

# A General Approach to Recovering Market Expectations from Futures Prices With an Application to Crude Oil by Christiane Baumeister and Lutz Kilian

Jing Cynthia Wu  
Chicago Booth & NBER

# Motivation

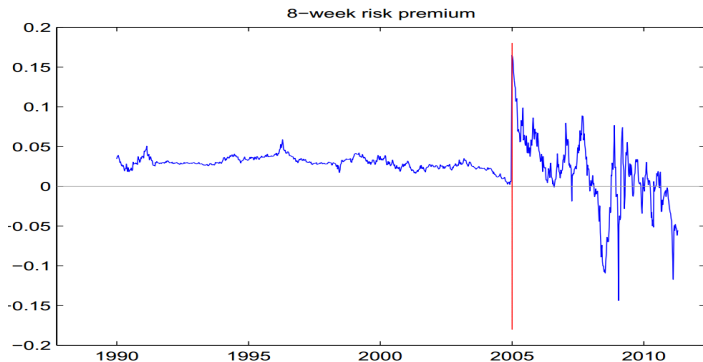
The is a great paper and worth reading!

- ▶ Object of interest
  - ▶ Infer market expectation of future spot price  $E_t[S_{t+n}]$  from futures prices
  - ▶ Expectation can be expressed as

$$E_t(S_{t+n}) = F_t^n - RP_t^n$$

- ▶ This is equivalent to computing the risk premium  $RP_t^n$

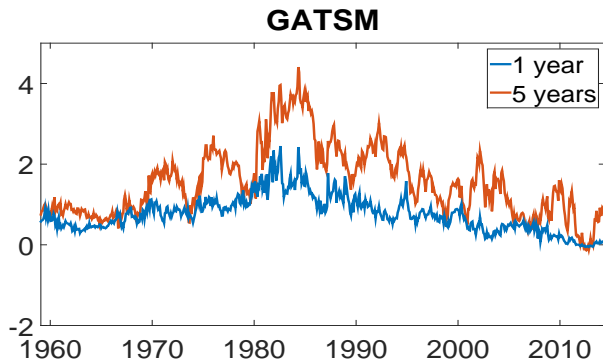
## Time variation in oil future risk premium



$\tilde{f}_t^8 - f_t^8$ , where  $f_t^8$  is log future price

Source: Hamilton and Wu (2014)

# Time variation in bond risk premium



$$tp_t^n = y_t^n - \frac{1}{n} E_t[y_t^1 + \dots + y_{t+n-1}^1]$$

Source: Creal and Wu (2016)

y-axis: in percentage points

## Methods in the literature

$$E_t(S_{t+n}) = F_t^n - RP_t^n$$

- ▶ Use futures prices as  $E_t(S_{t+n})$
- ▶ Return forecasting regression for  $\frac{S_{t+n} - F_t^n}{F_t^n}$ 
  - ▶ Fama and French (1987) with one regressor
  - ▶ Augment with more regressors
- ▶ Hamilton and Wu (2014)
  - ▶ extract  $RP_t^h$  from the term structure
  - ▶ a factor structure driving log futures prices
  - ▶ compute expectations of factors, and hence log prices
  - ▶  $RP_t^h = \text{actual price} - \text{rational expectations}$

## Hamilton and Wu (2014)

Log price linear in  $(m \times 1)$  factors  $x_t$

$$f_t^n = \log F_t^n = \alpha_n + \beta_n' x_t.$$

Factor dynamics

$$x_{t+1} = c + \rho x_t + \Sigma u_{t+1} \quad u_{t+1} \sim \text{i.i.d. } N(0, I_m)$$

Risk prices in affine form

$$\lambda_t = \lambda + \Lambda x_t.$$

Factor loadings are analogous to ATSM recursion

$$\begin{aligned}\beta_n' &= \beta_{n-1}' \rho - \beta_{n-1}' \Lambda \\ \alpha_n &= \alpha_{n-1} + \beta_{n-1}' c + (1/2) \beta_{n-1}' \Sigma \Sigma' \beta_{n-1} - \beta_{n-1}' \lambda.\end{aligned}$$

# Contribution

Contribution: a framework to evaluate methods to compute expectations in the literature

- ▶ Criterion 1: MSPE metric

$$E[S_{t+n} - E_t(S_{t+n})]^2$$

- ▶ Criterion 2: longer-term expectations should evolve smoothly
- ▶ Criterion 3: event studies
- ▶ Criterion 4: out-of-sample comparisons

## Comment 0

What I like about this paper

- ▶ The authors motivate the importance of understanding expectations nicely.
- ▶ The authors have done an elaborate study comparing the methods in the literature.



## Comment 1

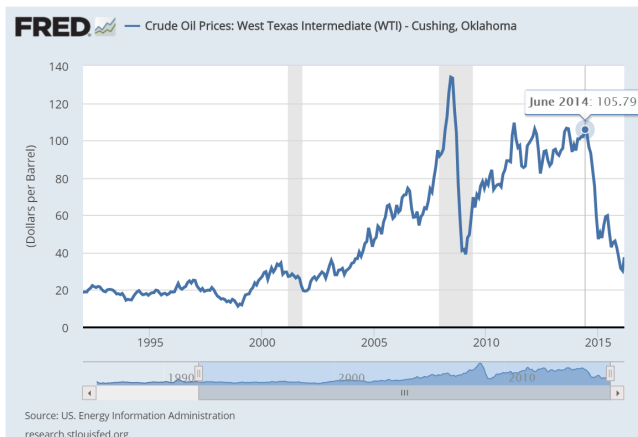
The econometric analysis of model comparison is elaborate

- ▶ Can you tie back to the economic motivations?
  - ▶ What kind of different economic implications do these models make?
  - ▶ What are the policy implications?
  - ▶ A neat example would be the policy makers did  $P^A$  during this episode because they used method  $M^A$  to infer the expectations. If they take the alternative better method  $M^B$ , they should implement an alternative policy  $P^B$  instead.

## Comment 2

Oil price has experienced a dramatic drop since June 2014

- ▶ What is the implication for the recent episode of low oil prices?
- ▶ How does the risk premium look?



## Comment 3

What scale should we work on?

- ▶ level of the price
  - ▶ more intuitive, and risk premium is denoted in dollars
  - ▶ the price series is exponential
  - ▶ The calculation might be heavily influenced by the episode with high prices
- ▶ log price
  - ▶ relatively equal weights over the periods with low and high prices
  - ▶ difference has a return interpretation
  - ▶ closer to definition of risk premium in the bond market, for example

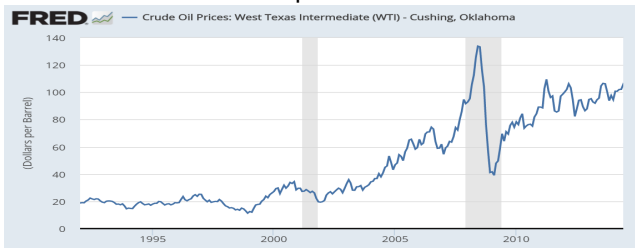
$$tp_t^n = y_t^n - \frac{1}{n} E_t[y_t^1 + \dots + y_{t+n-1}^1]$$

where

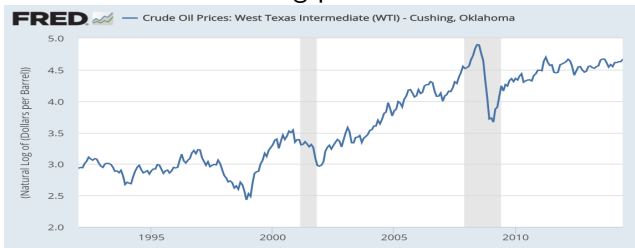
$$y_t^n = -\frac{1}{n} \log P_t^n$$

# Comment 3

price



log price



# Conclusion

Overall, this is a very interesting paper!

- ▶ The authors provide a comprehensive framework to assess various methods to extract market expectations
- ▶ Comment 1: policy implication
- ▶ Comment 2: recent episode of price drop
- ▶ Comment 3: level vs. log