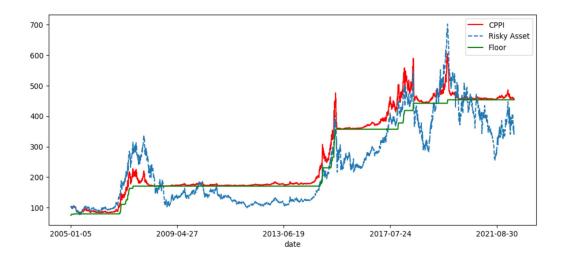
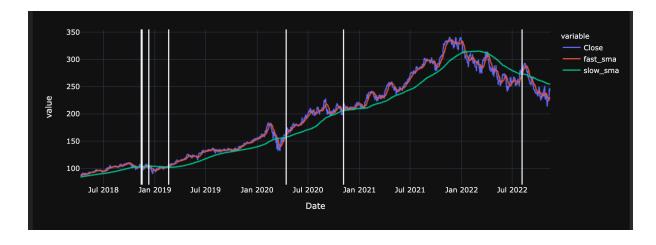


Part 1: Strategies and ML

• CPPI



• Moving Average

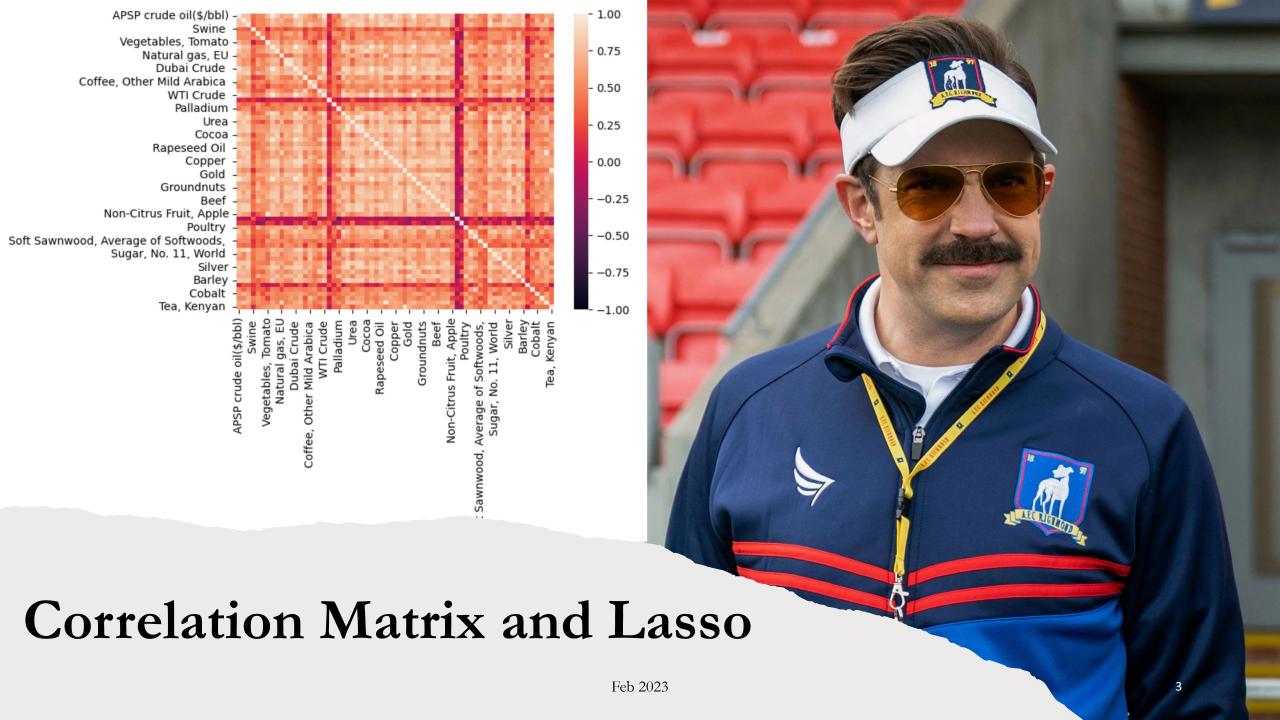


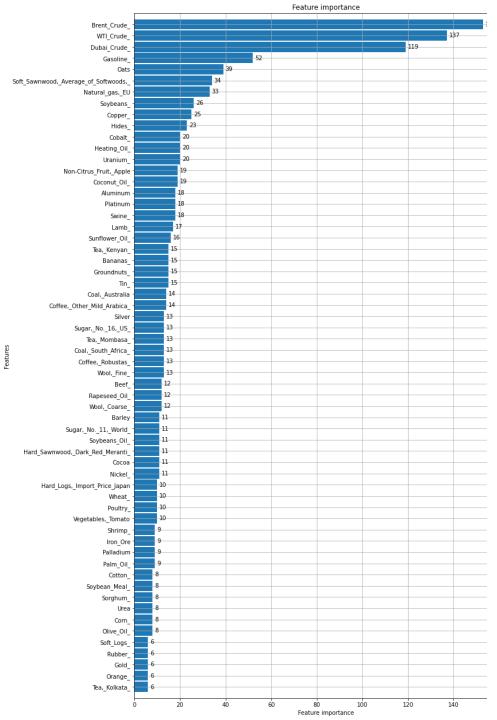
Macro

IMF Commodities Data 1990-2022

'APSP crude oil(\$/bbl)







Running lightgbm to determine features importance

- Oats
- Softwood
- Soybeans
- Copper
- Hides

- Cobalt
- Heating Oil
- Uranium
- Apples
- Coconut Oil

Tests

- F-testT-test
- VIF Score

```
from statsmodels.stats.outliers_influence import variance_inflation_factor
def run vif(df):
 vif = pd. DataFrame()
 vif['features'] = df.columns
 vif['vif'] = [variance_inflation_factor(df.values, i) for i in range(df.shape[1])]
 return(vif)
```

	features	vif
1	Uranium	4.003692
5	Poultry	12.019783
4	Coffee, Robustas	13.976314
3	Lamb	17.993315
2	Swine	20.740740
0	Hides	22.985027

	features	vit
0	Uranium	3.602549
1	Swine	15.335815
2	Lamb	13.090992
3	Coffee, Robustas	9.895261
4	Poultry	10.329726

	features	vif
0	Uranium	3.526962
1	Lamb	8.060172
2	Coffee, Robustas	9.223281
3	Poultry	9.219645

Predictive Models

Linear Regression

- R2: 0.16424671451356254
- MAPE: 0.27659548231379666

Gradient Boosting

- R2: -0.40962697675375837
- MAPE: 0.40345644928973884

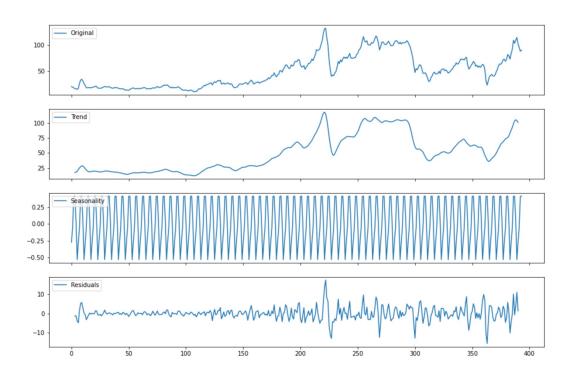
```
df_chosen2 = df[['Uranium', 'Lamb', 'Coffee, Robustas', 'Poultry']]

X, y = df[['Uranium', 'Lamb', 'Coffee, Robustas', 'Poultry']], df['APSP crude oil($/bbl)']

X_train, X_test, y_train, y_test = X[X.index < norwgd], X[X.index >= norwgd], y[y.index < norwgd], y[y.index >= norwgd]
```

Future developments

Seasonality



Macroeconomics Footprint

- ESG
- Covid's impact



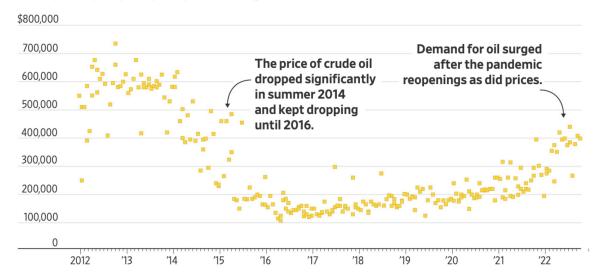


Why ESG?

Affects drilling

Drillship day rates

Drillship lease price per day, computed from signed contracts



Source: Westwood Global Energy Group Camille Bressange/The Wall Street Journal

Affects prices and business



Covid's Impact

China's zero covid policies and Crude's price explosions

Deaths and Government spendings

Healthcare Trends

