the “healthy” firms cause their market-betas to rise to a level higher than the “distressed” firms.
- Conversely, “distressed” (unconditionally high $PD_t^{\mathbb{P}}$) take on less debt in expansions
 - Thus, their betas are lower in the worst (highest marginal utility) economic times.

Conditional CAPM logic:

From Jagannathan and Wang (2007) and Lewellen and Nagel (2006):

- The conditional CAPM states:

$$\mathbb{E}_t[r_{i,t+t}] = \beta_{i,t} \mathbb{E}_t[r_{t+1}^m]$$

- Take unconditional expectations of each side, and note that $\mathbb{E}[xy] = \mathbb{E}[x]\mathbb{E}[y] + cov(x, y)$:

$$\mathbb{E}[r_{i,t+t}] = \mathbb{E}[\beta_{i,t}]\mathbb{E}[r_{t+1}^m] + \underbrace{cov(\beta_{i,t}, \mathbb{E}_t[r_{t+1}^m])}_{\approx \hat{\alpha}_i^{CAPM,u}}$$

- Over this sample, $\hat{\mathbb{E}}[r_t^m] \approx 7\%/yr$, and for the high-PD portfolio, $\hat{\mathbb{E}}[r_{i,t}] \approx 0$, $\hat{\beta}^u = 1.59$, $\hat{\alpha}_i^{CAPM,u} \approx -11\%/yr$
- $cov(\beta_{i,t}, \mathbb{E}_t[r_{t+1}^m]) = \sigma_{\beta_i} \sigma_{E[r^m]} \rho_{\beta, E[r^m]}$, so if $\sigma_{E[r^m]} = 7\%/yr$, then

$$\sigma_{\beta_i} \geq \frac{|cov(\beta_{i,t}, \mathbb{E}_t[r_{t+1}^m])|}{\sigma_{E[r^m]}} = \frac{11}{7} = 1.57$$

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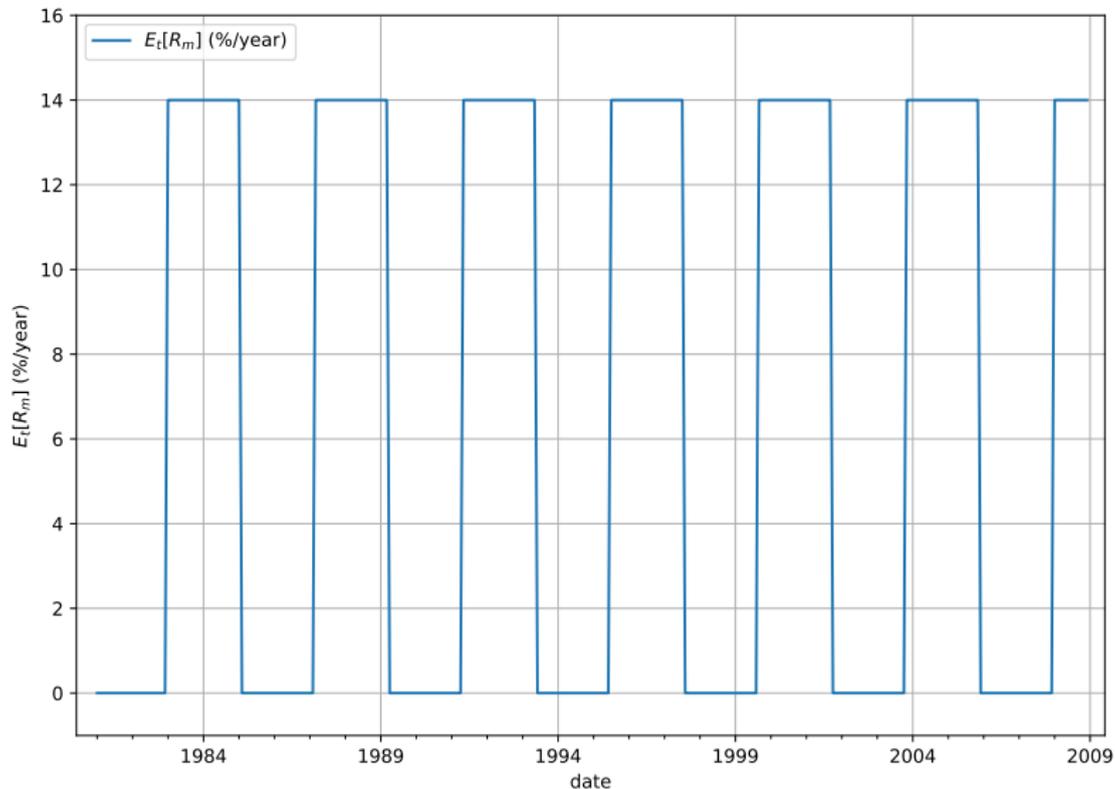
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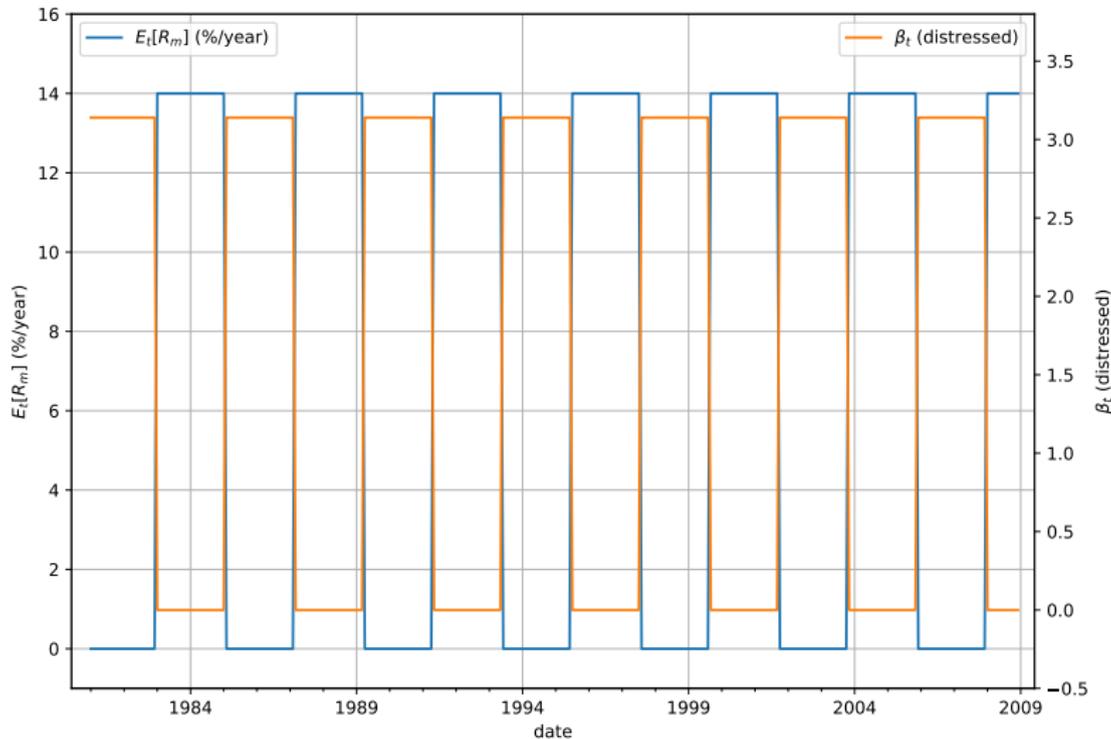
Simulated $\mathbb{E}_t[R_m]$ s and β_{tS}

Expected Market Returns:



Simulated $\mathbb{E}_t[R_m]$ s and β_{ts}

... and Distressed Firm Betas:



Estimation of $cov(\beta_{i,t}, \mathbb{E}_t[r_{t+1}^m])$

Table 7, Panel B:

Panel B. Model-Implied Excess Returns and Alphas											
	L(ow)	2	3	4	5	6	7	8	9	H(igh)	H-L
$\beta_{i,t}^E$	0.99	0.96	0.99	1.04	1.08	1.14	1.22	1.30	1.36	1.29	0.30
$E[\beta_{i,t}^E]E[r_t^m]$	6.89	6.66	6.88	7.24	7.50	7.90	8.47	9.07	9.49	9.00	2.11
$cov(\beta_{i,t}^E, r_t^m)$	-0.62	0.02	0.04	-0.60	0.18	-1.90	-1.09	-3.20	-3.95	-5.96	-6.58
$r_{i,t}^{ex}$	7.50	6.67	6.92	6.64	7.68	6.00	7.37	5.87	5.54	3.04	-4.47
α^u	1.19	0.52	0.41	-0.49	0.12	-2.29	-1.72	-4.28	-5.30	-7.96	-9.15

- Since $\mathbb{E}_t[r_{t+1}^m]$ is unobservable, the authors instead estimate:

$$cov(\hat{\beta}_{i,t+1}, r_{t+1}^m) = cov(\beta_{i,t}, \mathbb{E}_t[r_{t+1}^m]) + cov(\hat{\beta}_{i,t+1}, \epsilon_{t+1}^m)$$

using monthly returns and β_t s, estimated using daily data.

- Lewellen and Nagel (2006, p. 307) note this will be biased if these "... firms covary more [with r_m] in down markets."
 - Note that this would lead to a negative estimated covariance.
- Thus, LN "... report this first estimate primarily as a benchmark rather than as a perfect estimate." (p. 307)

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Estimating $\mathbb{E}_t[R_m]$ and $\hat{\beta}_t$

- Let's estimate this β variation using the other approach suggested by LN. Specifically, run time-series regressions:

$$\tilde{r}_{i,t} = \alpha_i(1 + \gamma_{0,i}Z_{t-1}) + (\beta_{0,i} + \beta_{1,i} \cdot Z_{t-1})\tilde{r}_{m,t} + \tilde{\epsilon}_{i,t}$$

for a set of instruments Z that forecast market returns.

- Then:

$$\hat{\beta}_{i,t-1} = \hat{\beta}_{0,i} + \hat{\beta}_{1,i} \cdot Z_{t-1}$$

- We'll use as instruments:
 - NBER Recession indicator
 - $-1 \times$ Past 2-year market return
 - $-1 \times$ Index of Leading Economic Indicators (LEI)
 - VIX
 - TERM Spread (10Y-3M)
- These variables are all high in periods of distress, so expected market returns should be positively related to these variables.*

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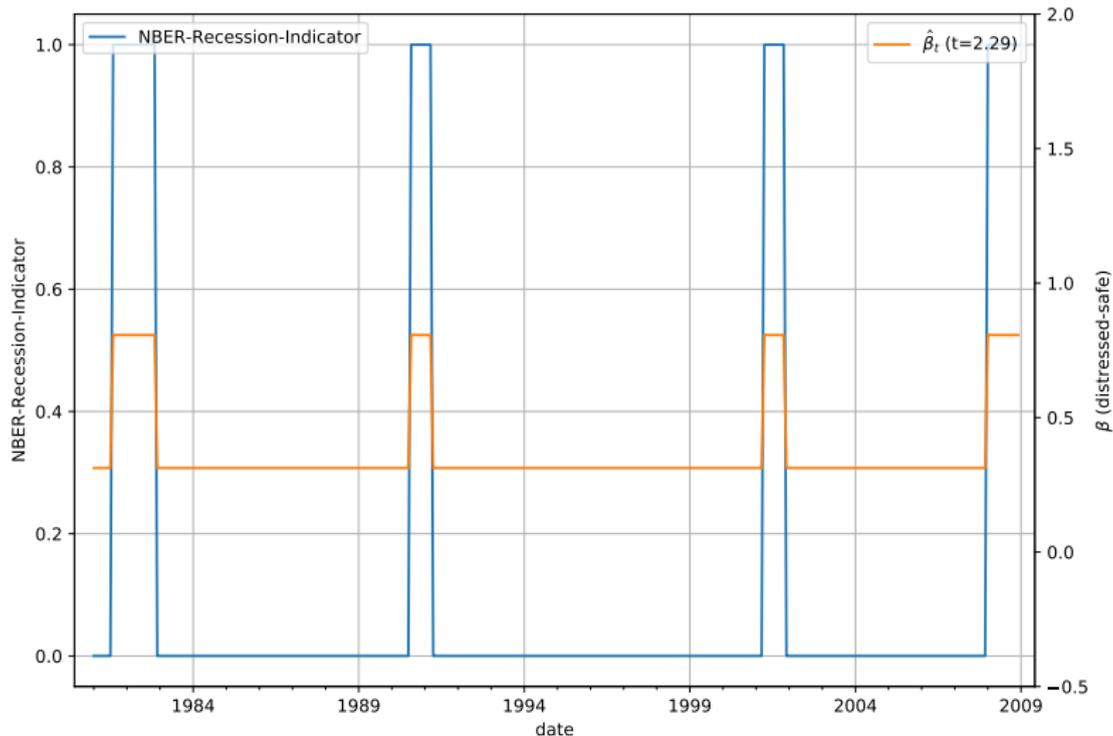
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$$\hat{\beta}_{i,t-1} = \hat{\beta}_{0,i} + \hat{\beta}_{1,i} \cdot Z_{t-1}$$

- We'll use as instruments:

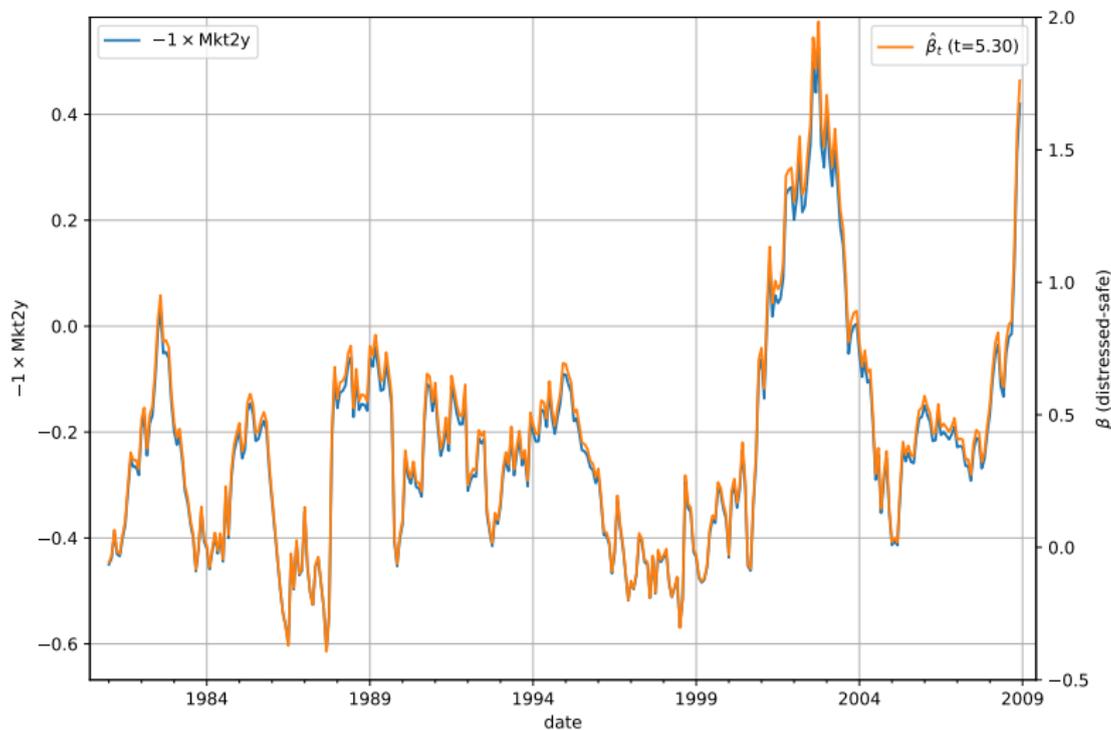
- 1 NBER Recession indicator
- 2 $-1 \times$ Past 2-year market return
- 3 $-1 \times$ Index of Leading Economic Indicators (LEI)
- 4 VIX
- 5 TERM Spread (10Y-3M)

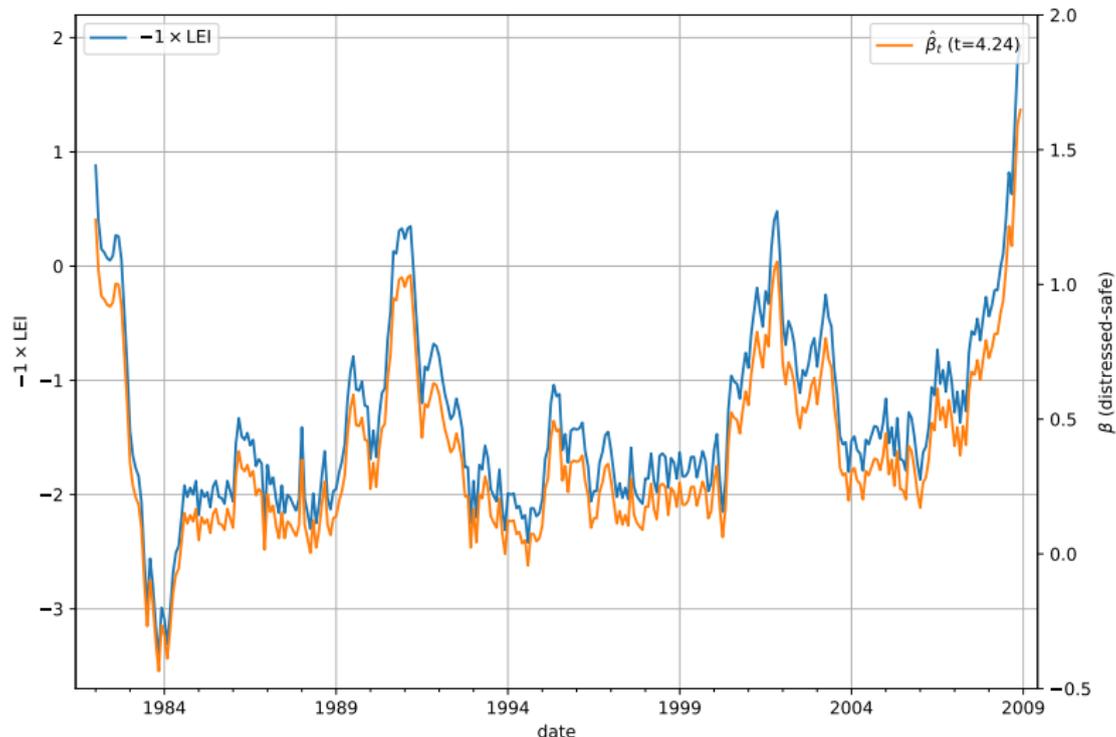
- These variables are all high in periods of distress, so expected market returns should be positively related to these variables.*

Time variation in $\hat{\beta}_{LS}$ $Z_{t-1} =$ NBER Recession indicator

Time variation in $\hat{\beta}_{LS}$

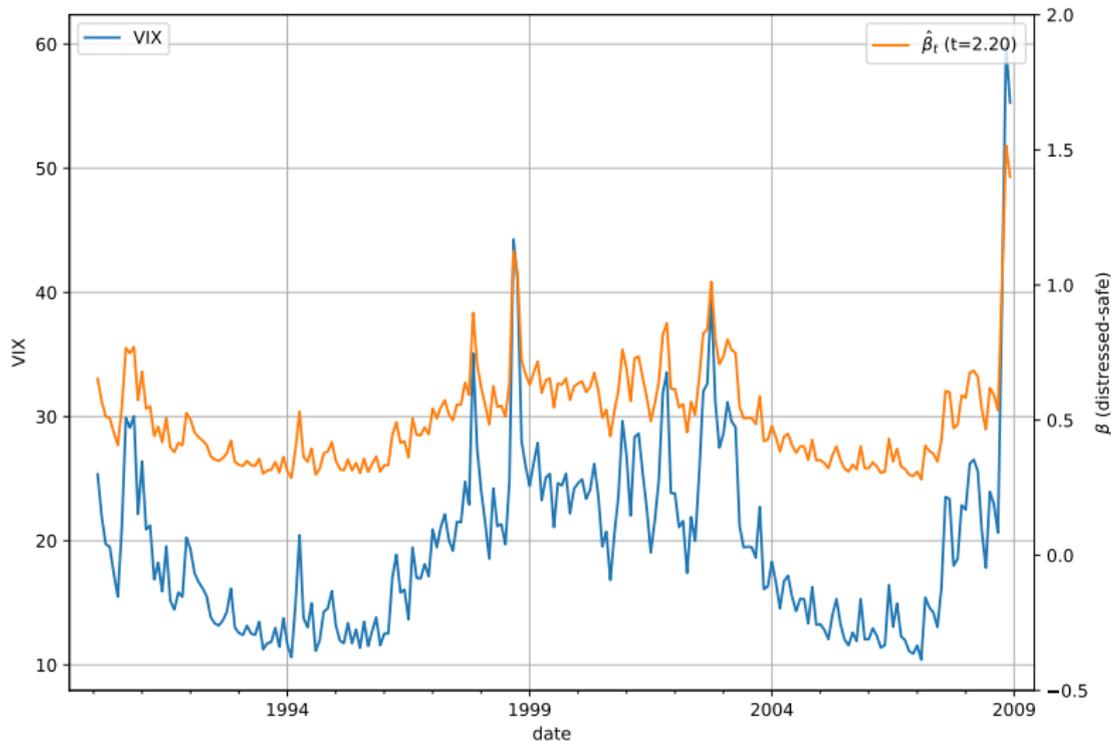
$Z_{t-1} = -1 \times 2\text{-year log Market Return}$

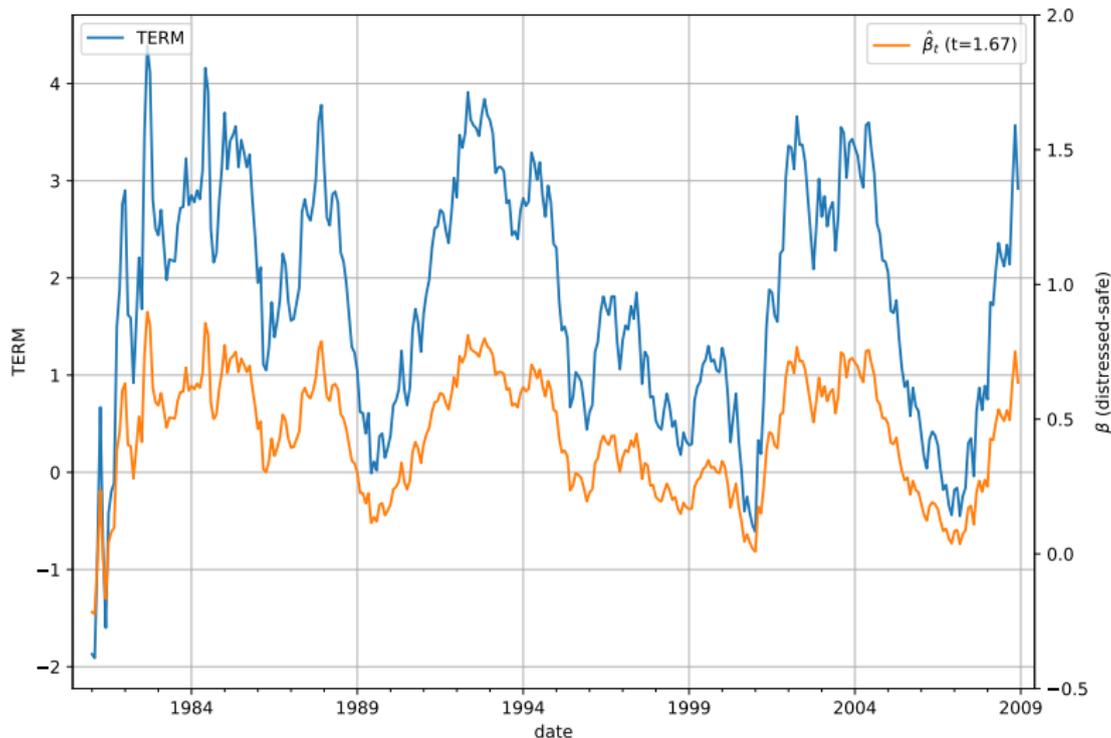


Time variation in $\hat{\beta}_{LS}$ $Z_{t-1} = -1 \times \text{Index of Leading Economic Indicators (LEI)}$ 

Time variation in $\hat{\beta}_{LS}$

$$Z_{t-1} = \text{VIX}$$



Time variation in $\hat{\beta}_{LS}$ $Z_{t-1} = \text{TERM Spread (10Y-3M)}$ 

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