Stochastic Convenience Yield and the Pricing of Oil Contingent Claims

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ABSTRACT

This paper develops and empirically tests a two-factor model for pricing financial and real assets contingent on the price of oil. The factors are the spot price of oil and the instantaneous convenience yield. The parameters of the model are estimated using weekly oil futures contract prices from January 1984 to November 1988, and the model's performance is assessed out of sample by valuing futures contracts over the period November 1988 to May 1989. Finally, the model is applied to determine the present values of one barrel of oil deliverable in one to ten years time.

CRUDE OIL IS A strategic natural resource which nowadays serves as "the underlying asset" to many financial instruments such as futures, options on futures, as well as oil-linked bonds. While most financial instruments cover the short to medium term maturity range, there is a large amount of real long term oil-linked assets such as oil lease contracts, oil reserves, etc., where pricing could be approached using the contingent claims framework. Until recently, most option valuation models aimed at valuing natural resources have been based on the assumption that there is a single source of uncertainty related to the price (or net operating value) of the commodity (reserve). In this study our objective is to present a more general approach which can easily be applied to the pricing of real and financial oil contingent claims. For that purpose we assume that the spot price of oil is a fundamental, but not unique, determinant of these latter claims' prices. In addition, we also allow for a stochastic convenience yield of crude oil in order to develop a two-factor oil contingent claims pricing model. The notion of a convenience yield, viewed as a net "dividend" yield accruing to the owner of the physical commodity at the margin, has already proven to drive the relationship between futures and spot prices of many commodities. The theory of storage posits an inverse relationship between the level of inventories and the net convenience yield which suggests that a constant convenience yield assumption will only hold under very restrictive assumptions. Moreover, Gibson and Schwartz (1989) refute this assumption in the case of crude oil, showing that the mean reverting tendency as well as the variability of its changes requires a stochastic representation in order to price oil-linked securities accurately. In this paper we derive a more realistic two-factor pricing model and subsequently analyze its performance in valuing short as well as long term oil contracts.

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1 See, in particular, Brennan and Schwartz (1985) and Paddock, Siegel, and Smith (1988).


3 See Brennan (1986).
The main empirical results of the paper show that the model performs well in valuing short term contracts such as futures. Since there are no traded long term oil “contracts,” we can only provide illustrative examples. The computed theoretical present values of a 1–10 years ahead deliverable oil barrel seem to be low and hence suggest that the risk premium for long term oil investments is high.

Second, the two-factor model is able to explain the “intrinsic” difference in price volatility between spot and futures contracts as well as its decreasing maturity pattern observed among the latter.

Finally, we show that, although we apply the model to financial securities whose payoff structure is linear in the spot price of crude oil, it can easily be extended to any more complex payoff structure characterizing the option feature(s) of real and financial oil claims.

The paper is organized as follows. In Section I, we derive the two-factor model and show that since one of the state variables is not a traded asset it requires the estimation of an exogenous parameter, namely the market price of convenience yield risk. In Section II, we estimate the parameters of the joint stochastic process followed by the spot price and the convenience yield of crude oil over a 5-year reference period, from January 1984 to November 1988. The prices of all futures contracts traded on the New York Mercantile Exchange over this period are then used in conjunction with the model’s theoretical prices to estimate in Section III the assumed constant market price of convenience yield risk, $\lambda$. In Section IV, we test the model’s out-of-sample performance in valuing futures contracts over the period starting November 18, 1988 and ending May 6, 1989. The model’s pricing accuracy decreases as the maturity of the futures contracts is lengthened, but we show that this maturity “bias” as well as the overpricing can be substantially reduced when one uses monthly updated estimates of $\lambda$. We then apply the model, in Section V, to determine the present values of one barrel of oil deliverable in 1–10 years time and to compute the hedge ratios with respect to the two state variables of these present value factors. We show that their sign and magnitude confirms previous empirical findings on the relationship between futures and spot prices’ volatility. Finally, Section VI concludes the paper.

I. The Two-Factor Oil Contingent Claim Pricing Model

In order to derive the general pricing equation which applies to any oil contingent claim, we shall first of all assume that its price depends only upon the spot price of oil $S$, the instantaneous net convenience yield of oil $\delta$ and time to maturity $\tau$ ($=T-t$).

Moreover, we shall assume that the spot price of oil and the net convenience yield follow a joint diffusion process specified as

$$dS/S = \mu dt + \sigma_1 dz_1,$$  
$$d\delta = k(\alpha - \delta)dt + \sigma_2 dz_2.$$  

$dz_1$ and $dz_2$ are correlated increments to standard Brownian processes and $dz_1 \cdot dz_2 = \rho dt$, where $\rho$ denotes the correlation coefficient between the two Brownian motions.
The form of (1) is based on the conjecture that the spot price of oil has a lognormal-stationary distribution. We shall discuss its realism in Section II, where we provide empirical evidence on the time series properties of the spot price of oil.

The form of (2) is motivated by Gibson and Schwartz's (1989) study of the time series properties of the forward convenience yields of crude oil, where they find strong empirical evidence in favor of their mean reverting pattern. Assuming that the price \( B(S, \delta, \tau) \) of the oil contingent claim is a twice continuously differentiable function of \( S \) and \( \delta \), we can use Itô's Lemma to define its instantaneous price change as follows:

\[
\begin{align*}
\frac{dB}{B} & = dS + B_t d\delta - B_t dt + \frac{1}{2} B_t \sigma^2 (dS)^2 + \frac{1}{2} B_t (d\delta)^2 + B_t dS d\delta, \\
\frac{dB}{B} & = [-B_s - \frac{1}{2} B_{ss} \sigma^2 S^2 + B_{s\delta} \rho \sigma_1 \sigma_2 + \frac{1}{2} B_{\delta\delta} \sigma^2 + B_s \mu S + B_\delta (k(\alpha - \delta))] dt \\
& + \sigma_1 B_S dz_1 + \sigma_2 B_\delta dz_2.
\end{align*}
\]

Abstracting from interest rate uncertainty and invoking the standard perfect market assumptions, it can be shown that in the absence of arbitrage the price of this claim must satisfy the following partial differential equation:

\[
\frac{1}{2} B_{ss} S^2 \sigma^2 + \frac{1}{2} B_{s\delta} \sigma^2 + B_{s\delta} \rho \sigma_1 \sigma_2 + B_s S (r - \delta) + B_\delta (k(\alpha - \delta) - \lambda \sigma_2) - B_t - r B = 0,
\]

where \( \lambda \) denotes the market price per unit of convenience yield risk and is at most a function of \( S, \delta, \alpha, \) and \( t \). By the same no-arbitrage argument, it can be

\[\lambda = \frac{\mu - r}{\sigma_1} \quad \text{and} \quad \tau = \frac{B_{s\delta}}{\sigma_2},\]

Then, recognizing that a spot contract on oil must also satisfy (1) and that the total expected return \( \mu_e \) to the owner of the oil derives from two sources, namely the convenience yield \( \delta \) and the expected oil price change \( \mu \), one can define the market price per unit of oil price risk \( \lambda' \) by solving the partial differential equation for \( S \) (since the analytical expression of all derivatives is then known). Hence,

\[\lambda' = \frac{\mu_e - r}{\sigma_1} = \frac{\mu + \delta - r}{\sigma_1},\]

from which (4) is then easily derived for any general oil contingent claim.

Brennan (1986) derives a two-factor model for the pricing of commodities in which \( \lambda \) is a constant. For his result to hold, it suffices that the representative investor has logarithmic utility and that the covariance between the return on aggregate wealth and the change in the instantaneous convenience yield be proportional to \( \sigma_2 \).
shown that the price of a futures contract \( F(S, \delta, \tau) \) on one barrel of crude oil deliverable at time \( T \) will satisfy the following partial differential equation:

\[
\frac{1}{2}F_{SS}S^2\sigma^2 + \frac{1}{2}F_{\delta\delta}\sigma^2 + F_{\delta\tau}\sigma_1 + F_{S}\sigma_2 + F_S S (r - \delta) \\
+ F_{\delta}(k(\alpha - \delta) - \lambda\sigma_2) - F_{\tau} = 0, \quad (50)
\]

subject to the initial condition:

\[ F(S, \delta, 0) = S. \quad (6) \]

Any other spot claim on oil will, under this framework, satisfy equation (4) subject to the appropriate boundary conditions. More specifically, the present value of one barrel of oil deliverable at time \( T \), \( B(S, \delta, T) \) satisfies (4) subject to the initial condition (6). The computation of \( B(S, \delta, T) \) is the starting point to any capital budgeting decision based on the present value of future oil-linked cash flows.

As far as the pricing of other financial securities is concerned, it is trivial to modify the boundary conditions to allow for quantity adjustments, for constant interim payoffs\(^5\) and for oil-linked interim payoffs. Finally, the price of any oil contingent security will also satisfy (4), and only the boundary conditions will have to be modified according to its specific exercise features. For example, if we are using (4) to price a European call option entitling its owner to buy one barrel of spot crude oil at time \( T \) at an exercise price of \( K \), the initial condition reads as follows:

\[ C(S, \delta, 0) = \max[0, S - K], \quad (7) \]

where \( C(S, \delta, 0) \) denotes the price of the European call at maturity. In most situations, however, there are no analytical solutions to the partial differential equations (4) and (5). We have therefore used a numerical technique\(^6\) to compute the present value factors \( B(S, \delta, \tau) \) and the futures prices \( F(S, \delta, \tau) \).

In order to apply the model we still need to estimate the market price of convenience yield risk \( \lambda \) as well as the parameters \( k, \alpha, \sigma_2, \rho, \) and \( \sigma_1 \) of the joint stochastic process followed by the spot price and the convenience yield of crude oil. Let us first turn to the estimation of these latter parameters since we shall need them to estimate \( \lambda \) in Section III.

II. Estimation of the Joint Stochastic Process

First of all, we shall define the two proxies we have chosen for \( S \) and \( \delta \) since neither of these two state variables can actually be observed. As Gibson and Schwartz (1989) point out, there is no true "spot" market for crude oil, and we have therefore identified this state variable with the settlement price of the

\(^5\) It suffices to add the amount \( D \) of the constant payoff in the partial differential equation (4) if it has the form of a fixed continuously compounded coupon rate. If the contract consists of a series of discrete oil-linked cash flows, we use the "discount factors" \( B(S, \delta, \tau) \) to price the security as we would price a discrete coupon bond.

\(^6\) More specifically, we have used an Alternative Direction Implicit Method (ADI) as described by McKee and Mitchell (1979).
closest maturity crude oil futures contract trading on the New York Mercantile Exchange. The procedure which has been used to compute the instantaneous convenience yield of crude oil relies on the well known relationship between the futures and the spot price of a commodity when there is neither interest rate nor convenience yield uncertainty, namely\textsuperscript{10}

\[ F(S, T) = S_0 e^{(r - c)(T - 1)}. \]  

(8)

This allows us to determine the annualized monthly forward convenience yields by using pairs of adjacent monthly maturities futures contracts’ prices according to the following formula:

\[ \delta_{T-1,T} = r_{T-1,T} - 12 \ln \left( \frac{F(S, T)}{F(S, T - 1)} \right), \]

(9)

where \( \delta_{T-1,T} \) denotes the \( T - 1 \) periods ahead annualized one month forward convenience yield, and \( r_{T-1,T} \) denotes the \( T - 1 \) periods ahead annualized one month riskless forward interest rate. Due to the absence of spot crude oil contracts, we used the two closest maturity futures contracts prices as well as the two T-bill rates with maturities as close as possible to the futures contracts’ ones in order to compute \( \delta_{1,2} \), the one month ahead annualized one month forward convenience yield. The latter has then been identified with the instantaneous convenience yield \( \delta_{0,1} \) of crude oil for the estimation and pricing purposes of this study. In Figures 1–3 we illustrate the joint as well as individual evolution of the two state variables over a 5-year period. They illustrate that the convenience yield of crude oil has a tendency to revert to its long run mean value and that it is a highly volatile parameter, while the spot price of crude oil is less volatile and seems to follow a random walk.

We first analyzed the time series properties of the chosen proxy of the spot price of crude oil in order to examine whether the lognormal distribution assumption was supported by the data. For that purpose, we collected nearly 5 years of weekly, every Friday, price data covering the period of January 6, 1984 to November 18, 1988 and regressed the logarithm of the price ratios on their lagged value. The results reported in Table I tend to support the conjectured hypothesis since they show significant evidence neither of a mean reverting tendency nor of first order serial correlation in the residuals. Furthermore, the historical volatility \( \hat{\delta}_1 \) of the logarithmic returns appears to be fairly stable across subperiods and does not indicate the existence of important volatility shifts over the reference period covered.

As already mentioned, the specification of the convenience yield’s stochastic process has been motivated by the strong mean reverting tendency of the 2–6 month ahead annualized forward convenience yields pointed out in Gibson and Schwartz (1989). Moreover, it is consistent with the theory of storage’s emphasis on an inverse relationship between the level of inventories and the relative net convenience yield.\textsuperscript{11} Since crude oil inventories fluctuate much more than other

\textsuperscript{10} For further details on the pricing of a futures contract when the relative marginal net convenience yield is constant, see Brennan and Schwartz (1985).

\textsuperscript{11} See, in particular, Kaldor (1938), Working (1949), Brennan (1986), and Fama and French (1988).
unfinished or finished goods inventories, the mean reverting process assures that the convenience yield will nevertheless remain finite. Changes in the convenience yield will also be more pronounced during oil market turmoils than over a steady state period in which the convenience yield is closer to its long run mean value of $\alpha$.

12 See Verleger (1982), Chapter 5.
Pricing of Oil Contingent Claims

![Graph showing the evolution of oil price and convenience yield.](image)

Figure 3. Evolution of oil price and convenience yield.

<table>
<thead>
<tr>
<th>Period</th>
<th>$b$</th>
<th>$t$</th>
<th>$DW^*$</th>
<th>$R^2$</th>
<th>$\sigma_t$</th>
<th>$N^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jan. 84-Nov. 88</td>
<td>0.002</td>
<td>0.03</td>
<td>1.98</td>
<td>0.00</td>
<td>35.34%</td>
<td>253</td>
</tr>
<tr>
<td>Jan. 84-May 86</td>
<td>0.099</td>
<td>1.97</td>
<td>1.97</td>
<td>0.00</td>
<td>32.30%</td>
<td>125</td>
</tr>
<tr>
<td>May 86-Nov. 88</td>
<td>0.060</td>
<td>0.66</td>
<td>1.97</td>
<td>0.00</td>
<td>33.80%</td>
<td>126</td>
</tr>
</tbody>
</table>

OLS model: $\ln(S_t/S_{t-1}) = a + b \ln(S_t/S_{t-1}) + \epsilon_t$

$^a$ $DW$ denotes the Durbin-Watson statistic.

$^b$ $\sigma_t$ denotes the annualized standard deviation of $\ln(S_t/S_{t-1})$ over the period.

$^c$ $N$ denotes the number of observations.

Furthermore, since the stochastic processes of the spot price of crude oil and of the forward convenience yields have correlated residuals, we have used a seemingly unrelated regression model to estimate the parameters $k$, $\alpha$, $\sigma_2$, and $\rho$. We ran it by using the linear discretized approximation of (2), namely

$$\delta_t - \delta_{t-1} = \alpha k + k \delta_{t-1} + \epsilon_t,$$

in conjunction with the following unrestricted regression model for $\ln(S_t/S_{t-1})$, namely

$$\ln(S_t/S_{t-1}) = a + b \ln(S_t/S_{t-1}) + \epsilon_t,$$

where we know, given the results in Table I, that $\delta = 0$ is strongly supported by the data.\(^{19}\)

\(^{19}\) Notice that the seemingly unrelated regression model with the restricted version of (11) led to almost identical estimates of $k$, $\alpha$, $\sigma_1$, and $\rho$. 
Table II

<table>
<thead>
<tr>
<th>Period</th>
<th>$k^b$</th>
<th>$t(k)$</th>
<th>$\alpha$</th>
<th>$t(\alpha)$</th>
<th>$\sigma_2$</th>
<th>$\rho$</th>
<th>$DW^c$</th>
<th>$R^2$</th>
<th>$N^*$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Jan. 84-Nov. 88</td>
<td>16.0747</td>
<td>6.74</td>
<td>0.1861</td>
<td>4.51</td>
<td>1.1211</td>
<td>0.3199</td>
<td>2.20</td>
<td>0.1527</td>
<td>253</td>
</tr>
<tr>
<td>Jan. 84-Nov. 86</td>
<td>16.8530</td>
<td>5.42</td>
<td>0.2020</td>
<td>3.36</td>
<td>1.3236</td>
<td>0.3610</td>
<td>2.24</td>
<td>0.1714</td>
<td>148</td>
</tr>
<tr>
<td>Nov. 86-Nov. 88</td>
<td>13.0572</td>
<td>3.32</td>
<td>0.1732</td>
<td>2.92</td>
<td>0.7429</td>
<td>0.1867</td>
<td>2.02</td>
<td>0.1052</td>
<td>103</td>
</tr>
</tbody>
</table>

* The seemingly unrelated regression model was fitted to estimate the coefficients of $\Delta S_t$ and $\ln(S_t/S_{t-1})$ jointly regressing, respectively, the former variable on $S_{t-1}$ and the latter on its lagged value.

* The estimates of $k$ and $\alpha$ have been annualized.

* The Durbin-Watson statistic applies to the first state variable—see equation (10)—regression component of the S.U.R. model.

* $R^2$ applies to the entire system which has been jointly estimated.

* $N$ denotes the total number of observations.

Since two extreme values of the convenience yield occurred—at the end of May 86 (see Figure 2)—exactly in the middle of the observation period, we felt it somehow arbitrary to present the subperiods’ estimates by cutting the data in the middle and hence presenting one strongly and one hardly mean reverting subperiod—by allowing overlapping observations—or simply two weekly mean reverting subperiods by allowing for non-overlapping data. This is the reason which led us to report subperiods’ estimates based on two unequal length time intervals.

In Table II, we report the annualized parameter estimates as well as summary statistics about the explanatory power of the seemingly unrelated regression model which has been applied to weekly data over the period January 6, 1984 to November 18, 1988.

The results show a very strong mean reverting pattern—high value of $k$—of the convenience yield over the entire period and the two subperiods.\(^{14}\) The value of the long run mean convenience yield $\alpha$ is fairly stable—around 18%—across time, whereas the volatility $\sigma_2$, although at very high levels, seems to be a function of the number of sharp oil price declines or increases observed. Finally, the correlation coefficient between the two processes’ residuals supports our initial conjecture about the positive relationship between the unexpected changes in the spot price and in the convenience yield of crude oil.

For the remainder of this study, we shall be using the value of the coefficients as estimated over the entire period since we feel that this is a more appropriate way to capture mean reversion on the one hand and to abstract from short term variations in the volatility of the convenience yield on the other. We summarize the relevant parameter values below.\(^{15}\)

\[
\begin{array}{ccccccc}
  & k & \alpha & \sigma_2 & \rho & \sigma_1 \\
 16.0747 & 0.1861 & 1.1211 & 0.3200 & 0.3534 \\
\end{array}
\]

Before plugging these parameter values into the partial differential equations (5) and (6) to test the two-factor model, we still need to specify and to estimate a last coefficient, namely the market price of stochastic convenience yield risk, $\lambda$.

\(^{14}\) According to Figures 1 and 2, we see that the first subperiod was characterized by a higher oil price level and by heterogeneous fluctuations, while the second led to a downward-smoother—trend in crude oil prices.

\(^{15}\) Notice that $\sigma_1$, the standard deviation of $\ln(S_t/S_{t-1})$, has been reported in Table I.
III. Estimating the Market Price of Convenience Yield Risk

As in any partial equilibrium model where one of the state variables is not the price, or yield, of the traded asset or portfolio, we are left with the delicate task of estimating the exogenously specified market price per unit of this state variable’s risk. For the purpose of this study, we shall assume that the market price of convenience yield risk \( \lambda \) is an intertemporal constant,\(^1\) a hypothesis which we shall further analyze on the basis of our empirical results.

In order to estimate \( \lambda \), we have used the market prices of all crude oil futures contracts traded on the NYMEX during the period January 6, 1984, to November 18, 1988, and compared them to their theoretical prices computed by solving numerically the partial differential equation (5) subject to the initial condition (6). More precisely, we started with three arbitrary values of \( \lambda \), computed for each of them the sum of squared errors,\(^1\) and then estimated \( \lambda^* \) assuming that the sum of squared errors is a second order polynomial in \( \lambda \). Hence, \( \lambda^* \) was set equal to the value minimizing the latter function. This procedure was then repeated until two successive optimizing values \( \lambda^*_r \) and \( \lambda^*_r+1 \) led to respective mean root squared pricing errors which differed by less than one cent. Following this procedure, the optimal value of \( \lambda \), estimated over the entire—January 6, 1984 to November 18, 1988—period from a total of 2,180 weekly futures prices and from the total sample parameter estimates\(^1\) defined in Section II, was found to be equal to \(-1.786\). The latter value led to a within sample mean relative pricing error\(^9\) of \(-0.0835\) and to a root mean squared error of \(0.68932\).

It is worthwhile to emphasize that the negative value of the latter parameter suggests that the excess expected return for convenience yield risk exposure is positive, since \( B_t \) is negative,\(^1\) and accordingly that it “pays” to bear convenience yield risk. Looking at the partial differential equations (4) and (5), the negative value of \( \lambda \) also translates into a higher risk-adjusted drift of the convenience yield \( \lambda(\alpha - \delta) - \lambda \sigma^2 \) under the equivalent martingale measure. This suggests that the covariance between expected relative changes in aggregate wealth and expected changes in the convenience yield is negative.\(^\text{21} \)

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\(^1\) See footnote 7 and Brennan (1986) for sufficient conditions leading to the latter result.

\(^2\) Notice that, when minimizing the sum of squared errors (SSE), this nonlinear estimation procedure relies on the assumption that the futures contracts pricing errors have a normal independent and identical distribution. A more general procedure which would have accounted for both cross-sectional as well as serial correlation in the residuals has been considered (relying on the construction and rebalancing of equal maturity futures portfolios and on a generalized least squares estimation procedure) would be one way to cope with this problem but finally ruled out since we were not focusing on the statistical significance of the parameter.

\(^3\) We have to point out that in order to estimate \( \lambda \) these coefficients were treated as known values.

\(^4\) Given that the errors are defined by subtracting the actual futures price from its theoretical value, this result points to a slight underpricing averaging eight cents.

\(^5\) See footnote 6, equation (1). This result is fairly intuitive and holds for any firm or optional commitment to a long position in crude oil. We shall later present numerical evidence on the negative value of such assets’ hedge ratio with respect to \( \delta \).

\(^6\) In a Cox, Ingersoll, and Ross (1986) general equilibrium framework, where the representative individual has a lognormal utility function. See footnote 7 also.
IV. The Performance of the Model in Valuing Futures Contracts

With the estimated value of $\lambda$ and of the coefficients of the joint stochastic process obtained over a 5-year reference period, we were first of all concerned by the model's actual out-of-sample performance in valuing short term financial instruments linked to the price of crude oil. For that purpose, we tested its ability to price the NYMEX crude oil futures contracts over the period ending November 23, 1988 and ending May 5, 1989, given that only the spot price of oil, the convenience yield, and the annualized Treasury bill rate with maturity as close as possible to that of the futures contract to be valued were updated every week when deriving the theoretical prices of these instruments.\(^{22}\) In order to analyze the degree of accuracy provided by the model, we report in Table III the mean pricing error and the root mean squared error computed over this 25-week period as well as a decomposition of these statistics with respect to the maturity of the contracts. For that purpose we have formed three different maturity groups, namely:

1) group 1, consisting of futures contracts with maturities up to 17 weeks;
2) group 2, consisting of futures contracts with maturities greater than 17 weeks but less than or equal to 32 weeks; and
3) group 3, consisting of all futures contracts whose maturity exceeds 32 weeks.

This decomposition has enabled us to analyze the errors by distinguishing among short, medium, and long term futures contracts while preserving a sufficiently large number of observations in each group. Moreover, it was designed to track the pricing errors with respect to a liquidity consideration since the contracts belonging to group 1 are the most actively traded, and their open interest averages between 60,000 and 10,000 contracts from the shorter to the longer delivery date represented, while those in group 3 seldom exceed 1,000 contracts in open interest and thus define the thinly traded segment of the crude oil futures market.\(^{23}\)

In Table III, we observe that the mean pricing error and the root mean squared error for the 240 contracts which have been valued reach $0.89 and $1.11, respectively. Given an average spot price for crude oil of $18.50 during this period, the magnitude of the overpricing accounts for 4.81% of the latter price. The pricing performance improves considerably as the maturity of the futures contracts is shortened: indeed, the mean pricing error and the root mean squared error shrink by a factor of more than 50% for group 1. There are two possible explanations for the fact that the performance of the model decreases as the maturity of the priced contingent claim increases. The first one stems from the structural property of the crude oil futures market resulting in a decreasing degree of liquidity—and, hence, in less reliable price quotes—over its longer term segment. Although the quality of the data provides an explanation for the maturity structure of the mispricing, it cannot be responsible for its persistence.

\(^{22}\) As before, the theoretical prices of the futures contracts were obtained by solving numerically the partial differential equation (5) subject to the initial condition (6).

\(^{23}\) See Gunnin (1984) for further evidence on the fact that only short term crude oil futures contracts trade actively.
Table III
Summary Statistics on the Two Factor Model's Pricing Errors in Valuing Futures Contracts

<table>
<thead>
<tr>
<th></th>
<th>MPE*</th>
<th>RMSE*</th>
<th>N</th>
</tr>
</thead>
<tbody>
<tr>
<td>All contracts</td>
<td>0.89</td>
<td>1.11</td>
<td>240</td>
</tr>
<tr>
<td>By maturity (τ):</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Group 1 (τ ≤ 17 weeks)</td>
<td>0.39</td>
<td>0.51</td>
<td>72</td>
</tr>
<tr>
<td>Group 2 (17 &lt; τ ≤ 32)</td>
<td>0.86</td>
<td>1.03</td>
<td>81</td>
</tr>
<tr>
<td>Group 3 (τ &gt; 32 weeks)</td>
<td>1.31</td>
<td>1.98</td>
<td>87</td>
</tr>
</tbody>
</table>


* MPE refers to the mean pricing error in dollars, namely:

\[ MPE = \frac{1}{N} \sum_{i=1}^{N} (\bar{F}_i - F_i) \]

where

\( N \) denotes the total number of price observations,

\( \bar{F}_i \) denotes the theoretical futures price,

\( F_i \) denotes the actual futures price.

* RMSE refers to the root mean squared error in dollars, namely:

\[ RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (\bar{F}_i - F_i)^2} \]

* The stochastic processes parameters are as reported for the total estimation period in Section III and all errors are computed for the estimated value of \( \lambda = -1.796 \).

in the very short term segment, namely in group 1. Hence, a second possible explanation is more likely to arise from a misspecification either of the state variables' stochastic processes or of the market price of risk, \( \lambda \). Since the latter parameter is "arbitrarily" specified in all partial equilibrium pricing models and since we have assumed, for simplicity, that it is an intertemporal constant, it appeared more natural to question the relevance of this assumption first.

Accordingly, we examined whether the model's performance could be improved by relaxing the assumption that \( \lambda \) is stationary. For that purpose we employed a "rolling over" strategy aimed at estimating \( \lambda \) over a shorter time period and using it over the subsequent time period to test the model. More precisely, we created six equal length, 5 weeks, reference periods starting October 21, 1988 and ending May 5, 1989. We then computed the optimal value of \( \lambda \) for each of the first five subperiods according to the procedure already described in Section III. In the last stage, each of the five estimates of \( \lambda \) has been used in order to price the futures contracts over the period subsequent to the one used for its estimation. Hence, \( \lambda \) estimated over the first period has been used to test the model over the second period, etc.

In Table IV, we report the estimated values of \( \lambda \) obtained over the first five periods together with their within-sample root mean squared errors. Indisputably, \( \lambda \) is a nonstationary parameter which can fluctuate dramatically among successive short time intervals (it suffices to compare \( \lambda_1 \) to \( \lambda_2 \), \( \lambda_2 \) to \( \lambda_3 \), or \( \lambda_1 \) or \( \lambda_3 \), although
Table IV
Estimation of $\lambda$ over Short-Term Periods

<table>
<thead>
<tr>
<th>Period No.</th>
<th>$\lambda_j$</th>
<th>$S_j$</th>
<th>RMSE</th>
<th>$N$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>-0.8163</td>
<td>13.96</td>
<td>0.2183</td>
<td>45</td>
</tr>
<tr>
<td>2</td>
<td>-1.9585</td>
<td>15.76</td>
<td>0.2717</td>
<td>39</td>
</tr>
<tr>
<td>3</td>
<td>-3.1268</td>
<td>17.97</td>
<td>0.3092</td>
<td>47</td>
</tr>
<tr>
<td>4</td>
<td>-3.0175</td>
<td>17.96</td>
<td>0.2427</td>
<td>50</td>
</tr>
<tr>
<td>5</td>
<td>-3.6873</td>
<td>19.84</td>
<td>0.3520</td>
<td>58</td>
</tr>
</tbody>
</table>

* Each period lasts five weeks. The first one starts October 21, 1986, and the last one ends April 7, 1989.

$\lambda_j, j = 1, \ldots, 5,$ denotes the estimated value of the market price of risk obtained during the $j$th period.

$S_j$ denotes the average spot price of crude oil observed in period $j$.

$RMSE$ denotes the root mean squared error as defined in Table III.

$N$ denotes the number of futures contract prices (and hence pricing errors) used in the period $j$ to estimate $\lambda_j$.

it remains negative throughout. Moreover, if we compare each estimate of $\lambda$ with the average spot price observed over its estimation period, we observe that its value tends to be more negative the higher the level of the spot price of crude oil. This suggests that a functional relationship between $\lambda$ and one or both state variable(s) should be further explored in order to enhance the performance of the model.24

We then examined whether the model based on an updated estimate of $\lambda$ is more successful in pricing futures contracts. In Table V, we report for each of the five “test” subperiods, numbered 2–6, the mean pricing error, the root mean squared error, and their decomposition by maturity group.

Indisputably, the results for all maturities futures contracts are improved when we account for the nonstationarity of $\lambda$. During every subperiod but the third, the mean pricing error and the root mean squared error remain substantially lower than in Table III, where we used the 5-year historical estimate of $\lambda$ equal to $-1.796$. Only in period 3, where we observed a dramatic shift in $\lambda$ from $-1.958$ to $-3.126$, did we observe a mean pricing error exceeding 50 cents and accounting for 3.5% of the average spot price of oil, which was primarily a consequence of the medium and the long term futures contracts mispricing. Globally, the overpricing tendency has been substantially reduced for all contracts as well as by maturity groups leading even to a slight underpricing of the long term futures contracts in period 4. Although the mispricing is still an increasing function of the futures contracts maturity, this pattern has also been substantially reduced and even reversed (in periods 3 and 6). These results indicate that the out-of-sample performance of the model can be greatly enhanced by taking into account the time, and presumably state variable, dependent evolution of $\lambda$.

24 We should, however, be aware of the fact that, given the partial equilibrium nature of the model, any such exogenous specification must remain consistent with the no-arbitrage condition. See Cox, Ingersoll, and Ross (1985) on that subject.
Table V

<table>
<thead>
<tr>
<th>Period</th>
<th>λ_1</th>
<th>λ_2</th>
<th>λ_3</th>
<th>λ_4</th>
<th>λ_5</th>
<th>Group 1 MPE</th>
<th>RMSE N</th>
<th>Group 2 MPE</th>
<th>RMSE N</th>
<th>Group 3 MPE</th>
<th>RMSE N</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>0.47</td>
<td>0.56</td>
<td>0.28</td>
<td>0.25</td>
<td>0.35</td>
<td>0.57</td>
<td>0.65</td>
<td>0.59</td>
<td>0.68</td>
<td>7</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>0.63</td>
<td>0.70</td>
<td>0.42</td>
<td>0.49</td>
<td>0.16</td>
<td>0.76</td>
<td>0.80</td>
<td>0.72</td>
<td>0.77</td>
<td>15</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>0.90</td>
<td>0.00</td>
<td>0.25</td>
<td>0.18</td>
<td>0.14</td>
<td>0.24</td>
<td>0.31</td>
<td>-0.21</td>
<td>0.30</td>
<td>15</td>
<td></td>
</tr>
<tr>
<td>5</td>
<td>0.47</td>
<td>0.59</td>
<td>0.25</td>
<td>0.33</td>
<td>0.15</td>
<td>0.55</td>
<td>0.63</td>
<td>0.55</td>
<td>0.68</td>
<td>26</td>
<td></td>
</tr>
<tr>
<td>6</td>
<td>0.43</td>
<td>0.22</td>
<td>0.61</td>
<td>0.34</td>
<td>0.11</td>
<td>0.48</td>
<td>0.53</td>
<td>0.02</td>
<td>0.35</td>
<td>21</td>
<td></td>
</tr>
</tbody>
</table>

a The five-week periods are indexed by j + 1, j = 1, ..., 6, to maintain consistency with the estimation periods in Table IV. Period 2 starts November 23, 1988, and Period 6 ends May 5, 1989.
b λ refers to the estimated value of the market price over the period j preceding the test period j + 1.
c S_j refers to the average spot price observed during the j + 1 period.
d MPE and RMSE refer to the mean pricing error and the root mean squared error as defined in Table III.

In general, the results reported in this section suggest that the two-factor model is a quite satisfactory tool to price short term oil-linked instruments such as futures contracts. Indeed, the results reported in Table V suggest that its performance in valuing—especially in the shortest—NYMEX crude oil futures contracts is remarkable given that we used monthly updates of λ and estimates of the stochastic processes’ parameters based on a 5-year historical period. The pricing errors observed are quite comparable to those that previous authors have obtained in pricing options using daily or weekly updated volatility estimates. Hence, it is our conjecture that the pricing performance of the two-factor model can, for practical purposes, be further enhanced by using weekly or even daily updated estimates of λ.

V. The Pricing of Long Term Oil Contingent Claims

An interesting and important application of the model is the computation of the present value of one barrel of crude oil deliverable at any time in the future. For that purpose, it suffices to solve the partial differential equation (4) subject to the boundary condition (6). We can then interpret the present value factors B(S, δ, τ) as the prices of hypothetical “spot” contracts on one barrel of oil deliverable at time T. The latter contracts would involve cash transfer at inception; that is, B(S, δ, τ) define the prices one would have to pay today in order to receive one barrel of oil in τ years time. Clearly, their computation represents the starting

55 See, in particular, Bodurtha and Courtnado (1987) and Macheth and Merville (1980) and Scott (1987) in the case of foreign currency and stock written options, respectively.
56 Since November 1986, the latter can also be applied to options written on futures contracts traded on the NYMEX. In order to price such options, one must solve the partial differential equation (4) subject to the specific boundary conditions applying to each option category. However, it is necessary to value the futures contract first since it represents “the underlying asset” of each option.
57 Contrarily to the futures contracts valued in Section IV. We shall therefore call them “spot” contracts to emphasize that they involve an initial cash outflow.
point to any capital budgeting decision involving long term firm commitment(s) to oil-linked cash flows. For example, a long term contract involving monthly delivery of $Q_j$ barrels during $J$ months has a present value $V_0$ which satisfies
\[ V_0 = \sum_{j=1}^{J} Q_j B(S, \delta, \tau_j), \tag{12} \]
where $B(S, \delta, \tau_j)$ defines the present value of one barrel of oil deliverable in month $j$. Alternatively, the same approach can be extended to price contingent commitments or contracts written on future oil-linked cash flows by solving the partial differential equation (4) subject to the appropriate boundary conditions. Since it is straightforward to modify the model in order to account for the other elements of the cash flows, such as extraction, development, or exploration costs, we shall in the present study limit ourselves to the valuation of the future output per se and leave to further specific applications the task of extending this general approach to the valuation of oil leases, oil deposits, and other long term oil-linked real or financial assets.\(^{26}\)

In Table VI, we reproduce the theoretical present value of one barrel of oil deliverable in $\tau = 1, 2, \ldots, 10$ years computed with the model using the 5-year historical estimates of the joint stochastic process described in Section II, and the estimate of the market price of risk derived over the same period. The spot price of oil ($S$), the convenience yield ($\delta$), and the risk-free interest rate ($r$)\(^{29}\) are as observed (in the *Wall Street Journal*) or computed on three distinct observation dates chosen to emphasize possible combination of the two state variables' level.

The theoretical prices computed in Table VI suggest that, using the historical parameter estimates, crude oil loses roughly 50% of its value within 5 years and 75% of its value within 10 years.

In other words, the discount factor for oil-linked investments seems to be fairly low. Such a high required rate of return also suggests that, contrary to Hotelling's Valuation Principle,\(^{30}\) the spot price of crude oil's risk adjusted growth rate lies below the interest rate, a phenomenon which could be related to the non-cooperative oligopolistic structure of the OPEC, to the crude oil extraction costs' structure, and to expected substitution effects and technological improvements. The figures in Table VI must, however, be interpreted with caution. Indeed, as we already pointed out in Section IV, the market price of convenience yield risk is a nonstationary parameter which has been estimated by minimizing the sum of squared errors arising in the pricing of short term futures contracts. Thus, if a unit of convenience yield risk is less rewarded\(^{31}\) over the long run, the value of

\(^{26}\) For previous studies using the option pricing approach—under the single state variable assumption—to value oil resources, see in particular Bjerkland and Ekern (1969) and Paddock, Siegel, and Smith (1968).

\(^{29}\) We used the longest Treasury bill yield to compute the prices of the long term "spot" contracts at each of these dates.

\(^{30}\) See Miller and Upton (1986) and Sundaresan (1984) for a detailed explanation of Hotelling's Principle and of its implications for valuing natural resources.

\(^{31}\) In particular, due to the possibility of replenishing inventories and monitoring supply and demand over the long run, the expected mean reversion in the convenience yield might become the dominant component of its evolution over the long run.
Pricing of Oil Contingent Claims

Table VI

**Present Value \( B(S, \delta, \tau) \) of One Barrel of Oil**

The present value factors are given in dollars per barrel and are the numerical solutions of the partial differential equation (4) subject to the initial condition (6) on those three different observation dates.

<table>
<thead>
<tr>
<th></th>
<th>7/6/84</th>
<th>3/21/86</th>
<th>5/23/89</th>
</tr>
</thead>
<tbody>
<tr>
<td>Deliverable in</td>
<td>( S = 29.65 )</td>
<td>( S = 13.94 )</td>
<td>( S = 19.05 )</td>
</tr>
<tr>
<td>( \delta )</td>
<td>3.30%</td>
<td>13.71%</td>
<td>85.50%</td>
</tr>
<tr>
<td>( \tau )</td>
<td>11.80%</td>
<td>6.90%</td>
<td>8.50%</td>
</tr>
<tr>
<td>No. of Years</td>
<td>1</td>
<td>2</td>
<td>3</td>
</tr>
<tr>
<td>( \lambda )</td>
<td>26.12</td>
<td>12.39</td>
<td>16.14</td>
</tr>
<tr>
<td>( \delta )</td>
<td>2</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>( \tau )</td>
<td>20.02</td>
<td>9.48</td>
<td>12.98</td>
</tr>
<tr>
<td>( \delta )</td>
<td>4</td>
<td>17.54</td>
<td>8.29</td>
</tr>
<tr>
<td>( \tau )</td>
<td>15.36</td>
<td>7.26</td>
<td>10.85</td>
</tr>
<tr>
<td>( \delta )</td>
<td>6</td>
<td>13.46</td>
<td>6.35</td>
</tr>
<tr>
<td>( \tau )</td>
<td>11.78</td>
<td>5.55</td>
<td>7.29</td>
</tr>
<tr>
<td>( \delta )</td>
<td>8</td>
<td>10.33</td>
<td>4.36</td>
</tr>
<tr>
<td>( \tau )</td>
<td>9.04</td>
<td>4.25</td>
<td>5.50</td>
</tr>
<tr>
<td>( \delta )</td>
<td>10</td>
<td>7.82</td>
<td>3.71</td>
</tr>
</tbody>
</table>

\( \lambda \), equal to \(-1.796\), used to compute the present value factors is understated and the latter hence become underestimated.

Finally, an important feature of the model is its consistence with Samuelson's (1965) hypothesis regarding the decreasing pattern of futures contracts prices' volatility with respect to maturity. Single factor contingent claims pricing models rest on the assumption that the prices of any claim and of the underlying asset are perfectly correlated, and, if the state variable follows a geometric Brownian motion, the model implies that any futures and the underlying asset have equal return volatility. Empirical studies\(^{32}\) on futures contracts have rejected this hypothesis and hence raise doubts about applying such models to price futures contracts or "spot" contracts tied to the price of the commodity. Within the context of this model, we are able to emphasize the imperfect correlation between the claim and the spot crude oil prices as well as the decreasing pattern in the claims' return volatility as their maturity is lengthened. In Table VII, we illustrate the latter point by computing the hedge ratios with respect to the spot price (\( \Delta_S \)) and to the convenience yield of crude oil (\( \Delta_c \)) for the long term contracts\(^{33}\) valued in Table VI. Clearly, the hedge ratio with respect to the spot price is, as expected, less than unity and decreases as the maturity of the contract is lengthened. Overall, the difference in the hedge ratios \( \Delta_S \), holding maturity constant, is rather small within scenarios and decreases with maturity. According to the model, a $1.00 change in the spot price of crude oil will, ceteris paribus,\(^{34}\) lead to

\(^{32}\) See, in particular, Fama and French (1988) and Duffie (1989).

\(^{33}\) Although this decaying pattern has originally been pointed out for futures prices, it applies identically to the present value factors \( B(S, \delta, \tau) \) since their value differs only with respect to the time value of money.

\(^{34}\) This is a purely mathematical comparative static analysis since a change in \( \delta \) will generally be expected to occur simultaneously and since, due to the correlation of the two factors, their effects on the prices of the contracts cannot be isolated in practice.
Table VII
Hedge Ratios of the Present Value Factors with Respect to \( S \) and \( \delta \)
The spot price, the interest rate, the convenience yield, and the present value factors \( B(S, \delta, \tau) \) for each observation date are as defined in Table VI.

<table>
<thead>
<tr>
<th>( \tau ) (Years)</th>
<th>7/6/84</th>
<th>3/21/86</th>
<th>5/23/89</th>
<th>7/6/84</th>
<th>3/21/86</th>
<th>5/23/89</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>0.881</td>
<td>0.887</td>
<td>0.847</td>
<td>-1.636</td>
<td>-0.766</td>
<td>-1.000</td>
</tr>
<tr>
<td>2</td>
<td>0.772</td>
<td>0.776</td>
<td>0.740</td>
<td>-1.423</td>
<td>-0.658</td>
<td>-0.877</td>
</tr>
<tr>
<td>3</td>
<td>0.676</td>
<td>0.678</td>
<td>0.647</td>
<td>-1.247</td>
<td>-0.586</td>
<td>-0.762</td>
</tr>
<tr>
<td>4</td>
<td>0.580</td>
<td>0.588</td>
<td>0.587</td>
<td>-1.092</td>
<td>-0.513</td>
<td>-0.672</td>
</tr>
<tr>
<td>5</td>
<td>0.518</td>
<td>0.520</td>
<td>0.497</td>
<td>-0.953</td>
<td>-0.447</td>
<td>-0.588</td>
</tr>
<tr>
<td>6</td>
<td>0.463</td>
<td>0.454</td>
<td>0.435</td>
<td>-0.834</td>
<td>-0.391</td>
<td>-0.516</td>
</tr>
<tr>
<td>7</td>
<td>0.398</td>
<td>0.398</td>
<td>0.382</td>
<td>-0.730</td>
<td>-0.343</td>
<td>-0.452</td>
</tr>
<tr>
<td>8</td>
<td>0.347</td>
<td>0.348</td>
<td>0.334</td>
<td>-0.638</td>
<td>-0.300</td>
<td>-0.396</td>
</tr>
<tr>
<td>9</td>
<td>0.304</td>
<td>0.304</td>
<td>0.293</td>
<td>-0.560</td>
<td>-0.264</td>
<td>-0.347</td>
</tr>
<tr>
<td>10</td>
<td>0.266</td>
<td>0.266</td>
<td>0.256</td>
<td>-0.490</td>
<td>-0.231</td>
<td>-0.304</td>
</tr>
</tbody>
</table>

\( \Delta_5^a \) is the first derivative of \( B(S, \delta, \tau) \) with respect to \( S \) computed numerically. 
\( \Delta_\delta^b \) is the first derivative of \( B(S, \delta, \tau) \) with respect to \( \delta \) computed numerically.

an instantaneous change of 88 to 26 cents for contracts of one to ten years maturities, respectively.

The decaying structure of the volatility of the present value factors is further reinforced by the fact that the hedge ratios with respect to the convenience yield are, as expected, negative but smaller in absolute magnitude as maturity increases. Hence, an increase in the convenience yield by 0.01, from 3.3% to 4.3% on July 6, 1984, would have induced a decrease of 1.092 (0.01 \( \times \) \(-1.092\)) cents in the value of \( B(S, \delta, 4) \) and of 0.49 cents only in the value of \( B(S, \delta, 10) \). The difference between the hedge ratios \( \Delta_\delta \) across scenarios does not—as with \( \Delta_5 \)—completely vanish for longer maturities contracts.

We thus can conclude by saying that the sign and the maturity pattern of both hedge ratios are consistent with the economical interpretations of the theory of storage. The price of a futures or “spot” contract will, ceteris paribus, increase with an increase in the spot price of oil. However, since the latter increase is generally associated with a higher convenience yield, the final impact will, even for shorter maturity contracts, be attenuated with respect to its predicted magnitude under a single state variable model. In this respect, the two-factor model leads to less variability in futures contracts prices,\(^{36}\) and this result, coupled with the maturity decaying volatility pattern it induces, allows the model to be fairly robust with respect to the empirical evidence on the relationship between futures and spot commodity prices.

VI. Conclusion

In this study we have presented a two-factor model aimed at valuing oil-linked assets under the assumption that the spot price of oil and the instantaneous net

\(^{36}\) Especially at high spot price, high convenience yield, and hence low inventory levels as predicted by the theory of storage. See, in particular, Fama and French (1988).
convenience yield of oil follow a joint stochastic process. The model seems to be a reliable instrument for the purpose of valuing short term financial instruments, such as futures contracts, when we update the estimated value of the market price of risk $\lambda$. Clearly, another very promising area lies in its application to value real assets indexed on the price of oil. In this perspective, we have shown how to compute the present value of one barrel of crude deliverable at any arbitrary future date. From there on, the possible applications of this partial equilibrium model extend to the pricing of oil reserves and oil leases and to the timing of the decisions to extract, to develop, or to explore an oil field, simply by adding additional technological and financial data to determine the cash flows and by solving the partial differential equation (4) under the appropriate set of boundary conditions.

Finally, we wish to conclude this paper by emphasizing how important the concept of a convenience yield is for a non speculative commodity. In the case of crude oil, this is obvious in light of the strategic nature of the benefits it provides to its owner and given the large fluctuations in crude oil inventories. Nevertheless, it might be useful to examine whether the convenience yield of other commodities or natural resources also evolves stochastically, which could—at least partially—explain the empirically observed relationship between futures and spot prices and between their volatilities. An affirmative answer would hence suggest that the structure of this two-factor—partial equilibrium—crude oil contingent claim pricing model could be generalized for the purpose of valuing and hedging other types of commodity-linked cash flows.

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