Procedure for Evaluating the Simple Papell Model

Objective: Adjust the hyperparameters of the proposed simple Papell model in order to maximize the model's accuracy.

- · Evaluation Metric: AIC
- ullet Evaluation Procedure: Cross-validation on a moving time window of size h.

For a set of parameters (h, λ) :

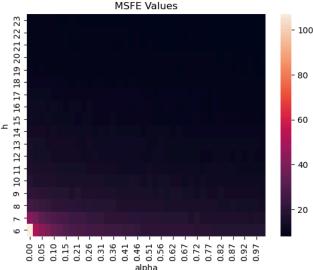
- Information from t-h to t. Calculated with a simple <code>Nowcasting</code> procedure.
 - Nowcasting $\mathbb{E}_{t-1}[\pi_t]$.
 - Nowcasting $\mathbb{E}_{t-1}[\tilde{\pi}_t]$.

return src.rolling_msfe_model_(
 data = data_nowcasted,
 alpha = alpha,
 h = h,
 bootstrap=True.

agg=lambda x: np.mean(x**2)

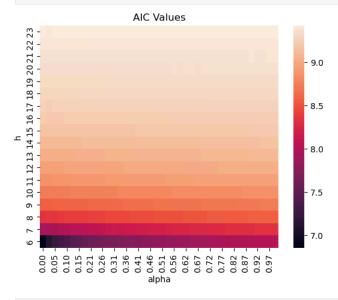
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• Nowcasting \mathbb{E}_{t-1}[y_t].
               • Nowcasting \mathbb{E}_{t-2}[	ilde{y}_t].
In [ ]: import matplotlib.pyplot as plt
         import pandas as pd
         import toolz as tz
         import numpy as np
         import src
         import itertools as itt
         import matplotlib.pyplot as plt
         import seaborn as sns
         import multiprocessing
In [ ]: data = pd.read_csv('data/processed_data.csv', index_col=0, parse_dates=True)
In [ ]: # Defining the variables to be used in the nowcasting procedure and the lags
         # to update.
         variables = {
              'pi': {
                  'lag': 1,
                  'nowcast': src.pi_nowcast_
              'pi_star': {
    'lag': 1,
                  'nowcast': src.pi_nowcast_
                   'lag': 1,
                  'nowcast': src.gap_nowcast_
             'lag': 2,
                  'nowcast': src.gap_nowcast_
In []: # Defining the rolling AIC function to be used in the grid search procedure.
         def rolling_aic(X: pd.DataFrame, y: pd.Series, h: int, alpha: float) -> float:
              data_nowcasted = src.nowcasting_(X, y, lags=variables)
              data_papell = tz.pipe(
                  data nowcasted.
                  lambda x: x.assign(diff_inf = x.iloc[:, 2] - x.iloc[:, 1]),
lambda x: x.assign(diff_y = x.iloc[:, 4] - x.iloc[:, 3]),
                  lambda x: x.iloc[:, [0, 5, 6]]
              return src.rolling_aic_model_(
                  data = data_nowcasted,
                  alpha = alpha,
                  h = h,
                  agg=np.mean,
                  bootstrap=True
In [ ]: #Defining the rolling MSFE function to be used in the grid search procedure.
         #Bootstrap is set to True.
         def rolling_msfe(X: pd.DataFrame, y: pd.Series, h: int, alpha: float) -> float:
              data_nowcasted = src.nowcasting_(X, y, lags=variables)
              data_papell = tz.pipe(
                  data_nowcasted,
                  lambda x: x.assign(diff_inf = x.iloc[:, 2] - x.iloc[:, 1]),
lambda x: x.assign(diff_y = x.iloc[:, 4] - x.iloc[:, 3]),
lambda x: x.iloc[:, [0, 5, 6]]
```

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In [ ]: # Grid search for the hyperparameters h and alpha.
        grid_ = {
   'h':list(range(6, 24)),
             'alpha': np.linspace(0, 1, 40)
        grid_ = list(itt.product(grid_['h'], grid_['alpha']))
In [ ]: # Parallelizing the grid search.
        def map_rolling_aic(args):
            x, h, alpha = args
             return rolling_aic(data.drop(columns='S'), data['S'], h, alpha)
In [ ]: def map_rolling_msfe(args):
             x, h, alpha = args
             return rolling_msfe(data.drop(columns='S'), data['S'], h, alpha)
In [ ]: with multiprocessing.Pool() as p:
             msfe_values = p.map(
                 map_rolling_msfe
                 [(data, x[0], x[1]) for x in grid_]
In [ ]: with multiprocessing.Pool() as p:
            aic_values = p.map(
                 map_rolling_aic,
                 [(data, x[0], x[1]) for x in grid_]
In [ ]: # Dataframe with MSFE values and hyperparameters.
        msfe_values = pd.DataFrame(
            {
                 'h': [x[0] for x in grid_],
                 'alpha': [x[1] for x in grid_],
                 'msfe': msfe_values
In [ ]: # Dataframe with AIC values and hyperparameters.
        aic_values = pd.DataFrame(
                 'h': [x[0] for x in grid_],
                 'alpha': [x[1] for x in grid_],
                 'aic': aic_values
In [ ]: aic_values.to_csv('data/aic_values.csv', index=False)
In [ ]: # Heatmap for the MSFE values.
        msfe_values_heatmap = msfe_values.pivot(index='h', columns='alpha', values='msfe')
        msfe_values_heatmap = msfe_values_heatmap.sort_index(ascending=False)
        fig. ax = plt.subplots()
        sns.heatmap(msfe_values_heatmap, ax=ax)
x_tick_labels = [item.get_text() for item in ax.get_xticklabels()]
         formatted_labels = [f"{float(label):.2f}" for label in x_tick_labels]
        {\tt ax.set\_xt\"{i}cklabels(formatted\_labels,\ rotation=90)} \quad \textit{\# Apt\"{i}car \ \=las\ etiquetas\ formateadas\ al\ eje\ X}
        ax.set_title('MSFE Values')
        plt.show()
                                 MSFE Values
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In []: # Heatmap for the AIC values.
aic_values_heatmap = aic_values.pivot(index='h', columns='alpha', values='aic')
aic_values_heatmap = aic_values_heatmap.sort_index(ascending=False)

fig, ax = plt.subplots()
sns.heatmap(aic_values_heatmap, ax=ax)
x_tick_labels = [item.get_text() for item in ax.get_xticklabels()]
formatted_labels = [if"{float(label):.2f}" for label in x_tick_labels]
ax.set_xticklabels(formatted_labels, rotation=90) # Aplicar las etiquetas formateadas al eje X
ax.set_title('AIC Values')
plt.show()
```



In []: