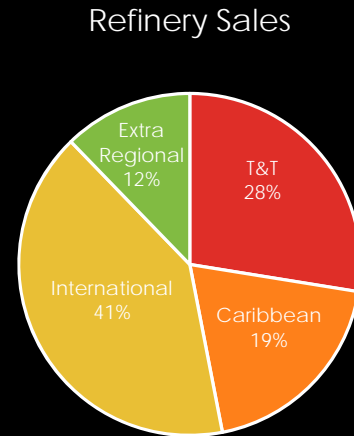
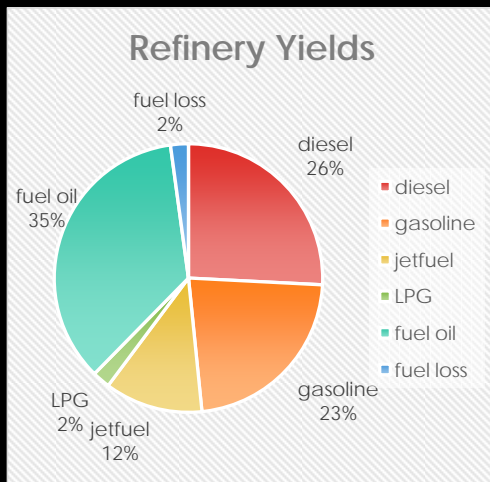


EXPORT PRICE VOLATILITY OF REFINED PETROLEUM PRODUCTS FROM A SMALL HYDROCARBON BASED ECONOMY

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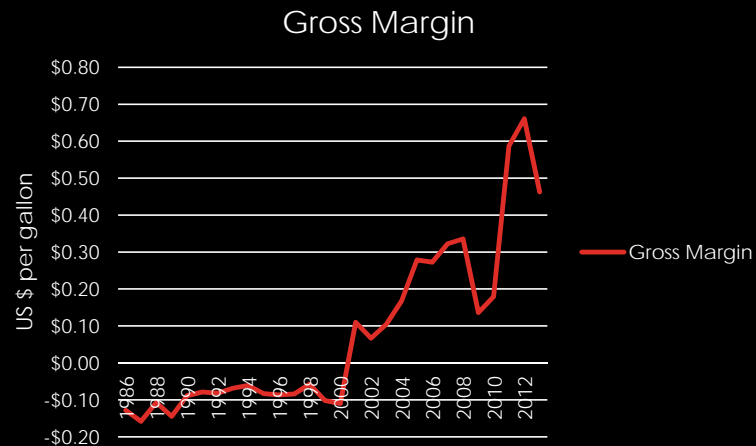
EXPORT PRICE VOLATILITY OF REFINED PETROLEUM PRODUCTS FROM A SMALL HYDROCARBON BASED ECONOMY

- Trinidad and Tobago (T&T) has 1 oil refinery situated at Point-a-Pierre. Crude oil is refined into the following products: : gasoline, diesel, fuel oil, jet fuel, and Liquid Petroleum Gas (LPG).



EXPORT PRICE VOLATILITY OF REFINED PETROLEUM PRODUCTS FROM A SMALL HYDROCARBON BASED ECONOMY

- The Gross margin is the weighted average price of refinery fractions less the price of crude oil. Within the recent years, the refinery margins have begun to decline. This has negatively impacted the profitability of Petrotrin's refinery.



The profitability of the refinery is of interest to T&T because: provide employment for 5,000 persons, sources services from 2,000 registered contractors, maintain roads and bridges that connects to its operations, and implements a number of positive externality projects

EXPORT PRICE VOLATILITY OF REFINED PETROLEUM PRODUCTS FROM A SMALL HYDROCARBON BASED ECONOMY

- Since refinery margins are influenced by the prices of the refinery fractions this study seeks to model the volatility of refinery fraction prices.
- Crude oil prices and refinery fraction prices are volatile. Volatility is the tendency of a series to fluctuate.
- Volatility is a also measure of risk. It is the tendency for the price of a commodity to deviate from its mean price. We are concerned about volatility because it measures the tendency of the price of refinery fractions to decline and thus refinery margins would get smaller.

EXPORT PRICE VOLATILITY OF REFINED PETROLEUM PRODUCTS FROM A SMALL HYDROCARBON BASED ECONOMY

- Objective: This study seeks to study the volatility of prices of the following commodities: gasoline, diesel, heating oil, and jet fuel. This study is distinguished from other studies by:
 - The length of the study and the utilization of the most recent data;
 - The utilization of both real and nominal data, as previous research neglected real data;
 - The consideration of a pre & post financial crisis (US Shale boom era);
 - Including oil prices in the mean equation;
 - Testing for volatility spillover for diesel, gasoline, heating oil and jet fuel;
 - Implications of volatility for a small country refinery

DATA

- Source of data:
- Both nominal and real data on the energy commodities are collected from the US Energy Information Administration (EIA). Weekly nominal data is collected from 19th April, 1996 to 1st August, 2014. Such period is used because data was available for all variables within such period. Weekly data is used over annual data since it will produce 955 observations per variable.
- Real data is collected over the period January 1979 to August 2014. The monthly data generates 428 observations per variable. Weekly data was preferred but an appropriate deflator could not be found for weekly data.

METHODOLOGY

- Test are performed for stationarity. The ADF, PP and KPSS test are used. Since all variables (gasoline, diesel, heating oil and jet fuel) were all I(1) they were converted into continuously compounded returns to make them stationary.
- $R_t = \log\left(\frac{P_t}{P_{t-1}}\right)$
- All series were also tested for ARCH effects to determine if GARCH type models were appropriate.
- GARCH and the EGARCH models were considered to model each refinery fraction prices
- A GARCH model is a model that simulates the conditional variance of a time series. It models variance as a function of the squared residual of the mean equation, and the lagged variance

METHODOLOGY

- For both GARCH and EGARCH a mean equation is specified. The mean equation was $P_t = \beta_1 WTI + \varepsilon_t$. WTI was used as a proxy for oil price. This was to allow oil prices to affect refinery fraction prices in the mean equation.

| Model | Equation |
|---------------|---|
| ARCH (1) | $\sigma^2 = \alpha_0 + \sum_{i=1}^q \alpha_i u_{t-i}^2$ |
| GARCH (1, 1) | $\sigma^2 = \alpha_0 + \alpha_1 u_{t-1}^2 + \beta \sigma_{t-1}^2$ |
| TARCH (1, 1) | $\sigma^2 = \alpha_0 + \alpha_1 u_{t-1}^2 + \beta \sigma_{t-1}^2 + \gamma u_{t-1}^2 I_{t-1}$ |
| EGARCH (1, 1) | $\log(\sigma^2) = \omega + \sum_{i=1}^q \alpha_i \left \frac{u_{t-i}}{\sigma_{t-i}} \right + \sum_{k=1}^r \lambda_k \frac{u_{t-k}}{\sigma_{t-k}} + \sum_{j=1}^p \beta_j \log(\sigma_{t-j}^2)$ |

The EGARCH model was used to test for asymmetric effects. The sign (leverage) effect is if negative shocks increase volatility more than positive shocks. The size effect is if large shocks increase volatility more than small shocks.

DIAGNOSTIC TESTS

- The MSE is given by:

$$\bullet \text{MSE} = \frac{1}{T-(T_1-1)} \sum_{t=T}^T (y_{t+s} - f_{t,s})^2$$

(11)

- The MAE is given by:

$$\bullet \text{MSE} = \frac{1}{T-(T_1-1)} \sum_{t=T}^T |y_{t+s} - f_{t,s}|$$

(12)

- The Theil U Statistic is given by:

$$\bullet U = \frac{\sqrt{\sum_{t=T}^T \left(\frac{y_{t+s} - f_{t,s}}{y_{t,s}} \right)^2}}{\sqrt{\sum_{t=T}^T \left(\frac{y_{t+s} - fb_{t,s}}{y_{t+s}} \right)^2}}$$

The smaller the value of the MSE and the MAE, the better the predictive accuracy

As U tends to zero, the better the predictive accuracy, and as U tends to 1, the more unreliable the predictive accuracy

RESULTS

Table 3: Stationary results

| Nominal data | | Weekly from 19 th April 1996 to 1 st August 2014 | | | | |
|--------------|-----------|--|----------|-------------------------|-----------|--------------------------|
| Variable | ADF level | ADF 1 st Diff | PP Level | PP 1 st diff | KPS level | KPS 1 st diff |
| Ln gasoline | 0.4876 | 0.0000 | 0.6184 | 0.0000 | 3.527408 | 0.031156 |
| Ln diesel | 0.7088 | 0.0000 | 0.7172 | 0.0000 | 3.467929 | 0.058297 |
| Ln heat oil | 0.7562 | 0.0000 | 0.7957 | 0.0000 | 3.529947 | 0.053852 |
| Ln jet fuel | 0.6850 | 0.0000 | 0.7288 | 0.0000 | 3.519684 | 0.047815 |
| Ln lpg | 0.3751 | 0.0000 | 0.3945 | 0.0000 | 3.004895 | 0.047909 |
| Ln WTI | 0.7026 | 0.0000 | 0.7753 | 0.0000 | 3.540030 | 0.053420 |
| Real data | | Monthly from January 1979 to August 2014 | | | | |
| Ln gasoline | 0.3826 | 0.0000 | 0.5671 | 0.0000 | 0.683519 | 0.144884 |
| Ln diesel | 0.3010 | 0.0000 | 0.3722 | 0.0000 | 0.587386 | 0.118427 |
| Ln heat oil | 0.4860 | 0.0000 | 0.6271 | 0.0000 | 0.802313 | 0.136739 |
| Ln WTI | 0.2331 | 0.0000 | 0.4471 | 0.0000 | 0.673827 | 0.112796 |

All variables are I(1)

Source: Computed

RESULTS

Table 4: Normality results

| Variable | Nominal data results | Real data results | Variable | Nominal data results | Real data results |
|------------------|--|--|---------------------|--|---|
| Gasoline returns | JB 485.83 prob. 0.0000 Kurtosis 6.49 Skewness -0.01 | JB 1019.74 prob. 0.0000 Kurtosis 10.28 Skewness -1.02 | Heating oil returns | JB 2767.24 prob. 0.0000 Kurtosis 11.33 Skewness 0.13 | JB 1526.62 prob. 0.0000 Kurtosis 12.09 Skewness 0.88 |
| Diesel returns | JB 307.68 prob. 0.0000 Kurtosis 5.76 Skewness 0.17 | JB 183.05 prob. 0.0000 Kurtosis 6.11 Skewness -0.39 | Propane returns | JB 4991.09 prob. 0.0000 Kurtosis 14.18 Skewness -0.34 | |
| Jet fuel returns | JB 220.7 prob. 0.0000 Kurtosis 5.33 Skewness -0.14 | | WTI returns | JB 402.65 prob. 0.0000 Kurtosis 6.12 Skewness -0.31 | JB 379.21 prob. 0.0000 Kurtosis 7.45 Skewness -0.61 |

Source: Computed

Low JB probabilities result in the rejection of the null hypothesis that the series are normally distributed. Thus all variables are not Gaussian distributed. Thus ARCH type modeling is appropriate.

RESULTS

Table 5: ARCH effects results

| Variable | Test Statistic | Nominal data | Real data |
|-----------------|---------------------|--------------|-----------|
| Diesel returns | Prob. F(7,939) | 0.0000 | 0.0001 |
| | Prob. Chi-Square(7) | 0.0000 | 0.0001 |
| Log gasoline | Prob. F(7,939) | 0.0000 | 0.0000 |
| | Prob. Chi-Square(7) | 0.0000 | 0.0000 |
| Log jet fuel | Prob. F(7,939) | 0.0000 | |
| | Prob. Chi-Square(7) | 0.0000 | |
| Log heating oil | Prob. F(7,939) | 0.0002 | 0.0000 |
| | Prob. Chi-Square(7) | 0.0002 | 0.0000 |
| Log propane | Prob. F(7,939) | 0.0000 | |
| | Prob. Chi-Square(7) | 0.0000 | |
| Log WTI | Prob. F(7,939) | 0.0000 | 0.0000 |
| | Prob. Chi-Square(7) | 0.0000 | 0.0000 |

Low probabilities result in the rejection of the null of no ARCH effects are present. Thus GARCH type modeling is appropriate.

Table 6: GARCH and EGARCH results

| GARCH | Nominal diesel | Coefficient | Std. Error | z-Statistic | Prob. |
|--------------|-----------------------------|---------------|---------------|---------------|---------------|
| Mean Eq. | RW | 0.7028 | 0.0200 | 35.1294 | 0.0000 |
| Variance Eq. | C | 0.0000 | 0.0000 | 2.5014 | 0.0124 |
| | ARCH(-1) | 0.2054 | 0.0407 | 5.0502 | 0.0000 |
| | GARCH(-1) | 0.7991 | 0.0292 | 27.3702 | 0.0000 |
| EGARCH | Nominal diesel | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.7051 | 0.0186 | 37.8089 | 0.0000 |
| Variance Eq. | C(2) constant | -0.5589 | 0.1113 | -5.0192 | 0.0000 |
| | C(3) size effect | 0.3718 | 0.0542 | 6.8631 | 0.0000 |
| | C(4) leverage effect | 0.0834 | 0.0335 | 2.4880 | 0.0128 |
| | C(5) GARCH term | 0.9589 | 0.0137 | 69.9322 | 0.0000 |
| GARCH | Nominal gasoline | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.8378 | 0.0238 | 35.1395 | 0.0000 |
| Variance Eq. | C | 0.0001 | 0.0000 | 3.2345 | 0.0012 |
| | ARCH(-1) | 0.2372 | 0.0460 | 5.1549 | 0.0000 |
| | GARCH(-1) | 0.7094 | 0.0464 | 15.2734 | 0.0000 |
| EGARCH | Nominal gasoline | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.8569 | 0.0227 | 37.8288 | 0.0000 |
| Variance Eq. | C(2) constant | -0.7742 | 0.1734 | -4.4636 | 0.0000 |
| | C(3) size effect | 0.3393 | 0.0568 | 5.9707 | 0.0000 |
| | C(4) leverage effect | 0.1320 | 0.0363 | 3.6342 | 0.0003 |
| | C(5) GARCH term | 0.9240 | 0.0227 | 40.7086 | 0.0000 |
| GARCH | Nominal heat oil | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.8113 | 0.0153 | 52.9572 | 0.0000 |
| Variance Eq. | C | 0.0001 | 0.0000 | 3.7531 | 0.0002 |
| | ARCH(-1) | 0.1977 | 0.0401 | 4.9355 | 0.0000 |
| | GARCH(-1) | 0.7309 | 0.0410 | 17.8254 | 0.0000 |
| EGARCH | Nominal heat oil | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.8184 | 0.0150 | 54.6137 | 0.0000 |
| Variance Eq. | C(2) constant | -0.7667 | 0.1678 | -4.5697 | 0.0000 |
| | C(3) size effect | 0.3329 | 0.0512 | 6.5008 | 0.0000 |
| | C(4) leverage effect | 0.0717 | 0.0343 | 2.0929 | 0.0364 |
| | C(5) GARCH term | 0.9319 | 0.0196 | 47.5188 | 0.0000 |
| GARCH | Nominal jet fuel | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.7837 | 0.0173 | 45.3244 | 0.0000 |
| Variance Eq. | C | 0.0000 | 0.0000 | 3.4583 | 0.0005 |
| | ARCH(-1) | 0.2249 | 0.0431 | 5.2190 | 0.0000 |
| | GARCH(-1) | 0.7328 | 0.0350 | 20.9512 | 0.0000 |
| EGARCH | Nominal jet fuel | Coefficient | Std. Error | z-Statistic | Prob. |
| Mean Eq. | RW | 0.7835 | 0.0170 | 46.1252 | 0.0000 |
| Variance Eq. | C(2) constant | -0.8047 | 0.1633 | -4.9290 | 0.0000 |
| | C(3) size effect | 0.3892 | 0.0554 | 7.0297 | 0.0000 |
| | C(4) leverage effect | 0.0490 | 0.0365 | 1.3446 | 0.1788 |
| | C(5) GARCH term | 0.9295 | 0.0201 | 46.3412 | 0.0000 |

RESULTS

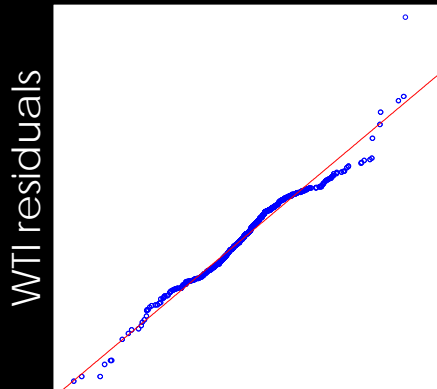
In all models, WTI is significant in the mean equation, suggesting oil prices affect refinery fraction prices.

The leverage effect for gasoline, diesel and heating oil were positive and significant. Thus negative shocks increase their volatility more than positive shocks. The leverage effect for jet fuel was insignificant, suggesting symmetric in sign of shock effects.

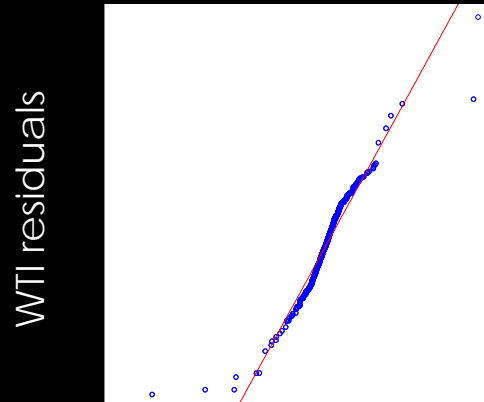
For all models the GARCH term is large and over 0.71 suggesting the persistence of shocks.

DIAGNOSTIC TESTS

Nominal data QQ plot of diesel residuals

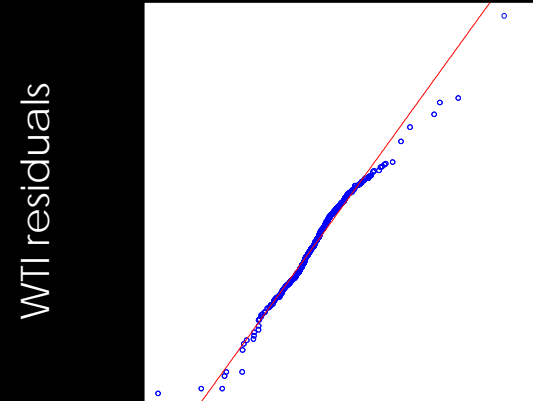


Diesel residuals



Heating oil residuals

QQ plot of gasoline residuals



Gasoline residuals

For all series, the refinery fraction residuals matched WTI residuals, however, positive and negative shocks caused the occasional deviation of a few data points

The QQ plot shows how the distribution of 1 series matches the distribution of another series.

RESULTS

Table 7: Volatility spillover effects.

| Equation | Probability | Equation | Probability |
|----------------------|-------------|----------------------|-------------|
| Diesel-gasoline | 0.0279 | Gasoline-Diesel | 0.0249 |
| Diesel-heating oil | 0.0000 | Gasoline-heating oil | 0.0048 |
| Diesel-jet fuel | 0.0000 | Gasoline-jet fuel | 0.0000 |
| Diesel-WTI | 0.0025 | Gasoline-WTI | 0.0767 |
| Heating oil-diesel | 0.0000 | Jet fuel-Diesel | 0.0000 |
| Heating oil-gasoline | 0.0113 | Jet fuel-gasoline | 0.0000 |
| Heating oil-jet fuel | 0.0000 | Jet fuel-heating oil | 0.0000 |
| Heating oil-WTI | 0.0001 | Jet fuel-WTI | 0.0023 |

For each series, the variance term was significant in the volatility spill over equation. This suggested volatility spillover effects. E.g. It means that the volatility of diesel prices affects the volatility of gasoline prices.

DIAGNOSTIC RESULTS

| Nominal data | diesel | gasoline | Heating oil | Jet fuel |
|------------------|---------------|---------------|---------------|---------------|
| GARCH (1,1) | MSE: 0.0393 | MSE: 0.0418 | MSE: 0.0316 | MSE: 0.0315 |
| | MAE: 0.0272 | MAE: 0.0282 | MAE: 0.0192 | MAE: 0.0204 |
| | Thiel: 0.5156 | Thiel: 0.4679 | Thiel: 0.4024 | Thiel: 0.4109 |
| TARCH (1,1) | MSE: 0.0393 | MSE: 0.0418 | MSE: 0.0316 | MSE: 0.0315 |
| | MAE: 0.0272 | MAE: 0.0282 | MAE: 0.0192 | MAE: 0.0204 |
| | Thiel: 0.5152 | Thiel: 0.4656 | Thiel: 0.4021 | Thiel: 0.4104 |
| EGARCH(1,1) | MSE: 0.0393 | MSE: 0.0418 | MSE: 0.0316 | MSE: 0.0315 |
| | MAE: 0.0272 | MAE: 0.0282 | MAE: 0.0192 | MAE: 0.0204 |
| | Thiel: 0.5151 | Thiel: 0.4642 | Thiel: 0.4014 | Thiel: 0.4109 |
| Real data | | | | |
| GARCH (1,1) | MSE: 0.0403 | MSE: 0.0369 | MSE: 0.0390 | |
| | MAE: 0.0289 | MAE: 0.0265 | MAE: 0.0248 | |
| | Thiel: 0.2638 | Thiel: 0.4957 | Thiel: 0.6615 | |
| TARCH (1,1) | MSE: 0.0403 | MSE: 0.0369 | MSE: 0.0390 | |
| | MAE: 0.0289 | MAE: 0.0265 | MAE: 0.0248 | |
| | Thiel: 0.2639 | Thiel: 0.4951 | Thiel: 0.6615 | |
| EGARCH(1,1) | MSE: 0.0403 | MSE: 0.0369 | MSE: 0.0388 | |
| | MAE: 0.0289 | MAE: 0.0265 | MAE: 0.0246 | |
| | Thiel: 0.2638 | Thiel: 0.4902 | Thiel: 0.6381 | |

Low values of the MSE and the MAE suggested relatively good predictive accuracy

POLICY RECOMMENDATIONS

- In summary, gasoline, diesel, heating oil and jet fuel were all $I(1)$. Their normality test results indicated that each series was not normally distributed. ARCH effects test results all indicated a presence of ARCH effects, thus GARCH type models were appropriate for modeling.
- There was volatility spillover between all series. This suggests that each refinery fraction's volatility is affected by the volatility of the other refinery fractions. Given that refinery margins may be slim in the future, refiners need to take a number of steps to ensure their survival.

POLICY RECOMMENDATIONS

- A key factor to ensure the survival of refineries outside of the US is the controlling of costs. Such refineries need to cut their operational costs and find more to do with less. Refinery firms may redesign their core processes and automation, try to minimize shut in productions, better organization of shifts for workers, and concentrating on work that drives performance.
- The local refinery should try to increase oil production and use more locally produced crude as this will decrease refinery costs.
- The refinery firm can consider hedging with forwards contracts for both crude oil and refinery fractions. This is where they enter a contract today to pay an agreed price of oil, or get an agreed price of refinery products, but the delivery of the commodity will be at a future date.

CONCLUSION

- Thank you