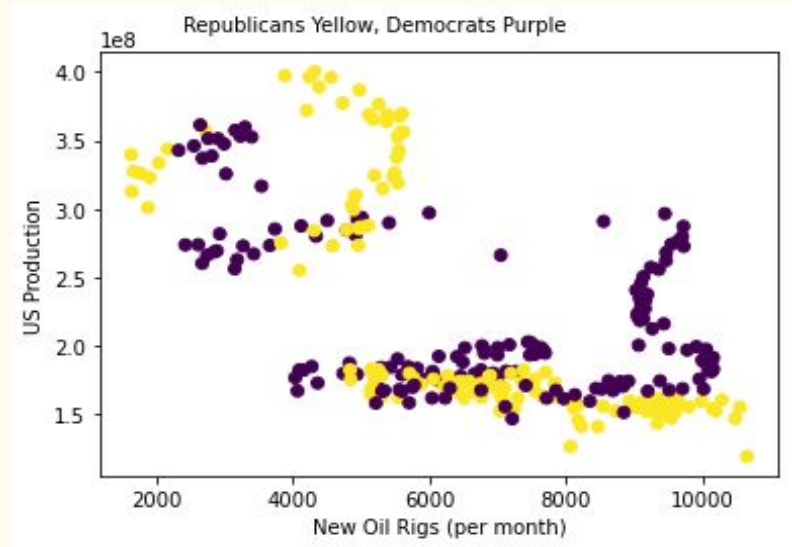
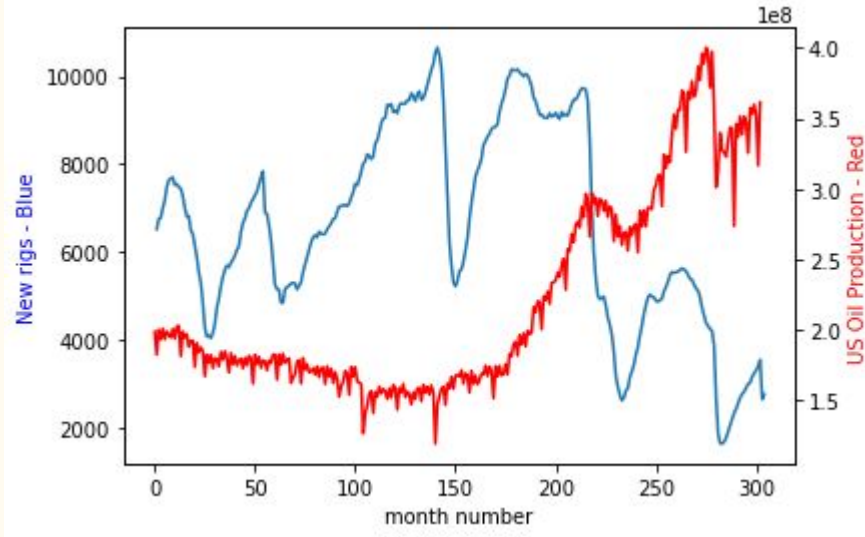


# Oil and Gas Analysis

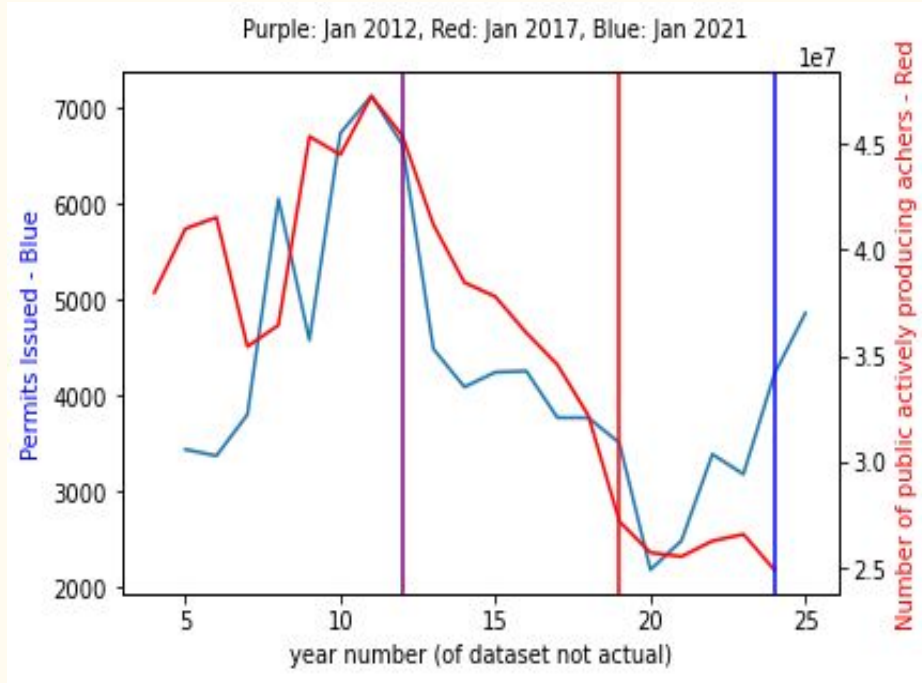
—

# US Oil Health: Construction of New Oil Rigs

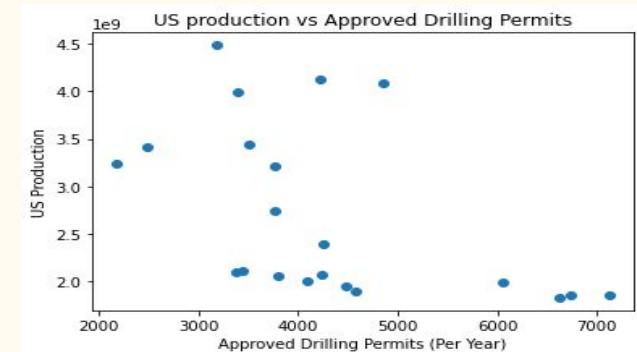
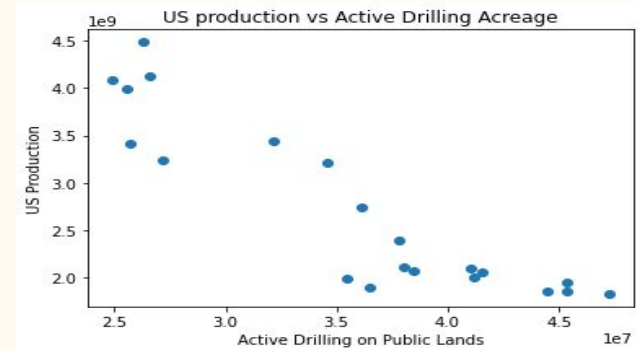


- No correlation between new oil rigs and US oil production
- No correlation between new oil rigs and presidential party (1997-2021)

# US Oil Health: Gov't Drilling Permits and Active Acreage

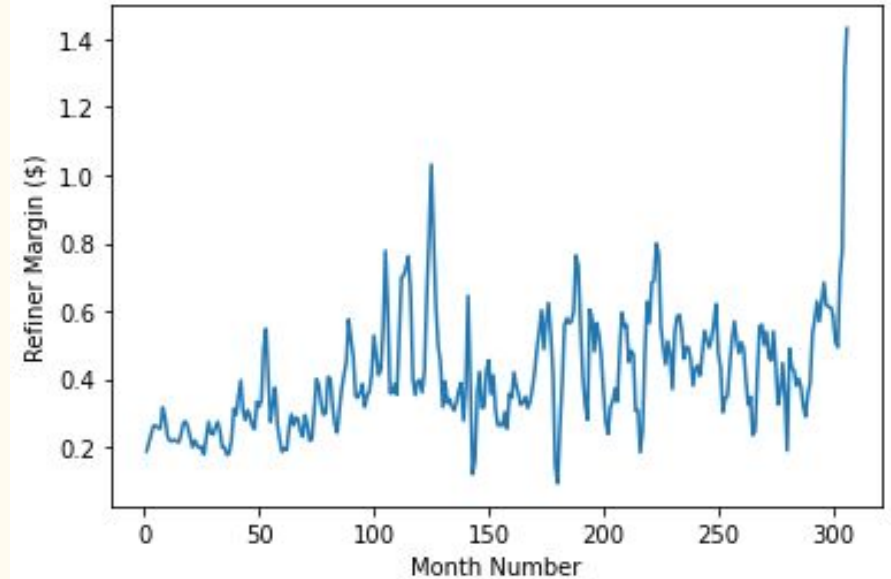
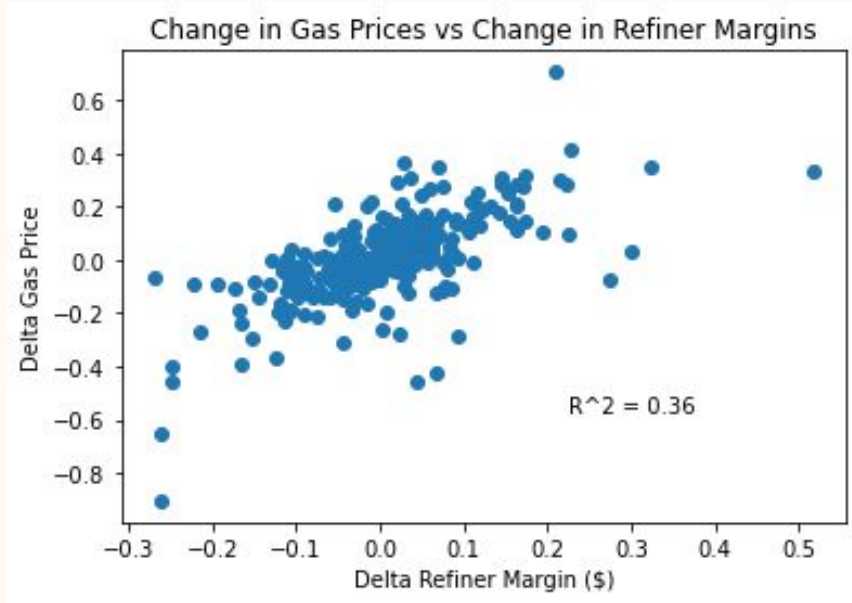


- Permits correlate to active drilling
- Active Drilling is negatively correlated with production (not all wells produce & data for public drilling only)



- No correlation between approved permits and US production
- Biden approved more Drilling permits than Trump YoY

# Refiner Profit Margin Analysis



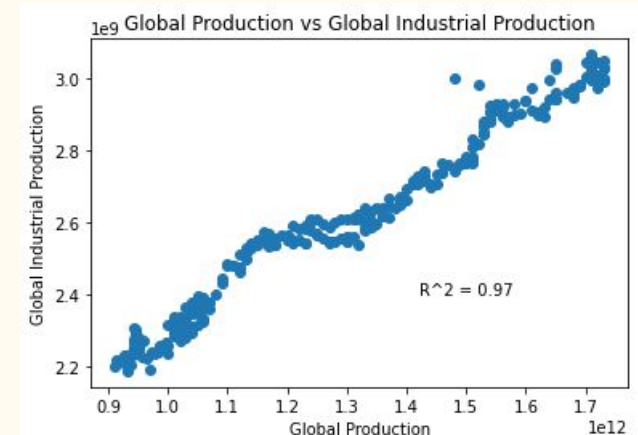
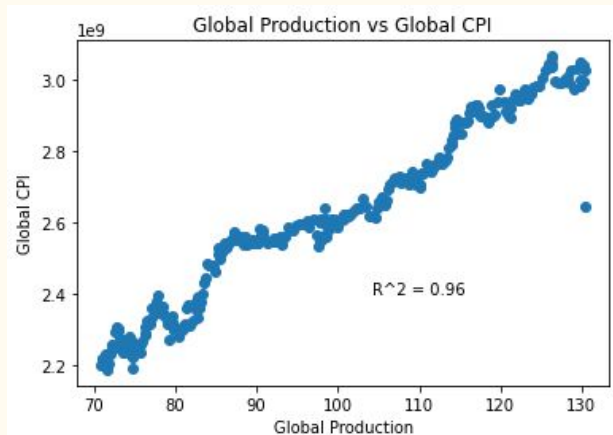
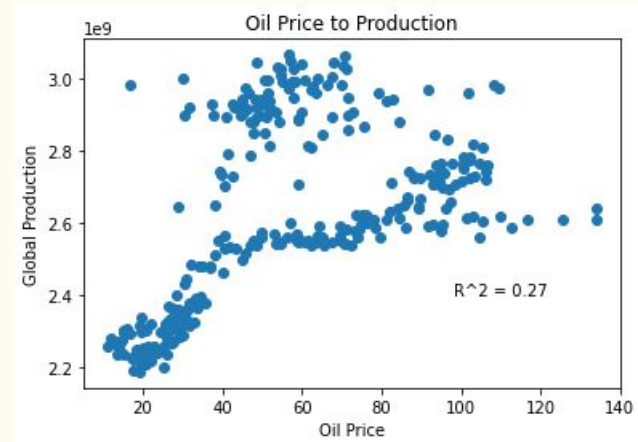
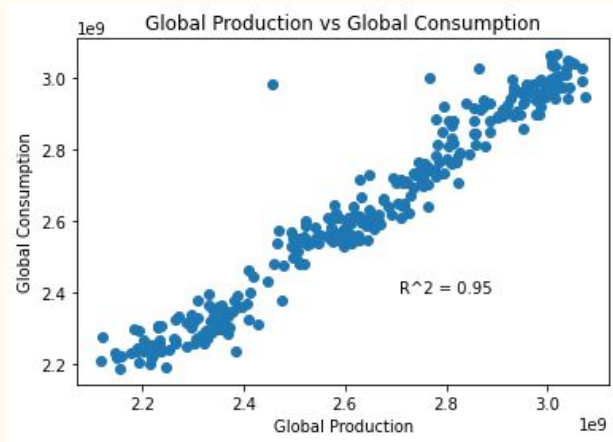
- Change in refinery profit margins correlate to change in gas prices (1 cent in margins = 3 cents in gas price on average)
- Highest refinery margins since at least 1997 have occurred in every month of 2022 (Data ends in May)

# Oil Production Correlations

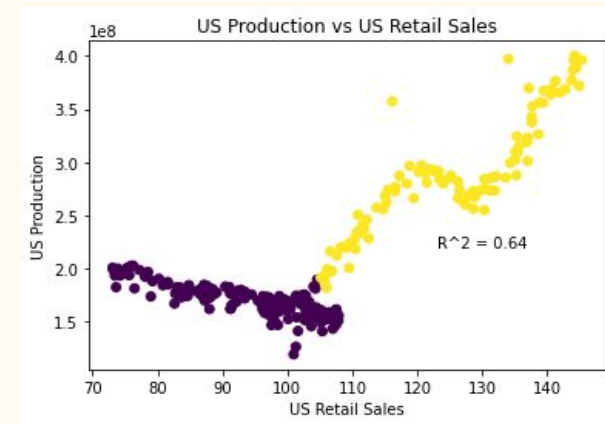
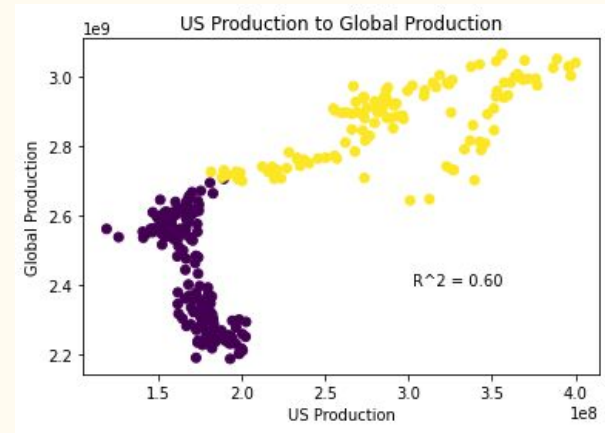
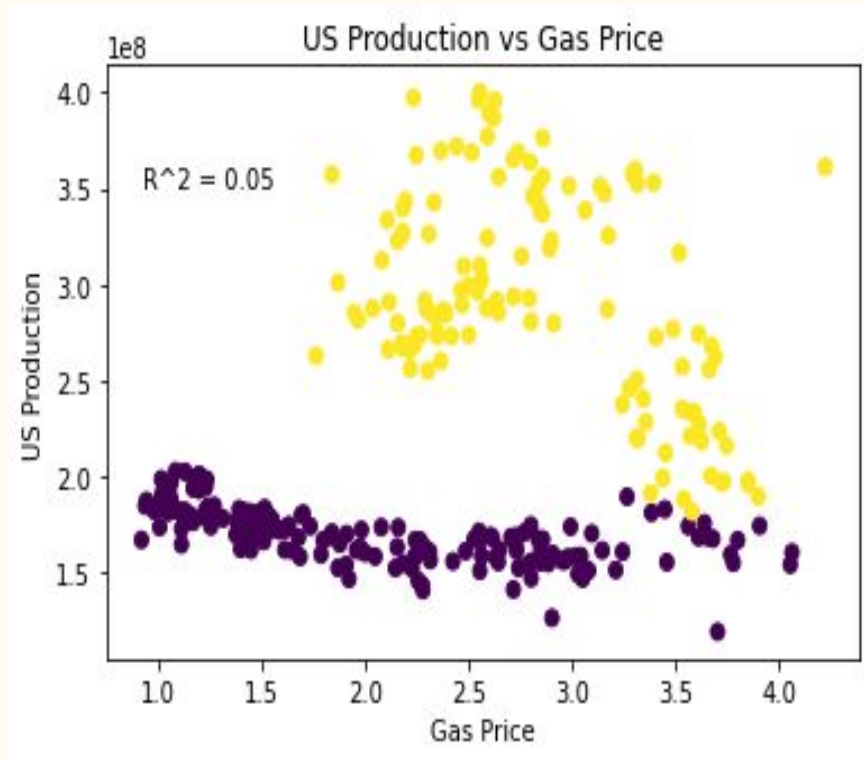
Index	Global Production ▼	US Production	Global Production Ind	US Production Ind
glob_prod_Mbbld	1	0.77072	True	True
glob_prod	1	0.77072	True	True
indust_prod	0.98425	0.761299	True	True
core_cpi	0.982567	0.640456	True	True
world_cpi	0.979158	0.781785	True	True
us_cpi	0.978998	0.757001	True	True
other_prod	0.975828	0.730539	True	True
nonus_prod	0.973788	0.605587	True	True
us_indust_prod	0.97304	0.702612	True	True
glob_con	0.972591	0.722943	True	True
glob_con_Mbbld	0.972591	0.722943	True	True
month_no	0.955457	0.793276	True	True
us_retail_sales	0.955146	0.801938	True	True
year	0.954066	0.792457	True	True
gas_margin	0.852076	0.763995	True	True
us_core_cpi	0.819912	0.588247	True	True
opec_prod_Mbbld	0.79091	0.293632	True	False
opec_prod	0.79091	0.293632	True	False
us_prod_Mbbld	0.770733	0.999999	True	True
us_prod	0.77072	1	True	True

Index	Global Production ▼	US Production	Global Production Ind	US Production Ind
retail_sales	0.723486	0.212515	True	False
gas_raw	0.65915	0.227268	True	False
refiner_sale	0.589584	0.135331	True	False
refiner_cost	0.544375	0.0944077	True	False
oil_price	0.523936	0.0551597	True	False
refiner_margin	0.500384	0.255492	True	False
us_con_Mbbld	0.124893	0.0172744	False	False
us_con	0.124893	0.0172744	False	False
republican	0.0281541	-0.00205397	False	False
glob_deficit_Mbbld	0.0192602	0.12507	False	False
glob_deficit	0.0192602	0.12507	False	False
margin_dif	0.00559738	-0.00520498	False	False
gas_adj	-0.00478892	-0.107354	False	False
gas_dif	-0.0180716	-0.0116632	False	False
oced_gdp	-0.0539358	-0.177773	False	False
new_rigs	-0.212452	-0.583333	False	True
unem_rate	-0.496772	-0.507057	False	True
running_def_Mbbl	-0.914923	-0.684189	True	True
running_def	-0.914923	-0.684189	True	True

# Oil Production Correlations: Global Production

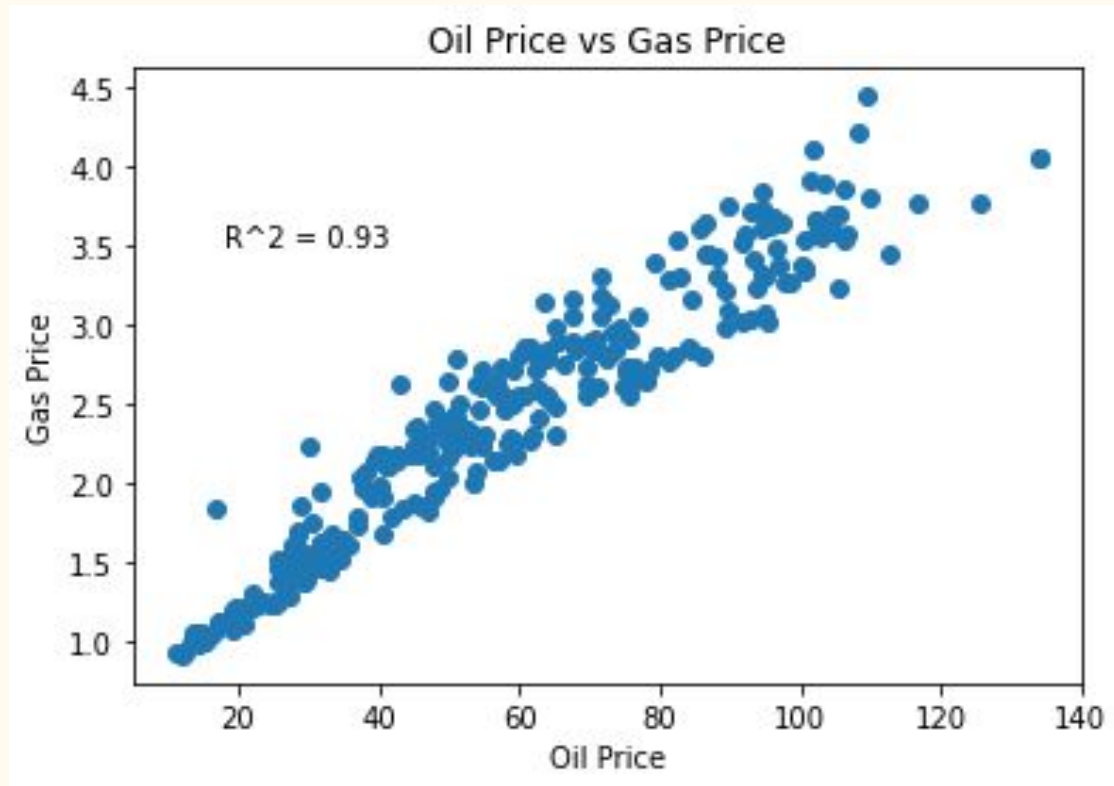


# Oil Production Correlations: US Production Purple is pre 2012





# What Drives Gas Price? Oil Price!

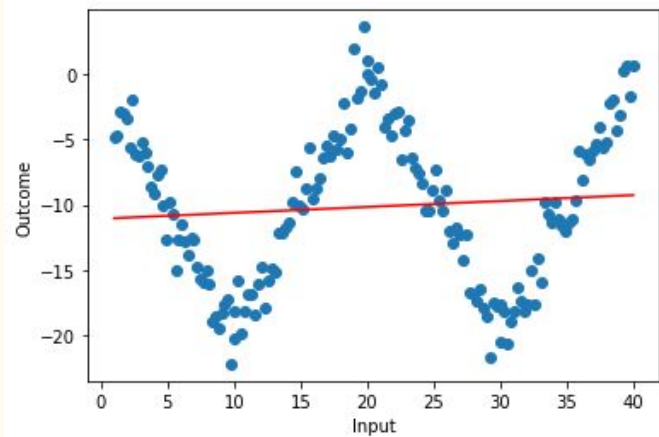




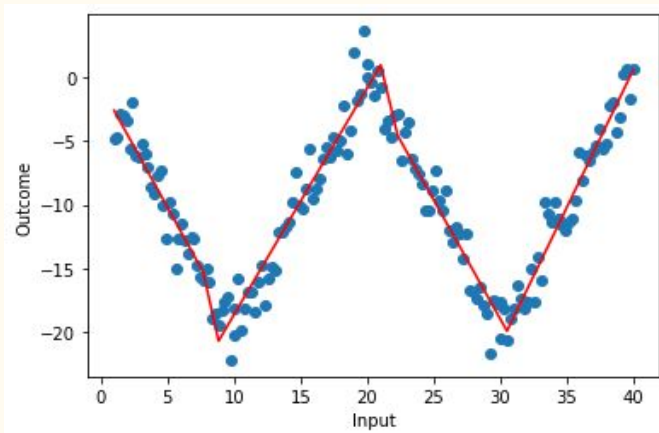
# MARS Introduction

- MARS = Multivariate Adaptive Regression Splines
  - Machine Learning method to rapidly identify both linear and non-linear trends
  - “Splines” are knots in otherwise linear data (vertices of the W)
  - MARS can find:
    - Linear Trends
    - Non-linear combinations (think variable \* variable)
    - Linear Splines

Standard Linear Regression



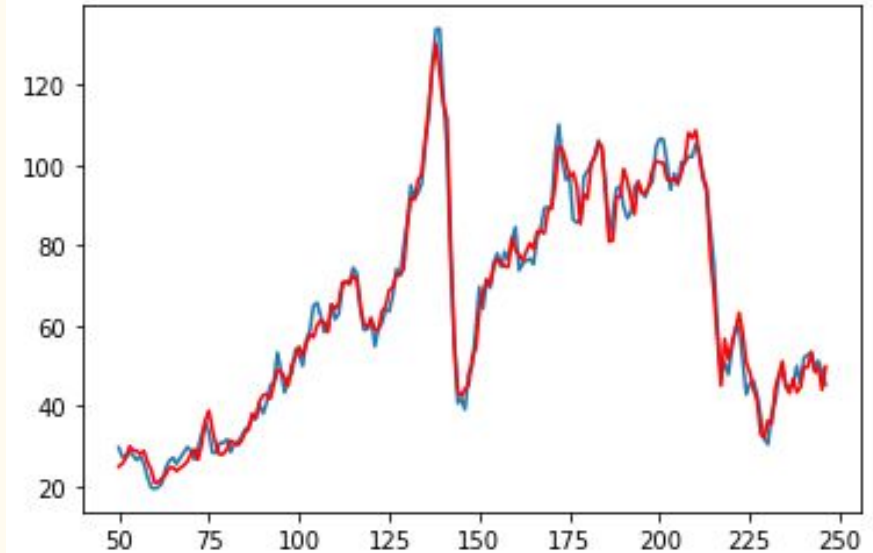
MARS Solution



# MARS Data Mining

- Training over whole dataset MARS fits nearly perfectly
- MARS alone struggled to deal with **out of sample** large drops (overfit to the data - next slide)
- Can MARS be used as a data mining application?
  - Run 200 80-20 Train/Discard splits and assess frequency of each feature incorporated into the model

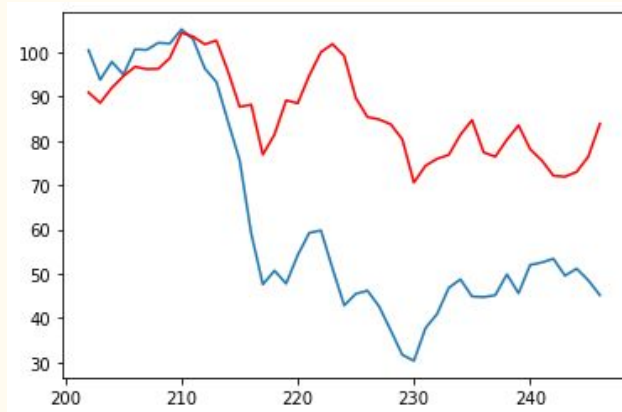
Full Model Fit with MARS (Blue is actual oil price red is MARS prediction)



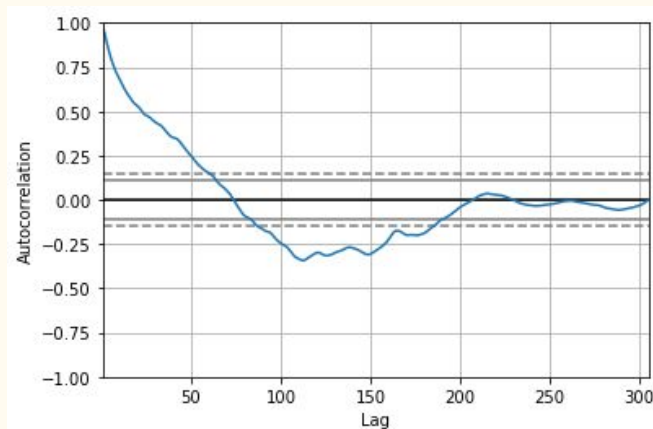
# Multiple Regression Optimization

- Used MARS feature frequency to narrow down feature options
  - 4,000 possible combinations instead of 4.1 Million
  - Added autoregressive term to 1000 with lowest Mean Squared Error (MSE)
    - Autoregressive term set arbitrarily at  $R = 0.75$  which correspond to regression over last 7 months
- Final Model chosen with lowest net MSE
  - Net MSE = MSE on oil price prediction + MSE when piped into regression to predict gas price (next slide)
  - Autoregression ultimately stabilized predictions rather than dictated them

MARS Out of Sample

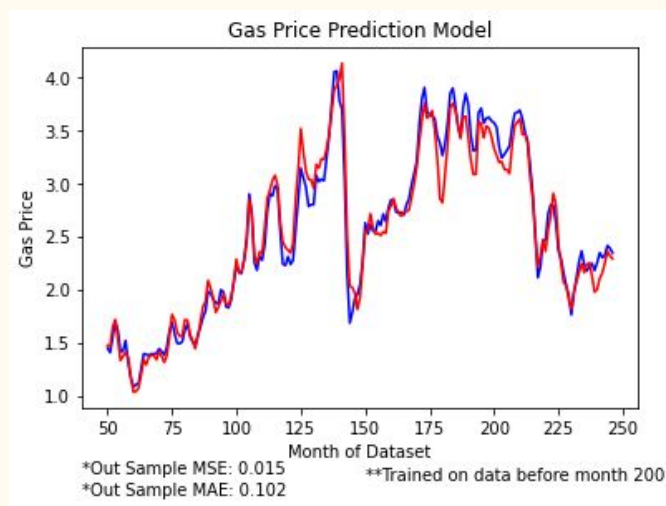
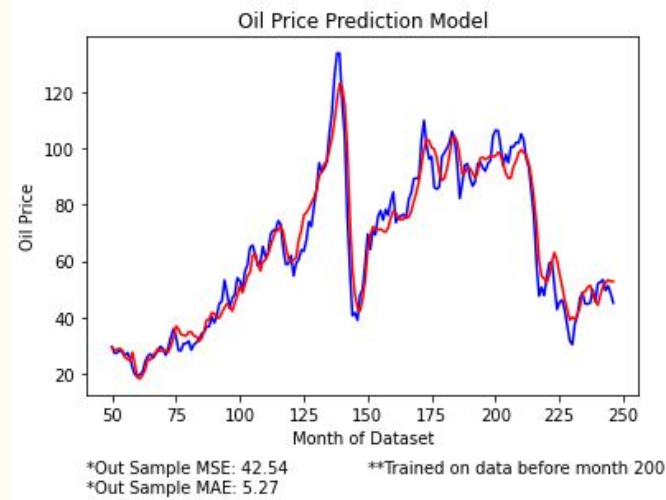


Oil Price Autocorrelation

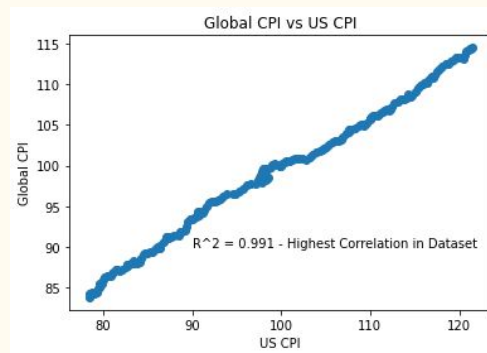
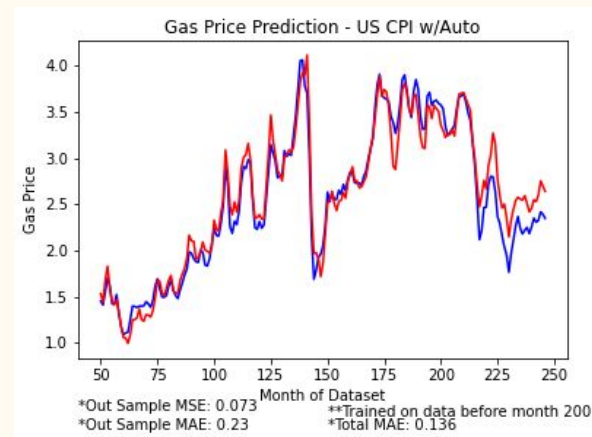
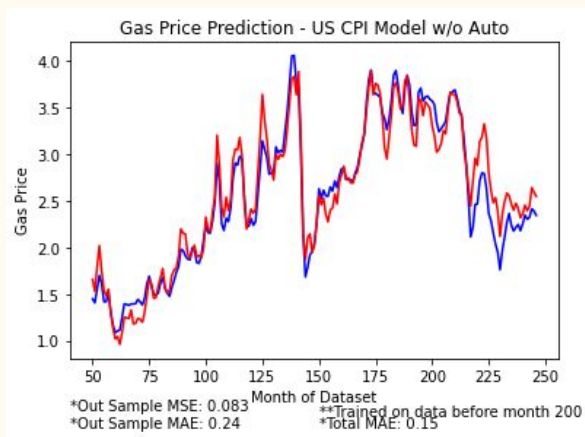
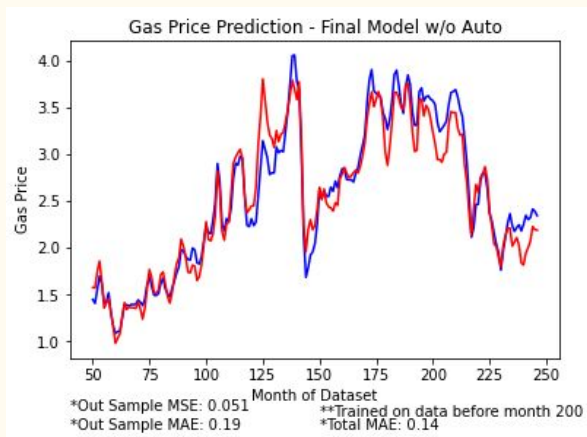


# Final Model

- Most Predictive Features:
  - Running Deficit (Oil production - Oil consumption from 1997 to 2022)
    - Crude estimation of supply/demand
  - US and Global Industrial Production
  - US Retail Sales
  - Global Overall CPI
  - Global Core CPI
- Final Model for Oil Price results piped into gas regression
  - Gas Price = Oil Price + Refinery Profit Margins
  - $R^2 = 0.933 \rightarrow 0.972$  (Just oil  $\rightarrow$  oil + refiner margins)
- Whole Model Average Absolute Errors:
  - Oil: \$4.63
  - Gas: \$0.11
- Total Absolute Errors:
  - Oil: \$912.27 (7.2%)
  - Gas: \$21.12 (4.2%)



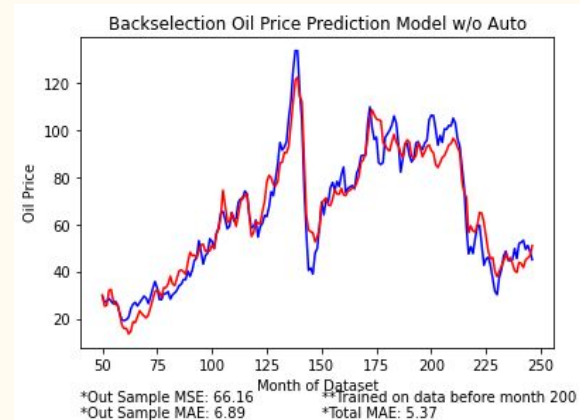
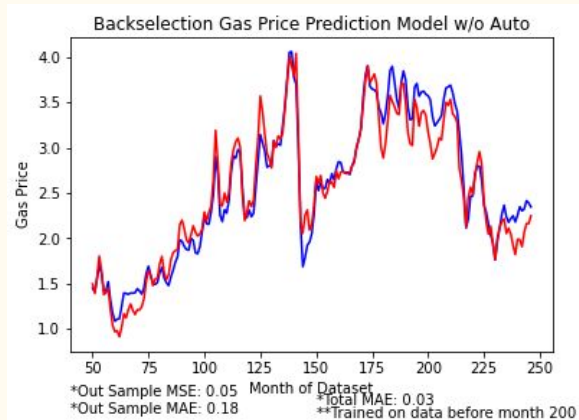
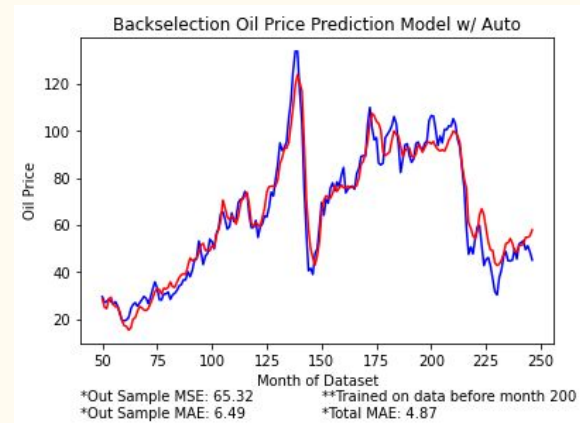
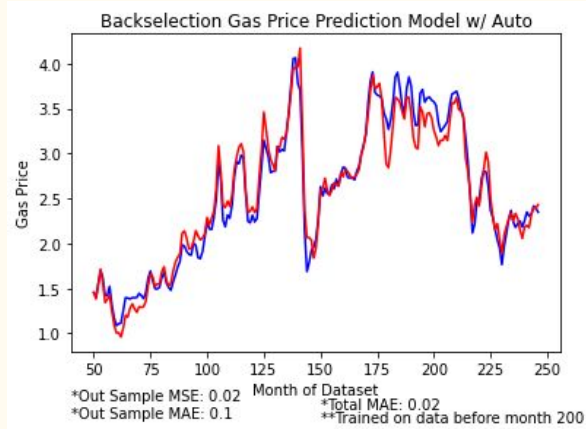
# Other Models - US CPI preforms mildly worse, but US CPI nearly perfectly correlated to global CPI



\*\* Auto = autoregressive term

\*\* Auto = autoregressive term

## Other Models - Standard backwards model selection performs worse than MARS data mining solution



# Correlation is not Causation

- The final model is the most accurate of all the candidates for both oil and gas predictions
- However many models perform similarly
- Does Oil price cause global inflation?
- Do **any** of the variables cause oil price changes?
  - If so, how many are correlated to each other?
- Is there any other factor tying them all together?
- Causation does require correlation