

# Technical Analysis: Mean-Reverting Oil Price Dynamics

## 1. Model Selection: Why Ornstein-Uhlenbeck?

Standard equity models typically utilize Geometric Brownian Motion (GBM), which assumes a constant upward drift. However, commodity markets such as Brent Crude are fundamentally different due to supply and demand equilibrium. We select the \*\*Ornstein-Uhlenbeck (OU) process\*\* because it accounts for **Mean Reversion**—the tendency of price levels to return to a long-term equilibrium ( $\theta$ ) after a shock.

## 2. Stochastic Differential Equation (SDE)

The spot price  $X_t$  is modeled by the following SDE:

$$dX_t = \kappa(\theta - X_t)dt + \sigma dW_t \quad (1)$$

Where:

- $\kappa$ : Reversion speed (calculated via regression or observation).
- $\theta$ : The long-term mean price, empirically derived as **\$79.75**.
- $\sigma$ : Annualized volatility, empirically derived as **33.93%**.
- $dW_t$ : Increment of a standard Wiener process.

## 3. Mathematical Proof using Ito's Lemma

To solve for the price at time  $t$ , we apply \*\*Ito's Lemma\*\* to the transformation  $f(X_t, t) = X_t e^{\kappa t}$ . Applying the Lemma:

$$d(X_t e^{\kappa t}) = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial X} dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial X^2} (dX_t)^2 \quad (2)$$

Calculating partial derivatives:

- $\frac{\partial f}{\partial t} = \kappa X_t e^{\kappa t}$
- $\frac{\partial f}{\partial X} = e^{\kappa t}$
- $\frac{\partial^2 f}{\partial X^2} = 0$

Substituting Eq. (1) into the expansion:

$$d(X_t e^{\kappa t}) = \kappa X_t e^{\kappa t} dt + e^{\kappa t} [\kappa(\theta - X_t)dt + \sigma dW_t] \quad (3)$$

Simplifying the drift terms:

$$d(X_t e^{\kappa t}) = \kappa \theta e^{\kappa t} dt + \sigma e^{\kappa t} dW_t \quad (4)$$

## 4. Numerical Discretization

For the Python implementation in `src/ou_model.py`, we utilize the **Euler-Maruyama method**. The discrete update rule is:

$$X_{t+\Delta t} = X_t + \kappa(\theta - X_t)\Delta t + \sigma\sqrt{\Delta t}\epsilon \quad (5)$$

Where  $\epsilon \sim N(0, 1)$ . This allows for the high-resolution simulation of paths that oscillate around the equilibrium mean.