

m-assignment-1-jupyter-python-code

April 4, 2024

```
[ ]: # Load the dataset
import pandas as pd
import numpy as np

df = pd.read_csv('/content/current.csv')

# Clean the DataFrame by removing the row with transformation codes
df_cleaned = df.drop(index=0)
df_cleaned.reset_index(drop=True, inplace=True)
df_cleaned['sasdate'] = pd.to_datetime(df_cleaned['sasdate'], format='%m/%d/%Y')
df_cleaned
```

```
[ ]:
```

	sasdate	RPI	W875RX1	DPCERA3M086SBEA	CMRMTSPLx	\
0	1959-01-01	2583.560	2426.0	15.188	2.766768e+05	
1	1959-02-01	2593.596	2434.8	15.346	2.787140e+05	
2	1959-03-01	2610.396	2452.7	15.491	2.777753e+05	
3	1959-04-01	2627.446	2470.0	15.435	2.833627e+05	
4	1959-05-01	2642.720	2486.4	15.622	2.853072e+05	
..	
776	2023-09-01	19111.748	15741.9	116.594	1.507530e+06	
777	2023-10-01	19145.402	15784.6	116.663	1.505477e+06	
778	2023-11-01	19213.108	15859.9	117.127	1.514733e+06	
779	2023-12-01	19251.946	15899.0	117.773	1.530296e+06	
780	2024-01-01	19377.558	15948.8	117.639	NaN	

	RETAILx	INDPRO	IPFPNSS	IPFINAL	IPCONGD	...	\
0	18235.77392	21.9665	23.3891	22.2688	31.7011	...	
1	18369.56308	22.3966	23.7048	22.4617	31.9337	...	
2	18523.05762	22.7193	23.8483	22.5719	31.9337	...	
3	18534.46600	23.2032	24.1927	22.9026	32.4374	...	
4	18679.66354	23.5528	24.3936	23.1231	32.5925	...	
..	
776	705304.00000	103.2096	101.0935	101.3665	102.1034	...	
777	703528.00000	102.3722	100.5292	100.5527	101.1664	...	
778	703336.00000	102.6710	100.9362	101.2159	101.8557	...	
779	706180.00000	102.6715	100.8332	101.2843	101.9884	...	
780	700291.00000	102.5739	100.9984	101.7258	102.6235	...	

	DNDGRG3M086SBEA	DSERRG3M086SBEA	CES06000000008	CES20000000008	\
0	18.294	10.152	2.13	2.45	
1	18.302	10.167	2.14	2.46	
2	18.289	10.185	2.15	2.45	
3	18.300	10.221	2.16	2.47	
4	18.280	10.238	2.17	2.48	
..	
776	120.395	123.976	29.90	34.55	
777	120.040	124.228	29.97	34.67	
778	119.325	124.551	30.26	34.96	
779	119.193	124.917	30.45	35.01	
780	118.745	125.662	30.56	35.21	

	CES30000000008	UMCSENTx	DTCOLNVHFM	DTCTHFM	INVEST	VIXCLSx
0	2.04	NaN	6476.00	12298.00	84.2043	NaN
1	2.05	NaN	6476.00	12298.00	83.5280	NaN
2	2.07	NaN	6508.00	12349.00	81.6405	NaN
3	2.08	NaN	6620.00	12484.00	81.8099	NaN
4	2.08	95.3	6753.00	12646.00	80.7315	NaN
..
776	26.62	67.9	508808.61	913938.95	5074.6108	15.0424
777	26.65	63.8	513229.64	918210.64	5015.5456	19.0462
778	26.89	61.3	517434.30	922552.40	4999.7208	13.8563
779	27.14	69.7	522366.13	928336.14	5077.4222	12.6960
780	27.22	NaN	NaN	NaN	5105.3504	13.3453

[781 rows x 128 columns]

```
[ ]: # Extract transformation codes
transformation_codes = df.iloc[0, 1:].to_frame().reset_index()
transformation_codes.columns = ['Series', 'Transformation_Code']

# Function to apply transformations based on the transformation code
import numpy as np
def apply_transformation(series, code):
    if code == 1:
        # No transformation
        return series
    elif code == 2:
        # First difference
        return series.diff()
    elif code == 3:
        # Second difference
        return series.diff().diff()
    elif code == 4:
        # Log
```

```

        return np.log(series)
    elif code == 5:
        # First difference of log
        return np.log(series).diff()
    elif code == 6:
        # Second difference of log
        return np.log(series).diff().diff()
    elif code == 7:
        # Delta ( $x_t/x_{t-1} - 1$ )
        return series.pct_change()
    else:
        raise ValueError("Invalid transformation code")

# Applying the transformations to each column in df_cleaned based on
↳ transformation_codes
for series_name, code in transformation_codes.values:
    df_cleaned[series_name] = apply_transformation(df_cleaned[series_name].
↳ astype(float), float(code))

df_cleaned = df_cleaned[2:]
df_cleaned.reset_index(drop=True, inplace=True)
df_cleaned.head()

```

```

[ ]:
      sasdate      RPI      W875RX1  DPCERA3M086SBEA  CMRMTSPLx  RETAILx  \
0 1959-03-01  0.006457  0.007325      0.009404 -0.003374  0.008321
1 1959-04-01  0.006510  0.007029     -0.003622  0.019915  0.000616
2 1959-05-01  0.005796  0.006618      0.012043  0.006839  0.007803
3 1959-06-01  0.003068  0.003012      0.003642 -0.000097  0.009064
4 1959-07-01 -0.000580 -0.000762     -0.003386  0.012155 -0.000330

      INDPRO  IPFPNSS  IPFINAL  IPCONGD  ...  DNDGRG3M086SBEA  \
0  0.014306  0.006035  0.004894  0.000000  ...      -0.001148
1  0.021075  0.014338  0.014545  0.015650  ...       0.001312
2  0.014955  0.008270  0.009582  0.004770  ...     -0.001695
3  0.001141  0.007034  0.007128 -0.004767  ...       0.003334
4 -0.024240  0.001168  0.008249  0.013054  ...     -0.001204

      DSERRG3M086SBEA  CES06000000008  CES20000000008  CES30000000008  UMCSNTx  \
0      0.000292      -0.000022      -0.008147      0.004819      NaN
1      0.001760      -0.000022      0.012203     -0.004890      NaN
2     -0.001867     -0.000021     -0.004090     -0.004819      NaN
3      0.001946     -0.004619      0.003992      0.004796      NaN
4     -0.000013      0.000000     -0.004040     -0.004796      NaN

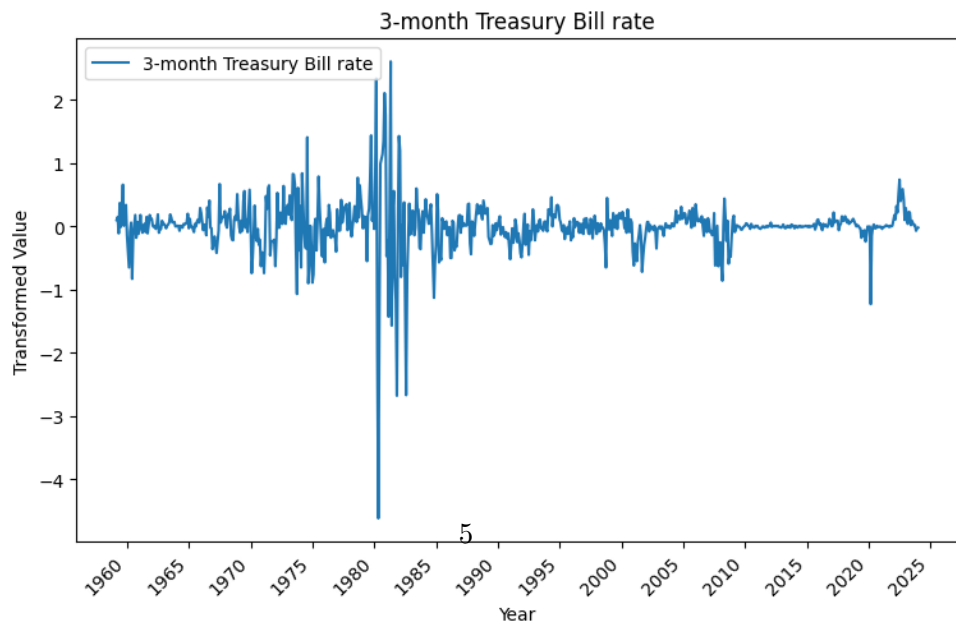
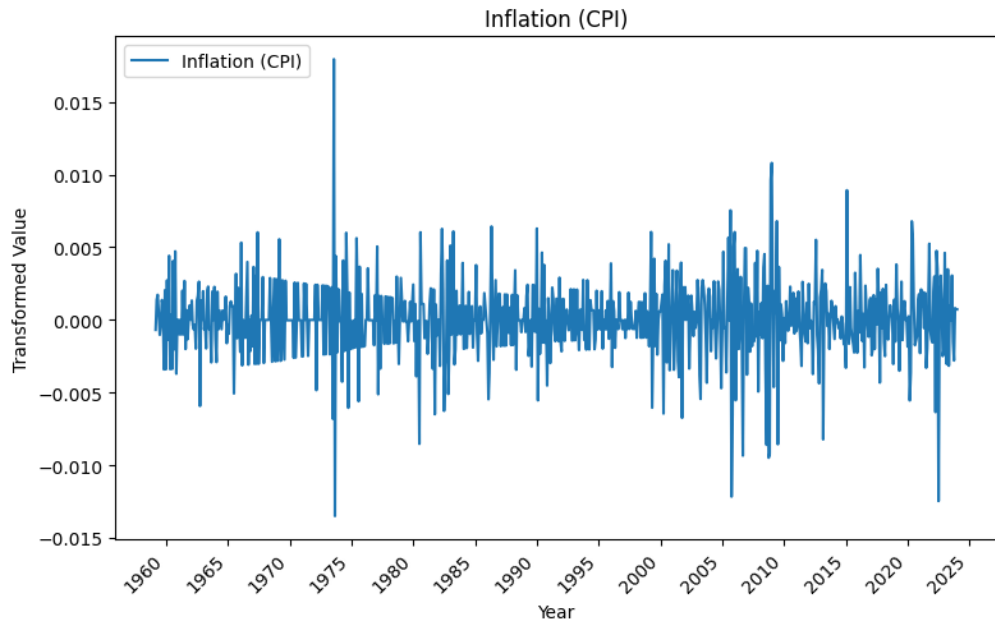
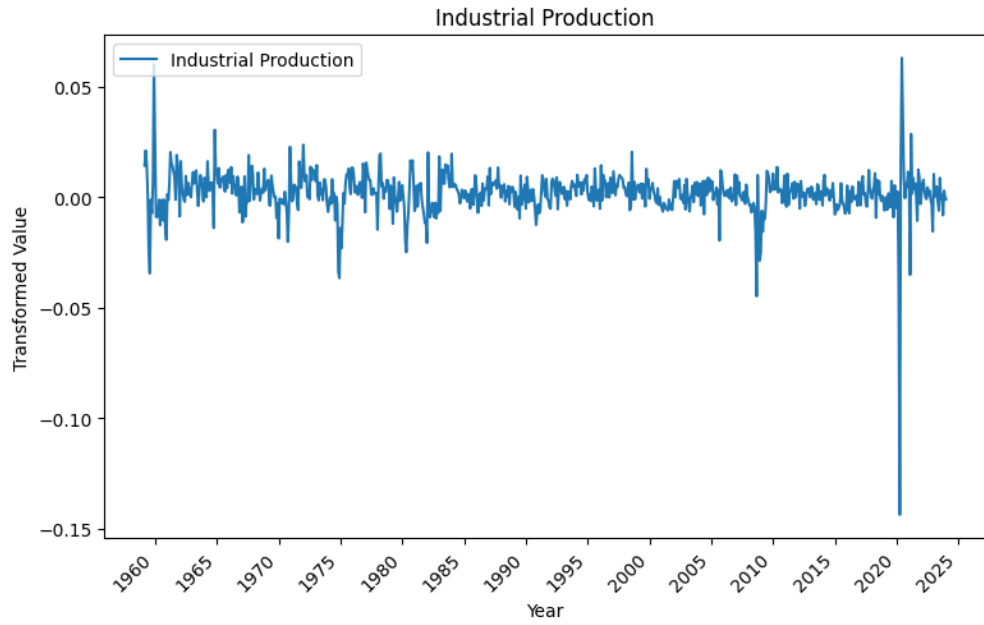
      DTCOLNVHFM  DTCTHFM  INVEST  VIXCLSx
0      0.004929  0.004138 -0.014792      NaN
1      0.012134  0.006734  0.024929      NaN

```

2	0.002828	0.002020	-0.015342	NaN
3	0.009726	0.009007	-0.012252	NaN
4	-0.004631	-0.001000	0.029341	NaN

[5 rows x 128 columns]

```
[ ]: # Plotting
import matplotlib.pyplot as plt
import matplotlib.dates as mdates
series_to_plot = ['INDPRO', 'CPIAUCSL', 'TB3MS']
series_names = ['Industrial Production', 'Inflation (CPI)', '3-month Treasury_
↳Bill rate']
fig, axs = plt.subplots(len(series_to_plot), 1, figsize=(8, 15))
for ax, series_name, plot_title in zip(axs, series_to_plot, series_names):
    if series_name in df_cleaned.columns:
        dates = pd.to_datetime(df_cleaned['sasdate'], format='%m/%d/%Y')
        ax.plot(dates, df_cleaned[series_name], label=plot_title)
        ax.xaxis.set_major_locator(mdates.YearLocator(base=5))
        ax.xaxis.set_major_formatter(mdates.DateFormatter('%Y'))
        ax.set_title(plot_title)
        ax.set_xlabel('Year')
        ax.set_ylabel('Transformed Value')
        ax.legend(loc='upper left')
        plt.setp(ax.xaxis.get_majorticklabels(), rotation=45, ha='right')
    else:
        ax.set_visible(False) # Hide plots for which the data is not available
plt.tight_layout()
plt.show()
```



```
[22]: # Prepare the data for ARX model
Y = df_cleaned['INDPRO'].dropna()
X = df_cleaned[['CPIAUCSL', 'FEDFUNDS']].dropna()

# Set the number of lags for Y and X
h = 1 # One-step ahead
p = 4
r = 4

# Prepare the target variable, lagged variables, and exogenous variables
Y_target = Y.shift(-h).dropna()
Y_lagged = pd.concat([Y.shift(i) for i in range(p+1)], axis=1).dropna()
X_lagged = pd.concat([X.shift(i) for i in range(r+1)], axis=1).dropna()

#Forecasting with ARX

Yraw = df_cleaned['INDPRO']
Xraw = df_cleaned[['CPIAUCSL', 'TB3MS']]
num_lags = 4 ## this is p
num_leads = 1 ## this is h
X = pd.DataFrame()

## Add the lagged values of Y
col = 'INDPRO'
for lag in range(0,num_lags+1):
    # Shift each column in the DataFrame and name it with a lag suffix
    X[f'{col}_lag{lag}'] = Yraw.shift(lag)
for col in Xraw.columns:
    for lag in range(0,num_lags+1):
        # Shift each column in the DataFrame and name it with a lag suffix
        X[f'{col}_lag{lag}'] = Xraw[col].shift(lag)

## Add a column on ones (for the intercept)
X.insert(0, 'Ones', np.ones(len(X)))
## X is now a DataFrame
X.head()
```

```
[22]:
```

	Ones	INDPRO_lag0	INDPRO_lag1	INDPRO_lag2	INDPRO_lag3	INDPRO_lag4	\
0	1.0	0.014306	NaN	NaN	NaN	NaN	
1	1.0	0.021075	0.014306	NaN	NaN	NaN	
2	1.0	0.014955	0.021075	0.014306	NaN	NaN	
3	1.0	0.001141	0.014955	0.021075	0.014306	NaN	
4	1.0	-0.024240	0.001141	0.014955	0.021075	0.014306	

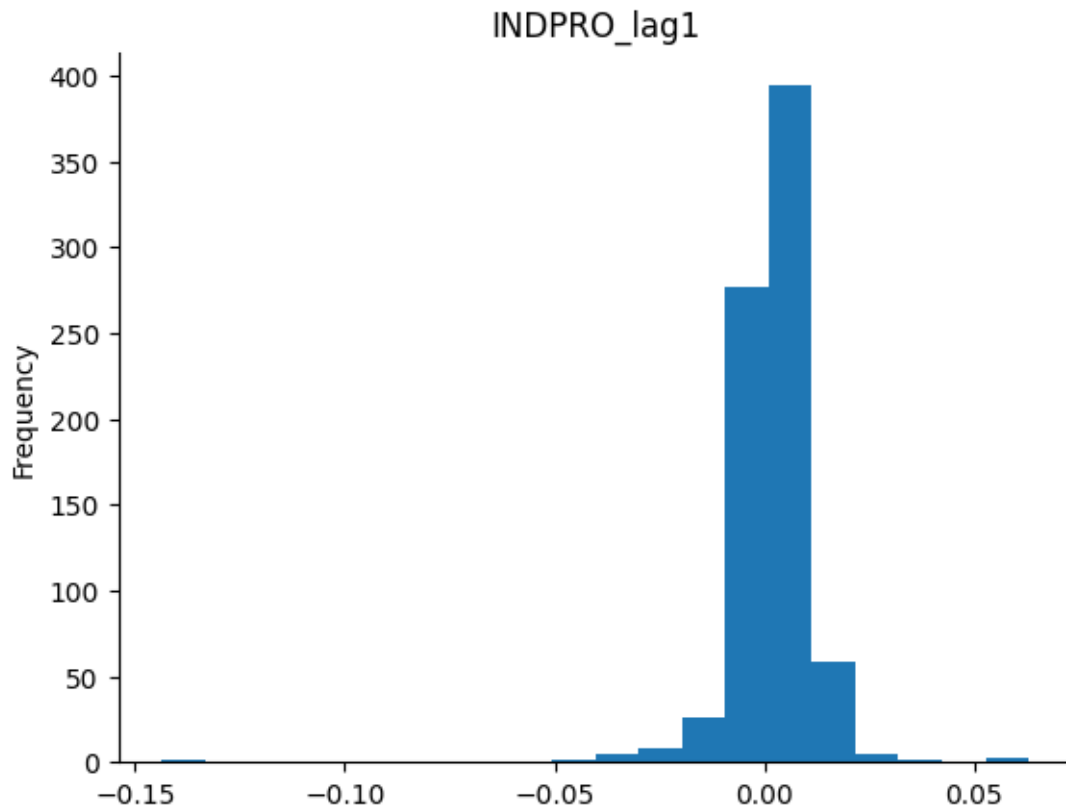
	CPIAUCSL_lag0	CPIAUCSL_lag1	CPIAUCSL_lag2	CPIAUCSL_lag3	CPIAUCSL_lag4	\
--	---------------	---------------	---------------	---------------	---------------	---

0	-0.000690	NaN	NaN	NaN	NaN
1	0.001380	-0.000690	NaN	NaN	NaN
2	0.001723	0.001380	-0.000690	NaN	NaN
3	0.000339	0.001723	0.001380	-0.00069	NaN
4	-0.001034	0.000339	0.001723	0.00138	-0.00069

	TB3MS_lag0	TB3MS_lag1	TB3MS_lag2	TB3MS_lag3	TB3MS_lag4
0	0.10	NaN	NaN	NaN	NaN
1	0.15	0.10	NaN	NaN	NaN
2	-0.11	0.15	0.10	NaN	NaN
3	0.37	-0.11	0.15	0.10	NaN
4	-0.01	0.37	-0.11	0.15	0.1

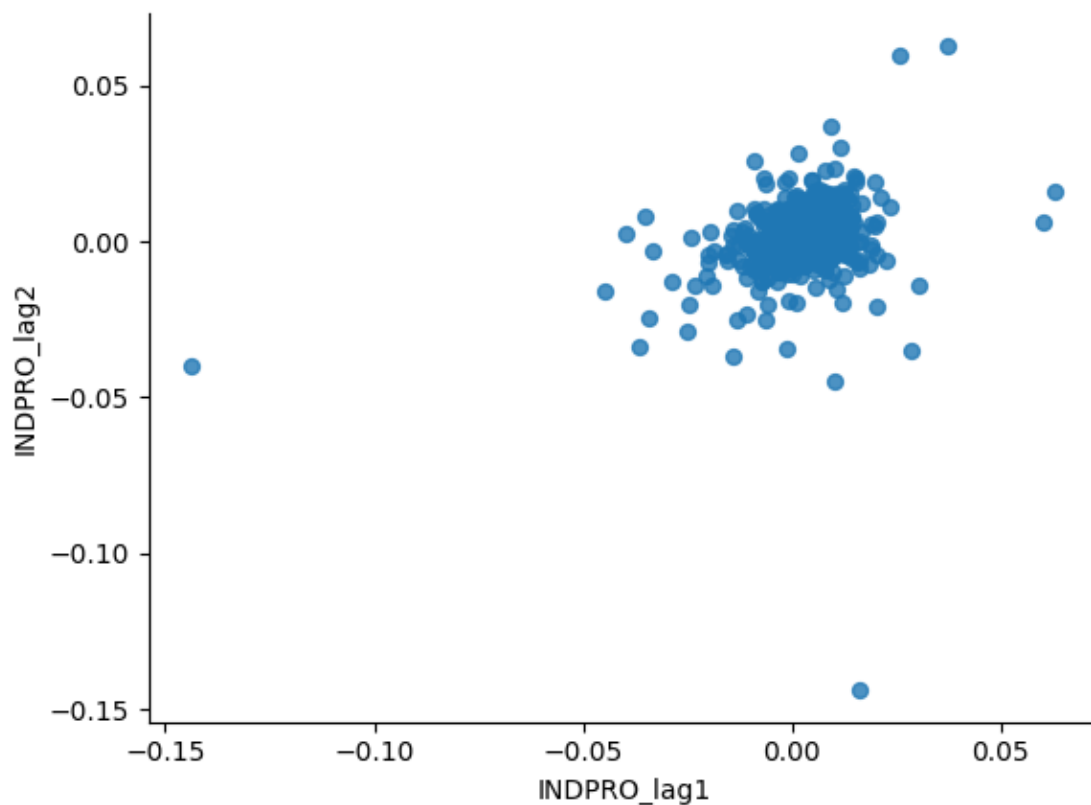
```
[ ]: # @title INDPRO_lag1
```

```
from matplotlib import pyplot as plt
X['INDPRO_lag1'].plot(kind='hist', bins=20, title='INDPRO_lag1')
plt.gca().spines[['top', 'right']].set_visible(False)
```



```
[ ]: # @title INDPRO_lag1 vs INDPRO_lag2
```

```
from matplotlib import pyplot as plt
X.plot(kind='scatter', x='INDPRO_lag1', y='INDPRO_lag2', s=32, alpha=.8)
plt.gca().spines[['top', 'right']].set_visible(False)
```



```
[ ]: # @title Ones vs INDPRO_lag1
```

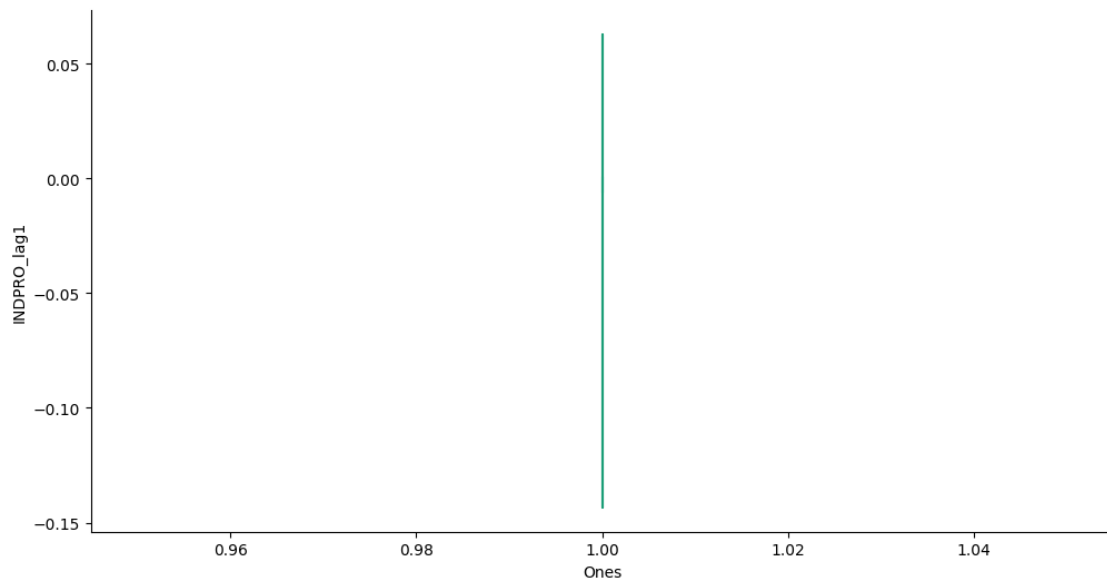
```
from matplotlib import pyplot as plt
import seaborn as sns
def _plot_series(series, series_name, series_index=0):
    palette = list(sns.palettes.mpl_palette('Dark2'))
    xs = series['Ones']
    ys = series['INDPRO_lag1']

    plt.plot(xs, ys, label=series_name, color=palette[series_index % len(palette)])

fig, ax = plt.subplots(figsize=(10, 5.2), layout='constrained')
df_sorted = X.sort_values('Ones', ascending=True)
_plot_series(df_sorted, '')
```

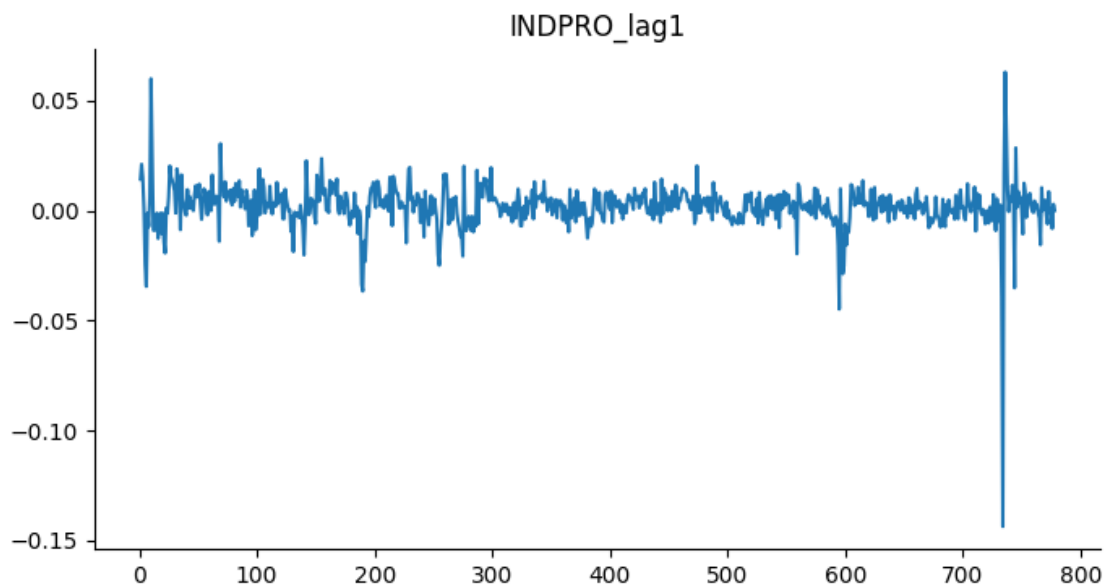


```
sns.despine(fig=fig, ax=ax)
plt.xlabel('Ones')
_ = plt.ylabel('INDPRO_lag1')
```



```
[ ]: # @title INDPRO_lag1

from matplotlib import pyplot as plt
X['INDPRO_lag1'].plot(kind='line', figsize=(8, 4), title='INDPRO_lag1')
plt.gca().spines[['top', 'right']].set_visible(False)
```



```
[ ]: y = Yraw.shift(-num_leads)
      y
```

```
[ ]: 0      0.021075
      1      0.014955
      2      0.001141
      3     -0.024240
      4     -0.034465
      ...
      774    -0.008147
      775     0.002915
      776     0.000005
      777    -0.000951
      778         NaN
      Name: INDPRO, Length: 779, dtype: float64
```

```
[ ]: ## Save last row of X (converted to numpy)
      X_T = X.iloc[-1:].values
      ## Subset getting only rows of X and y from p+1 to h-1
      ## and convert to numpy array
      y = y.iloc[num_lags:-num_leads].values
      X = X.iloc[num_lags:-num_leads].values

      ##
      X_T
```

```
[ ]: array([[ 1.00000000e+00, -9.51056709e-04,  4.86991246e-06,
              2.91450984e-03, -8.14668061e-03,  9.25729878e-04,
              7.21400503e-04,  7.26467817e-04,  8.11330254e-04,
              -2.79891559e-03, -1.51527417e-03, -2.00000000e-02,
              -3.00000000e-02, -7.00000000e-02,  2.00000000e-02,
              2.00000000e-02]])
```

```
[ ]: ##Estimation
      from numpy.linalg import solve
      # Solving for the OLS estimator beta: (X'X)^{-1} X'Y
      beta_ols = solve(X.T @ X, X.T @ y)

      ## Produce the One step ahead forecast
      ## % change month-to-month INDPRO
      forecast = X_T@beta_ols*100
      forecast
```

```
[ ]: array([0.08445815])
```

```
[ ]: #Forecasting Exercise
def calculate_forecast(df_cleaned,
                      p=4,
                      H=[1, 4, 8],
                      end_date='12/1/1999',
                      target='INDPRO',
                      xvars=['CPIAUCSL', 'TB3MS']):

    # Subset df_cleaned to use only data up to end_date
    rt_df = df_cleaned[df_cleaned['sasdate'] <= pd.Timestamp(end_date)]

    # Get the actual values of the target at different steps ahead
    Y_actual = []
    for h in H:
        os = pd.Timestamp(end_date) + pd.DateOffset(months=h)
        Y_actual.append(df_cleaned[df_cleaned['sasdate'] == os][target] * 100)
        ## Now Y contains the true values at T+H (multiplying * 100)

    Yraw = rt_df[target]
    Xraw = rt_df[xvars]
    X = pd.DataFrame()

    # Add the lagged values of Y
    for lag in range(0, p):
        # Shift each column in the DataFrame and name it with a lag suffix
        X[f'{target}_lag{lag}'] = Yraw.shift(lag)
    for col in Xraw.columns:
        for lag in range(0, p):
            X[f'{col}_lag{lag}'] = Xraw[col].shift(lag)

    # Add a column of ones (for the intercept)
    X.insert(0, 'Ones', np.ones(len(X)))

    # Save the last row of X (converted to a `numpy` array)
    X_T = X.iloc[-1:].values

    Yhat = []
    for h in H:
        y_h = Yraw.shift(-h)
        # Subset getting only rows of X and y from p+1 to h-1
        y = y_h.iloc[p:-h].values
        X_ = X.iloc[p:-h].values
        # Solving for the OLS estimator beta:  $(X'X)^{-1} X'Y$ 
        beta_ols = solve(X_.T @ X_, X_.T @ y)
        # Produce the One step ahead forecast
        # % change month-to-month INDPRO
        Yhat.append(X_T @ beta_ols * 100)
```

```

    # Calculate the forecasting error and return
    return np.array(Y_actual) - np.array(Yhat)

t0 = pd.Timestamp('12/1/1999')
e = []
T = []
for j in range(0, 10):
    t0 = t0 + pd.DateOffset(months=1)
    print(f'Using data up to {t0}')
    ehat = calculate_forecast(df_cleaned, p=4, H=[1, 4, 8], end_date=t0)
    e.append(ehat.flatten())
    T.append(t0)

# Create a pandas DataFrame from the list
edf = pd.DataFrame(e)

# Calculate the RMSFE, that is, the square root of the MSFE
np.sqrt(edf.apply(np.square).mean())

```

```

Using data up to 2000-01-01 00:00:00
Using data up to 2000-02-01 00:00:00
Using data up to 2000-03-01 00:00:00
Using data up to 2000-04-01 00:00:00
Using data up to 2000-05-01 00:00:00
Using data up to 2000-06-01 00:00:00
Using data up to 2000-07-01 00:00:00
Using data up to 2000-08-01 00:00:00
Using data up to 2000-09-01 00:00:00
Using data up to 2000-10-01 00:00:00

```

```

[ ]: 0    0.337110
     1    0.512690
     2    0.624035
     dtype: float64

```