

```
In [ ]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import os
from tqdm import tqdm
from sklearn.metrics import mean_squared_error, r2_score
```

```
In [ ]: # hide warning messages
import warnings
warnings.filterwarnings("ignore")
```

```
In [ ]: # better plots
sns.set(rc={'figure.figsize':(12,8)});
```

```
In [ ]: directory = os.path.dirname(os.getcwd())
directory
```

```
Out[ ]: 'd:\\github\\AssignmentEconometricsIV'
```

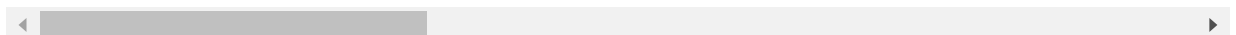
```
In [ ]: # read the data
input_path = f'{directory}\\data\\stacionarized_cpi.csv'
df = pd.read_csv(input_path)
df['date'] = pd.to_datetime(df['date'])
df = df.set_index('date')
```

```
In [ ]: df.head()
```

```
Out[ ]:
```

	RPI	W875RX1	DPCERA3M086SBEA	RETAILx	INDPRO	IPFPNSS	IPFINAL	IPCONGD
date								
1959-03-01	0.643011	0.735934	0.941009	0.832120	1.430253	0.603609	0.489927	0.000000
1959-04-01	0.649412	0.704864	-0.363947	0.061571	2.107741	1.433803	1.454234	1.565338
1959-05-01	0.576311	0.661646	1.200535	0.780340	1.495024	0.826920	0.958304	0.476849
1959-06-01	0.310244	0.297379	0.370829	0.906434	0.114438	0.703445	0.712642	-0.476849
1959-07-01	-0.058921	-0.076384	-0.342687	-0.033018	-2.423797	0.116693	0.824692	1.305596

5 rows × 104 columns



```
In [ ]: # set the amount of lags
lags = 2
```

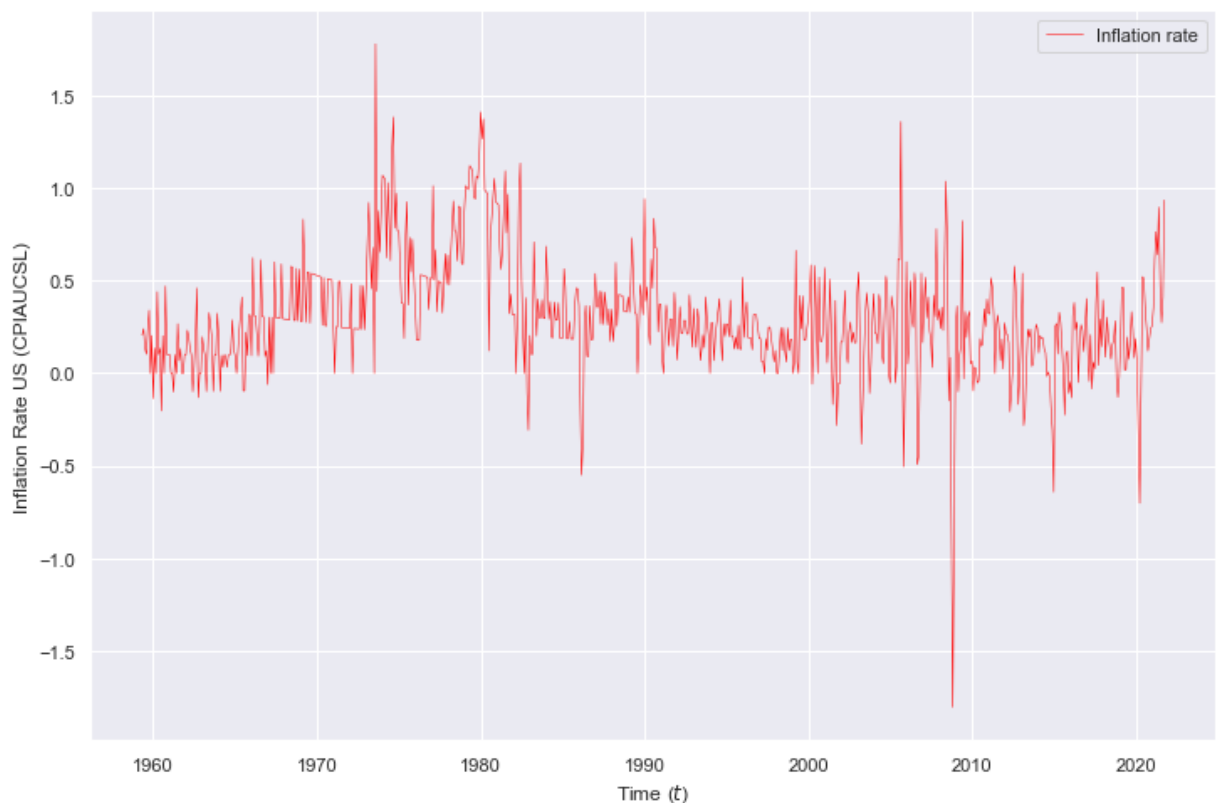
```
In [ ]: # create lagged variables
        for col in df.columns:
            for i in range(lags):
                name_col = col + f"(-{i+1})"
                df[name_col] = df[col].shift(i+1)
```

```
In [ ]: # inflation rate ahead
        df['CPIAUCSL(+1)'] = df['CPIAUCSL'].shift(-1)
```

```
In [ ]: # drop nan rows
        df = df.dropna()
```

```
In [ ]: # plot Inflation rate of US
        plt.plot(df['CPIAUCSL'], color='red', label='Inflation rate', linewidth = 0.5)

        plt.xlabel(f'Time ($t$)')
        plt.ylabel('Inflation Rate US (CPIAUCSL)')
        plt.legend()
        plt.show()
```



```
In [ ]: import glmnet_python.glmnet_python
        from cvglmnet import cvglmnet
        from cvglmnetPredict import cvglmnetPredict
        from cvglmnetPlot import cvglmnetPlot
        from cvglmnetCoef import cvglmnetCoef

        from functions.linear_models import Ridge, LASSO
```

```
In [ ]: # set variables
        y = df["CPIAUCSL(+1)"]
```

```
X = df.drop("CPIAUCSL(+1)", axis=1)
```

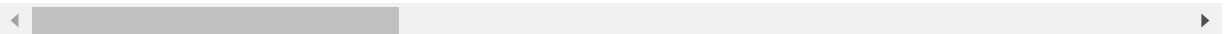
In []:

```
df
```

Out[]:

	RPI	W875RX1	DPCERA3M086SBEA	RETAILx	INDPRO	IPFPNSS	IPFINAL	IPCONG
date								
1959-05-01	0.576311	0.661646	1.200535	0.780340	1.495024	0.826920	0.958304	0.4768
1959-06-01	0.310244	0.297379	0.370829	0.906434	0.114438	0.703445	0.712642	-0.4768
1959-07-01	-0.058921	-0.076384	-0.342687	-0.033018	-2.423797	0.116693	0.824692	1.3055
1959-08-01	-0.563656	-0.574751	0.600333	0.636421	-3.446532	-0.702622	-0.234750	0.1178
1959-09-01	0.072136	0.000000	1.001845	-1.315700	-0.120914	-0.470898	-0.353804	-0.3537
...
2021-06-01	-0.268168	0.246839	0.588917	0.848804	0.546727	0.002528	0.114008	-0.2029
2021-07-01	0.803329	0.376098	-0.313442	-1.637075	0.767226	1.461057	1.668025	0.9707
2021-08-01	-0.041013	-0.094940	0.727968	1.152697	-0.136496	-0.232743	-0.485054	-0.4072
2021-09-01	-1.333562	0.169571	0.264718	0.738929	-1.018960	-0.235388	-0.539534	-0.5550
2021-10-01	-0.226827	0.002789	0.740685	1.769110	1.664607	1.015773	1.119566	1.0356

750 rows × 313 columns



In []:

```
# add forecast columns
df["CPIAUCSL_estimated_Ridge"] = np.nan
df["CPIAUCSL_estimated_LASSO"] = np.nan
```

In []:

```
# set some parameters
rolling_window = 490 - lags
T = len(y) - rolling_window
```

In []:

```
# List of forecast squared errors
errors_Ridge = []
errors_LASSO = []

for t in tqdm(range(T), desc='Processing for time'):
    # predict date
    date = y[[rolling_window+t]].index

    # estimation sets
    X_train = X[t:(rolling_window+t)]
```

```

y_train = y[t:(rolling_window+t)]

# forecast sets
X_test = X.iloc[[rolling_window+t]]
y_test = y[rolling_window+t]

# estimations
y_pred_Ridge, error_pred_Ridge = Ridge(X_train, y_train, X_test, y_test)
y_pred_LASSO, error_pred_LASSO = LASSO(X_train, y_train, X_test, y_test)

# fill forecast columns
df["CPIAUCSL_estimated_Ridge"][date] = y_pred_Ridge
df["CPIAUCSL_estimated_LASSO"][date] = y_pred_LASSO

# append forecast squared errors
errors_Ridge.append(error_pred_Ridge)
errors_LASSO.append(error_pred_LASSO)

```

Processing for time: 0%| | 0/262 [00:00<?, ?it/s]Processing for time: 100%|██████████| 262/262 [58:23<00:00, 13.37s/it]

In []: df[["CPIAUCSL", "CPIAUCSL_estimated_Ridge", "CPIAUCSL_estimated_LASSO"]][rolling_win

Out []: CPIAUCSL CPIAUCSL_estimated_Ridge CPIAUCSL_estimated_LASSO

date			
2000-01-01	0.295334	0.095432	0.161590
2000-02-01	0.411765	0.347451	0.410884
2000-03-01	0.584795	0.393511	0.467927
2000-04-01	-0.058514	0.016565	0.052578
2000-05-01	0.175234	0.221143	0.177611
...
2021-06-01	0.896742	0.271468	0.484639
2021-07-01	0.471599	0.089117	0.225026
2021-08-01	0.273614	0.157016	0.208878
2021-09-01	0.410742	0.237001	0.323774
2021-10-01	0.934505	0.375331	0.512426

262 rows × 3 columns

In []: # prediction dates
prediction_dates = list(X.iloc[(rolling_window):].index)

In []: # treat the error list
errors_Ridge_ = [item[0] for item in errors_Ridge]
errors_LASSO_ = [item[0] for item in errors_LASSO]

compute cumulated mse
cum_errors_Ridge = list(np.cumsum(errors_Ridge_))
cum_errors_LASSO = list(np.cumsum(errors_LASSO_))

```
In [ ]: errors = {'Cumulated_MSE_Ridge': cum_errors_Ridge, 'Cumulated_MSE_LASSO': cum_errors_LASSO}
mse = pd.DataFrame(errors, index=prediction_dates)
```

```
In [ ]: mse
```

```
Out[ ]:
```

	Cumulated_MSE_Ridge	Cumulated_MSE_LASSO
2000-01-01	0.100067	0.062587
2000-02-01	0.156399	0.092832
2000-03-01	0.360726	0.369972
2000-04-01	0.385901	0.385016
2000-05-01	0.515197	0.547513
...
2021-06-01	22.719491	17.702431
2021-07-01	22.753531	17.704792
2021-08-01	22.817907	17.745541
2021-09-01	23.304419	18.118534
2021-10-01	23.462633	18.186481

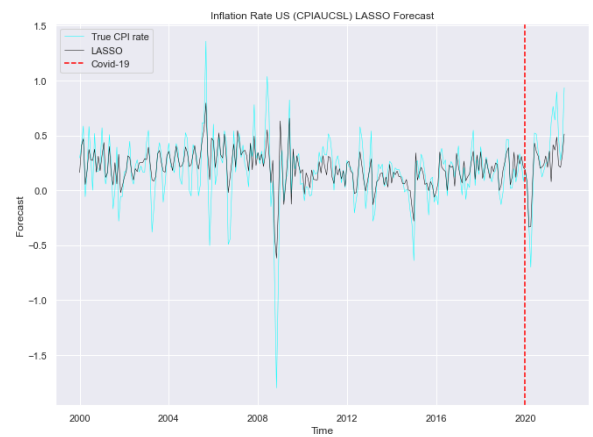
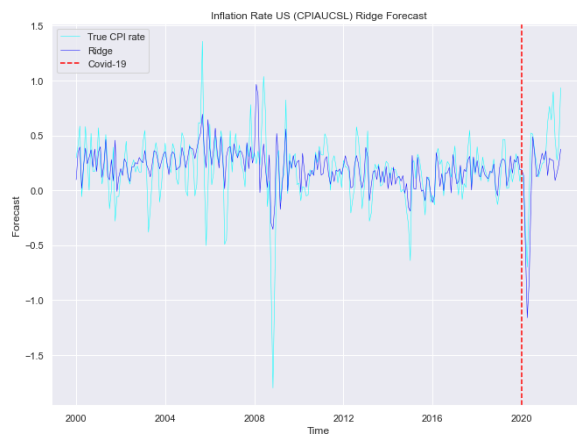
262 rows × 2 columns

```
In [ ]: fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(24, 8))

# True CPI
axes[0].plot(df['CPIAUCSL'][rolling_window:], color='cyan', label='True CPI rate', linewidth=2)
# Ridge forecasts
axes[0].plot(df["CPIAUCSL_estimated_Ridge"], color='blue', label='Ridge', linewidth=2)
# adding a vertical line at 2020, January
axes[0].axvline(x=mse.index[240], color='red', linestyle='--', label='Covid-19')
# Labels
axes[0].set_xlabel('Time')
axes[0].set_ylabel('Forecast')
axes[0].set_title('Inflation Rate US (CPIAUCSL) Ridge Forecast')
axes[0].legend()

# True CPI
axes[1].plot(df['CPIAUCSL'][rolling_window:], color='cyan', label='True CPI rate', linewidth=2)
# LASSO forecasts
axes[1].plot(df["CPIAUCSL_estimated_LASSO"], color='black', label='LASSO', linewidth=2)
# adding a vertical line at 2020, January
axes[1].axvline(x=mse.index[240], color='red', linestyle='--', label='Covid-19')
# Labels
axes[1].set_xlabel('Time')
axes[1].set_ylabel('Forecast')
axes[1].set_title('Inflation Rate US (CPIAUCSL) LASSO Forecast')
axes[1].legend()
```

```
Out[ ]: <matplotlib.legend.Legend at 0x1fd0b4e8910>
```



In []:

```
plt.plot(df['CPIAUCSL'][rolling_window:], color='cyan', label='True CPI rate', linewidth=0.5)
plt.plot(df["CPIAUCSL_estimated_Ridge"], color='blue', label='Ridge', linewidth = 0.5)
plt.plot(df["CPIAUCSL_estimated_LASSO"], color='black', label='LASSO', linewidth = 0.5)

# adding a vertical line at 2020, January
plt.axvline(x=mse.index[240], color='red', linestyle='--', label='Covid-19')

plt.xlabel('Time')
plt.ylabel('Inflation Rate US (CPIAUCSL) Forecasts')
plt.legend()
plt.show()
```



In []:

```
# plot of cumulated MSE
plt.plot(mse["Cumulated_MSE_Ridge"], color='blue', label="Ridge", linewidth = 0.5)

# plot of cumulated MSE
plt.plot(mse["Cumulated_MSE_LASSO"], color='black', label="LASSO", linewidth = 0.5)

# adding a vertical line at 2020, January
plt.axvline(x=mse.index[240], color='red', linestyle='--', label='Covid-19')
```

```
plt.xlabel('Time')
plt.ylabel('Cumulated MSE')
plt.title('Inflation Rate US (CPIAUCSL) Cumulated MSE')
plt.legend()
plt.show()
```

