## LP problem, Bakery Horvat VT, profit maximization

## May 18, 2021

```
[9]: import numpy as np
     import pandas as pd
     from sklearn.neighbors import KNeighborsRegressor
     from sklearn.neighbors import KNeighborsClassifier
     from sklearn.metrics import confusion_matrix, accuracy_score
     from sklearn.model_selection import cross_val_score
     from sklearn.metrics import mean_squared_error
     import matplotlib.pyplot as plt
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import MinMaxScaler
     from sklearn.model_selection import GridSearchCV
     from sklearn.preprocessing import LabelEncoder
     %matplotlib inline
     import matplotlib as mpl
     import matplotlib.pyplot as plt
     from sklearn.pipeline import Pipeline
     import numpy as np
     import pandas as pd
     import statsmodels.api as sm
     from sklearn.preprocessing import StandardScaler
     from sklearn.preprocessing import PolynomialFeatures
     from sklearn.model_selection import train_test_split, cross_val_score
     from sklearn.linear_model import LinearRegression
     from sklearn.linear_model import SGDRegressor
     from sklearn.metrics import mean_squared_error, r2_score
     from sklearn.pipeline import Pipeline
     from sklearn.model_selection import GridSearchCV
     from sklearn.svm import LinearSVC
```

```
from sklearn.svm import SVC
import sklearn.linear_model as skl lm
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.model_selection import train_test_split, LeaveOneOut, KFold,
⇒StratifiedKFold, cross_val_score
from sklearn.preprocessing import PolynomialFeatures
from sklearn.model_selection import GridSearchCV
#scikit-learn packages for machine learning algorithms
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import LogisticRegression
from sklearn.pipeline import Pipeline
from sklearn.linear_model import SGDRegressor
from sklearn.preprocessing import scale
from sklearn.model_selection import train_test_split
from sklearn.linear_model import Ridge, RidgeCV, Lasso, LassoCV
from sklearn.linear_model import ElasticNet
from sklearn.metrics import mean_squared_error
import numpy as np
import matplotlib.pyplot as plt
from sklearn import linear model
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import confusion_matrix, classification_report, __
→precision_score, roc_curve, roc_auc_score, accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeRegressor, DecisionTreeClassifier
from sklearn.tree import export_text, plot_tree, export_graphviz
from sklearn.ensemble import RandomForestRegressor, GradientBoostingRegressor
from sklearn.metrics import confusion_matrix, mean_squared_error
from sklearn.model_selection import GridSearchCV
#from dtreeviz.trees import dtreeviz
# install graphviz : pip install graphviz
# download graphviz-2.38 and place it into C:/Program Files
import graphviz
import os
os.environ["PATH"] += os.pathsep + 'C:/Program Files/graphviz-2.38/release/bin/'
import matplotlib.pyplot as plt
from IPython.core.interactiveshell import InteractiveShell
```

```
InteractiveShell.ast_node_interactivity = 'all'
[10]: import pulp as p
[11]: Lp_prob = p.LpProblem('Profit_maximization', p.LpMaximize)
[12]: x1 = p.LpVariable("x1", lowBound = 0)
            x2 = p.LpVariable("x2", lowBound = 0)
            x3 = p.LpVariable("x3", lowBound = 0)
            x4 = p.LpVariable("x4", lowBound = 0)
            x5 = p.LpVariable("x5", lowBound = 0)
            x6 = p.LpVariable("x6", lowBound = 0)
            x7 = p.LpVariable("x7", lowBound = 0)
            x8 = p.LpVariable("x8", lowBound = 0)
[13]: Lp_prob += 4.71 * x1 + 2.39 * x2 + 2.77 * x3 + 2.39 * x4 + 1.99 * x5 + 5.35 *_
             \rightarrowx6 + 3.44 * x7 + 5.49 * x8
            Lp_prob += 0.5 * x1 + 0.042 * x2 + 0.084 * x3 + 0.055 * x4 + 0.058 * x5 + 0.1 *_{\sqcup}
             \rightarrowx6 + 0.084 * x7 + 0.084 * x8 <= 3450
            Lp_prob += 0.3 * x1 + 0.025 * x2 + 0.05 * x3 + 0.03 * x4 + 0.032 * x5 + 0.06 *_{\sqcup}
              \rightarrowx6 + 0.05 * x7 + 0.05 * x8 <= 10000
            Lp\_prob += 0.01 * x1 + 0.0008 * x2 + 0.0016 * x3 + 0.0008 * x4 + 0.0008 * x5 + 0.000
             \hookrightarrow0.003 * x6 + 0.0016 * x7 + 0.0016 * x8 <= 70
            Lp prob += 0 * x1 + 0 * x2 + 0 * x3 + 0.001 * x4 + 0.0008 * x5 + 0 * x6 + 0 *_{11}
             \rightarrow x7 + 0 * x8 <= 5
            Lp prob += 0.01 * x1 + 0.0008 * x2 + 0.0016 * x3 + 0.001 * x4 + 0.001 * x5 + 0.
              \rightarrow002 * x6 + 0.0016 * x7 + 0.0016 * x8 <= 70
            Lp prob += 0 * x1 + 0.00025 * x2 + 0.0005 * x3 + 0.008 * x4 + 0.007 * x5 + 0.
             \rightarrow0007 * x6 + 0.0005 * x7 + 0.0005 * x8 <= 20
            Lp_prob += 0 * x1 + 0.00025 * x2 + 0.0005 * x3 + 0.0006 * x4 + 0.0008 * x5 + 0.
             \rightarrow 0005 * x6 + 0.0005 * x7 + 0.0005 * x8 <= 6
            Lp prob += 0 * x1 + 0 * x2 + 0 * x3 + 0.25 * x4 + 0.25 * x5 + 0 * x6 + 0 * x7 +
             →0 * x8 <= 400
            Lp prob += 0 * x1 + 0 * x2 + 0 * x3 + 0.03 * x4 + 0.04 * x5 + 0 * x6 + 0 * x7 + 11
             →0 * x8 <= 55
            Lp prob += 0 * x1 + 0 * x2 + 0 * x3 + 0 * x4 + 0 * x5 + 0.04 * x6 + 0.025 * x7_{\bot}
             →+ 0 * x8 <= 100
            Lp_prob += 0 * x1 + 0 * x2 + 0 * x3 + 0 * x4 + 0 * x5 + 0.04 * x6 + 0.025 * x7_{\square}
             →+ 0.025 * x8 <= 120
            Lp\_prob += 0 * x1 + 0 * x2 + 0 * x3 + 0 * x4 + 0 * x5 + 0.025 * x6 + 0 * x7 + 0
             →* x8 <= 40
            Lp_prob += 0 * x1 + 0 * x2 + 0 * x3 + 0 * x4 + 0 * x5 + 0.0002 * x6 + 0 * x7 + 0.0002
             \rightarrow 0 * x8 <= 0.5
            Lp_prob += 0 * x1 + 0 * x2 + 0 * x3 + 0 * x4 + 0 * x5 + 0 * x6 + 0 * x7 + 0.04
             →* x8 <= 40
```

```
 \begin{array}{l} \text{Lp\_prob} \ += \ 0 \ * \ x1 \ + \ 0 \ * \ x2 \ + \ 0 \ * \ x3 \ + \ 0 \ * \ x4 \ + \ 0 \ * \ x5 \ + \ 0 \ * \ x6 \ + \ 0 \ * \ x7 \ + \ 0.015 \\ \longrightarrow * \ x8 \ <= \ 15 \\ \end{array}
```

## [14]: print(Lp\_prob)

Profit maximization:

MAXIMIZE

4.71\*x1 + 2.39\*x2 + 2.77\*x3 + 2.39\*x4 + 1.99\*x5 + 5.35\*x6 + 3.44\*x7 + 5.49\*x8 + 0.0

SUBJECT TO

\_C1:  $0.5 \times 1 + 0.042 \times 2 + 0.084 \times 3 + 0.055 \times 4 + 0.058 \times 5 + 0.1 \times 6 + 0.084 \times 7 + 0.084 \times 8 <= 3450$ 

\_C2:  $0.3 \times 1 + 0.025 \times 2 + 0.05 \times 3 + 0.03 \times 4 + 0.032 \times 5 + 0.06 \times 6 + 0.05 \times 7 + 0.05 \times 8 \le 10000$ 

\_C3:  $0.01 \times 1 + 0.0008 \times 2 + 0.0016 \times 3 + 0.0008 \times 4 + 0.0008 \times 5 + 0.003 \times 6 + 0.0016 \times 7 + 0.0016 \times 8 <= 70$ 

\_C4:  $0.001 \times 4 + 0.0008 \times 5 \le 5$ 

\_C5:  $0.01 \times 1 + 0.0008 \times 2 + 0.0016 \times 3 + 0.001 \times 4 + 0.001 \times 5 + 0.002 \times 6 + 0.0016 \times 7 + 0.0016 \times 8 <= 70$ 

\_C6:  $0.00025 \times 2 + 0.0005 \times 3 + 0.008 \times 4 + 0.007 \times 5 + 0.0007 \times 6 + 0.0005 \times 7 + 0.0005 \times 8 \le 20$ 

\_C7:  $0.00025 \times 2 + 0.0005 \times 3 + 0.0006 \times 4 + 0.0008 \times 5 + 0.0005 \times 6 + 0.0005 \times 7 + 0.0005 \times 8 \le 6$ 

\_C8:  $0.25 \times 4 + 0.25 \times 5 \le 400$ 

\_C9:  $0.03 \times 4 + 0.04 \times 5 \le 55$ 

\_C10:  $0.04 \times 6 + 0.025 \times 7 \le 100$ 

\_C11:  $0.04 \times 6 + 0.025 \times 7 + 0.025 \times 8 \le 120$ 

\_C12: 0.025 x6 <= 40

\_C13: 0.0002 x6 <= 0.5

\_C14: 0.04 x8 <= 40

\_C15: 0.015 x8 <= 15

VARIABLES

```
x1 Continuous
x2 Continuous
x3 Continuous
x4 Continuous
x5 Continuous
x6 Continuous
x7 Continuous
x8 Continuous
x8 Continuous

(15]: status = Lp_prob.solve()
    print(p.LpStatus[status])

Optimal

[18]: print(p.value(x1), p.value(x2), p.value(x3), p.value(x4), p.value(x5), p.
```

4832.8 18800.0 0.0 0.0 0.0 1600.0 0.0 1000.0 81744.488

[]:

→value(x6), p.value(x7), p.value(x8), p.value(Lp\_prob.objective))