

Market Efficiency of Oil Spot and Futures: A Stochastic Dominance Approach

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Abstract

This paper examines the market efficiency of oil spot and futures prices by using a stochastic dominance (SD) approach. As there is no evidence of an SD relationship between oil spot and futures, we conclude that there is no arbitrage opportunity between these two markets, and that both market efficiency and market rationality are not rejected in the oil spot and futures markets.

Keywords: Stochastic dominance, risk averter, risk seeker, futures market, spot market.

JEL Classifications: C14, G12, G15.

1. Introduction

Crude oil is an important commodity for the world economy. With the increasing fluctuations and tension of crude oil prices, oil futures have become one of the popular derivatives to hedge the risk of oil price hikes or crashes. Spot and futures prices of oil have been investigated over an extended period. Substantial research has been undertaken to analyze the relationship between spot and futures prices, and their associated returns. The efficient market hypothesis is crucial for understanding optimal decision-making with regard to hedging and speculation. It is also important for making financial decisions about the optimal allocation of portfolios of assets with regard to their multivariate returns and associated risks.

Research on the relationships between spot and futures prices of petroleum products has examined issues such as market efficiency and price discovery. Bopp and Sitzer (1987) find that futures prices have a significant positive contribution to past price changes, even when crude oil prices, inventory levels, weather, and other important variables are accounted for. Serletis and Banack (1990) use daily data for spot, two-month futures crude oil prices, and prices of gasoline and heating oil traded on the New York Mercantile Exchange (NYMEX), to test market efficiency, and they find evidence in support of the market efficiency hypothesis. In addition, Crowder and Hamid (1993) use co-integration analysis to test the simple efficiency hypothesis and the arbitrage condition for crude oil futures. Their results support the simple efficiency hypothesis that the expected returns from futures speculation in the oil futures market are zero.

Studies conducted during different time periods also provide insight. Between 1990 and 2000, Taback (2003) tests whether Brent spot and futures prices contain a unit root, and finds that both spot prices and futures prices are non-stationary. During the period 1989-2003, Coimbra and Esteves (2004) test the stationarity of Brent crude oil spot and futures prices which omit the impact

of the Gulf war from January 1992 to December 2003. For both of these time periods, the null hypothesis of a unit root in crude oil prices cannot be rejected.

Postali and Picchetti (2006) apply unit root tests to international oil prices. They find that the traditional unit root tests reject the unit root null for the entire sample of more than one century of annual data. Recently, Maslyuk and Smyth (2008) employ LM unit root tests with one and two structural breaks to reveal that oil spot and futures markets are efficient in the weak form. Their result suggests that future spot and futures prices cannot be predicted on the basis of past prices.

Examining the price discovery process for the crude oil market using monthly data, Quan (1992) finds that the futures price does not play an important role in this process. Using daily data from NYMEX closing futures prices, Schwartz and Szakmary (1994) find that futures prices strongly dominate in the price discovery process relative to deliverable spots in all three petroleum markets. In addition, applying cointegration tests in a series of oil markets with pairwise comparisons on post-1990 data, Gulen (1999) concludes that oil markets have grown more unified during the period of 1994-1996 as compared with 1991-1994.

Silvapulle and Moosa (1999) examine the daily spot and futures prices of WTI crude by using both linear and non-linear causality testing. They find that linear causality testing reveals that futures prices lead spot prices, whereas non-linear causality testing reveals a bi-directional effect. Bekiros and Diks (2008) test the existence of linear and nonlinear causal lead-lag relationships between spot and futures prices of West Texas Intermediate (WTI) crude oil. They discover strong bi-directional Granger causality between spot and futures prices, and that the pattern of leads and lags changes over time.

Lin and Tamvakis (2001) investigate information transmissions between the NYMEX and London's International Petroleum Exchange, and find that NYMEX is a true leader in the crude oil

market. Investigating information transmissions among NYMEX WTI crude prices, NYMEX gasoline prices, NYMEX heating oil prices, and among international gasoline spot markets, including the Rotterdam and Singapore markets, Hammoudeh et al. (2003) conclude that the NYMEX gasoline market is the true leader. In addition, Hammoudeh and Li (2004) show that the NYMEX gasoline price is the gasoline leader in both pre- and post-Asian crisis periods.

Empirical studies indicate that commodity prices can be extremely volatile at times, and sudden changes in volatility are quite common in commodity markets. For example, using an iterative cumulative sum-of-squares approach, Wilson et al. (1996) document sudden changes in the unconditional variance in daily returns on one-month through six-month oil futures and relate these changes to exogenous shocks, such as unusual weather, political conflicts and changes in OPEC oil policies. Fong and See (2002) conclude that regime switching models provide a useful framework for studying factors behind the evolution of volatility and short-term volatility forecasts. In addition, Fong and See (2003) show that the regime switching model outperforms the standard GARCH model on all commonly-used evaluation criteria for short-term volatility forecasts.

Most of the existing literature has employed conventional parametric tests, such as mean-variance (MV) criterion and CAPM statistics. These approaches are derived under the assumptions of a von Neumann-Morgenstern (1944) quadratic utility function and returns being normally distributed (Feldstein, 1969; Hanoch and Levy, 1969). Thus, the reliability of performance comparisons using the MV criterion and CAPM analysis depends on the degree of non-normality of the returns data and the nature of the (non-quadratic) utility functions (Beedles, 1979; Schwert, 1990; Fung and Hsieh, 1999).

The stochastic dominance (SD) approach differs from conventional parametric approaches in comparing the performance of different prospects. It endorses the minimum assumptions on

investors' utility functions. The advantage of SD analysis over parametric tests becomes apparent when the asset returns distributions are non-normal. As the SD approach does not require any assumption about the nature of the distributions, it can be used for any type of distribution. In addition, SD rules offer superior criteria on prospects investment decisions since SD incorporates information on the entire returns distribution, rather than just the first two moments, as are used in the MV and CAPM methodologies. The SD approach has been regarded as one of the most useful tools to rank investment prospects (see, for example, Levy 1992) as the ranking of the assets has been shown to be equivalent to utility maximization for the preferences of risk averters and risk seekers (Tsefatsion, 1976; Stoyan, 1983; Li and Wong, 1999).

Consider a utility-maximizing investor who holds a portfolio of two assets, namely oil spot and oil futures. The objective is to rank preferences of these two assets to maximize expected wealth and/or expected utility. In this paper, we use the SD test proposed by Linton et al. (2005) to investigate the characteristics of the entire distributions of oil futures and spot returns, rather than considering only the mean and standard deviation, as are used in much of the existing literature.

This paper contributes to the energy economics literature in several ways. This is the first paper that discusses oil prices from the investors' perspective using the SD approach. Second, a more robust decision tool is used for investment decisions under uncertainty to the oil spot and futures markets. Third, greater information and inferences on investors' behavior can be made, including the identification of any arbitrage opportunities in these markets, tests of market efficiency and market rationality in these markets, and an examination of the preferences of risk averters in these markets. Finally, we examine the impacts of OPEC's decision on reduction of production capacity in 1999, the effects of the 2003 Iraq War on these markets, and the diversification effects on these markets.

2. Data and Methodology

We examine the efficiency of the spot-futures market by investigating the SD relationship between oil spot and its futures for the period January 1, 1989 to June 30, 2008. We first investigate the daily closing prices of Brent Crude oil spot and futures with one month maturity, which are obtained from *Datastream*. As it is well known (see, for example, Ripple and Moosa (2005, 2007) and Serletis (1992)) that different maturities have an impact on market investment, hedging, efficiency and predictability, we will analyse the spot-future relationship for different maturities. However, because the data for Brent Crude oil futures with other maturities is not available from the same data source, we collect the WTI spot prices together with its futures at maturities of 1, 2, 3 and 4 months from the Energy Information Administration for the same sample period, and analyze their relationships as a complement to the Brent Crude data to check the effects of different maturities.

As is standard, the daily log returns, $R_{i,t}$, for the oil spot and futures prices are defined as $R_{i,t} = \ln(P_{i,t} / P_{i,t-1})$, where $P_{i,t}$ is the daily price at day t for asset i , with $i = S$ (spot) and F (futures), respectively. We further examine the effects of two major oil crises (OPEC's decision on reduction of capacity in 1999 and the 2003 Iraq War) by examining two pairs of sub-periods. The first pair of sub-periods is the pre-OPEC sub-period (Pre-OPEC) and the sub-period thereafter (OPEC), using October 29, 1999 as a cut-off point, while the second pair of sub-periods is the pre-Iraq-War sub-period (pre-Iraq War) and the sub-period thereafter (Iraq War), using March 20, 2003 as the cut-off point.¹

We display Figure 1 for the plots of Brent Crude oil spot and futures prices with the corresponding cut-off points, and Figure 2 for the plots of WTI spot and futures prices with the corresponding cut-off points. The plots show that these markets could be efficient. In order to test this claim formally, we further analyse their relationship by the mean-variance criterion, CAPM statistics, and the stochastic dominance approach. For computing the CAPM statistics, we use the

¹ We have examined other crises. Their effects on oil are similar to OPEC's decision and the 2003 Iraq War, but the magnitudes of their effects are less significant. Since OPEC's decision and the 2003 Iraq War are more strongly related to oil markets, the effects of only these crises are analysed in this paper.

3-month U.S. T-bill rate and the Morgan Stanley Capital International index (MSCI) to approximate the risk-free rate and the global market index, respectively.

2.1 Mean-Variance criterion and CAPM statistics

For comparative purposes, we first apply the MV and CAPM statistics to analyse the data. The MV model developed by Markowitz (1952) and Tobin (1958), and the CAPM statistics developed by Sharpe (1964), Treynor (1965) and Jensen (1969), are commonly used to compare investment prospects.² For any two investment prospects, with variables of returns Y_i and Y_j , means μ_i and μ_j , and standard deviations σ_i and σ_j , respectively, Y_j is said to dominate Y_i by the MV rule if $\mu_j \geq \mu_i$ and $\sigma_j \leq \sigma_i$ significantly (Markowitz, 1952; Tobin, 1958; Wong, 2007). CAPM statistics include the beta, Sharpe ratio, Treynor's index and Jensen (alpha) index to compare the performance of different prospects³.

2.2 Stochastic Dominance Test

The stochastic dominance (SD) theory, initially developed by Hadar and Russell (1969), Hanoch and Levy (1969) and Rothschild and Stiglitz (1970), is one of the most useful tools in investment decision-making under uncertainty to rank investment prospects. Let X and Y represent spot and futures, respectively, defined on the common support $[a, b]$, where $a < b$ with their cumulative

² We note that recently Leung and Wong (2008) have developed a multivariate Sharpe ratio statistic to test the hypothesis of the equality of multiple Sharpe ratios, whereas Bai et al. (2009a,b) have developed new bootstrap-corrected estimators of the optimal returns for the Markowitz mean-variance optimization.

³ The formulae for the Sharpe ratio, Treynor index, and Jensen index are $S_i = \frac{R_i - R_f}{\sigma_i}$, $T_i = \frac{R_i - R_f}{\beta_i}$, and $J_i = \alpha_i = (R_i - R_f) - \beta_i(R_m - R_f)$, respectively (see Sharpe (1964), Treynor (1965) and Jensen (1969) for further information on these statistics).

distribution functions (CDFs), F and G , and their corresponding probability density functions (PDFs), f and g , respectively. We define⁴

$$H_0 = h, H_j(x) = \int_a^x H_{j-1}(t) dt \quad (1)$$

for $h = f, g$; $H = F, G$; and $j = 1, 2, 3$. We call the integral H_j the j^{th} order cumulative distribution function (CDF).

The most commonly used SD rules that correspond with three broadly defined utility functions are first-, second- and third-order SD, denoted as FSD, SSD and TSD, respectively. All investors are non-satiated (that is, they prefer more to less) under FSD, non-satiated and risk-averse under SSD; and non-satiated, risk-averse and possessing decreasing absolute risk aversion (DARA) under TSD. We define the SD rules as follows (see Quirk and Saposnik, 1962; Fishburn, 1964; Hanoch and Levy, 1969; Sriboonchita, et al., 2009):

X dominates Y by FSD (SSD, TSD), denoted by $X \succ_1 Y$ ($X \succ_2 Y$, $X \succ_3 Y$) if and only if $F_1(x) \leq G_1(x)$ ($F_2(x) \leq G_2(x)$, $F_3(x) \leq G_3(x)$) for all possible returns x , and the strict inequality holds for at least one value of x .

The theory of SD is important as it is related to utility maximization (Quirk and Saposnik 1962, Hanoch and Levy 1969, Li and Wong 1999). The existence of SD implies that risk-averse investors always obtain higher expected utilities when holding dominant assets than when holding dominated assets.⁵ Consequently, dominant assets are preferred by investors. We note that a hierarchical relationship exists in SD: FSD implies SSD, which in turn implies TSD. However, the converse is not true: the existence of SSD does not imply the existence of FSD. Likewise, the existence of TSD does not imply the existence of SSD or FSD. Thus, only the lowest dominance

⁴ See Wong and Chan (2008) for further discussion regarding notation.

⁵ The SD theory could be extended further to satisfy non-expected utilities (see Wong and Ma (2008) and the references contained therein for further details).

order of SD is reported.

Finally, we note that, under certain regularity conditions⁶, investment X stochastically dominates investment Y in first-order, if and only if there is an arbitrage opportunity between X and Y , such that investors will increase their expected wealth, as well as their expected utility, if their investments are shifted from Y to X (Bawa, 1978; Jarrow, 1986; Wong et al., 2008). In this situation, they could make huge profits by setting up zero-dollar portfolios to exploit this opportunity. On the other hand, if FSD does not exist between X and Y , one could conclude that both markets display market efficiency and market rationality (Bernard and Seyhun, 1997; Larsen and Resnick, 1999; Sriboonchita, et al., 2009). We will discuss this issue in detail in the next subsection.

The advantages presented by SD have motivated prior studies using SD techniques to analyze many financial puzzles. There are two broad classes of SD tests: one is the minimum/maximum statistic, while the other is based on distribution values computed on a set of grid points. McFadden (1989) develops a SD test using the minimum/maximum statistic, followed by Klecan et al. (1991) and Kaur et al. (1994). Barrett and Donald (2003) develop a Kolmogorov-Smirnov-type test, and Linton et al. (LMW, 2005) extend their work to relax the iid assumption. On the other hand, the SD tests developed by Anderson (1996, 2004) and Davidson and Duclos (2000) (hereafter DD) compare the underlying distributions at a finite number of grid points. The DD test is found to be one of the most powerful tests (see for example, Lean et al., (2008)), and the LMW test is also found to be efficient. However, the DD test requires the iid assumption for the observations being analysed, whereas the LMW test allows general dependence among the prospects and also non-iid observations. As Tables 2A and 2B show that spot and futures are non-iid for both Brent Crude and WTI spots and futures, we adopt the LMW test in this paper.

⁶ See Jarrow (1986) for the conditions.

The SD test developed by Linton et al. (2005) is based on sub-sampling, and the resulting tests are consistent and powerful against some $N^{-1/2}$ local alternatives. The test statistic is:

$$T_j = \min \sup_x \sqrt{N \left[\hat{F}_j(x) - \hat{G}_j(x) \right]}, \quad \hat{H}_j(x) = \frac{1}{N(j-1)!} \sum_{i=1}^N (x - z_i)_+^{j-1}, \quad H = F, G.$$

The LMW test evaluates the following two sets of null and alternative hypotheses:

$$H_0 : F_j(x_i) \leq G_j(x_i) \text{ for all } x; \text{ and}$$

$$H_1 : F_j(x_i) > G_j(x_i) \text{ for some } x.$$

$$H'_0 : G_j(x_i) \leq F_j(x_i) \text{ for all } x; \text{ and}$$

$$H'_1 : G_j(x_i) > F_j(x_i) \text{ for some } x.$$

The null hypothesis in H_0 states that the spot index dominates the futures index, while the null hypothesis in H'_0 states that the futures index dominates the spot index. The alternative hypothesis is the SD relationship fails at some points. If we do not reject the first H_0 and reject the second H'_0 , this means that spot stochastically dominates futures at the j order. On the other hand, if we reject the first H_0 and do not reject the second H'_0 , this means that futures stochastically dominates spot at the j order. In addition, if we do not reject both H_0 and H'_0 , this says that there is no dominance between spot and futures, and the distributions of spot and futures are not rejected to be the same. Finally, if we reject both H_0 and H'_0 , this suggests that spot does not dominate futures and futures does not dominate spot, but the distributions of spot and futures may not be the same.

2.3. Market Efficiency and Market Rationality

The conventional theory of market efficiency states that a market is considered inefficient and irrational if one is able to earn an abnormal return. Our focus here is how market efficiency and market rationality can be inferred by using SD rules to examine the existence of arbitrage

opportunities, market efficiency and the rationality of investors, without identifying any risk index or specific model. By examining market data, SD answers the following queries: (a) Can investors increase their (expected) wealth by switching their portfolio choice, say from the oil spot to the oil futures or vice-versa? (b) Can risk-averse investors who switch from oil spot to oil futures increase their expected utility?

If all non-satiated investors can switch among their investment choices, say by selling spot and longing futures, and increase their (expected) wealth, then independently of their specific preferences, investors can benefit, and hence we could infer the market to be inefficient and irrational. Jarrow (1986) and Falk and Levy (1989) claim that, if FSD exists, under certain conditions arbitrage opportunities exist, and investors will increase their wealth and expected utility if they shift from holding the dominated asset to the dominant one. On the other hand, Wong et al. (2008) claim that, if FSD exists statistically, arbitrage opportunities may not exist, but investors can increase their expected wealth and expected utility if they shift from holding the dominated asset to the dominant one.

In addition, if the market is not ‘complete,’ even if FSD exists, investors may not be able to exploit any arbitrage opportunities.⁷ If the SD test detects FSD of a particular asset over another, but the dominance only lasts for a short period, the results cannot be used to reject market efficiency or market rationality.⁸ In general, FSD should not last for a very long period of time because market forces induce adjustments to a condition of no FSD if the market is rational and efficient. For example, if oil futures stochastically dominate oil spot at the first order, then investors would buy oil futures and sell oil spot. This will drive up the price of oil futures relative to oil spot until the market price of oil futures relative to oil spot is high enough to make the marginal investor indifferent between them. If new information is either made public quickly or is anticipated, the opportunity to use the new information to earn abnormal returns is of limited value. This idea changes slightly in a world where utility functions and returns distributions are not as severely

⁷ See Jarrow (1986), Wong et al. (2008), and Sriboonchita, et al. (2009) for further discussion.

⁸ See Falk and Levy (1989), Bernard and Seyhun (1997) and Larsen and Resnick (1999) for further discussion.

circumscribed. If the FSD does not last for a long period of time, we infer that the market is still efficient and rational. However, in a situation where the FSD holds for a long period of time and all investors increase their expected wealth by switching their asset choices, the market would be neither efficient nor rational.

On the other hand, Falk and Levy (1989) claim that, given two assets, F and S, if by switching from S to F (or by selling S short and holding F long), an investor can increase expected utility, so that the market is inefficient. SSD does not imply any arbitrage opportunity, but implies the preference of one asset over another by risk-averse investors. For example, if oil futures dominate oil spot by SSD, one would not make an expected profit by switching from spot to futures, but switching would allow risk-averse investors to increase their expected utility. A similar argument can be made for the TSD criterion, which assumes that all investors' utility functions exhibit non-satiation, risk aversion, and decreasing absolute risk aversion (DARA).

If oil futures TSD oil spot, one would not make an expected profit by switching from spot to futures, but switching would allow risk-averse investors with DARA to increase their expected utility. Therefore, one could claim that the market is inefficient if investors are assumed to be risk averse and possess DARA. If no SSD is found in the market containing S and F, this suggests that risk-averse investors are indifferent between S and F, so they will not switch S to F, or vice-versa, to increase their expected utility. In this situation, we claim that the market is rational and efficient. Similarly, if no TSD is found in the market containing S and F, this says that risk-averse investors who possess DARA are indifferent between S and F. In this situation, we claim that the market is both rational and efficient.

3. Empirical Results and Discussion

[Insert Table 1 here]

Table 1 provides the descriptive statistics for the daily returns of both oil spot and futures prices for the entire sample period. Panel A shows that the mean of Brent Crude spot daily returns is slightly higher than that of futures, whereas the standard deviation of Brent Crude futures daily returns is slightly smaller than that of futures, implying the Brent Crude spot dominates its futures according to the mean-variance criterion. On the other hand, Panel B shows the reverse result that the daily returns of WTI oil futures have a higher mean and smaller standard deviation than those of WTI oil spot, especially for longer maturity, implying WTI oil futures dominate their spot according to the mean-variance criterion, especially for longer maturity. However, the unreported paired t tests reveal that the mean differences of the spot returns and their corresponding futures returns are insignificant, while the F statistic shows that the standard deviations of the spot returns and their corresponding futures returns are also insignificant. These results indicate that the mean-variance criterion does not imply any dominance between spot and futures for Brent Crude and WTI.

For the CAPM measures, all betas are negative and are less than one in absolute value. The magnitude of the beta of Brent Crude oil spot returns is smaller than that of futures. Based on the annualized Sharpe ratio, the Brent Crude oil spot outperforms its futures, while the WTI futures outperform spot, especially for longer maturity. However, the Sharpe ratio test (Leung and Wong, 2008) shows that their differences are insignificant. Similarly, unreported test statistics reveal that both the Treynor and Jensen indices of the spots and their corresponding futures are insignificant for both Brent Crude and WTI, suggesting that the CAPM statistics do not demonstrate any preference between the spot and futures markets. The inference drawn from the MV and CAPM statistics suggests that the spot and futures markets are efficient and rational.

[Insert Tables 2 and 3 here]

However, so far there is no strong linkage between market efficiency and the inferences drawn from MV and CAPM. In order to obtain a more accurate inference, we use the stochastic

dominance (SD) approach to examine the spot and futures markets. The results of the Ljung-Box statistics based on levels and squared levels of returns of spot and futures displayed in Table 2A, and the results of the Lo-MacKinlay variance ratio test statistics displayed in Table 2B, show that both spot and futures are non-iid for both Brent Crude and WTI. Thus, we cannot employ the SD test developed by Davidson and Duclos (DD, 2000) to analyse the spot and futures returns because DD test relies on the iid assumption. In this connection, we adopt the SD test developed by Linton et al. (LMW, 2005) in the paper as this test can be applied to both iid and non-iid observations.

The results of the LMW test are displayed in Table 3 Panel A for Brent Crude oil and Panel B for WTI oil, respectively. As the p-values are all bigger than the 10% significance level for both H_0 and H'_0 , this shows that (1) there is no arbitrage opportunity between spot and futures oil, (2) spot does not dominate futures significantly and vice versa, (3) investors are indifferent from investing in spot or futures, and (4) the spot and futures oil markets are efficient and rational for both Brent Crude and WTI.

3.3 The Impact of Oil Crises

The oil market is very sensitive not only to news, but also to the expectation of news (Maslyuk and Smyth, 2008). For example, when the OPEC countries agreed to reduce the combined production of crude oil in 1999, oil prices increased further. Similarly, the Iraq War, otherwise known as the second Gulf War, occurred in March 2003, also caused oil futures prices to increase further due to the fear that the Iraqi oil fields and pipelines might be destroyed during the war.

We use regression analysis, with the cut-off points of the crises being stated in the previous section as dummies, and find that the dummies affect both spot and futures in the Iraq War crisis but not in the OPEC crisis, indicating that the impact of war is greater for both spot and futures

markets.⁹ However, the impact of the war could not be used to draw a reference for the preferences and the performance between spot and futures and draw inference on market efficiency. To this end, we use the SD tests to analyse the returns series for the pre- and OPEC, and pre- and Iraq-War, sub-periods.

[Insert Table 4 here]

Before we conduct SD tests on the oil market, we first apply the MV criterion and CAPM statistics on the series. The results are reported in Table 4, in which the results for Brent Crude oil are displayed in Table 4A, while those for WTI oil are given in Table 4B for each subperiod. As most of the results of the MV criterion and CAPM statistics for all the sub-periods are similar to those for the entire full sample period, we discuss only those results that are different from the full sample period. First, as compared with the pre-OPEC sub-period, the means for both spot and futures returns in the OPEC sub-period dramatically increased five-fold. However, as compared with the pre-Iraq-War sub-period, both Brent Crude oil spot and futures returns in the Iraq-War sub-period were reduced by 90%. For WTI oil, the spot and futures returns in the Iraq-War sub-period dramatically increased more than six-fold. Nonetheless, the differences between the means of spot and futures in each sub-period are not significant. In addition, the standard deviations for the returns of spot and futures are also not significantly different in each of the sub-periods. Thus, similar to the inferences for the entire sample, the MV criterion is unable to indicate any preference between the spot and futures markets. In addition, the CAPM statistics are unable to indicate any preference between the spot and futures markets.

We now apply SD to examine the performance of the spot and futures markets in all the sub-periods. The results from Table 3 show that all the p-values of the LMW test are greater than for the 10% significance level, thereby leading to the same conclusion as for the entire period. Thus,

⁹ Detailed results are available on request.

there is no arbitrage opportunity between spot and futures oil; spot and futures do not dominate each other; investors are indifferent from investing in spot or futures; and the spot and futures oil markets are efficient and rational for both Brent Crude oil and WTI oil in any of the sub-periods.

3.4 Robust Test on Diversification

[Insert Table 5 here]

Academics and practitioners are interested in examining investor's diversification preferences (Samuelson, 1967; Egozcue and Wong, 2010) in oil spot and futures markets. In order to achieve this purpose, we examine the dominance of spot or futures with the portfolios of different convex combinations of spot and futures, and report the p-values of the corresponding LMW test results in Table 5¹⁰.

We compare the full 100% of oil futures as one portfolio, with another portfolio consisting of different weights, from 10% to 90%, of oil spot and futures. If the weight of oil spot is $x\%$, then the weight of oil futures is $(100-x)\%$. We also compare the full 100% of oil spot as one portfolio, with another portfolio consisting of different weight of oil spot and futures, from 10% to 90%. The same weight method is applied. The first row, second column shows the pairwise comparison for 100% of oil futures, with 10% oil spot plus 90% oil futures, and so on. The results are reported in Table 5. From this table, we draw the same conclusion as in comparing spot and futures, namely that we cannot find any significant evidence of SD between any pair of portfolios. In short, the diversification results in Table 5 are consistent with the results of spot and futures without diversification. This provides evidence that the spot and futures oil markets are efficient.

4. Conclusions

¹⁰ As the results are qualitatively similar, we only report the results for Brent Crude oil. Results for WTI are available upon request.

This paper introduces the SD approach to examine the performance of spot and futures, and investors' behaviour in these markets, by analysing the entire period and the sub-periods, as well as different convex combinations of the portfolios of spot and futures. Our empirical findings suggest that there is no arbitrage opportunity between spot and futures oil, spot and futures do not dominate one another, investors are indifferent from investing in spot or futures, and the spot and futures oil markets are efficient and rational for both the Brent Crude oil and WTI oil markets.

We note that Moosa and Al-Luoghani (1995) show that both arbitrage and speculation play a role in determining oil futures prices, but the role of arbitrage is dominant. Our result of no arbitrage opportunity in these markets is contrary to Moosa and Al-Luoghani (1995). This could arise from the different methodology used by Moosa and Al-Luoghani (1995), or it may be due to the shorter period they examined, namely January 1986 to December 1991. As we have discussed in Section 2.3, in a short period, there may exist arbitrage opportunities. If the market is efficient, arbitrage opportunities will disappear in the long run.

The SD approach introduced in this paper provides useful information to investors for decision making in oil markets. We note that investors could also apply other techniques to study the market to provide additional information. For example, Silvapulle and Moosa (1999) find a bidirectional nonlinear causality effect between oil spot and futures prices, thereby suggesting that both markets react simultaneously to new information. We note that SD does not provide such information, while causality does not provide information drawn from the SD approach. Thus, if one would like to draw a more complete picture about oil markets, they should apply a wider range of tools to analyse the market. In particular, it would seem useful to apply the SD approach introduced in this paper to obtain information which other methods may not be able to obtain to assist in a better understanding of the oil market.

Table 1: Descriptive Statistics for Returns of Spot and Futures, 1989-2008

Variable	Brent Crude Oil		WTI				
	Spot	Futures	Spot	F1	F2	F3	F4
Mean (%)	0.0435 [*]	0.0432	0.0417	0.0417	0.0425	0.043	0.0433 [*]
Std Dev	0.0186	0.0219	0.0243	0.0234	0.0204	0.0187	0.0177
Skewness	-0.9201 ^{***}	-1.6782 ^{***}	-1.1932 ^{***}	-1.2878 ^{***}	-1.5021 ^{***}	-1.3012 ^{***}	-1.0969 ^{***}
Kurtosis	12.9542 ^{***}	32.0111 ^{***}	20.7355 ^{***}	21.4867 ^{***}	27.7857 ^{***}	21.2559 ^{***}	16.3703 ^{***}
Jarque-Bera (J-B)	21711.86 ^{***}	180710.47 ^{***}	90907 ^{***}	97721 ^{***}	162983 ^{***}	95691 ^{***}	56924 ^{***}
Beta	-0.0153	-0.1617	-0.1544	-0.1749	-0.1292	-0.1133	-0.1079
Sharpe Ratio	3.68	3.04	2.6571	2.7846	3.2604	3.6188	3.8651
Treynor Index	-4.625	-0.425	-0.425	-0.375	-0.525	-0.6	-0.65
Jensen Index	0.075	0.075	0.075	0.075	0.075	0.075	0.075
F Statistics	0.7221			1.0808	1.4176	1.6937	1.8856

Note: *, **, *** denote significance at the 10%, 5%, and 1% levels, respectively, and F1, F2, F3 and F4 refer to oil futures with 1, 2, 3 and 4 month's maturity date, respectively. The F statistic tests for the equality of variances between spot and futures. Readers may refer to footnote 4 for the formulae of the Sharpe Ratio, Treynor Index, and Jensen Index. The reported values of the Sharpe Ratio, Treynor Index, and Jensen Index are all annualized.

Table 2A: Results of Ljung-Box tests for the Returns of Spot and Futures

		Brent Crude Oil		WTI				
		Spot	Futures	Spot	F1	F2	F3	F4
lag=5	LB test	133.93	14.88	35.60	34.48	16.78	12.39	10.94
	p-value	0.00	0.01	0.00	0.00	0.00	0.03	0.05
	LB ² test	1082.79	114.39	198.00	278.94	124.04	64.06	92.56
	p-value	0.00	0.00	0.00	0.00	0.00	0.00	0.00
lag=10	LB test	138.02	30.67	46.00	49.44	32.06	25.16	23.90
	p-value	0.00	0.00	0.00	0.00	0.00	0.01	0.01
	LB ² test	1264.03	164.40	259.75	310.71	155.22	80.43	118.27
	p-value	0.00	0.00	0.00	0.00	0.00	0.00	0.00

Note: F1, F2, F3 and F4 refer to the oil futures with 1, 2, 3 and 4 month's maturity date, respectively. LB and LB² are the Ljung-Box statistic based on the levels and squared levels of the time series respectively. Both of them are asymptotically chi-square distributed with degree of freedom equals to the lag length.

Table 2B: Lo-MacKinlay variance ratio test statistics for the returns of Spot and Futures

		Brent Crude Oil		WTI				
k		Spot	Futures	Spot	F1	F2	F3	F4
5		5.313 ^{***}	-2.930 ^{***}	-4.153 ^{***}	-3.545 ^{***}	-2.129 ^{**}	-1.719 [*]	-2.605 ^{***}
10		2.181 ^{**}	-3.522 ^{***}	-4.960 ^{***}	-4.390 ^{***}	-2.947 ^{***}	-2.527 ^{***}	-3.012 ^{***}
20		2.018 ^{**}	-2.243 ^{**}	-4.186 ^{***}	-3.685 ^{***}	-2.410 ^{***}	-1.993 ^{**}	-2.272 ^{***}
30		2.278 ^{**}	-1.383	-3.264 ^{***}	-2.812 ^{***}	-1.652 [*]	-1.213	-1.406

Note: *, **, *** represent significance levels of 10%, 5% and 1%, respectively. k is the duration period. Under the null hypothesis of iid, the Lo-MacKinlay variance ratio statistic follows the standard normal distribution asymptotically for any duration period k.

Table 3: Results of LMW Test for the Returns of Spot and Futures

A: Brent Crude Oil				
	S > F		F > S	
	FSD	SSD	FSD	SSD
Whole Period	0.6523	0.5265	0.7353	0.4915
Pre-OPEC	0.6264	0.5055	0.6204	0.5555
OPEC	0.5375	0.6494	0.6134	0.5215
Pre-Iraq	0.6294	0.5045	0.6534	0.5385
Iraq War	0.5305	0.7153	0.5824	0.5215
B: WTI				
	S > F1		F1 > S	
	FSD	SSD	FSD	SSD
Whole Period	0.8182	0.4665	0.7862	0.5195
Pre-OPEC	0.8501	0.5534	0.7463	0.4965
OPEC	0.9620	0.6114	0.9401	0.5345
Pre-Iraq	0.8681	0.5684	0.7782	0.4885
Iraq War	0.8941	0.7203	0.9660	0.6284
	S > F2		F2 > S	
	FSD	SSD	FSD	SSD
Whole Period	0.7393	0.4965	0.6653	0.4995
Pre-OPEC	0.7692	0.5195	0.7592	0.4905
OPEC	0.7992	0.5305	0.8771	0.4975
Pre-Iraq	0.8362	0.5115	0.7333	0.4865
Iraq War	0.8771	0.5335	0.8881	0.5115
	S > F3		F3 > S	
	FSD	SSD	FSD	SSD
Whole Period	0.7792	0.5065	0.6713	0.5005
Pre-OPEC	0.62138	0.5185	0.7572	0.4815
OPEC	0.8162	0.5155	0.7992	0.4955
Pre-Iraq	0.6234	0.4945	0.7443	0.5035
Iraq War	0.8511	0.5345	0.8142	0.5015
	S > F4		F4 > S	
	FSD	SSD	FSD	SSD
Whole Period	0.7582	0.5095	0.6983	0.5045
Pre-OPEC	0.6294	0.5275	0.6484	0.4935
OPEC	0.7722	0.5115	0.7572	0.4935
Pre-Iraq	0.6074	0.4835	0.6943	0.5075
Iraq War	0.7522	0.5285	0.7223	0.5115

Note: The table displays the p-values of the LMW test. Readers may refer to Linton et al. (2005) for the LMW SD test statistics. F1, F2, F3 and F4 refer to oil futures with 1, 2, 3 and 4 month's maturity date, respectively.

Table 4A: Descriptive Statistics for the returns of Brent Crude Oil Spot and Futures in Sub-Periods

Variable	Pre-OPEC		OPEC		Pre-Iraq War		Iraq War	
	Spot	Futures	Spot	Futures	Spot	Futures	Spot	Futures
Mean (%)	0.0129	0.0119	0.0819**	0.08242*	0.0157	0.0134	0.0012***	0.0012***
Std Dev	0.0197	0.0224	0.0172	0.0213	0.0196	0.0228	0.0159	0.0193
Skewness	-1.0881***	-2.6108***	-0.5726	-0.3245	-1.02***	-2.035***	-0.2882***	0.0224
Kurtosis	17.4315***	51.5032***	2.576	2.1657	14.2659***	37.0708***	1.4916***	0.7288***
J-B	25063*	280027***	140	105	20252***	181905***	149***	296***
Beta	0.0112	-0.3738	-0.0337	-0.0005	0.0238	0.1861	-0.1645	-0.0781
Sharpe Ratio (annualize)	-0.8875	-1.0375	10.35	8.45	0.3443	-0.65	17.24	14.35
Treynor Index	-0.0065	0.0003	-0.0216	-1.5764	0.0012	0.0003	-0.0068	-0.0145
Jensen Index	-7.4*10 ⁻⁵	-3.9*10 ⁻⁵	0.0007	0.0007	-2.66*10 ⁻⁵	-7.09*10 ⁻⁵	0.0012	0.0012
F Statistics	0.7726		0.6523		0.7338		0.67425	

Table 4B: Descriptive Statistics for the returns of WTI Oil Spot and Futures in Sub-Periods

	Variable	Spot	F1	F2	F3	F4
Pre-OPEC	Mean (%)	0.0078	0.0080	0.0093	0.0100	0.0099
	Std Dev	0.0248	0.0240	0.0201	0.0180	0.0167
	Skewness	-1.6449***	-1.8059***	-2.5034***	-2.3260***	-1.9327***
	Kurtosis	31.7345***	33.1100***	51.1408***	42.4780***	33.3643***
	J-B	117864***	128450***	305742***	211439***	130627***
	Beta	-0.3950	-0.3964	-0.3144	-0.2910	-0.2751
	Sharpe Ratio (annualize)	-1.3967	-1.3639	-1.4367	-1.4738	-1.5606
	Treynor Index	0.0004	0.0003	0.0004	0.0004	0.0004
	Jensen Index	-0.0001	-0.0001	-0.0001	-0.0001	-0.0001
	F Statistics		1.0706	1.5265	1.9022	2.2063
OPEC	Mean (%)	0.0839*	0.0837*	0.0838*	0.0841**	0.0848**
	Std Dev	0.0237	0.0226	0.0208	0.0195	0.0189
	Skewness	-0.5397***	-0.5105***	-0.3894***	-0.3048***	-0.3800***
	Kurtosis	4.0572***	2.9623***	2.5129***	1.7668***	2.8456***
	J-B	1637***	912***	643***	324***	806***
	Beta	0.0283	-0.0065	0.0117	0.0221	0.0196
	Sharpe Ratio (annualize)	8.0095	8.3306	8.9527	9.5226	9.9091
	Treynor Index	0.0272	-0.1179	0.0648	0.0342	0.0388
	Jensen Index	0.0008	0.0008	0.0008	0.0008	0.0008
	F Statistics		1.0951	1.2917	1.4731	1.5733

	Variable	Spot	F1	F2	F3	F4
Pre-Iraq War	Mean (%)	0.0150	0.0149	0.0154	0.0152	0.0151
	Std Dev	0.0252	0.0243	0.0208	0.0187	0.0175
	Skewness	-1.3887***	-1.5326***	-1.9440***	-1.7571***	-1.5000***
	Kurtosis	23.7929***	24.9110***	35.4246***	28.5700***	22.5287***
	J-B	87292***	95832***	193201***	126050***	78579***
	Beta	-0.1841	-0.2122	-0.1557	-0.1382	-0.1301
	Sharpe Ratio (annualize)	-0.4513	-0.4482	-0.4607	-0.5164	-0.5583
	Treynor Index	0.0003	0.0002	0.0003	0.0003	0.0003
	Jensen Index	-0.0001	-0.0001	0.0000	0.0000	0.0000
	F Statistics		1.0744	1.4733	1.8091	2.0805
Iraq War	Mean (%)	0.1135*	0.1138**	0.1154**	0.1177**	0.1191**
	Std Dev	0.0217	0.0207	0.0195	0.0185	0.0183
	Skewness	-0.3074***	-0.1354*	-0.0398	-0.0329	-0.1666**
	Kurtosis	3.7087***	1.0915***	0.5379***	0.4749***	2.4133***
	J-B	799***	72***	17***	13***	336***
	Beta	-0.0497	-0.0421	-0.0365	-0.0263	-0.0307
	Sharpe Ratio (annualize)	12.4559	13.1195	14.0045	14.9236	15.2669
	Treynor Index	-0.0220	-0.0261	-0.0303	-0.0426	-0.0369
	Jensen Index	0.0011	0.0011	0.0011	0.0011	0.0011
	F Statistics		1.1047	1.2477	1.3774	1.4102

Note: ***, **, and * denote significance at the 1%, 5%, and 10% levels, respectively. The F statistic tests the equality of variances. Readers may refer to footnote 4 for the formulae of the Sharpe Ratio, Treynor Index, and Jensen Index, and for further information about these statistics.

Table 5: Results of LMW Test for the Portfolio of Oil Spot and Futures

% of Oil Spot	100% Oil Futures		100% Oil Spot	
	P > F	F > P	P > S	S > P
10	0.5125	0.5295	0.5095	0.5225
20	0.5065	0.5185	0.5644	0.5456
30	0.5095	0.5075	0.4775	0.5275
40	0.5145	0.4945	0.6693	0.5335
50	0.5125	0.4955	0.6703	0.5345
60	0.7003	0.4835	0.6653	0.5415
70	0.5125	0.5005	0.6723	0.5395
80	0.5145	0.5005	0.6663	0.5505
90	0.5175	0.4975	0.7413	0.5844

Notes: The table reports the p-values of the LMW test for SSD of the portfolios of oil spot and futures (P) with oil spot (S) or futures (F) alone. Readers may refer to Linton et al. (2005) for the LMW SD test statistics. The weight of oil spot in the portfolios is shown in the first column.

Figure 1: Brent Crude Oil Spot and Futures Indices

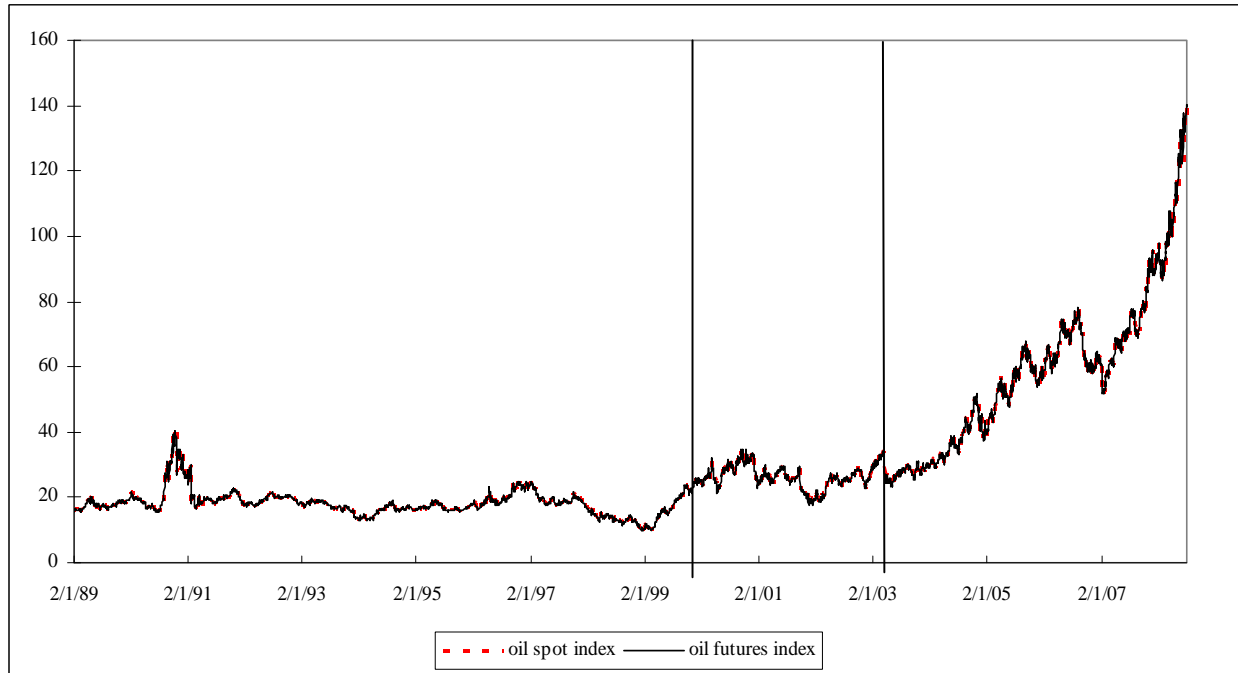
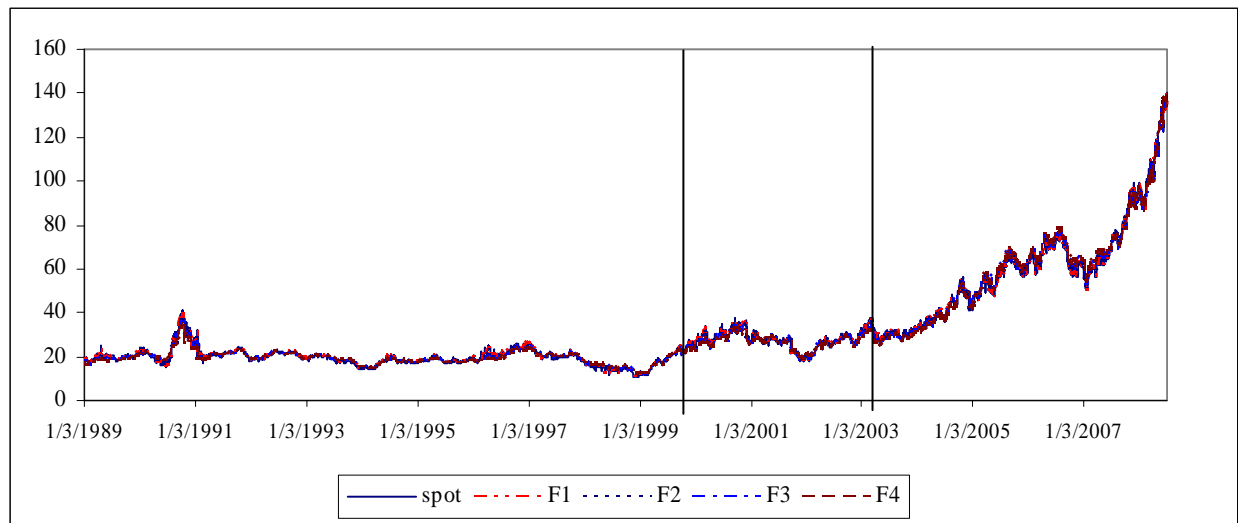


Figure 2: WTI Spot and Futures (F1, F2, F3 and F4)



Notes: These figures show the time series plots of oil spot and futures indices from January 1, 1989 to June 30, 2008. The first vertical line located at October 29, 1999 represents the cut-off point of the OPEC crisis, while the second vertical line located at March 20, 2003 represents the cut-off point of the Iraq War (see Section 2 for further details).

References

- Akerlof, G.A., 1970. The Market for 'Lemons:' Quality Uncertainty and the Market Mechanism. *Quarterly Journal of Economics* 84, 488-500.
- Anderson, G., 1996. Nonparametric Tests of Stochastic Dominance in Income Distributions. *Econometrica* 64, 1183 – 1193.
- Anderson, G., 2004. Toward an Empirical Analysis of Polarization. *Journal of Econometrics* 122, 1-26.
- Bai, Z.D., Liu, H.X., Wong, W.K., 2009a. On the Markowitz Mean-Variance Analysis of Self-Financing Portfolios. *Risk and Decision Analysis* 1(1), 35-42.
- Bai, Z.D., Liu, H.X., Wong, W.K., 2009b. Enhancement of the Applicability of Markowitz's Portfolio Optimization by Utilizing Random Matrix Theory. *Mathematical Finance* 19(4), 639-667.
- Barrett, G., Donald, S., 2003. Consistent Tests for Stochastic Dominance. *Econometrica* 71, 71-104.
- Bawa, Vijay S., 1978. Safety-First, Stochastic Dominance, and Optimal Portfolio Choice. *Journal of Financial and Quantitative Analysis* 13, 255-271.
- Beedles, W.L. 1979. Return, Dispersion and Skewness: Synthesis and Investment Strategy. *Journal of Financial Research*, 71-80.
- Bekiros, Stelios D., Diks, Cees G.H., 2008. The Relationship between Crude Oil Spot and Futures Prices: Cointegration, Linear and Nonlinear Causality. *Energy Economics* 30(5), 2673-2685.
- Bernard, V.L., Seyhun, H.N., 1997. Does Post-Earnings-Announcement Drift in Stock Prices Reflect a Market Inefficiency? A stochastic dominance approach. *Review of Quantitative Finance and Accounting* 9, 17-34.
- Bishop, J.A., Formly, J.P., Thistle, P.D., 1992. Convergence of the South and Non-South Income Distributions. *American Economic Review* 82, 262-272.
- Bopp, A.E., Sitzer, S., 1987. Are Petroleum Futures Prices Good Predictors of Cash Value? *Journal of Futures Markets* 7, 705-719.
- Coimbra, C., Esteves, P.S., 2004. Oil Price Assumptions in Macroeconomic Forecasts: Should We

Follow Futures Market Expectations?, *OPEC Review* 28, 87–106.

Crowder, W.J., Hamid, A., 1993. A Co-integration Test for Oil Futures Market Efficiency. *Journal of Futures Markets* 13, 933-941.

Davidson, R., Duclos, J-Y., 2000. Statistical Inference for Stochastic Dominance and for the Measurement of Poverty and Inequality. *Econometrica* 68, 1435-1464.

Egozcue, M., Wong, W.K. 2010. Gains from Diversification: A Majorization and Stochastic Dominance Approach. *European Journal of Operational Research* 200, 893–900.

Falk, H., Levy, H., 1989. Market Reaction to Quarterly Earnings' Announcements: A Stochastic Dominance Based Test of Market Efficiency. *Management Science* 35, 425-446.

Feldstein, M.S., 1969. Mean Variance Analysis in the Theory of Liquidity Preference and Portfolio Selection, *Review of Economics Studies*, 36, 5-12.

Fishburn, P.C., 1964. *Decision and Value Theory*, New York.

Fong, W.M., Lean, H.H., Wong, W.K., 2008. Stochastic Dominance and Behavior Towards Risk: The Market for Internet Stocks. *Journal of Economic Behavior and Organization* 68(1), 194-208.

Fong, W.M., See, K.H., 2002. A Markov Switching Model of the Conditional Volatility of Crude Oil Futures Prices. *Energy Economics* 24, 71-95.

Fong, W.M., See, K.H., 2003. Basis Variations and Regime Shifts in the Oil Futures Market. *European Journal of Finance* 9, 499–513.

Fong, W.M., Wong, W.K., Lean, H.H., 2005. International Momentum Strategies: A Stochastic Dominance Approach. *Journal of Financial Markets* 8, 89–109.

Fung, W., Hsieh, D.A., 1999. Is Mean-Variance Analysis Applicable to Hedge Funds?, *Economics Letters*, 62, 53-58.

Gasbarro, D., Wong, W.K., Zumwalt, J.K., 2007. Stochastic Dominance Analysis of iShares. *European Journal of Finance* 13, 89-101.

Gertler, M.L., 1988. Financial Structure and Aggregate Economic Activity: An Overview. *Journal of Money, Credit and Banking* 20, 559-588.

Gulen, S.G., 1999. Regionalization in World Crude Oil Markets: Further Evidence. *Energy Journal*

20, 125-139.

Hadar J., Russell, W.R., 1969. Rules for Ordering Uncertain Prospects. *American Economic Review* 59, 25-34.

Hanoch, G., Levy, H., 1969. The Efficiency Analysis of Choices Involving Risk. *Review of Economic studies* 36, 335-346.

Hammond, J.S., 1974. Simplifying the Choice Between Uncertain Prospects Where Preference is Nonlinear. *Management Science* 20(7), 1047-1072.

Hammoudeh, S., Li, H., Jeon, B., 2003. Causality and Volatility Spillovers Among Petroleum Prices of WTI, Gasoline and Heating Oil in Different Locations. *North American Journal of Economics and Finance* 14, 89-114.

Hammoudeh, S., Li, H., 2004. The Impact of the Asian Crisis on the Behavior of US and International Petroleum Prices. *Energy Economics* 26, 135–160.

Jarrow, R., 1986. The Relationship Between Arbitrage and First Order Stochastic Dominance. *Journal of Finance* 41, 915-921.

Jensen, M.C., 1969. Risk, the Pricing of Capital Assets and the Evaluation of Investment Portfolios. *Journal of Business* 42, 167-247.

Kahneman, D., Tversky, A., 1979. Prospect Theory of Decisions Under Risk. *Econometrica* 47(2), 263-291.

Kaur, A., Rao, B.L.S.P., Singh, H., 1994. Testing for Second Order Stochastic Dominance of Two Distributions. *Econometric Theory* 10, 849 – 866.

Klecan, L., McFadden, R., McFadden, D., 1991. A Robust Test for Stochastic Dominance. Working Paper, MIT & Cornerstone Research.

Larsen G.A., Resnick, B.G., 1999. A Performance Comparison Between Cross-sectional Stochastic Dominance and Traditional Event Study Methodologies. *Review of Quantitative Finance and Accounting* 12, 103-112.

Lean, H.H., Wong, W.K., Zhang, X., 2008. The Sizes and Powers of Some Stochastic Dominance Tests: A Monte Carlo Study for Correlated and Heteroskedastic Distributions. *Mathematics and Computers in Simulation* 79, 30-48.

- Leung, P.L., Wong, W.K., 2008. On Testing the Equality of the Multiple Sharpe Ratios, with Application on the Evaluation of IShares. *Journal of Risk* 10(3), 1-16.
- Levy H., 1992. Stochastic Dominance and Expected Utility: Survey and Analysis. *Management Science*, 38, 555-593.
- Li, C.K., Wong, W.K., 1999. Extension of Stochastic Dominance Theory to Random Variables. *RAIRO Recherche Opérationnelle*, 33(4), 509-524.
- Lin, X.S., Tamvakis, M.N., 2001. Spillover Effects in Energy Futures Markets. *Energy Economics* 23, 43-56.
- Linton, O., Maasoumi, E., Whang, Y-J., 2005. Consistent Testing for Stochastic Dominance Under General Sampling Schemes. *Review of Economic Studies* 72, 735-765.
- Markowitz, H.M., 1952. Portfolio Selection. *Journal of Finance* 7, 77-91.
- Maslyuk, S., and Smyth R., 2008. Unit Root Properties of Crude Oil Spot and Futures Prices. *Energy Policy* 36, 2591-2600.
- McFadden, D., 1989. Testing for Stochastic Dominance. In T.B. Fomby and T.K. Seo, (Eds.), *Studies in the Economics of Uncertainty*. Springer Verlag, New York.
- Meyer, J., 1977. Second Degree Stochastic Dominance with Respect to a Function. *International Economic Review* 18, 476-487.
- Moosa, I.A., and Al-Loughani, N.E., 1995. The Effectiveness of Arbitrage and Speculation in the Crude Oil Futures Market. *Journal of Futures Markets* 15, 167-186.
- Post, T., Levy, H., 2005. Does Risk Seeking Drive Asset Prices? A Stochastic Dominance Analysis of Aggregate Investor Preferences and Beliefs. *Review of Financial Studies* 18(3), 925-953.
- Postali, F., Picchetti, P., 2006. Geometric Brownian Motion and Structural Breaks in Oil Prices: A Quantitative Analysis. *Energy Economics* 28, 506-522.
- Quan, J., 1992. Two Step Testing Procedure for Price Discovery Role of Futures Prices. *Journal of Futures Markets* 12, 139-149.
- Quirk J.P., Saposnik, R., 1962. Admissibility and Measurable Utility Functions. *Review of Economic Studies* 29, 140-146.

- Richmond, J., 1982. A General Method for Constructing Simultaneous Confidence Intervals. *Journal of the American Statistical Association* 77, 455-460.
- Rochschild, M., Stiglitz, J.E., 1970. Increasing Risk I. A Definition. *Journal of Economic Theory* 2, 225-243.
- Samuelson, P.A., 1967. General Proof That Diversification Pays. *Journal of Financial and Quantitative Analysis* 2(1), 1-13.
- Schwartz, T.V., Szakmary, A.C., 1994. Price Discovery in Petroleum Markets: Arbitrage, Cointegration and the Time Interval of Analysis. *Journal of Futures Markets* 14, 147-167.
- Schwert, G., William, 1990. Stock Returns and Real Activity: A Century of Evidence. *Journal of Finance* 45(4), 1237-1257.
- Serletis, A., Banack, D., 1990. Market Efficiency and Co-integration: \An Application to Petroleum Markets. *Review of Futures Markets* 9, 372-385.
- Sharpe, W.F., 1964. Capital Asset Prices: Theory of Market Equilibrium Under Conditions of Risk. *Journal of Finance* 19, 425-442.
- Silvapulle, P., Moosa I., 1999. The Relationship Between Spot and Futures Prices: Evidence from the Crude Oil Market. *Journal of Futures Markets* 19, 175-193.
- Sriboonchita, S., Wong, W.K., Dhompongsa, S., Nguyen, H.T., 2009. *Stochastic Dominance and Applications to Finance, Risk and Economics*, Chapman and Hall/CRC, Taylor and Francis, Boca Raton, Florida, USA.
- Stoline, M.R., Ury, H.K., 1979. Tables of the Studentized Maximum Modulus Distribution and an Application to Multiple Comparisons Among Means. *Technometrics* 21, 87-93.
- Stoyan, D., 1983. *Comparison Methods for Queues and Other Stochastic Models*. New York: Wiley.
- Taback, B.M., 2003. On the Information Content of Oil Future Prices. Working Paper no. 65, Banco Central de Brazil.
- Tesfatsion, L., 1976. Stochastic Dominance and Maximization of Expected Utility. *Review of Economic Studies* 43, 301-315.

Tobin, J., 1958. Liquidity Preference and Behavior Towards Risk. *Review of Economic Studies* 25, 65-86.

Treynor, J.L., 1965. How to Rate Management of Investment Funds. *Harvard Business Review* 43, 63-75.

Tse, Y.K., Zhang, X., 2004. A Monte Carlo Investigation of Some Tests for Stochastic Dominance. *Journal of Statistical Computation and Simulation* 74, 361-378.

von Neumann, J., Morgenstern, O., 1944. *Theory of Games and Economic Behavior*, Princeton University Press, Princeton N.J.

Wilson, B., Aggarwal, R., Inclan, C., 1996. Detecting Volatility Changes Across the Oil Sector. *Journal of Futures Markets* 16, 313–320.

Wong, W.K., 2007, Stochastic Dominance and Mean-Variance Measures of Profit and Loss for Business Planning and Investment. *European Journal of Operational Research* 182, 829-843.

Wong, W.K., Chan, R.H., 2008. Markowitz and Prospect Stochastic Dominances. *Annals of Finance* 4(1), 105-129.

Wong, W.K., Li, C.K., 1999. A Note on Convex Stochastic Dominance Theory. *Economics Letters* 62, 293-300.

Wong, W.K. Ma, C., 2008. Preferences Over Location-Scale Family. *Economic Theory* 37(1), 119-146.

Wong, W.K., Phoon, K.F., Lean, H.H., 2008. Stochastic Dominance Analysis of Asian Hedge Funds, *Pacific-Basin Finance Journal* 16(3), 204-223.

Wong, W K, Thompson, H.E., Wei, S., Chow, Y.F., 2006, Do Winners Perform Better Than Losers? A Stochastic Dominance Approach, *Advances in Quantitative Analysis of Finance and Accounting* 4, 219-254.