

# PD4

May 5, 2021

```
[1]: import pandas as pd
from sklearn.svm import SVC
from sklearn.datasets import load_wine
import dalex
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.model_selection import RandomizedSearchCV
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import accuracy_score
import warnings
warnings.filterwarnings('ignore')
```

## 1 Zbiór danych apartments

```
[2]: apartments = dalex.datasets.load_apartments()
print(apartments.info())
apartments.head()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1000 entries, 1 to 1000
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  -
0   m2_price               1000 non-null   int64
1   construction_year      1000 non-null   int64
2   surface                1000 non-null   int64
3   floor                  1000 non-null   int64
4   no_rooms                1000 non-null   int64
5   district               1000 non-null   object
dtypes: int64(5), object(1)
memory usage: 54.7+ KB
None
```

```
[2]:
```

	m2_price	construction_year	surface	floor	no_rooms	district
1	5897	1953	25	3	1	Srodmiescie
2	1818	1992	143	9	5	Bielany
3	3643	1937	56	1	2	Praga
4	3517	1995	93	7	3	Ochota
5	3013	1992	144	6	5	Mokotow

### 1.0.1 Encoding i podział na zbiory treningowy i testowy

```
[23]: le=LabelEncoder()
aparts['district'] = le.fit_transform(aparts['district'])

X = aparts.drop('district', axis = 1)
y = aparts['district']

X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, test_size=
↳ 0.3, random_state = 29)
```

### 1.0.2 Prosty model bez skalowania

```
[15]: model_svc = SVC()

model_svc.fit(X_train, y_train)
y_pred = model_svc.predict(X_test)

simple_model = accuracy_score(y_test, y_pred)
print('Accuracy score dla prostego modelu bez skalowania i strojenia_
↳ hiperparametrów wynosi: ', simple_model)
```

Accuracy score dla prostego modelu bez skalowania i strojenia hiperparametrów wynosi: 0.21666666666666667

### 1.0.3 Strojenie hiperparametrów bez skalowania

```
[24]: model_svc = SVC()

C = [i for i in range(10)]
C.append(20)
C.append(50)

parameters = {
    'C' : C,
    'gamma' : ['scale', 'auto', 3, 10],
    'degree': np.linspace(1, 5, 5)
```

```

}

svc_rand = RandomizedSearchCV(model_svc, parameters, n_iter=1000, n_jobs=-1)
best_svc = svc_rand.fit(X_train ,y_train)
best_svc.best_estimator_

```

[24]: SVC(C=1, degree=1.0)

```

[25]: model_svc = best_svc.best_estimator_
model_svc.fit(X_train, y_train)

y_pred = model_svc.predict(X_test)

hiper_model = accuracy_score(y_test, y_pred)
print('Accuracy score dla modelu bez skalowania ze strojeniem hiperparametrów
→wynosi: ', hiper_model)

```

Accuracy score dla modelu bez skalowania ze strojeniem hiperparametrów wynosi:  
0.21666666666666667

#### 1.0.4 Skalowanie bez strojenia hiperparametrów

```

[27]: scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.fit_transform(X_test)

model_svc = SVC()

model_svc.fit(X_train, y_train)
y_pred = model_svc.predict(X_test)

scal_model = accuracy_score(y_test, y_pred)
print('Accuracy score dla prostego modelu ze skalowaniem i bez strojenia
→hiperparametrów wynosi: ', scal_model)

```

Accuracy score dla prostego modelu ze skalowaniem i bez strojenia  
hiperparametrów wynosi: 0.31333333333333335

#### 1.0.5 Strojenie hiperparametrów ze skalowaniem

```

[30]: model_svc = SVC()

svc_rand = RandomizedSearchCV(model_svc, parameters, n_iter=1000, n_jobs=-1)
best_svm = svc_rand.fit(X_train ,y_train)
best_svm.best_estimator_

```

```
[30]: SVC(C=4, degree=1.0, gamma='auto')
```

```
[31]: model_svc = best_svm.best_estimator_  
  
model_svc.fit(X_train, y_train)  
  
y_pred = model_svc.predict(X_test)  
  
hiper_scal_model = mean_squared_error(y_test, y_pred, squared=True)  
print('RMSE dla modelu ze skalowaniem i strojeniem hiperparametrów wynosi: ',   
      ⇨hiper_scal_model)
```

RMSE dla modelu ze skalowaniem i strojeniem hiperparametrów wynosi:  
16.546666666666667

### 1.0.6 Podsumowanie

Skalowanie poprawiło wyniki modelu, za to strojenie hiperparametrów bez skalowania nie poprawiło wyniku, a po skalowaniu nawet pogorszyło, może to wynikać z tego, że RandomizedSearch nie trafił w optymalne rozwiązanie.

## 2 Zbiór danych Wine

```
[13]: wine_dict = load_wine()
```

```
[14]: X = pd.DataFrame(wine_dict['data'], columns = wine_dict['feature_names'])  
Y = wine_dict['target']  
  
X_train, X_test, y_train, y_test = train_test_split(X, Y, random_state = 29,   
      ⇨test_size = 0.3)
```

### 2.0.1 Prosty model bez skalowania

```
[15]: model_svc = SVC(random_state=29)  
  
model_svc.fit(X_train, y_train)  
y_pred = model_svc.predict(X_test)  
  
simple_model = accuracy_score(y_test, y_pred)  
  
print('Accuracy score dla prostego modelu bez skalowania i strojenia   
      ⇨hiperparametrów wynosi: ', simple_model)
```

Accuracy score dla prostego modelu bez skalowania i strojenia hiperparametrów wynosi: 0.5740740740740741

## 2.0.2 Strojenie hiperparametrów bez skalowania

```
[16]: model_svc = SVC()

parameters = {
    'C' : [np.linspace(1, 10, 10), 20, 50],
    'gamma' : ['scale', 'auto', 3, 10],
    'degree': np.linspace(1, 5, 5),
    'kernel': ['linear', 'poly', 'rbf']
}

svc_rand = RandomizedSearchCV(model_svc, parameters, n_iter=100)
best_svc = svc_rand.fit(X_train, y_train)
best_svc.best_estimator_
```

```
[16]: SVC(C=20, degree=4.0, gamma=3, kernel='poly')
```

```
[17]: model_svc = best_svc.best_estimator_

model_svc.fit(X_train, y_train)

y_pred = model_svc.predict(X_test)

hiper_model = accuracy_score(y_test, y_pred)
print('Accuracy score dla modelu bez skalowania ze strojeniem hiperparametrów_
↪wynosi: ', hiper_model)
```

Accuracy score dla modelu bez skalowania ze strojeniem hiperparametrów wynosi: 0.9259259259259259

## 2.0.3 Skalowanie bez strojenia hiperparametrów

```
[18]: scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.fit_transform(X_test)

model_svc = SVC(random_state=29)

model_svc.fit(X_train, y_train)
y_pred = model_svc.predict(X_test)

scal_model = accuracy_score(y_test, y_pred)
```

```
print('Accuracy score dla prostego modelu ze skalowaniem i bez strojenia_
↪hiperparametrów wynosi: ', scal_model)
```

Accuracy score dla prostego modelu ze skalowaniem i bez strojenia  
hiperparametrów wynosi: 0.9814814814814815

#### 2.0.4 Strojenie hiperparametrów ze skalowaniem

```
[19]: model_svc = SVC()
svc_rand = RandomizedSearchCV(model_svc, parameters, n_iter=100,
↪random_state=29)
best_svc = svc_rand.fit(X_train, y_train)
best_svc.best_estimator_
```

```
[19]: SVC(C=20, degree=4.0, gamma='auto', kernel='linear')
```

```
[20]: model_svc = best_svc.best_estimator_

model_svc.fit(X_train, y_train)

y_pred = model_svc.predict(X_test)

hiper_model = accuracy_score(y_test, y_pred)
print('Accuracy score dla modelu ze skalowaniem i strojeniem hiperparametrów_
↪wynosi: ', hiper_model)
```

Accuracy score dla modelu ze skalowaniem i strojeniem hiperparametrów wynosi:  
0.9259259259259259

#### 2.0.5 Podsumowanie

Skalowanie sprawiło, że model ma prawie 98% poprawności, jednak strojenie hiperparametrów nie poprawiło wyników.