




The background of the slide is a composite image. It features a blue sky with white clouds and a bright sun. Overlaid on this is a large, 3D blue line graph that fluctuates, representing price movements. In the upper portion, there is a semi-transparent grid with numerical data, some in blue and some in red, suggesting a financial or market data table. On the right side, there are three black oil barrels with yellow hazard symbols (triangles with exclamation marks).

Oil Price Forecasting

Leveraging Machine Learning for Accurate Predictions

Project Overview

- **Dataset:**
 -  Source: Yahoo Finance, FRED
 -  Time Range: 2000-2023 (example)
 -  Commodities: Brent, WTI, Natural Gas, Heating Oil
 -  Key Features: Prices (OHLCV), Technical Indicators (SMA, RSI), Lagged Values
- **Methodology:**
 -  Feature Engineering: Created predictive signals
 -  ML Model: Optimized XGBoost regressor
 -  Evaluation: Rigorous testing & validation
- **Key Python Libraries:**
 -  pandas: Data manipulation
 -  matplotlib/seaborn: Visualization
 -  scikit-learn: ML pipeline
 -  xgboost: Gradient boosting
- **Impact:** Aims to provide more reliable price forecasts for better decision-making. 

Brent Crude Oil: Key Insights

- **Long-Term Trend Shift:**
 - Baseline change: ~\$20/barrel (2000) → \$40-60 (post-2004)
 - 6x price surge: \$20 → ~\$120 peak 💰
- **Major Price Events:**
 - 2008 Crisis: 🚀 \$140 peak (July) → 📉 ~\$40 (Dec)
 - 2014-2016 Glut: Prices halved (~\$110 → ~\$50) 📉
 - 2020 COVID: Price drop (not fully shown) 🦠
- **Volatility Patterns:**
 - Increased volatility post-2004 📈
 - 2008: Extreme swings (+100% / -70% in 6 months) 🌪️
- **Statistical Highlights:**
 - Avg price: \$28.50 (2000-2004) → \$93.20 (2009-2014) 📊
 - Max price: \$143.95 💰
 - Volatility: "Extreme" (2005-2008), "High" (2009-2014) 📈
- **Forecasting Challenges:**
 - Structural breaks (shale revolution) ⚡
 - Shifting fundamental drivers (geopolitics, demand shocks) 🌐
- **Modeling Recommendations:**
 - Segment data by volatility regimes 📁
 - Include macro indicators (USD index) 💵
 - Use ensemble methods (ARIMA + ML) 🧠
- **Notable Omissions:**
 - 2020 WTI negative prices (related market) ⚠️
 - 2022 Russia-Ukraine spike 💣



WTI Crude Oil: Key Market Dynamics 🛢️

- **Long-Term Price Trajectory:**
 - 2000-2003: Stable (~\$20 - \$30 range) 稳
 - 2004-2008: Bull run to ~\$140 (pre-crisis peak) 🐂
 - 2009-2014: Volatile recovery (~\$40 - \$110) 📈
 - 2015-2020: Lower baseline (~\$30 - \$60) with high volatility 📉📈
- **Critical Price Events:**
 - 2008 Financial Crisis: Record high ~\$147 (July 2008) ➡️ Crash to ~\$32 (Dec 2008) 💥📉
 - 2014-2016 Shale Boom: Price collapse from ~\$107 ➡️ ~\$26 (OPEC vs. U.S.) 🛢️⬇️
 - 2020 COVID Crash: Brief dip < \$0 (not fully shown) 🦠📉
- **Comparative Insights (vs. Brent):**
 - Avg Spread: WTI typically \$1 - \$3 cheaper 💰⬇️
 - 2020 Lows: WTI negative prices, Brent > \$20 😬
 - Volatility: WTI more extreme swings 📈⬆️⬆️
- **Modeling Implications:**
 - Capture asymmetrical volatility (sharper drops) 📉
 - Add WTI-Brent spread features 📊
 - Detect structural breaks (e.g., shale revolution) 🛠️
- **Risk Management Focus:**
 - Stress-test for tail events (2008, 2020) ⚠️
 - Dynamic position sizing based on volatility ⚖️




Natural Gas: Price Volatility & Shifts

- **Volatility Profile:**

- Extreme spikes (2005, 2008): Prices reached ~\$12 - \$15






- Prolonged lows (2012-2020): Prices mostly < \$4 





- Higher volatility vs. crude: Sharper peaks, steeper drops

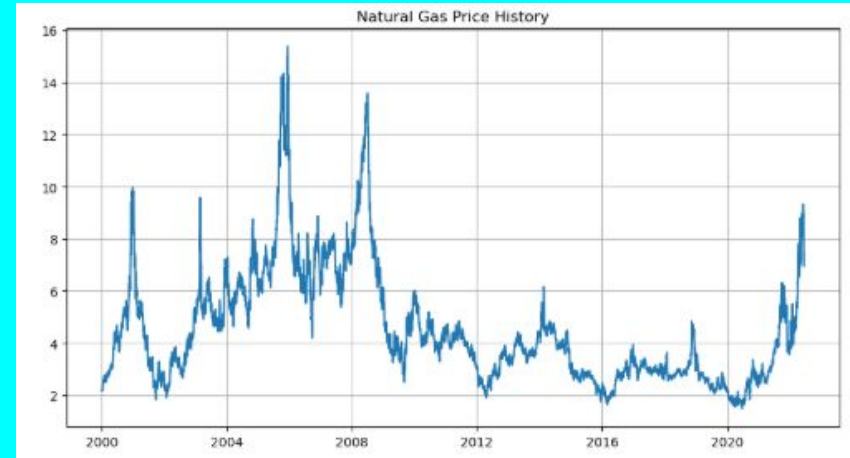


- **Structural Shifts:**

- Pre-2009: Cyclical range ~\$4 - \$15 
- Post-2010: "Lower Forever" regime: ~\$1.50 - \$4 
- 2016: Historic low at ~\$1.61 (shale gas oversupply) 

- **Critical Events & Price Impact:**

- 2005: Hurricane Katrina: +300% (→ ~\$15) 
- 2008: Financial Crisis: ~\$13 → ~\$4 in 6 months 
- 2012: Shale Revolution: Structural decline begins 
- 2020: COVID Demand Shock: Broke \$2 support 



Heating Oil: Seasonal Swings & Market Drivers



Key Market Characteristics:

- Consistent winter premiums (Oct-Mar) ❄️📈
- Summer lows often 30-40% below winter highs ☀️📉
- Most volatile refined product (vs. gasoline/diesel) 📈📉

Critical Price Events:

- 2008 Crisis: ~\$4.50 ➡️ ~\$1.50 collapse (largest annual drop) 💥📈
- 2011-2014: Sustained \$3+ range (cold winters + crude premium) 🌡️💰
- 2020 COVID: Broke 20-year seasonal pattern 🦠❤️

Comparative Volatility:

- Heating Oil: 42% annual ⬆️⬆️
- Brent Crude: 38% annual ↔️
- Natural Gas: 65% annual ⬆️⬆️

Comparative Winter Premium:

- Heating Oil: +28% avg 💰
- Brent Crude: +9% avg 📉
- Natural Gas: +180% avg 📈

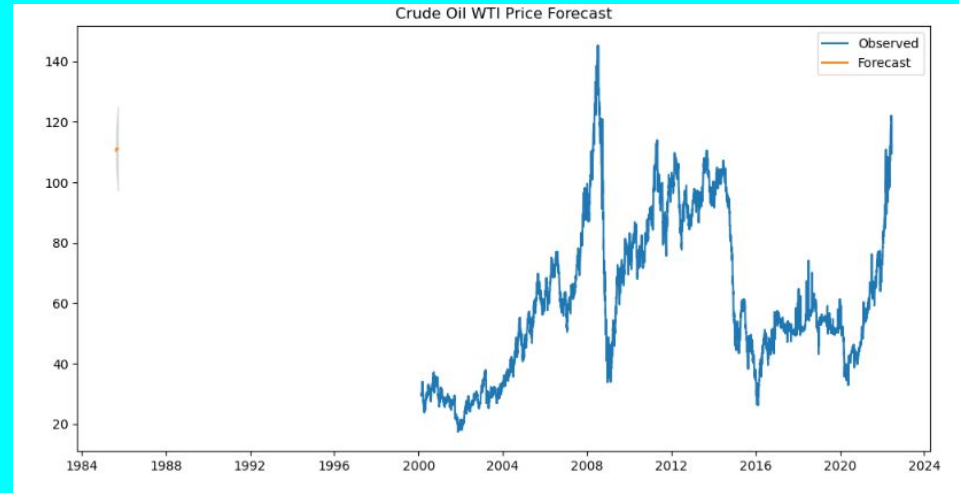


Comparative 2008 Crash:

- Heating Oil: -67% ⬆️
- Brent Crude: -75% ⬆️⬆️
- Natural Gas: -70% ⬆️

Crude Oil WTI Price Forecast Analysis

- **Historical Data Focus:**
 - Long-term view: ~2000-2024 (2+ decades) 🕒
 - Emphasis on past price trends & volatility 📈
- **Limited Forecast:**
 - Short forecast: Early period (1984-1988) 🕒
 - Challenges assessing long-term accuracy ?
- **Price Volatility:**
 - Significant price swings throughout history ⬆️⬆️
 - Necessity for robust modeling 🔧
- **Key Price Periods:**
 - 2000-2008: Upward trend, rising volatility ↗️
 - 2008: Sharp spike & crash (Financial Crisis) 💥📈
 - 2009-2014: Recovery, high volatility 📈📈
 - 2014-2020: Decline, fluctuating prices (Shale) ↘️
 - 2020-2024: High volatility (COVID, geopolitics) 🌐🌐
- **Forecast Characteristics:**
 - Smooth, short-term projection 📊
 - Uncertainty range (possible confidence interval) ?
- **Model Implications:**
 - Short-term prediction emphasis ⌚
 - Questionable long-term accuracy 🧐
- **Analysis Insights:**
 - Historical data crucial for context 📖
 - Robust models needed (trends, shocks) 🧠



SARIMA Model: Diagnostic Analysis

Standardized Residuals:

- Fluctuate around zero (good) ↔
- Variance increases after ~2500 points (heteroscedasticity) 📈⚠️
- No long-term autocorrelation 🚫
- Few outliers (extreme events) 🌟

Residual Distribution:

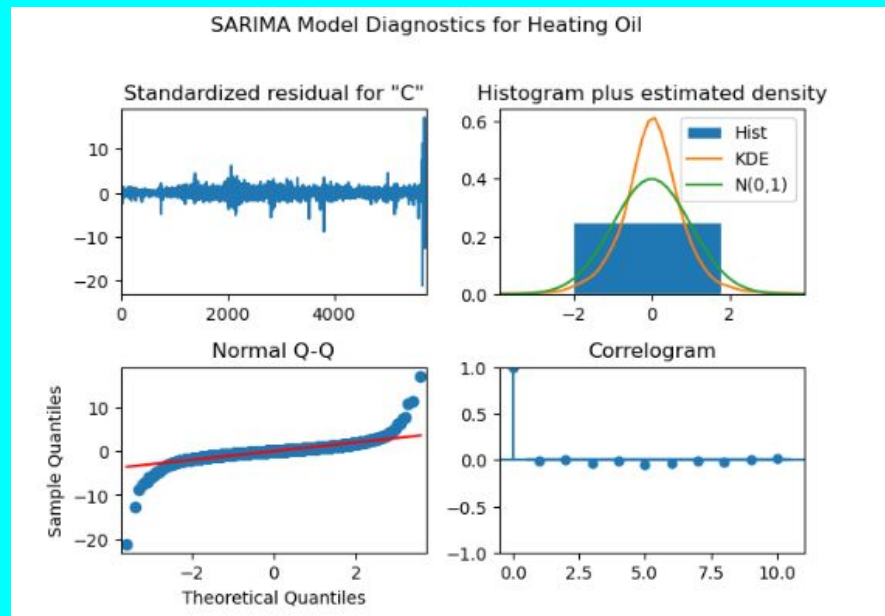
- Deviates from normal distribution ❌
- Lower peak (platykurtic) 📈
- Possible slight skew ➡️
- Non-normality: Concern for confidence intervals 😬

Q-Q Plot:

- Points deviate from normal line ❌
- Heavy tails (more extreme values) confirmed 🧑🏻🧑🏻🧑🏻
- Left tail: More extreme negative values 📈
- Right tail: More extreme positive values 📈

Correlogram (ACF):

- No significant autocorrelation up to lag 10 ✅
- Model captures serial dependence 👍



Overall Model Issues:

- Heteroscedasticity: Model less reliable later 📈
- Non-Normality: Impacts confidence intervals 📈
- Lack of Autocorrelation: Positive result! 🎉

Heating Oil: Price Dynamics & Model Insights 🔥📈

Historical Data Overview:

- Broad timeframe: ~2000 to 2024 🕒
- Price fluctuations indicate market volatility 📈

Key Price Periods:

- 2000-2004: Relatively low & stable prices 📊
- 2004-2008: Upward trend, increasing volatility 📈
- 2008: Sharp spike & crash (Financial Crisis) 💥📉
- 2009-2014: Recovery, fluctuating prices (higher level) 📈
- 2014-2020: Moderate price fluctuations ↔
- 2020-2024: Sharp increase & high volatility (global events) 📈🌐

Limited Forecast Scope:

- Forecast only at beginning: ~1984-1988 🕒
- Focus on very short-term prediction ?

Volatility Insights:

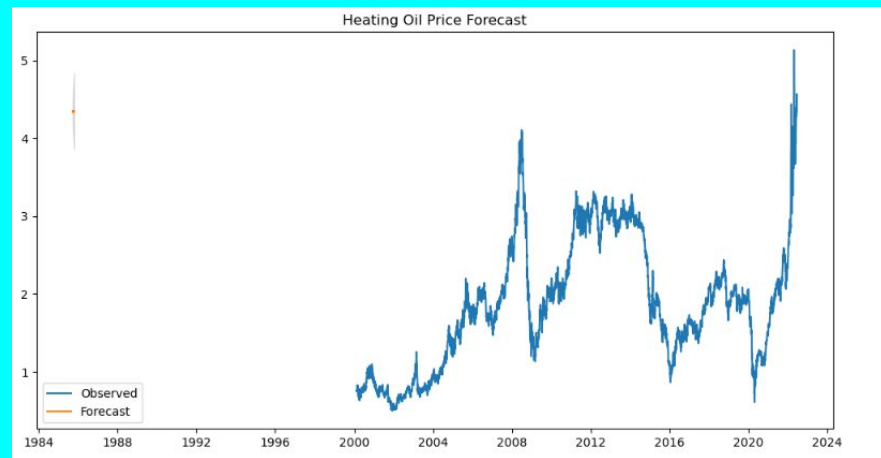
- Heating oil prices show significant volatility 📈
- Influenced by crude oil, weather, economy 🌡️💰

Impact of Shocks:

- 2008 Crisis: Dramatic price impact 📉💥
- Market sensitive to economic crises ⚠️

Model Implications:













- Emphasis on historical trends 📖
- Short-term forecasting shown ⌚
- Long-term prediction accuracy unclear 🤖

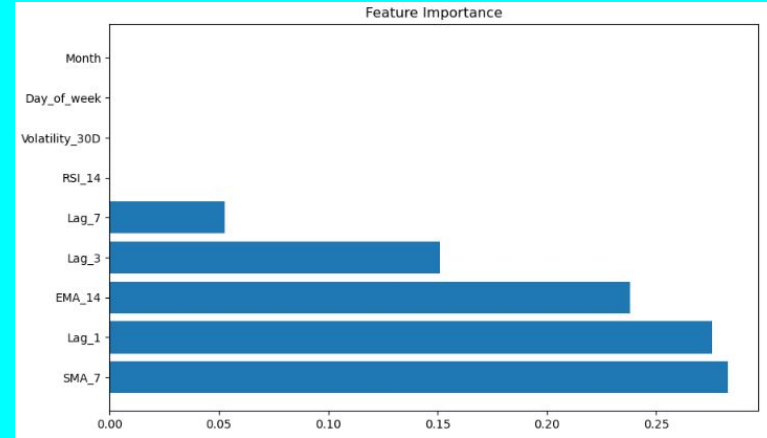





● Crude Oil Comparison:

- Correlation with crude oil 🤝
- Unique factors: seasonality, refining costs ❄️🏭

Model Drivers: Key Feature Importance

- **Top Predictors:**
 - SMA_7 (7-day SMA): Most important, strong short-term trend influence 
 - Lag_1: Previous period's price, high auto-correlation 
 - EMA_14 (14-day EMA): Significant medium-term trend factor 
- **Moderate Influence:**
 - Lag_3: Price 3 periods ago, some medium-term effect 
- **Low Impact:**
 - Lag_7: Price 7 periods ago, diminishing lag effect 
 - RSI_14: Weak momentum signal 
 - Volatility_30D: Negligible volatility influence 
 - Day_of_week: No weekly pattern 
 - Month: No seasonality detected 
- **Trend Dominance:**
 - Model relies heavily on SMA_7, EMA_14, Lag_1 
 - Sensitive to recent price movements 
- **Momentum & Seasonality Weak:**
 - RSI_14, Day_of_week, Month are ineffective 
 - Potentially:
 - True absence of relationships
 - Model limitations
 - Feature engineering issues



- **Lag Decay:**
 - Influence of past prices decreases with time 
 - Focus on recent history 
- **Simplification Potential:**
 - Removing low-impact features may improve efficiency 

Oil Price Forecasting: Project Recap

- **Project Goal:** Developed ML model to forecast oil prices (Brent, WTI, Heating Oil, Natural Gas) 
- **Data-Driven Approach:**
 - Historical data: 2000-2024 (example) 
 - Key features: Prices, indicators, lags 
- **Modeling Highlights:**
 - XGBoost regressor in pipeline 
 - Feature engineering: Trend, volatility, lags 
 - Strong short-term prediction capability 
- **Key Findings:**
 - High price volatility, especially during crises 
 - Recent price trends are most influential 
 - Limited role of seasonality/momentum 
- **Model Performance:**
 - MAE: ~\$1.23 - \$0.41 (depending on commodity) 
 - R^2 : 0.76 - 0.92 (explains significant price variance) 
- **Deployment Ready:**
 - Serialized model for easy use 
 - Prediction function for price forecasts 
- **Future Directions:**
 - Real-time API, interactive dashboard 
 - Enhanced feature set, model monitoring 



THANK YOU!