

# "Common Factors in Commodity Futures Curves" by Karstanje, van der Wel, and van Dijk (2015)

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# Background

- ▶ This paper is about modeling **commodity futures curves**
- ▶ **Commodity:** A standardized good, which is traded in bulk and whose units are interchangeable
  - ▶ Natural gas, Coffee, Copper
- ▶ **Futures contract:** Commitment to a transaction on a future date at a prearranged price
  - ▶ Pay \$80 for 1 barrel of oil 1 year in the future

## Background: Why are commodity futures curves interesting?

- ▶ Futures important for commodities producers to hedge price risks
- ▶ Investors use commodities for diversification purposes

# This paper

- ▶ Investigate **commonality** among factors driving many commodity futures curves
  - ▶ Comovement in price levels
  - ▶ Comovement in curve shapes
  
- ▶ **24 commodities** separated into **5 sectors**
  - ▶ Energy, Metals, Softs, Grains, and Meats

## What the authors do

- ▶ **Dynamic Nelson Siegel** framework as in **Diebold and Li (2006)**
  - ▶ Decomposition into level, slope, curvature
  
- ▶ **Common (market and sector-specific) components** drive individual commodity factors as in **Diebold, Li, and Yue (2008)** and **Kose, Otrok, and Whiteman (2003)**

## Preview of main results

- ▶ Important **common components** in level, slope, and curvature factors
  - ▶ On average  $\approx 60\%$  of variation for level factor
  - ▶ On average  $\approx 70\%$  of variation for slope and curvature factors
- ▶ Commonality mostly driven by **sector-specific components** as opposed to market components
  - ▶ On average  $\approx 40\%$  of variation for level factor
  - ▶ On average  $\approx 50\%$  of variation for slope and curvature factors
- ▶ Many other results that I will not go into today...
  - ▶ Economic interpretation of factors
  - ▶ Time-variation in importance of common components

# Model

## Model: Measurement equation

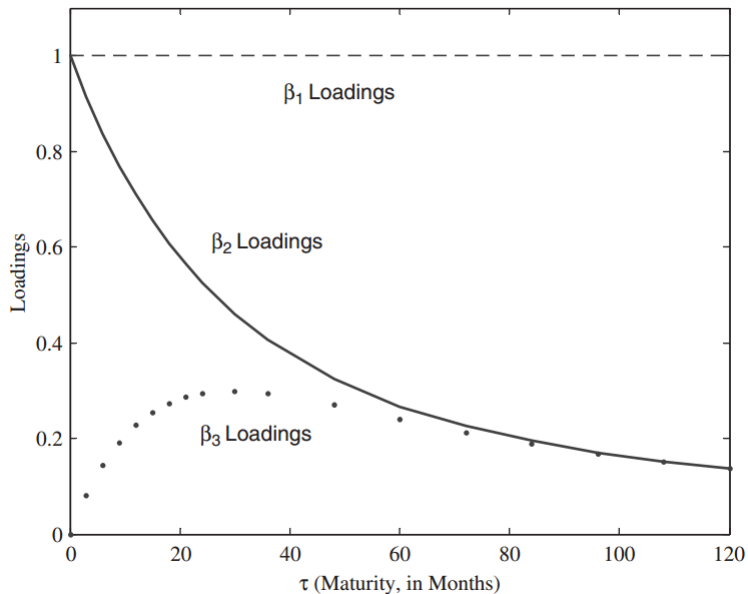
$$f_{i,t}(\tau) = l_{i,t} + s_{i,t} \left[ \left( \frac{1 - \exp^{-\lambda_i \tau}}{\lambda_i \tau} \right) - \left( \frac{1 - \exp^{-\lambda_i}}{\lambda_i} \right) \right] + \\ c_{i,t} \left[ \left( \frac{1 - \exp^{-\lambda_i \tau}}{\lambda_i \tau} - \exp^{-\lambda_i \tau} \right) - \left( \frac{1 - \exp^{-\lambda_i}}{\lambda_i} - \exp^{-\lambda_i} \right) \right] + \\ \kappa_i \cos(\omega(g_i(t, \tau) - \theta_i)) + \nu_{i,t}(\tau)$$

$$\nu_{i,t}(\tau) \sim_{iid} N(0, \sigma_{\nu,i}^2)$$

- ▶  $f_{i,t}(\tau)$  : Futures price for **commodity i** at **time t** with **maturity  $\tau$**
- ▶  $l_{i,t}, s_{i,t}, c_{i,t}$  : level, slope, curvature
- ▶  $\lambda_i$  : decay parameter
- ▶  $\kappa_i \cos(\omega(g_i(t, \tau) - \theta_i))$  : seasonal adjustment term



# Factor loadings (from Diebold and Li, 2006)



## Model: Common components

$$l_{i,t} = \alpha_i^L + \beta_i^L L_{market,t} + \gamma_i^L L_{sector,t} + L_{i,t}$$
$$s_{i,t} = \alpha_i^S + \beta_i^S S_{market,t} + \gamma_i^S S_{sector,t} + S_{i,t}$$
$$c_{i,t} = \alpha_i^C + \beta_i^C C_{market,t} + \gamma_i^C S_{sector,t} + C_{i,t}$$

- ▶  $X_{market,t}$  : Market component
- ▶  $X_{sector,t}$  : Sector-specific component
- ▶  $X_{i,t}$  : Individual commodity component

## Model: Transition equation

$$\begin{pmatrix} \Delta L_{y,t} \\ S_{y,t} \\ C_{y,t} \end{pmatrix} = \Phi \begin{pmatrix} \Delta L_{y,t-1} \\ S_{y,t-1} \\ C_{y,t-1} \end{pmatrix} + \begin{pmatrix} \eta_{y,t}^L \\ \eta_{y,t}^S \\ \eta_{y,t}^C \end{pmatrix}$$

$$\eta_{y,t}^X \sim iid N(0, \sigma_{\eta, X, y}^2)$$

- ▶  $y = \{market, sector, idiosyncratic\}$
- ▶  $\Phi$  is diagonal
- ▶ **Note:** level factor modelled as nonstationary

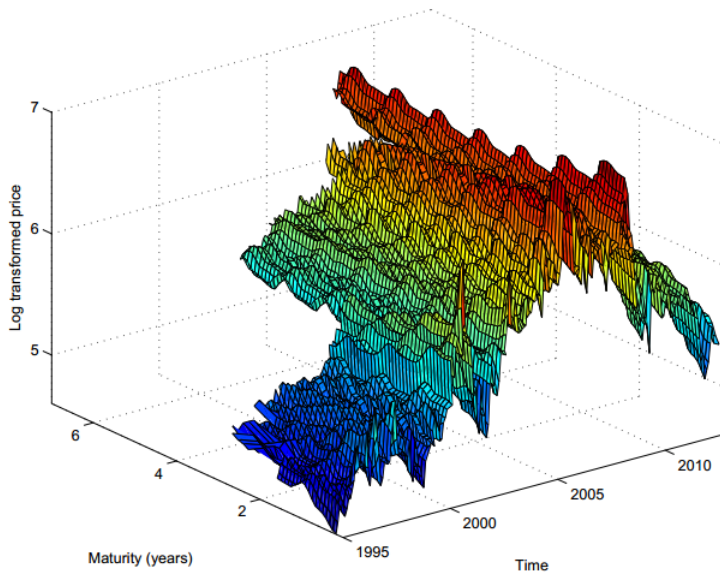
# Estimation

# Data

- ▶ 24 commodities separated into 5 sectors: Energy, Metals, Softs, Grains, and Meats
- ▶ Choose commodities based off of inclusion in **S&P Goldman Sachs Commodity Index**
- ▶ Monthly data from **Jan 1995** to **Sept 2012**
- ▶ Heterogeneity in data availability **between commodities** and **across time**

# Data example: natural gas

(a) Natural gas

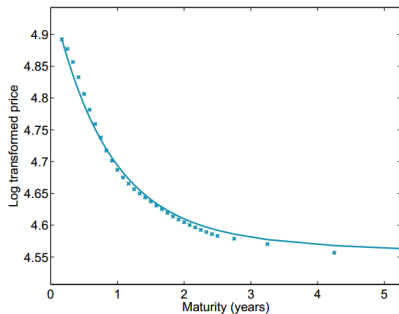


# Is the Dynamic Nelson Siegel framework appropriate?

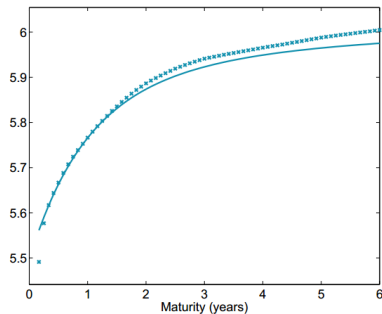
- ▶ Yield curve naturally decomposes into **level, slope, and curvature**
- ▶ Dynamic Nelson Siegel model has level, slope, and curvature factors
- ▶ **Question:** Does the same hold true for commodity futures curves?

# Examples of futures price curve

(c) WTI crude oil January 2000

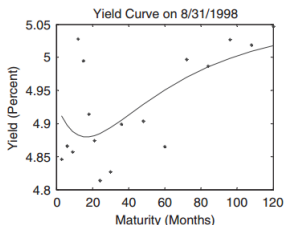
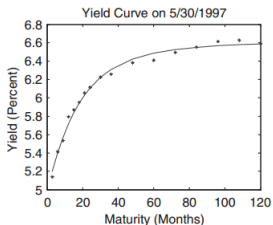
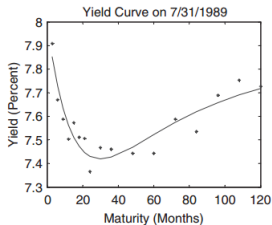
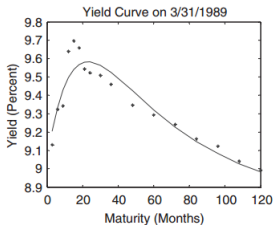


(d) WTI crude oil futures price curve November 2008





# Examples of yield curve (taken from Diebold and Li, 2006)



## Answer: As a statistical approximation... YES!

VERY simple case (nonstochastic)

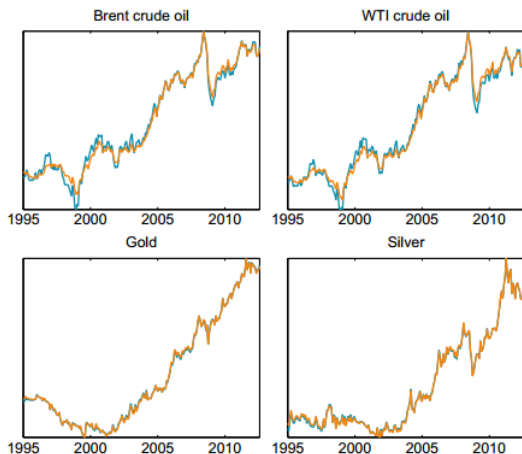
$$f_t(\tau) = s_t + (r - c)\tau$$

- ▶  $f_t(\tau)$  : futures price with **maturity**  $\tau$
- ▶  $s_t$  : spot price
- ▶  $r$  : risk free rate
- ▶  $c$  : convenience yield - the flow of services which accrues to the owner of a physical inventory but not to the owner of a contract for future delivery (**Brennan, 1991**)
- ▶ **Question:** Could there be a theoretical issue with DNS approximation as maturity goes to  $\infty$ ?

## Relationship with Schwartz (1997)

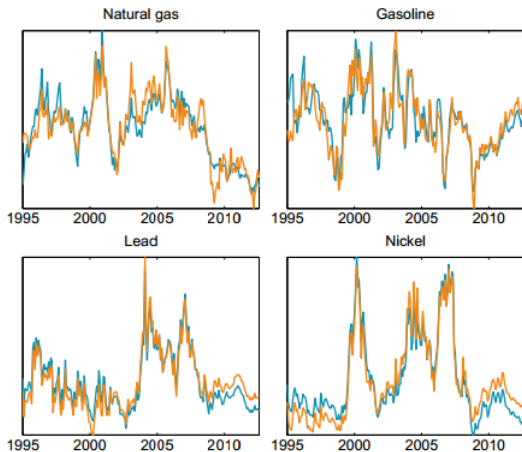
- ▶ **Schwartz (1997)** 3-factor model
  - ▶ **Arbitrage-free model** of commodities futures curve
  - ▶ Posits stochastic processes for **spot price, convenience yield, and risk free rate**
  - ▶ Uses no-arbitrage restrictions to compute commodities futures curve
  
- ▶ **Commodity-by-commodity:** estimate individual factor DNS model and 3-factor arbitrage-free model

# Comparison of DNS and Schwartz (1997) model estimates



- ▶ Orange: DNS level factor
- ▶ Blue: Spot price factor
- ▶ Average correlation across commodities: 0.67

# Comparison of DNS and Schwartz (1997) model estimates



- ▶ Orange: DNS slope factor
- ▶ Blue: Convenience yield factor
- ▶ Average correlation across commodities: 0.76

# Joint model factor estimates

## Autoregressive parameters

|        | $\Delta$ Level | Slope | Curvature |
|--------|----------------|-------|-----------|
| Market | 0.11           | 0.93  | 0.78      |
| Energy | 0.12           | 0.79  | 0.88      |
| Metals | 0.12           | 0.96  | --        |
| Softs  | 0.45           | 0.93  | 0.87      |
| Grains | -0.12          | 0.94  | 0.88      |
| Meats  | 0.05           | 0.88  | 0.72      |

Red numbers are significant at 10% level

► Factor estimates

# Variance decompositions

|        | $\Delta$ Level |        |      | Slope  |        |      | Curvature |        |      |
|--------|----------------|--------|------|--------|--------|------|-----------|--------|------|
|        | Market         | Sector | Idio | Market | Sector | Idio | Market    | Sector | Idio |
| Energy | 12.4           | 61.3   | 26.3 | 74.6   | 12.3   | 13.1 | 28.3      | 44.0   | 27.7 |
| Metals | 16.1           | 49.1   | 34.7 | 17.5   | 75.6   | 6.9  | —         | —      | —    |
| Softs  | 34.1           | 12.8   | 53.2 | 1.7    | 26.6   | 71.7 | 22.7      | 34.0   | 43.3 |
| Grains | 53.5           | 21.6   | 25.0 | 0.1    | 74.1   | 25.9 | 25.4      | 66.8   | 7.9  |
| Meats  | 7.3            | 25.2   | 67.4 | 7.2    | 53.3   | 39.5 | 15.4      | 61.8   | 22.7 |

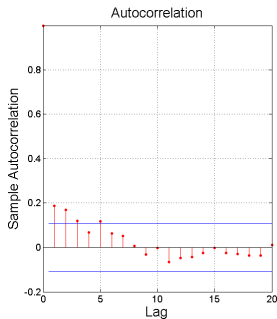
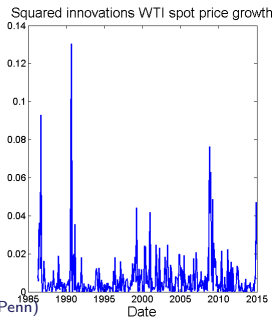
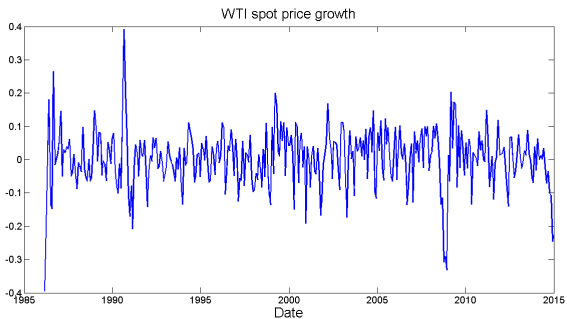
# Further comments/extensions



## Extension 1: Forecasting exercise

- ▶ DNS model has been shown to forecast well ([Diebold and Li, 2006](#)), especially when compared to affine term structure models
- ▶ Forecast comparison:
  - ▶ Individual-commodity DNS model
  - ▶ [Schwartz \(1997\)](#) 3-factor model
  - ▶ Benchmark forecasting models (i.e. random walk)

## Extension 2: Stochastic volatility



## Extension 3: Going beyond the U.S. economy

- ▶ Economic interpretation of futures curves in terms of **U.S. macroeconomic variables** ▶ Economic interpretation
- ▶ What about other countries?
- ▶ Importers versus exporters of commodities (**Peersman and van Robays, 2012**)

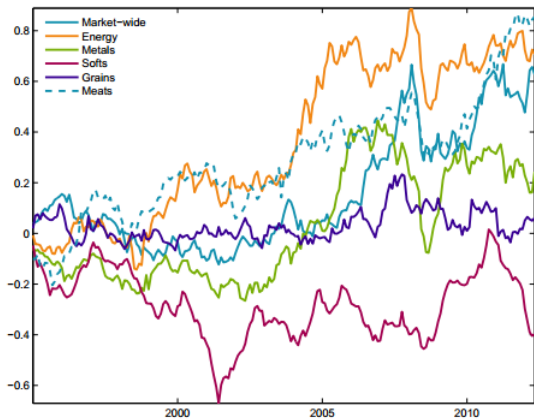
## Extension 4: Real effects of commodities futures curve uncertainty shocks

- ▶ How does uncertainty about the shape of commodity futures curves impact the real economy?
  - ▶ Overall economy versus commodity sectors
  - ▶ Commodity importers versus exporters
  - ▶ Monetary policy implications?
  
- ▶ **Creal and Wu (2014)**: Effects of short and long-term interest rate uncertainty on the macroeconomy

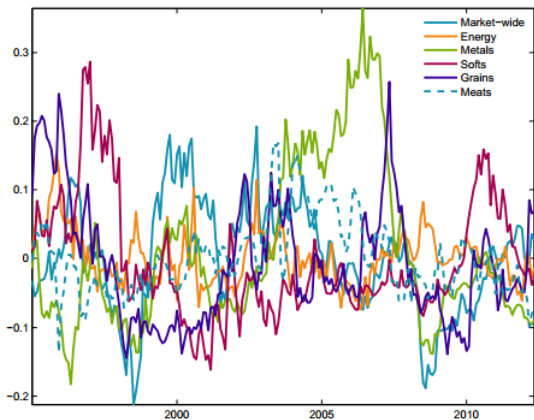
# Conclusion

- ▶ I really like this paper
  - ▶ Novel application of DNS methodology
  - ▶ **Lots** of clean empirical results
- ▶ Opens up a lot of interesting empirical questions (of which I have only listed 4)

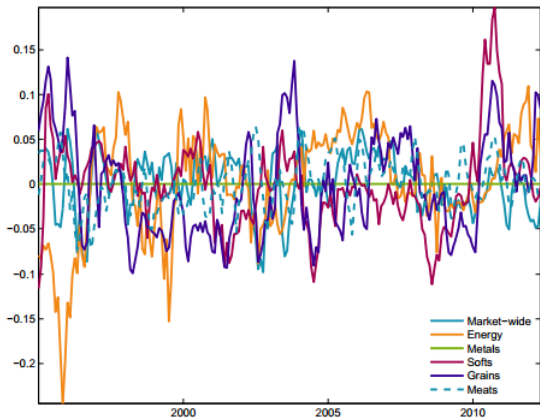
(a) Level - market-wide



(a) Slope - market-wide



### (a) Curvature - market-wide



▶ Return



# Economic interpretation of factors

- ▶ Theory of normal backwardation (**Keynes, 1930**)
  - ▶ Commodity producers and inventory holders hedge risk by selling commodity futures
  - ▶ Requires setting futures prices at discount to encourage speculators to take opposite long position
  - ▶ **Empirical implication:** relationship between **hedging pressure** and commodity futures curve
- ▶ Theory of storage (**Kaldor, 1939, Working, 1949**)
  - ▶ Relationship between futures price, spot price, inventory storage benefits, and inventory storage costs
  - ▶ **Empirical implication:** relationship between **inventory levels** and commodity futures curve

▶ Return

# Economic interpretation of factors

## Market-wide components:

- ▶  $\Delta$  Level
  - ▶ Hedging pressure (+), Equity returns (+), USD exchange rate returns (-)
- ▶ Slope
  - ▶ Business inventories (+), Hedging pressure (+), Leading economic indicators (+), Current economic indicators (-)
- ▶ Curvature
  - ▶ Business inventories (+), Hedging pressure (-), Interest rates (+)

# Time-variation in importance of factors

Investigate time-variation in importance of common components

- ▶ Weights  $\delta^{|k|}$  to each observation
- ▶  $\delta = 0.99$
- ▶  $k = \dots, -2, -1, 0, 1, 2, \dots$
- ▶ Centered at  $\Theta = 1, 2, \dots, T$

(a)  $\Delta$ Level

