Import Libraries

Import libraries and set date parameters for Regression Model.

```
In [19]: import pandas_datareader as pdr
    import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import statsmodels.api as sm
    import matplotlib.pyplot as plt
    import datetime as dt
    import statsmodels.formula.api as smf

# set date parameters
    start = dt.datetime(2017, 1, 1)
    end = dt.datetime.now()

print("Date Parameters:"
    ,"\nStart Date:", start
```

Date Parameters:

Start Date: 2017-01-01 00:00:00 End Date 2022-12-01 23:21:23.076011

Pull Crypto Data

Using Pandas Data Reader, retrieve Bitcoin, Ethereum Historical Data

```
In [2]: cryptodata = pdr.get_data_fred(['CBBTCUSD','CBETHUSD'], start, end)

#Seperate crypto data into their own dataframe
X =pdr.get_data_fred(['CBBTCUSD']) #Use variable X to define our indepdent(explanatory) Variable, in this case
Y =pdr.get_data_fred(['CBETHUSD']) #Use variable Y to define our dependent(response) Variable, in this case,

#X = sm.add_constant(X) #add constant to our Bitcoin dataset

print(cryptodata) # prevew our BTC-ETH dataset
#print(X) # preview our bitcoin dataset
```

	CBBTCUSD	CBETHUSD
DATE		
2017-01-01	992.95	8.18
2017-01-02	1011.45	8.38
2017-01-03	1020.67	9.59
2017-01-04	1130.30	11.17
2017-01-05	1007.00	10.23
• • •		• • •
2022-11-27	16428.77	1193.72
2022-11-28	16193.89	1166.69
2022-11-29	16441.63	1216.23
2022-11-30	17179.78	1295.80
2022-12-01	16980.55	1276.41

[2161 rows x 2 columns]

Execute Regression Model (Ordinary Least Squares Method)

```
In [3]: # Y(depdent variable) ~ X(indepdent variable)
       reg1='CBETHUSD~CBBTCUSD'
       model=smf.ols(reg1,cryptodata).fit()
       print(model.summary())
       #Believe this code was cleaning data to forward/backward fill
       X= X.fillna(method = "pad")
      Y = Y.fillna(method = "pad")
      X = pd.DataFrame(X.replace([np.inf, -np.inf], np.nan))
      X = X.fillna(method='ffill')
       X = X.fillna(method='bfill')
                              OLS Regression Results
       ______
       Dep. Variable:
                               CBETHUSD
                                        R-squared:
                                                                    0.863
       Model:
                                   0LS
                                        Adj. R-squared:
                                                                    0.863
       Method:
                           Least Squares F-statistic:
                                                                 1.365e+04
                        Thu, 01 Dec 2022
       Date:
                                        Prob (F-statistic):
                                                                     0.00
       Time:
                               23:08:59
                                        Log-Likelihood:
                                                                  -16155.
       No. Observations:
                                   2160
                                        AIC:
                                                                 3.231e+04
       Df Residuals:
                                   2158
                                        BIC:
                                                                 3.232e+04
       Df Model:
                                     1
       Covariance Type:
                              nonrobust
       ______
                                                         [0.025
                     coef
                            std err
                                          t
                                                P>|t|
                                                                   0.975]
                 -135.8108
                             13.336
                                     -10.183
                                                0.000
                                                        -161.964
       Intercept
                                                                  -109.657
       CBBTCUSD
                   0.0639
                             0.001
                                     116.833
                                                0.000
                                                          0.063
                                                                    0.065
       ______
       Omnibus:
                                543.583
                                        Durbin-Watson:
                                                                    0.013
       Prob(Omnibus):
                                  0.000 Jarque-Bera (JB):
                                                                 1934.306
       C1....
                                        D. - L / JD .
                                                                     ^ ^^
```

```
In [4]: print("Parameters: ", model.params) #obtain parameters (EFFR Coefficient, Intercept)
print("R-Squared: ", round(model.rsquared *100,2),"%") #obtain rsquared - validate model: 99%
```

Parameters: Intercept -135.810792

CBBTCUSD 0.063897

dtype: float64

R-Squared: 86.35 %

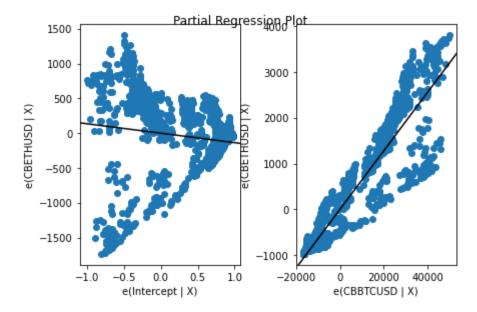
Residuals Analysis

- 1) Describe Residuals Dataset
- 2) Plot Residuals

Partial Regression Plot

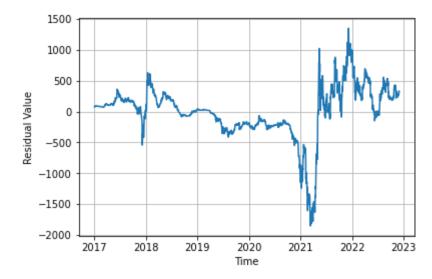
In [6]: fig = sm.graphics.plot_partregress_grid(model)

eval_env: 1
eval_env: 1



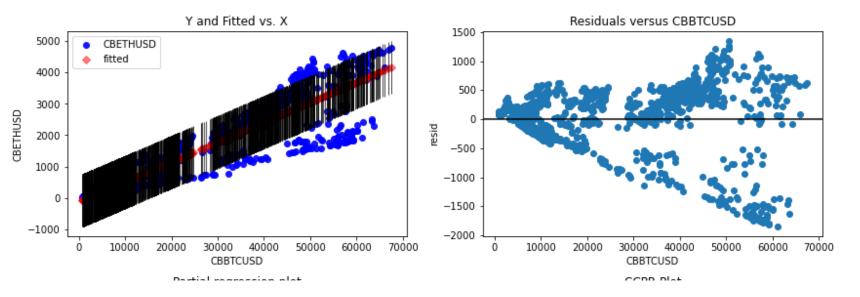
```
In [7]: plt.plot(uhat)  # Plot Residuals
    plt.xlabel('Time')
    plt.ylabel('Residual Value')
    plt.grid(True)
```

Out[7]: <function matplotlib.pyplot.show(close=None, block=None)>



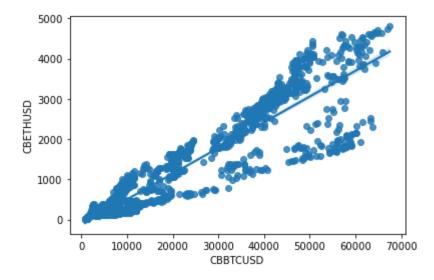
In [8]: fig = plt.figure(figsize=(12,8))
 eval_env: 1

Regression Plots for CBBTCUSD



In [9]: import seaborn as sns

Out[9]: <AxesSubplot:xlabel='CBBTCUSD', ylabel='CBETHUSD'>



```
In [10]: #ax = motif.plot(x='motifScore', y='expression', kind='scatter')
# plot regression line on the same axes, set x-axis limits
#ax.plot(x, p.const + p.motifScore * x)
```

Residuals Analysis Conclusion:

Time Frame: 2017 - 2021

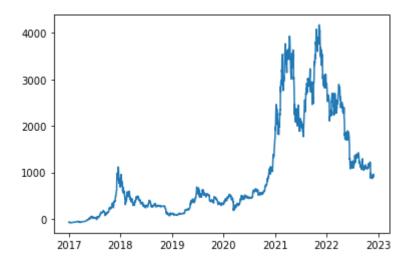
Our residuals during the time frame are consistently within our zero line and at times range from (- 500,500). Towards the end of this time frame, we begin to see our residuals stray further away from the zero line and a higher variablity begins during this time frame (end of 2020 - 2021).

Time Frame: 2022 - Today

Our Residuals range (-100, 600) is within a more normal range at this point.

Overall conclusion: Our high R-Squared results (86.35%) indicates that Ethereum moves relatively in line with the movement of Bitcoin.

Visualizing our Fitted Line against Actual ETH

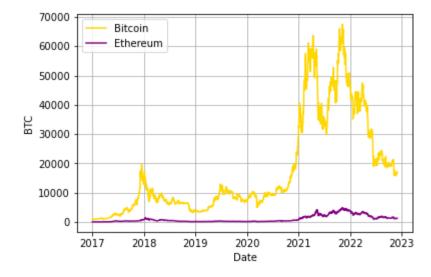


```
In [13]: plt.plot(cryptodata.CBETHUSD, label = "Actual ETH", color = 'purple')
          plt.plot(yhat, label = "Predicted ETH", color = 'dodgerblue')
          plt.xlabel('Date')
          plt.ylabel('ETH')
          plt.grid(True)
          legend=plt.legend()
Out[13]: <function matplotlib.pyplot.show(close=None, block=None)>
             5000
                      Actual ETH
                      Predicted ETH
             4000
             3000
             2000
             1000
                  2017
                         2018
                                2019
                                       2020
                                              2021
                                                     2022
                                                            2023
                                       Date
In [15]: sp = cryptodata.plot(x = 'CBBTCUSD', y = 'CBETHUSD', kind = 'scatter')
            Input In [15]
              print('Yhat = (x)',mode.params[CBBTCUSD'],model.params['Intercept']
          SyntaxError: invalid syntax
```

Bitcoin Actuals Vs Ethereum Actuals

```
In [16]: plt.plot(cryptodata.CBBTCUSD, label = "Bitcoin", color = 'gold')
    plt.plot(cryptodata.CBETHUSD, label = "Ethereum", color = 'purple')
    plt.xlabel('Date')
    plt.ylabel('BTC')
    plt.grid(True)
    legend=plt.legend()
```

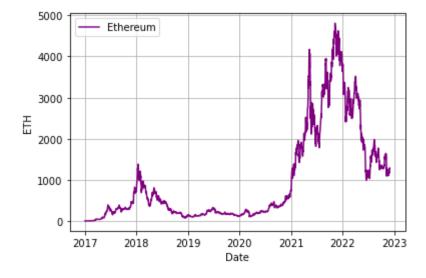
Out[16]: <function matplotlib.pyplot.show(close=None, block=None)>



Ethereum Actuals due to incompatability of scaling on previous plot

```
In [17]: plt.plot(cryptodata.CBETHUSD, label = "Ethereum", color = 'purple')
    plt.xlabel('Date')
    plt.ylabel('ETH')
    plt.grid(True)
    legend=plt.legend()
```

Out[17]: <function matplotlib.pyplot.show(close=None, block=None)>



Ehtereum Actuals to visually compare against Bitcoin Actuals. Demonstrating similr behavior to Bitcoin.