

Time-Varying Jump Intensities and the Interconnectedness of the North American Crude Oil Complex

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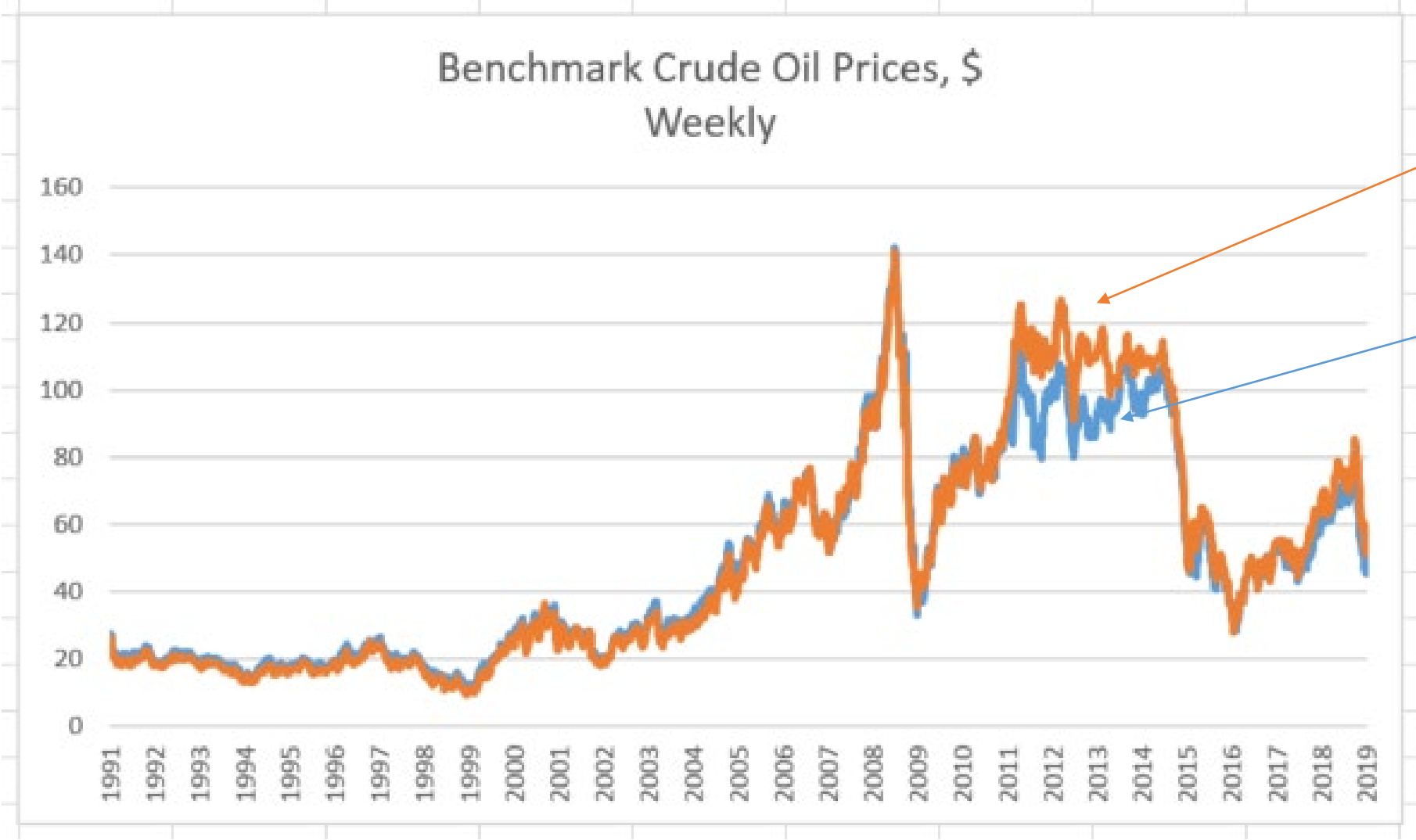
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AIEE / IAEE Dec 2020

Global Pool Hypothesis

- Adleman (1984)
 - *“The world oil market, like the world ocean, is one great pool.”*
- Large price differentials between crude streams have been observed recently.
- Dar (2019)
 - post-2010, the two markets [WTI, Brent] have been witnessing growing divergence, providing evidence that the two markets have moved toward ‘regionalization’.

West Texas Intermediate, Brent crude oil



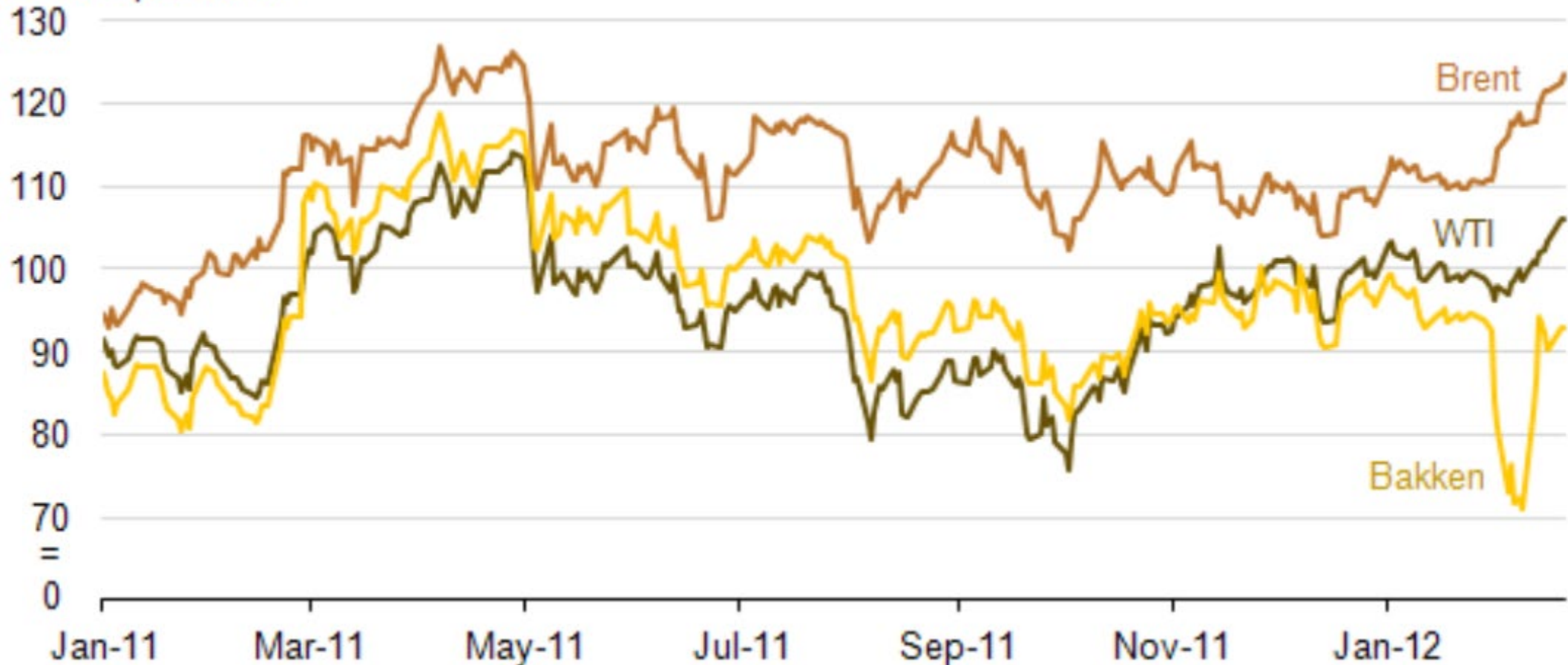
Brent

WTI

WTI, Brent and Bakken

Crude oil spot prices, January 3, 2011 - February 22, 2012

dollars per barrel



Source: U.S. Energy Information Administration, based on Bloomberg.

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Can we quantify the relationship between the crude oil streams, specifically within the North American complex?

Literature Review

- Global Pool Hypothesis
 - Adleman (1984), Ewing and Harter (2000), Hammoudeh *et al* (2008), Fattouh (2010), Wilmot (2013), Dar (2019)
- Discontinuous stochastic processes
 - Askari and Krichene (2008), Lee *et al* (2010), Wilmot and Mason (2013); Postali and Pichetti (2006)
- Volatility
 - Bollerslev (1986), Pindyck, R. (2004), Efimova and Serletis (2014)

Methodology

- Utilizing the autoregressive Jump Intensity (ARJI) model of Chan and Maheu (2002)

$$R_t = \mu + \sum_{i=1}^l \phi R_{t-i} + \sqrt{h_t} z_t + \sum_{k=1}^{n_t} Y_{t,k} \quad z_t \sim NID(0,1)$$

- The conditional jump size is given as $Y_{t,k} \sim N(\theta, \delta^2)$
- And the number of jumps arriving between $t - 1$ and t , is described by the discrete counting process, n_t
- And this process is distributed as a Poisson random variable, described by the jump intensity $\lambda_t > 0$

Methodology cont'd

- The process that describes the evolution of the conditional jump intensity must be specified.
- As in Chan and Maheu (2002), an AR(1) process describes the evolution of λ_t such as:

$$\lambda_t = \lambda_0 + \sum_{i=1}^r \rho_i \lambda_{t-i} + \sum_{i=1}^s \gamma_i \xi_{t-i}$$

Methodology cont'd

- The time series values of $\hat{\lambda}_t$ are obtained, and a GARCH model is constructed.
- Univariate Model

$$\hat{\lambda}_t = c_0 + \varepsilon_t, \quad \text{and} \quad \varepsilon_t \sim N(0, h_t)$$

- where

$$h_t = \alpha_0 + \alpha_1 \varepsilon_{t-1}^2 + \dots + \alpha_q \varepsilon_{t-q}^2 + \beta_1 h_{t-1} + \dots + \beta_p h_{t-p}$$

Methodology cont'd

- Bollerslev (1990) specifies a multivariate GARCH model that has a constant conditional correlations (CCC), with conditional covariance matrix

$$H_t = D_t \Gamma D_t$$

- For example, in the bivariate case

$$D_t = \begin{bmatrix} \sqrt{h_{11t}} & 0 \\ 0 & \sqrt{h_{22t}} \end{bmatrix} \text{ and } \Gamma = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}.$$

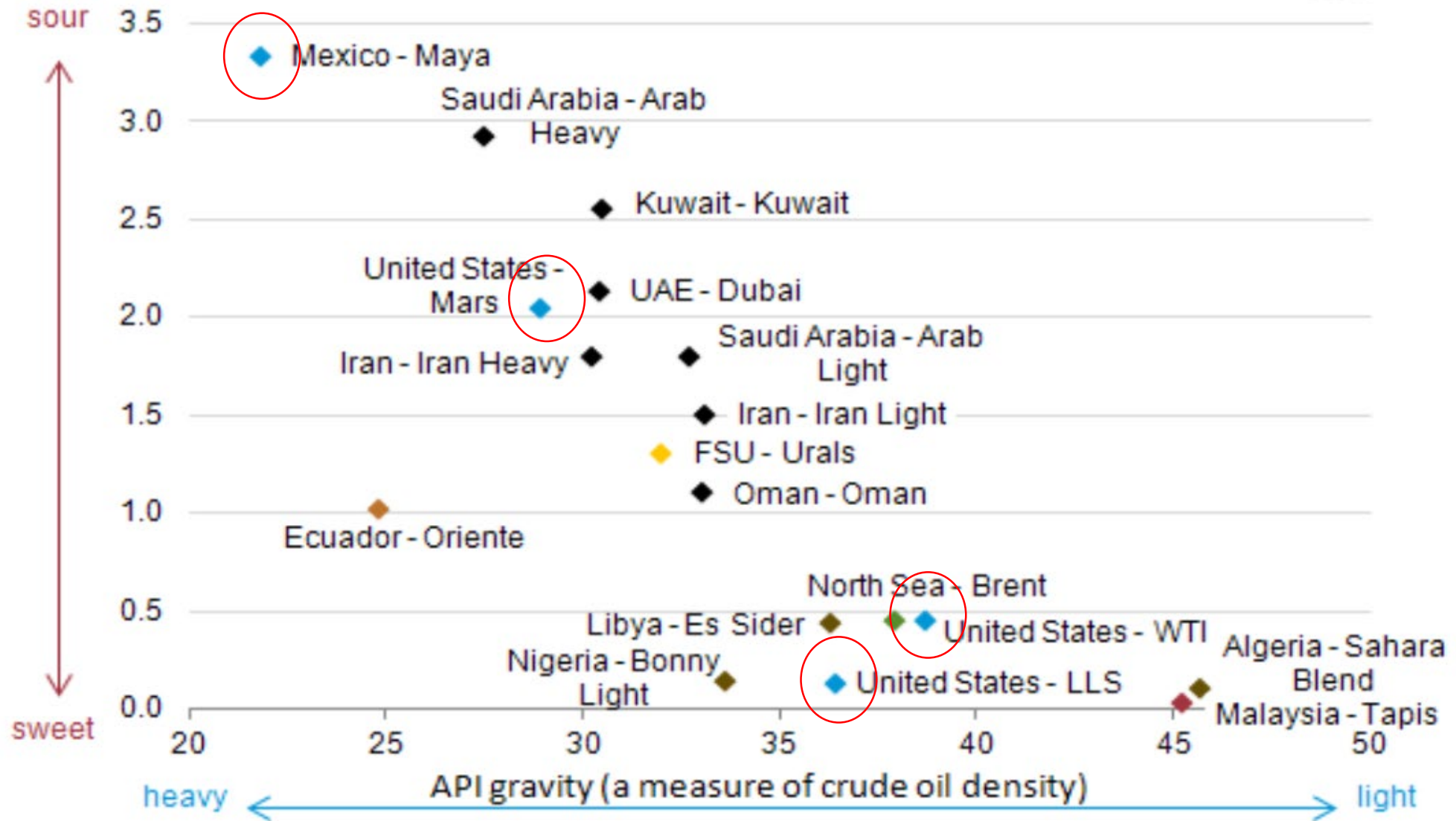
Data

The benchmark *WTI* series along with a mix of secondary blends, across *density* and *Sulphur* content

Table 2: Summary Statistics for Log Crude Oil Price Returns

Blend	ln <i>WTI</i>	ln <i>LLS</i>	ln <i>Mars</i>	ln <i>Lou. Heavy</i>	ln <i>Mex. Maya</i>
Start	01/05/2000	01/05/2000	01/05/2000	01/05/2000	01/05/2000
End	12/31/2018	12/31/2018	12/31/2018	12/31/2018	12/31/2018
Mean	0.010	0.014	0.016	0.014	0.020
Median	0.090	0.071	0.072	0.102	0.095
Maximum	21.28	15.437	15.431	15.028	13.274
Minimum	-16.550	-18.349	-18.619	-19.060	-18.437
Std. Dev.	2.420	2.356	2.646	2.437	2.285
Coefficient of Variation	220.80%	164.5%	161.9%	169.8%	116.5%
Skewness	-0.04	-0.085	-0.165	-0.168	-0.283
Kurtosis	4.51	3.78	3.65	3.71	4.71
JB Test	4030.7	2836.7	2688.2	2760.6	4469.9
<i>p</i> -value	<0.001	<0.001	<0.001	<0.001	<0.001
Count	4763	4763	4763	4763	4763
<u><i>Crude Characteristics</i></u>					
API Density	40	35.6	30.3	32.9	22
	<i>Light</i>	<i>Light</i>	<i>Light</i>	<i>Heavy</i>	<i>Heavy</i>
Sulphur content	0.3	0.37	1.91	0.35	3.3
%	<i>Sweet</i>	<i>Sweet</i>	<i>Sour</i>	<i>Sweet</i>	<i>Sour</i>

Crude Oil Blends (*EIA*)



Results: *Stationarity*

Table: Stationarity Tests

Series	Modified		KPSS	Lags
	ADF test	Lags		
<i>Prices</i>				
WTI	-1.529	5	2.29	31
LLS	-1.396	1	2.33	31
Mars	-1.495	1	2.30	31
Lou. Hvy	-1.428	1	2.29	31
Mex. Maya	-1.547	4	2.22	31
<i>Ln Returns</i>				
WTI	-3.732	31	0.0323	31
LLS	-3.473	30	0.0367	31
Mars	-3.529	30	0.036	31
Lou. Hvy	-3.308	30	0.0361	31
Mex. Maya	-7.23	30	0.0336	31

Note: Modified ADF test with ERS critical values -3.48 (1%), -2.89 (5%), -2.57 (10%). Trend.



Results: Jump Diffusion Process (*ARJI*)

Table: ARJI Results

	LLS		West Texas intermediate		Mars		Mex. Maya		Lou. Heavy	
	Constant	ARJI	Constant	ARJI	Constant	ARJI	Constant	ARJI	Constant	ARJI
μ	0.1074 *** 0.033	0.0889 *** 0.030	0.1149 *** 0.032	0.0906 *** 0.030	0.1787 *** 0.046	0.1408 *** 0.043	0.1238 *** 0.029	0.1352 *** 0.031	0.1134 *** 0.035	0.0935 *** 0.032
ω	0.0221 *** 0.010	0.0190 *** 0.007	0.0191 ** 0.008	0.0185 *** 0.006	0.0063 0.011	0.0129 0.008	0.0172 ** 0.007	0.0164 *** 0.006	0.0235 ** 0.010	0.0019 *** 0.007
α	0.0499 *** 0.007	0.0295 *** 0.005	0.0452 *** 0.006	0.0280 *** 0.005	0.0557 *** 0.007	0.0291 *** 0.008	0.0455 *** 0.006	0.0367 *** 0.005	0.0442 *** 0.007	0.0288 *** 0.005
β	0.9344 *** 0.009	0.9571 *** 0.007	0.9409 *** 0.008	0.9590 *** 0.007	0.9309 *** 0.009	0.9575 *** 0.009	0.9384 *** 0.008	0.9478 *** 0.007	0.9418 *** 0.009	0.9591 *** 0.006
ζ	2.8578 *** 0.373	2.9105 *** 0.319	3.0600 *** 0.401	3.1476 *** 0.357	2.2538 *** 0.330	2.4169 *** 0.298	3.1266 *** 0.358	2.9550 *** 0.331	3.1263 *** 0.418	3.1365 *** 0.363
η	-0.8883 *** 0.266	-0.6302 *** 0.208	-1.2451 *** 0.336	-0.7761 *** 0.245	-0.9096 *** 0.256	-0.5988 *** 0.175	-1.2078 *** 0.331	-1.1648 *** 0.263	-1.1474 *** 0.271	-0.7660 *** 0.249
λ	0.0980 *** 0.033	0.0417 *** 0.015	0.0847 *** 0.027	0.0303 ** 0.015	0.1887 *** 0.071	0.0303 * 0.018	0.0855 *** 0.023	0.0606 ** 0.025	0.0823 *** 0.029	0.0469 *** 0.017
ρ	.	0.6795 *** 0.115	.	0.7524 *** 0.143	.	0.8908 *** 0.065	.	0.4408 ** 0.190	.	0.5756 *** 0.102
γ	.	0.7373 *** 0.196	.	0.6004 *** 0.174	.	0.4932 *** 0.178	.	0.4496 *** 0.149	.	0.8221 *** 0.188

Results: GARCH(1,1) Process for $\hat{\lambda}_t$

Table: Univariate GARCH Model on Conditional Jump Intensities

	WTI	LLS	Mars	Mex. Maya	Lou. Heavy
c_0	0.0889 *** 0.0007	0.1001 *** 0.0007	0.2065 *** 0.0009	0.1006 *** 0.0007	0.0886 *** 0.0006
α_0	0.0038 *** 0.0001	0.0057 *** 0.0001	0.0040 *** 0.0001	0.0031 *** 0.0000	0.0064 *** 0.0000
 α_1	0.8294 *** 0.0260	0.9402 *** 0.0271	0.7780 *** 0.0267	0.6838 *** 0.0188	1.1734 *** 0.0293
 β_1	0.1022 *** 0.0114	0.0652 *** 0.0084	0.1474 *** 0.0181	0.0320 *** 0.0044	.
L	5070.6	4354.8	3345.4	6295.4	4418.7
AIC	-10133.2	-8701.7	-6682.8	-12582.8	-8831.5

Note: The standard errors are presented below the coefficients. L is the log-likelihood function. *, **, and *** represent the 10%, 5% and 1% level of significance, respectively. n = 4760

Results: Multivariate GARCH model; CCC

		CCC Model							
		WTI		WTI		WTI		WTI	
Coefficient	$i =$ $j =$	LLS		Mars		Maya		Lou Heavy	
GARCH Model Parameter Estimates									
ρ		0.8492 ***		0.8088 ***		0.7493 ***		0.8093 ***	
		0.004		0.005		0.006		0.005	
C_{11}		0.0041 ***		0.0031 ***		0.0027 ***		0.0037 ***	
		0.000		0.000		0.000		0.000	
C_{22}		0.0059 ***		0.0020 ***		0.0031 ***		0.0063 ***	
		0.000		0.000		0.000		0.000	
α_{11}		0.6292 ***		0.4589 ***		0.2942 ***		0.5264 ***	
		0.059		0.043		0.030		0.052	
α_{22}		0.6766 ***		0.1574 ***		0.3050 ***		0.9369 ***	
		0.062		0.018		0.051		0.075	
β_{11}		0.0792 ***		0.2673 ***		0.3775 ***		0.1468 ***	
		0.018		0.030		0.030		0.029	
β_{22}		0.0752 ***		0.5983 ***		0.1082 ***		0.0226 ***	
		0.020		0.044		0.035		0.008	



Implications of Multivariate GARCH model

- The ARJI models fits the data well.
 - Jumps are statistically important in the crude oil markets
- Time varying volatility found in the $\hat{\lambda}_t$ series.
- Based on the results of the CCC estimates, there appears to be a significant degree of correlation among the contemporaneous shocks to the jump intensities, in the series under study.

Future Work

- The CCC garch assumption of time invariant conditional correlation matrix is restrictive.
 - Dynamic conditional correlation (DCC) model, BEKK specification
- Structural breaks
 - The shale oil revolution, experienced in the US, could plausibly be responsible for a break in the models or relationships.
- Breadth of coverage
 - Bakken prices, Canadian prices, other regional prices