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Lab 3 Maryna Borovyk Task 2. Grid search for simple exponential smoothing.
 In [1]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
 In [2]: df = pd.read_csv('AlgeriaExport.txt', header = None)
         df['Year']=pd.RangeIndex(start=1960, stop=1960+len(df))
         df['Year'] = pd.to_datetime(df['Year'], format='%Y')
         df.set_index('Year', inplace=True)
         df.columns = ['Export']
         df.index.freq='YS'
         df.head()
 Out[2]:
                        Export
                Year
         1960-01-01 39.043173
         1961-01-01 46.244557
         1962-01-01 19.793873
         1963-01-01 24.684682
         1964-01-01 25.084059
 In [3]: df.plot()
 Out[3]: <Axes: xlabel='Year'>
        50
                 Export
        45
        40
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        25
        20
        15 ·
         1960
                    1970
                                1980
                                           1990
                                                       2000
                                                                  2010
                                          Year
 In [4]: train_data = df.iloc[0:45]
         test_data = df.iloc[45:]
 In [5]: train_data['Export'].plot(legend=True, label='train')
         test_data['Export'].plot(legend=True, label='test')
 Out[5]: <Axes: xlabel='Year'>
        40
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        20
         1960
                    1970
                                1980
                                           1990
                                                       2000
                                                                  2010
                                          Year
 In [6]: train_values=train_data['Export'].values
         alpha=0.7
         10=train_values[0]
         predicted_train_values=[10]
 In [7]: for i in range(0, len(train_values-1)):
             predicted_train_values.append(alpha*train_values[i]+(1-alpha)*predicted_train_values[i])
 In [8]: plt.plot(predicted_train_values)
         plt.plot(train_values)
 Out[8]: [<matplotlib.lines.Line2D at 0x12beeaea0>]
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                           10
 In [9]: def ses(train_values, alpha, 10):
             predicted_train_values = [10]
             for i in range(0, len(train_values)-1):
                predicted_train_values.append(alpha * train_values[i] + (1 - alpha) * predicted_train_values[i])
             return predicted_train_values
In [10]: pdt=ses(train_values, 0.7, train_values[0])
         plt.plot(pdt)
Out[10]: [<matplotlib.lines.Line2D at 0x12bf4ef90>]
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              0
                           10
                                        20
                                                     30
                                                                   40
In [11]: def sse(values, predictions):
             for i in range(len(values)):
                sum+=(values[i]-predictions[i])**2
             return sum
In [12]: sse(train_values, predicted_train_values)
Out[12]: 1663.92327922879
In [13]: def ses_fit(train_values, dalpha):
             10=train_values[0]
             alpha_grid=np.arange(0, 1, dalpha)
             sse_list=[]
             for alpha in alpha_grid:
                predicted_train_values=ses(train_values, alpha, 10)
                sse_list.append(sse(train_values, predicted_train_values))
             min_index=min(enumerate(sse_list), key=lambda x: x[1]) [0]
             best_alpha=alpha_grid[min_index]
             return best_alpha
In [14]: alpha=ses_fit(train_values,0.01)
         print(alpha)
       0.76
In [15]: def ses_predict(train_values, alpha, 10, h):
             predicted_train_values=ses(train_values, alpha, 10)
            one_step=alpha*train_values[-1]+(1-alpha)*predicted_train_values[-1]
             h_steps=[one_step]*h
             return one_step, h_steps, predicted_train_values
In [16]: one_step, h_steps, predicted_train_values=ses_predict(train_values, alpha, train_values[0], len(test_data))
In [17]: train_data['Fit']=predicted_train_values
         test_data['Predicted']=h_steps
         test_data
       A value is trying to be set on a copy of a slice from a DataFrame.
       Try using .loc[row_indexer,col_indexer] = value instead
       See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
         train_data['Fit']=predicted_train_values
       /var/folders/z6/r8xtfhp56gd9rzqzptq9vv880000gn/T/ipykernel_79542/1599865325.py:2: SettingWithCopyWarning:
       A value is trying to be set on a copy of a slice from a DataFrame.
       Try using .loc[row_indexer,col_indexer] = value instead
       See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
        test_data['Predicted']=h_steps
Out[17]:
                        Export Predicted
               Year
         2005-01-01 47.205193 39.484538
         2006-01-01 48.810688 39.484538
         2007-01-01 47.068164 39.484538
         2008-01-01 47.973345 39.484538
         2009-01-01 35.371651 39.484538
         2010-01-01 38.444548 39.484538
         2011-01-01 38.786954 39.484538
         2012-01-01 36.890548 39.484538
         2013-01-01 33.209898 39.484538
         2014-01-01 30.219117 39.484538
         2015-01-01 23.171778 39.484538
         2016-01-01 20.860011 39.484538
         2017-01-01 22.638887 39.484538
In [18]: train_data['Export'].plot(legend=True,label='train')
         test_data['Export'].plot(legend=True,label='test')
         train_data['Fit'].plot(legend=True,label='fit')
         test_data['Predicted'].plot(legend=True,label='prediction')
Out[18]: <Axes: xlabel='Year'>
        50
                                      — train
                                        test
        45
                                        – fit
                                       prediction
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Year

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