

3rd International Workshop on Empirical Methods in Energy Economics (EMEE2010)

Surrey Energy Economics Centre (SEEC)
University of Surrey, UK
24th – 25th June 2010

NOTE:

The following Presentation represents *Work in Progress* for discussion at the EMEE2010 workshop. It therefore must not be referred to without the consent of the author(s).

Sponsored by:



Oil Prices: Heavy Tails, Mean Reversion and the Convenience Yield

Jean-Thomas Bernard
GREEN, Université Laval

Lynda Khalaf
CIREQ, GREEN, Carleton University

Maral Kichian
Bank of Canada

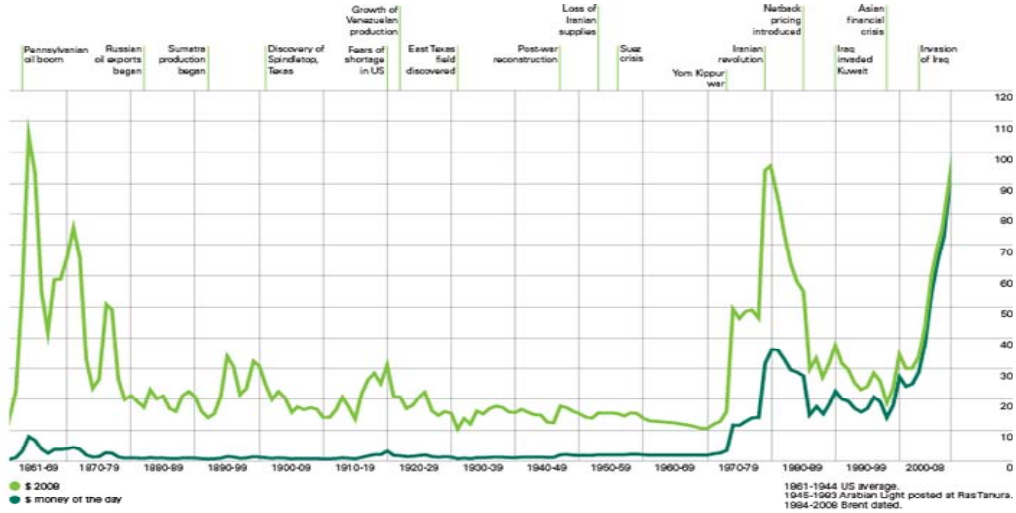
Sebastien McMahon
Fédération des Caisses Desjardins



Chart of crude oil prices since 1861

Crude oil prices 1861 - 2008

US dollar per barrel
World events



MOTIVATION

- Analysis of oil price → active area of research:
 - linkages: financial markets or macro-economy [e.g. Kilian (2008a,b,c; 2009), Hamilton (2009)]
 - price dynamics, forecasting [e.g. Lee, List & Strazi-
cich (2006), Cortazar & Naranjo (2006), Tabak &
Cajueiro (2007), Postali & Picchetti (2006)].
- Stylized facts: heavy tails & breaks \oplus statistical sup-
port for many models.
- Methodological disagreements:
 - deterministic/stochastic trends?
 - structural change / inherent fluctuations
 - unexpected discontinuities
 - non-constancy of variance
 - non-constancy of convenience yield.

RESEARCH QUESTION

- We examine mean-reversion in oil price from forecast-based perspective.
- We consider alternative classes of models with non-constancies in level or volatility.
- We do not take a stand on worth of unit root tests for the problem at hand, but we do believe that available evidence on the importance of parameter non-constancies, whether for refuting or for justifying the unit-root hypothesis, should be taken seriously.

- Theoretical reasons: unit-root not appropriate for natural resources or commodities:
 - Demand & Supply: when prices are higher (or lower) than some equilibrium level, high-cost producers will enter (or exit) the market, which pushes prices downward (or upward).
 - Relationship between (information in) future prices at different maturities & spot price ► convenience yield [**CY**].

- CY: flow of goods & services that accrues to owner of a spot commodity (**a physical inventory**) but not to owner of a futures contract (**a contract for future delivery**).
 - Random-walk [RW]: consistent with constant CY.
 - Mean Reversion [MR] & positive correlation between spot price & CY changes: consistent with theory of storage: when inventories ↓ (or ↑), spot price will ↑ (or ↓) & CY will also ↑ (or ↓) because futures prices will not ↑ (or ↓) as much as spot prices.
- [Pindyck (1999, 2001), Schwartz (1997), Schwartz & Smith (2000)]: refute constant CY & suggest MR to long-run equilibrium that itself can change randomly over time.

CONTRIBUTION

- We analyze a MR class of models from Schwartz (1997), Schwartz & Smith (2000), relative to various RW alternatives, with focus on forecast performance.
 - The MR structural form:
 - presumes a stochastic CY,
 - derives from joint behavior of spot & future prices
 - allows one to disentangle the persistent (or long-run [**LR**]) from the transitory (or short-run [**SR**]) component of price
 - LR** ► Brownian motion;
 - SR** ► Ornstein – Uhlenbeck
- ⇒ overall evolution \neq standard RW walk via ST deviation term.

- Non-MR class of models considers:
 - various (G)ARCH effects including asymmetric, conditionally non-normal (G)ARCH and (G)ARCH-in-mean [see Regnier (2007), Beck (2001), Sadorsky (2006)]
 - random jumps: an integral part of price process leading to relatively rare adjustments that can be distinguished from frequent and relatively "small" ordinary price fluctuations. Ait-Sahalia (2004), Drost, Nijman & Werker (1998).

- Future prices, from 1986 to 2007, weekly and monthly frequencies, and for various forecast horizons.
- Forecasting in real time: we use one-step-ahead out-of-sample forecasts, where parameter estimates are updated at every step of the procedure.
- In models with jumps, analytical formulae are not readily available for expected forecast errors, so we devise a simple **simulation-based** procedure to approximate these errors [Khalaf, Saphores & Bilodeau (2003); Bernard, Khalaf, Kichian & McMahon (2008)]
- Our results support the MR model over all forecast horizons considered.

MODELS

$y_t = \ln(Y_t) - \ln(Y_{t-1})$, $Y_t =$ nominal price level

► GARCH-M(1,1)

relationship between returns & time-varying risk, via ϕ

$$y_t = v_t + \beta h_t,$$

$$v_t = \mu + \sqrt{h_t} z_t,$$

$$h_t = \alpha_0 + \alpha_1 (y_{t-1} - \mu)^2 + \phi h_{t-1}$$

$$z_t \stackrel{i.i.d.}{\sim} N(0, 1).$$

► GARCH(1,1) with cond. non-normality

$$\beta = 0 \text{ \&}$$

$z_t \stackrel{i.i.d.}{\sim}$ student-t(τ) where τ is unknown.

EGARCH(1,1) sign of shocks is relevant: $\beta = 0$ &

$$\ln(h_t) = \alpha_0 + \phi \ln(h_{t-1}) + \gamma \left(\frac{y_{t-1} - \mu}{\sqrt{h_{t-1}}} \right) + \eta \left[\frac{|y_{t-1} - \mu|}{\sqrt{h_{t-1}}} - \frac{\sqrt{2}}{\sqrt{\pi}} \right].$$

► **Specification with Poisson jumps:**

$$y_t = v_t + \sum_{i=1}^{n_t} \ln P_{it},$$

n_t = number of jumps that occur between $t - 1$ & t ,

P_{it} = size of i th jump over this time interval.

Jumps follow a Poisson process with arrival rate λ
(a jump occurs on average, every $1/\lambda$ periods),

P_{it} i.i.d. lognormal(θ, δ^2).

Schwartz & Smith (2000): on commodities

- MR stems from holding inventories: producers make joint decisions on production & inventory levels, accounting for a spot (sales) price & a storage price determined from marginal CY
- Two markets interact, so **equilibria** in both markets are relevant: exogenous demand shock (or changes in spot price volatility) → reactions in spot prices AND inventory adjustments → reactions in price of holding such inventory [CY] → reflected in future prices, which allows prices to revert back to trend.
- CY = counterpart of dividend yield (stock): spot price = present value of discounted future prices. Holding oil is risky \Rightarrow spot prices \neq expected future prices, and difference [net of storage costs] leads to marginal CY.

- Tractable time series model: MR, but a time-varying CY implies that mean to which price reverts is, itself, time-varying.
- LR (equilibrium, persistent) component \sim Brownian motion, & SR (deviations, transitory) component \sim Ornstein-Uhlenbeck process that reverts towards zero:

$$\ln(Y_t) = \underbrace{\chi_t}_{SR} + \underbrace{\xi_t}_{LR}$$

$$\chi_t = e^{-\kappa} \chi_{t-1} + \epsilon_t^\chi,$$

$$\xi_t = \mu_\xi + \xi_{t-1} + \epsilon_t^\xi$$

κ = rate of speed at which price reverts to its equilibrium (rate at which SR deviations disappear), μ_ξ = mean of equilibrium price, normal shocks with volatilities σ_χ & σ_ξ correlation $\rho_{\chi\xi}$.

- Note: $\chi_t = \frac{1}{\kappa}(\delta_t - \alpha)$ where $\delta_t = \text{CY}$.

- The above specification leads to the following for future prices:

$$\ln(F_{n,t}) = e^{-kn} \chi_t + \xi_t + A(n),$$

where $F_{n,t}$ represents the market price, at time t , for a futures contract with time n until maturity

$$\xi_t = \mu_\xi + \xi_{t-1} + \epsilon_t^\xi,$$

$$\chi_t = e^{-\kappa} \chi_{t-1} + \epsilon_t^\chi,$$

$$A(n) = (\mu_\xi - \lambda_\xi) n - (1 - e^{-\kappa n}) \frac{\lambda_\chi}{\kappa} + \frac{1}{2} \left((1 - e^{-2\kappa n}) \frac{\sigma_\chi^2}{2\kappa} + \sigma_\xi^2 n + 2 (1 - e^{-\kappa n}) \frac{\rho_{\chi\xi} \sigma_\chi \sigma_\xi}{\kappa} \right).$$

- The system can then be written in a state-space form, estimation via Kalman filter.

Empirical Analysis

- Daily crude oil prices obtained from the U.S. Department of Energy, Energy Information Administration, for 1, 2, 3 and 4 month futures, from January 2, 1986 to January 9, 2007.
- Weekly & monthly prices: Wednesday values, & price on Wednesday \approx 15th day of month.
- MR model: uses four maturities

$$MAPE = \frac{1}{K} \sum_{k=1}^K |\ln(\hat{Y}_{T+k|T+k-1}) - \ln(Y_{T+k})|,$$

$$MSPE = \frac{1}{K} \sum_{k=1}^K \left[\ln(\hat{Y}_{T+k|T+k-1}) - \ln(Y_{T+k}) \right]^2.$$

- To check whether outcomes are driven by recent soaring markets, we repeat exercise, using data from January 2, 1986 to January 9, 2005 only. Qualitatively similar results

- For models with jumps:
 - We estimate parameters over sample of size T
 - Given the latter, we generate 1,000 simulated values of dependent variable \tilde{Y}_{T+1} .
 - The forecast value of $Y_{T+1} =$ average of these 1,000 \tilde{Y}_{T+1} ; the $T + 1$ forecast error is computed.
 - The observed value of the dependent variable, Y_{T+1} , is added to the sample, the model is re-estimated, and the entire simulation process is repeated. Thus, \tilde{Y}_{T+2} is obtained, as well as the forecast error for $T + 2$.
 - The above steps are repeated until $T + K$ forecast errors are obtained, which are then used to construct the MAPE and MSPE.

Horizon	1 year	3 years	5 years	
Unit root	.0115	.3558	.8256	<u>Daily</u>
GARCH	.0113	.3324	.8706	1 month
GARCH-M	.0118	.4259	1.711	
EGARCH	.0104	.3429	.7835	
GARCH-t	.0101	.3318	.7984	
GARCH + jumps	.0150	.3822	1.2272	
GARCH- t + jumps	.0103	.3444	.8443	
MR	.0006	.0016	.0003	
Unit Root	.0113	.4203	.7539	<u>Daily</u>
GARCH	.0102	.4083	.7601	4 months
GARCH-M	.0084	.3909	.9071	
EGARCH	.0105	.4200	.8004	
GARCH-T	.0096	.3998	.7504	
GARCH + jumps	.0126	.4635	.9787	
GARCH- t + jumps	.0101	.1483	.7901	
MR	.0001	.0250	.0001	

	1 year	3 years	5 years	
Unit Root	.0160	.2949	.7028	<u>Weekly</u>
GARCH	.0174	.3273	.6670	1 month
GARCH-M	<i>.0071</i>	<i>.1745</i>	<i>.4237</i>	
EGARCH	.0176	.3346	.6919	
GARCH- <i>t</i>	.5338	5.1347	10.7391	
GARCH + jumps	.0239	.4005	.8603	
GARCH- <i>t</i> + jumps	.0174	.3143	.6453	
MR	.0004	.0003	.0002	
Unit Root	.0190	.3572	.6492	<u>Weekly</u>
GARCH	.0196	.3827	.6183	4 months
GARCH-M	<i>.0158</i>	<i>.0996</i>	<i>.2980</i>	
EGARCH	.0195	.3809	.6118	
GARCH-T	.6181	6.7638	10.4958	
GARCH + jumps	.0270	.5022	.8310	
GARCH- <i>t</i> + jumps	.0197	.3739	.6081	
MR	.0001	.0001	.0001	

	1 year	3 years	5 years	
Unit Root	.0015	.2479	.8461	<u>Monthly</u>
GARCH	<i>.0014</i>	.2967	.8527	1 month
GARCH-M	.0098	<i>.2324</i>	<i>.7151</i>	
EGARCH	.0114	.3051	.8800	
GARCH- <i>t</i>	.0115	.3138	.8721	
GARCH + jumps	.0129	.3140	.9817	
GARCH- <i>t</i> + jumps	.0121	.2946	.8427	
MR	.0004	.0002	.0002	
Unit Root	.5363	.3196	.7431	<u>Monthly</u>
GARCH	.5319	.3459	.7594	4 months
GARCH-M	<i>.4961</i>	<i>.2515</i>	<i>.5483</i>	
EGARCH	.5320	.3453	.7556	
GARCH- <i>t</i>	.5319	.3445	.7593	
GARCH + jumps	.5470	.3949	.9374	
GARCH- <i>t</i> + jumps	.5330	.3391	.7325	
MR	.0001	.0001	.0001	

CONCLUSION

- Our analysis with future price data ranging from 1986 to 2007 suggests that imposing the random walk for oil prices has pronounced costs for out-of-sample forecasting.
- The mean reverting model we consider decomposes price into the sum of a Brownian motion for the long-run evolution, and an Ornstein-Uhlenbeck process for the short term deviations.
- Our results illustrate the importance of relying on a time series models for oil which fits [via the dynamics of the convenience yield] the inter-related equilibria in the spot and storage market.