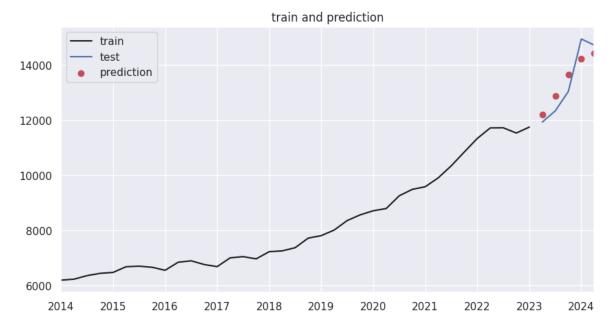
```
In [11]: %load ext autoreload
         %autoreload 2
         import sys
         sys.path.append("../")
         import pandas as pd
         import numpy as np
         import statsmodels.api as sm
         import statsmodels.graphics as sg
         import matplotlib.pylab as plt
         import seaborn as sns
         sns.set theme()
         from sklearn.metrics import r2 score, mean squared error
         from sklearn.linear model import LinearRegression
         import statsmodels.api as sm
         import statsmodels.formula.api as smf
         from statsmodels.stats.outliers influence import OLSInfluence
         from pygam import LinearGAM, s, l
         from pygam.datasets import wage
         import seaborn as sns
         import matplotlib.pyplot as plt
         from dmba import stepwise selection
         from dmba import AIC score
         %matplotlib inline
        The autoreload extension is already loaded. To reload it, use:
         %reload ext autoreload
In [12]: | df = pd.read_csv('.../data/nts.csv', index_col=0, parse_dates=True)
         df.shape, df.columns
Out[12]: ((42, 15),
          Index(['rynek', 'inflacja_r', 'inflacja_q', 'stopa_procentowa',
                 'liczba_kredytow', 'tempo_wzrostu', 'ufnosc', 'duze_zakupy',
                 'bezrobocie', 'spr_detaliczna', 'pkb', 'wynag', 'inflacja_q_ku
         m',
                 'ogolem', 'rpk'],
                dtype='object'))
'bezrobocie', 'spr_detaliczna', 'pkb', 'wynag', 'inflacja_q_kum',
                'ogolem', 'rpk']
         outcome = 'rynek'
         model = LinearRegression()
         bn = 37
```

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```
model.fit(X=df[predictors].iloc[:bn], y=df[outcome].iloc[:bn])
         print(f'Wyraz wolny: {model.intercept :.3f}')
         print('Współczynniki:')
         for name, coef in zip(predictors, model.coef ):
             print(f' {name}: {coef}')
        Wyraz wolny: 9241.832
        Współczynniki:
         inflacja_r: -27.007597094080783
         inflacja q: -38.893426250826906
         stopa procentowa: 13377.093887206072
         liczba kredytow: 0.01731614343817256
         tempo wzrostu: -59.27660706477059
         ufnosc: -23.08856390632795
         duze zakupy: -4.037395044947253
         bezrobocie: 6.33587746349118
         spr detaliczna: 16.144534333623447
         pkb: 8.85640126139158
         wynag: 2.4084176822627157
         inflacja q kum: -4034.808805515178
         ogolem: -0.004609262250147239
         rpk: 0.0
In [27]: | preds = model.predict(df[predictors].iloc[bn:])
         RMSE = np.sqrt(mean_squared_error(df[outcome].iloc[bn:], preds))
         r2 = r2 score(df[outcome].iloc[bn:], preds)
         print(f'RMSE: {RMSE:.0f}')
         print(f'r2: {r2:.4f}')
        RMSE: 517
        r2: 0.8229
In [39]: preds, df.index[bn:]
Out[39]: (array([12188.64894668, 12865.73931734, 13658.13377819, 14227.55739409,
                  14415.16922223]),
          DatetimeIndex(['2023-06-30', '2023-09-30', '2023-12-31', '2024-03-31',
                          '2024-06-30'],
                         dtype='datetime64[ns]', freq=None))
In [46]: | fig = plt.Figure()
         ax = fig.gca()
         df['rynek'].iloc[:bn].plot(style='k', figsize=(10,5), label = 'train', ax
         df['rynek'].iloc[bn:].plot(style='b', figsize=(10,5), label = 'test', ax=
         ax.scatter(x=df.index[bn:], y=preds, c='r', label = 'prediction')
         ax.set title('train and prediction')
         ax.legend()
         fig.savefig('../images/lr output.png')
         display(fig)
```

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```
In [41]: model = sm.OLS(df[outcome], df[predictors].assign(const=1))
    results = model.fit()
    print(results.summary())
```

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OLS Regression Results

==== Dep. Variable:		rynek	R-squared:		
0.996 Model:		0LS	Adj. R-squared:		
0.994 Method:	ا معد		F-statistic:		5
18.8	·				
Date: e-29			Prob (F-statistic):		4.78
Time: 9.30	11:35:32		Log-Likelihood:		-26
No. Observations: 68.6		42	AIC:		5
Df Residuals: 94.7		27	BIC:		5
Df Model:	14 nonrobust				
Covariance Type:					
=======				5 1	
0.975]	coef	std err	t	P> t	[0.025
inflacja_r 35.849	-42.8063	38.334	-1.117	0.274	-121.462
inflacja_q 53.747	-72.2687	61.416	-1.177	0.250	-198.284
stopa_procentowa	1.617e+04	5449.041	2.967	0.006	4986.618
2.73e+04 liczba_kredytow 0.034	0.0204	0.007	3.008	0.006	0.006
tempo_wzrostu 46.080	-49.4437	46.555	-1.062	0.298	-144.967
ufnosc 29.310	-31.3060	29.542	-1.060	0.299	-91.922
duze_zakupy 46.541	1.9986	21.709	0.092	0.927	-42.544
bezrobocie	5.0389	82.263	0.061	0.952	-163.750
173.828 spr_detaliczna 81.391	42.1357	19.132	2.202	0.036	2.881
pkb	-26.0818	28.737	-0.908	0.372	-85.045
32.881 wynag	2.4606	0.514	4.788	0.000	1.406
3.515 inflacja_q_kum	-3835.8214	3529.897	-1.087	0.287	-1.11e+04
3406.929 ogolem	-0.0115	0.018	-0.651	0.521	-0.048
0.025 rpk	-884.7307	227.921	-3.882	0.001	-1352.385
-417.076 const 2.62e+04	1.344e+04	6210.021	2.165	0.039	701.597
		=======			
Omnibus:		0.586	Durbin-Watso	n:	
1.617 Prob(Omnibus):		0.746 Jarque-Bera (JB):			

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1.42

0.701 Skew:

0.232 Prob(JB):

0.704

Kurtosis:

2.569 Cond. No.

e+07

====

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

 $\[2\]$ The condition number is large, 1.42e+07. This might indicate that ther e are

strong multicollinearity or other numerical problems.

In []:

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