

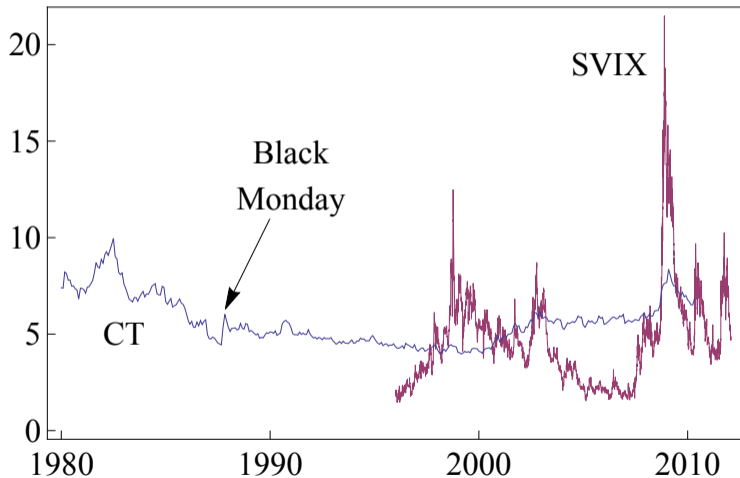
# Volatility, Valuation Ratios, and Bubbles: An Empirical Measure of Market Sentiment

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# Two views of the equity premium

Based on valuation ratios (Campbell and Thompson, *RFS*, 2008) and on index option prices (Martin, *QJE*, 2017)



# Outline

- Very roughly, think of  $D/P$  as revealing  $\mathbb{E}R - \mathbb{E}G$ , and interest rates and option prices as revealing  $\mathbb{E}R$ ; then the gap between the two reveals  $\mathbb{E}G$

Specifically, today:

- 1 Relate dividend yields to expected returns and dividend growth using a twist on the Campbell–Shiller methodology
- 2 Introduce a bound on expected returns based on interest rates and option prices
- 3 Derive a bound on expected dividend growth by playing off (1) against (2)

# Campbell–Shiller decomposition (1)

Notation: log dividend yield  $dp_t = \log(D_t/P_t)$ ; log return  $r_{t+1}$ ; log dividend growth  $g_{t+1}$

- Campbell and Shiller (1988) famously showed that, up to a linearization,

$$dp_t = \frac{k}{1 - \rho} + \sum_{i=0}^{\infty} \rho^i \mathbb{E}_t [r_{t+1+i} - g_{t+1+i}] \quad \text{where } \rho \approx 0.97$$

- These are expected **log returns**, not expected returns
- Low expected log returns may be consistent with *high* expected returns if returns are volatile, right-skewed, or fat-tailed
- All three plausibly true in late 1990s, so the distinction between log returns and simple returns matters

## Campbell–Shiller decomposition (2)

$$dp_t = \frac{k}{1-\rho} + \sum_{i=0}^{\infty} \rho^i (r_{t+1+i} - g_{t+1+i}) - \underbrace{\frac{\rho(1-\rho)}{2} \sum_{i=0}^{\infty} \rho^i (dp_{t+1+i} - \overline{dp})^2}_{\text{second order term} \approx -0.145 \text{ in late '90s}}$$

- In the late '90s  $dp_t$  was 2.2 sd below its mean (using CRSP data 1947–2017)
- Ignoring the **second order term** is equivalent to understating  $\mathbb{E}_t r_{t+1+i} - g_{t+1+i}$  by 14.5 pp for one year, 3.1 pp for five years, or 1.0 pp for 20 years
  - ▶ In long sample, 1871–2015, numbers are even bigger: 25.3 pp for one year, 5.5 pp for five years, 1.8 pp for 20 years, or 1.0 pp for ever
- Thus the CS decomposition may “cry bubble” too soon

## An alternative approach (1)

- Campbell and Shiller loglinearize

$$r_{t+1} - g_{t+1} = dp_t + \log \left( 1 + e^{-dp_{t+1}} \right)$$

- We start, instead, from

$$r_{t+1} - g_{t+1} = y_t + \log \left( 1 - e^{-y_t} \right) - \log \left( 1 - e^{-y_{t+1}} \right)$$

where

$$y_t = \log \left( 1 + \frac{D_t}{P_t} \right)$$

- $y_t$ , unlike  $dp_t$ , is in natural units: if  $D_t/P_t = 2\%$  then  $y_t = 1.98\%$  whereas  $dp_t = -3.91$

## An alternative approach (2)

### Result

We have the loglinearization

$$y_t = (1 - \rho) \sum_{i=0}^{\infty} \rho^i (r_{t+1+i} - g_{t+1+i})$$

where  $\rho = e^{-\bar{y}} \approx 0.97$ .

On average, this relationship holds **exactly**—no linearization needed:

$$\bar{y} = \bar{r} - \bar{g}$$

## An alternative approach (3)

- We have already seen that the Campbell–Shiller approximation may lead one to conclude too quickly that the market is bubbly, as

$$dp_t < -\frac{k}{1-\rho} + \sum_{i=0}^{\infty} \rho^i \mathbb{E}_t (r_{t+1+i} - g_{t+1+i})$$

- Our variant is a **conservative** diagnostic for bubbles. If  $y_t$  is *far from its mean* then

$$y_t \geq (1-\rho) \sum_{i=0}^{\infty} \rho^i \mathbb{E}_t (r_{t+1+i} - g_{t+1+i})$$

- ▶ *Far from its mean*:  $\mathbb{E}_t [(y_{t+i} - \bar{y})^2] \leq (y_t - \bar{y})^2$  for all  $i \geq 0$
- ▶ In AR(1) case, “*far*” means “one standard deviation”



## Information in valuation ratios (1)

- If  $y_t$  follows an AR(1) with autocorrelation  $\phi_y$ ,

$$\mathbb{E}_t(r_{t+1} - g_{t+1}) = \text{constant} + \frac{1 - \rho\phi_y}{1 - \rho}y_t$$

- In the unit root case  $\phi_y = 1$ , we have  $y_t = \mathbb{E}_t(r_{t+1} - g_{t+1})$
- So we use  $y_t$  to forecast  $r_{t+1} - g_{t+1}$
- We estimate the regression freely, but results are almost identical if we estimate  $\rho$  and  $\phi_y$  from time series, then use the formula above
- AR(1) is not critical: key is that we have a forecast of  $\mathbb{E}_t y_{t+1}$ . Will show AR(k) later

## Information in valuation ratios (2)

RHS <sub>t</sub>	LHS <sub>t+1</sub>	$\hat{a}_0$	s.e.	$\hat{a}_1$	s.e.	R <sup>2</sup>
$y_t$	$r_{t+1} - g_{t+1}$	-0.067	[0.049]	3.415	[1.317]	7.73%
	$r_{t+1}$	-0.018	[0.050]	3.713	[1.215]	10.51%
	$-g_{t+1}$	-0.049	[0.028]	-0.298	[0.812]	0.32%
$dp_t$	$r_{t+1} - g_{t+1}$	0.417	[0.146]	0.107	[0.042]	7.58%
	$r_{t+1}$	0.500	[0.138]	0.114	[0.041]	9.92%
	$-g_{t+1}$	-0.083	[0.085]	-0.007	[0.024]	0.19%

Table: S&P 500, annual data, 1947–2017, dividends reinvested monthly at CRSP 30-day T-bill rate. Hansen–Hodrick standard errors.

- Relative importance of  $r$  and  $g$  is sample specific:  $g$  more important in long sample. But coefficient estimates for  $r - g$  are stable

## Information in options (1)

- We start from an identity

$$\mathbb{E}_t r_{t+1} = \frac{1}{R_{f,t+1}} \mathbb{E}_t^* (R_{t+1} r_{t+1}) - \text{cov}_t (M_{t+1} R_{t+1}, r_{t+1})$$

- $M_{t+1}$  is an SDF. Risk-neutral  $\mathbb{E}_t^*$  satisfies  $\frac{1}{R_{f,t+1}} \mathbb{E}_t^* (X_{t+1}) = \mathbb{E}_t (M_{t+1} X_{t+1})$
- We assume that  $\text{cov}_t (M_{t+1} R_{t+1}, r_{t+1}) \leq 0$ 
  - ▶ Similar to the negative correlation condition of Martin (2017)
  - ▶ Loosely, requires that investors are sufficiently risk-averse wrt  $R_{t+1}$
  - ▶ Holds in Campbell–Cochrane (1999), Bansal–Yaron (2004), Barro (2006), Wachter (2013), Bansal et al. (2014), Campbell et al. (2016), ...
- We then have

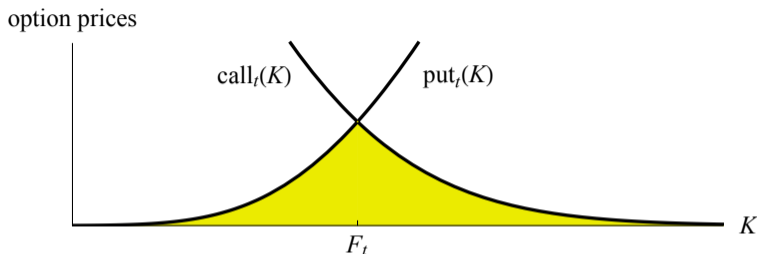
$$\mathbb{E}_t r_{t+1} \geq \frac{1}{R_{f,t+1}} \mathbb{E}_t^* (R_{t+1} r_{t+1})$$

## Information in options (2)

$$\mathbb{E}_t r_{t+1} \geq \frac{1}{R_{f,t+1}} \mathbb{E}_t^* (R_{t+1} r_{t+1})$$

- Doesn't require that the market is complete
- Doesn't require any distributional assumptions (eg lognormality)
- Allows for the presence of constrained and/or irrational investors
- Holds with equality for a log investor who chooses to hold the market
- This investor's perspective works well empirically for forecasting
  - ▶ the market as a whole (Martin, *QJE*, 2017)
  - ▶ individual stocks (Martin and Wagner, *JF*, 2019)
  - ▶ currencies (Kremens and Martin, *AER*, 2019)

## Information in options (3)



- Using the result of Breeden and Litzenberger (1978), we show

$$\frac{1}{R_{f,t+1}} \mathbb{E}_t^* (R_{t+1} r_{t+1}) = r_{f,t+1} + \underbrace{\frac{1}{P_t} \left\{ \int_0^{F_t} \frac{\text{put}_t(K)}{K} dK + \int_{F_t}^{\infty} \frac{\text{call}_t(K)}{K} dK \right\}}_{\text{LVIX}_t}$$

- This gives the lower bound  $\mathbb{E}_t r_{t+1} - r_{f,t+1} \geq \text{LVIX}_t$
- Bootstrapped  $p$ -value for the mean of  $r_{t+1} - r_{f,t+1} - \text{LVIX}_t$  being *negative* is 0.097

## A sentiment index

- Putting the pieces together,

$$\begin{aligned}\mathbb{E}_t g_{t+1} &= \mathbb{E}_t (r_{t+1} - r_{f,t+1}) + r_{f,t+1} - \mathbb{E}_t (r_{t+1} - g_{t+1}) \\ &\geq \underbrace{\text{LVIX}_t + r_{f,t+1} - \mathbb{E}_t (r_{t+1} - g_{t+1})}_{B_t}\end{aligned}$$

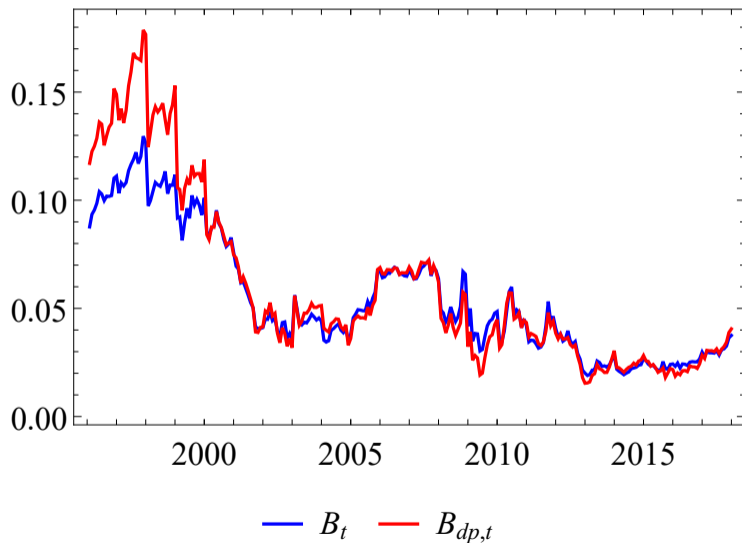
- We replace  $\mathbb{E}_t (r_{t+1} - g_{t+1})$  by the forecast based on  $y_t$ :

$$B_t = \text{LVIX}_t + r_{f,t+1} - (\hat{a}_0 + \hat{a}_1 y_t)$$

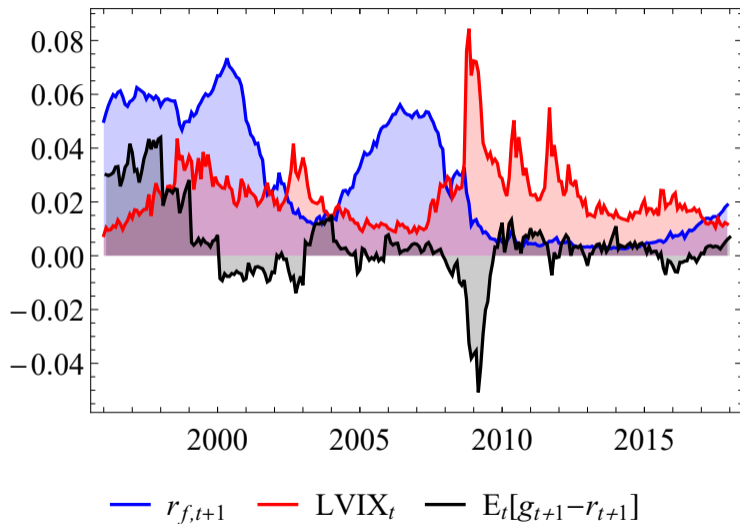
with  $\hat{a}_0$  and  $\hat{a}_1$  calculated on a rolling basis so  $B_t$  is observed at  $t$

- The bound  $\mathbb{E}_t g_{t+1} \geq B_t$  relies on two key assumptions:
  - ▶ the modified NCC
  - ▶ a stable statistical relationship between valuation ratios and  $r - g$

# The sentiment index

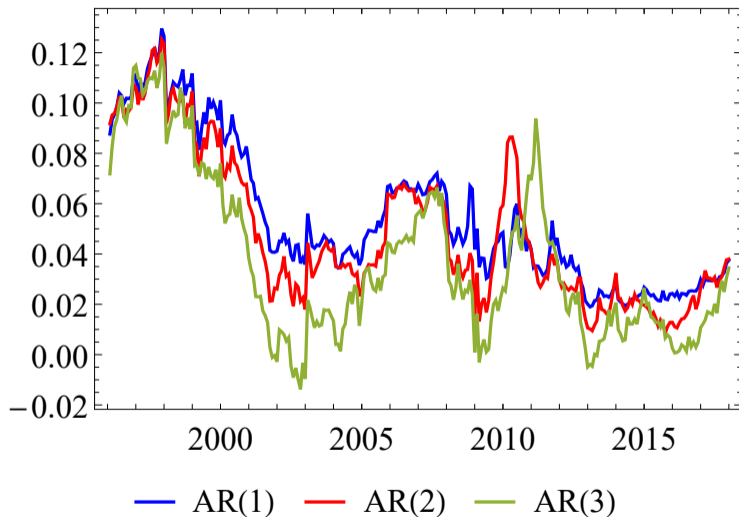


## The three components of the sentiment index, $B_t$

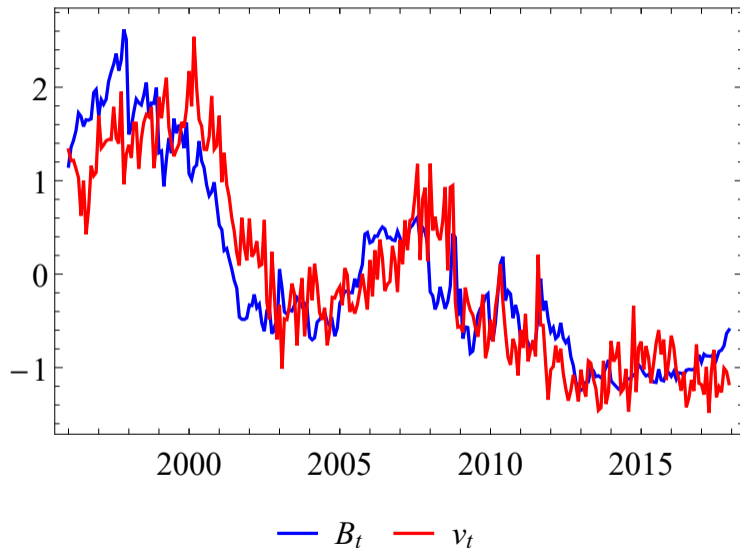




## Allowing $y_t$ to follow an AR( $k$ )



## Sentiment index vs. detrended volume (1)



## Sentiment index vs. detrended volume (2)

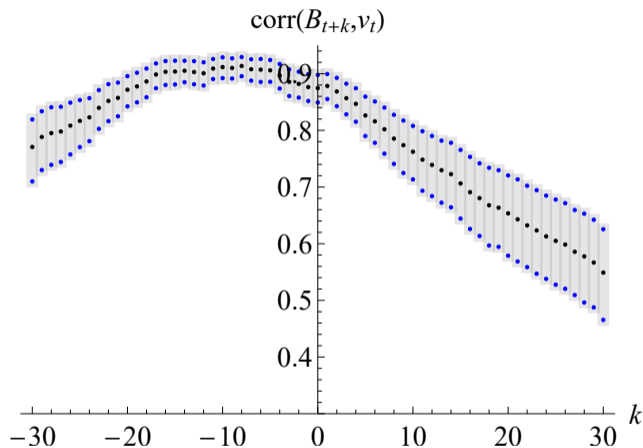


Figure: Correlation between  $B_{t+k}$  and detrended volume at time  $t$ .

## Sentiment index vs. crash probability index (1)

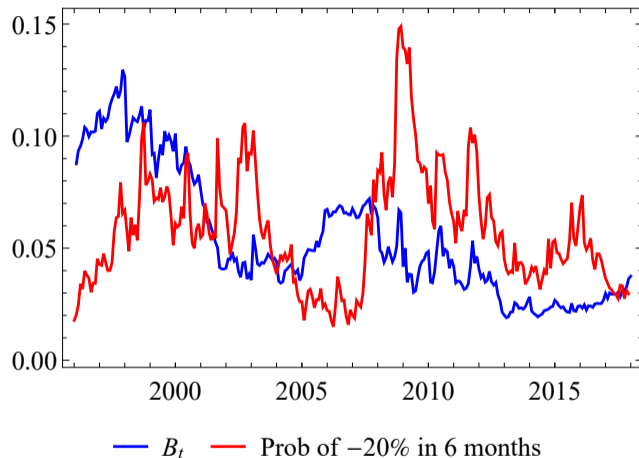


Figure:  $B_t$  and crash probability (Martin, 2017)

## Sentiment index vs. crash probability index (2)

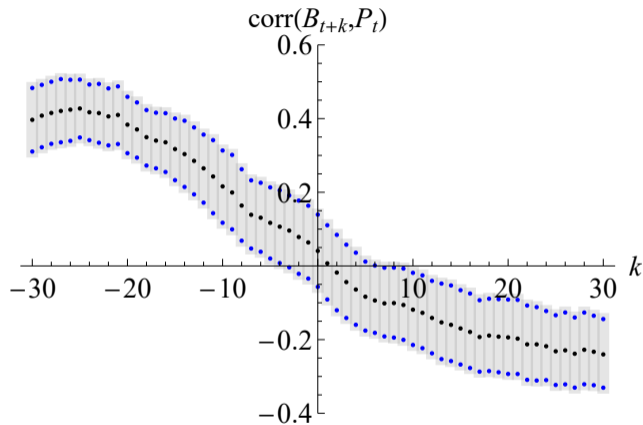


Figure: Correlation between  $B_{t+k}$  and crash probability at time  $t$ .

# Conclusion

- Volatility and valuation ratios have long been linked to bubbles
- We use some theory to make the link quantitative
- We have tried to make choices in a conservative way to avoid “crying bubble” prematurely, and/or overfitting
- Signature of a bubble: valuation ratios, volatility, and interest rates are simultaneously high