

## Introduction

In a previous section, you learned about how to use pipelines in scikit-learn to combine several supervised learning algorithms in a manageable pipeline. In this lesson, you will integrate PCA along with classifiers in the pipeline.

# **Objectives**

In this lab you will:

Integrate PCA in scikit-learn pipelines

## The Data Science Workflow

You will be following the data science workflow:

- 1. Initial data inspection, exploratory data analysis, and cleaning
- 2. Feature engineering and selection
- 3. Create a baseline model
- 4. Create a machine learning pipeline and compare results with the baseline model
- 5. Interpret the model and draw conclusions

# Initial data inspection, exploratory data analysis, and cleaning

You'll use a dataset created by the Otto group, which was also used in a Kaggle competition. The description of the dataset is as follows:

The Otto Group is one of the world's biggest e-commerce companies, with subsidiaries in more than 20 countries, including Crate & Barrel (USA), Otto.de (Germany) and 3 Suisses (France). They are selling millions of products worldwide every day, with several thousand products being added to their product line.

A consistent analysis of the performance of their products is crucial. However, due to their global infrastructure, many identical products get classified differently. Therefore, the quality of product analysis depends heavily on the ability to accurately cluster similar products. The better the classification, the more insights the Otto Group can generate about their product range.

In this lab, you'll use a dataset containing:

- A column id, which is an anonymous id unique to a product
- 93 columns feat\_1, feat\_2, ..., feat\_93, which are the various features of a product
- a column target the class of a product

The dataset is stored in the 'otto\_group.csv' file. Import this file into a DataFrame called data, and then:

- Check for missing values
- Check the distribution of columns
- ... and any other things that come to your mind to explore the data

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

```
data = pd.read_csv("otto_group.csv")

data.head()

<style scoped> .dataframe tbody tr th:only-of-type { vertical-align: middle; }

.dataframe tbody tr th {
    vertical-align: top;
}

.dataframe thead th {
    text-align: right;
}
```

#### </style>

	id	feat_1	feat_2	feat_3	feat_4	feat_5	feat_6	feat_7	feat_8
0	1	1	0	0	0	0	0	0	0
1	2	0	0	0	0	0	0	0	1
2	3	0	0	0	0	0	0	0	1
3	4	1	0	0	1	6	1	5	0
4	5	0	0	0	0	0	0	0	0
4					1		1		<b>•</b>

#### 5 rows × 95 columns

data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 61878 entries, 0 to 61877
```

```
RangeIndex: 61878 entries, 0 to 61877
Data columns (total 95 columns):
id
          61878 non-null int64
feat_1
          61878 non-null int64
          61878 non-null int64
feat 2
feat 3
          61878 non-null int64
feat_4
          61878 non-null int64
feat 5
          61878 non-null int64
feat_6
          61878 non-null int64
feat_7
           61878 non-null int64
```

feat_8	61878	non-null	int64
feat_9	61878	non-null	int64
feat_10	61878	non-null	int64
feat_11	61878	non-null	int64
feat_12	61878	non-null	int64
feat_13	61878	non-null	int64
feat_14	61878	non-null	int64
feat_15	61878	non-null	int64
feat_16	61878	non-null	int64
feat_17	61878	non-null	int64
feat_18	61878	non-null	int64
feat_19	61878	non-null	int64
feat_20	61878	non-null	int64
feat_21	61878	non-null	int64
feat_22	61878	non-null	int64
feat_23	61878	non-null	int64
feat_24	61878	non-null	int64
feat_25	61878	non-null	int64
feat_26	61878	non-null	int64
feat_27	61878	non-null	int64
feat_28	61878	non-null	int64
feat_29	61878	non-null	int64
feat_30	61878	non-null	int64
feat_31	61878	non-null	int64
feat_32	61878	non-null	int64
feat_33	61878	non-null	int64
feat_34	61878	non-null	int64
feat_35	61878	non-null	int64
feat_36	61878	non-null	int64
feat_37	61878	non-null	int64
feat_38	61878	non-null	int64
feat_39		non-null	
feat_40	61878		
feat_41	61878		
feat_42	61878		
feat_43			int64
feat_44	61878		
feat_45	61878	non-null	
feat 46	61878	non-null	
feat_47	61878		
feat_48	61878		
feat_49	61878	non-null	int64
feat_50	61878	non-null	int64
feat_51	61878		
feat_52		non-null	
feat_53	61878		
feat_54	61878		
feat_55		non-null	
feat_56	61878		
. 545_56	010/0	uii	Z.1.CO-7

```
feat 57
           61878 non-null int64
feat 58
           61878 non-null int64
feat 59
           61878 non-null int64
feat 60
           61878 non-null int64
feat 61
           61878 non-null int64
           61878 non-null int64
feat 62
feat_63
           61878 non-null int64
feat 64
           61878 non-null int64
feat 65
           61878 non-null int64
           61878 non-null int64
feat_66
feat 67
           61878 non-null int64
feat 68
           61878 non-null int64
           61878 non-null int64
feat 69
feat 70
           61878 non-null int64
           61878 non-null int64
feat 71
feat 72
           61878 non-null int64
feat 73
           61878 non-null int64
feat 74
           61878 non-null int64
feat 75
           61878 non-null int64
           61878 non-null int64
feat 76
feat 77
           61878 non-null int64
feat_78
           61878 non-null int64
           61878 non-null int64
feat 79
feat 80
           61878 non-null int64
           61878 non-null int64
feat_81
feat 82
           61878 non-null int64
feat 83
           61878 non-null int64
feat 84
           61878 non-null int64
feat 85
           61878 non-null int64
feat 86
           61878 non-null int64
           61878 non-null int64
feat 87
feat_88
           61878 non-null int64
feat_89
           61878 non-null int64
feat 90
           61878 non-null int64
           61878 non-null int64
feat_91
feat 92
           61878 non-null int64
feat 93
           61878 non-null int64
           61878 non-null object
target
dtypes: int64(94), object(1)
memory usage: 44.8+ MB
data.isna().any()
id
           False
feat 1
           False
```

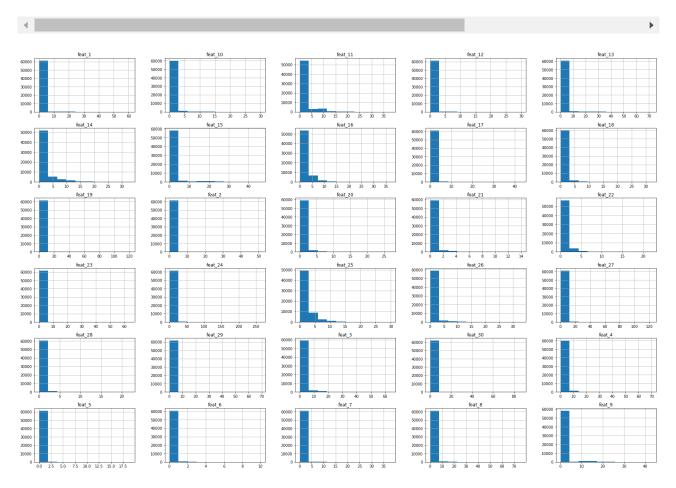
False

feat 2

feat_3	False
feat_4	False
feat_5	False
feat_6	False
feat_7	False
_	
feat_8	False
feat_9	False
feat_10	False
feat_11	False
feat_12	False
feat_13	False
feat_14	False
feat_15	False
feat_16	False
feat_17	False
feat_18	False
feat_19	False
feat_20	False
feat_21	False
feat_22	False
feat_23	False
feat_24	False
feat_25	False
feat_26	False
feat_27	False
Teal 27	FAISE
_	
feat_28	False
_	
feat_28 feat_29	False
feat_28 feat_29 feat_65	False False False
feat_28 feat_29 feat_65 feat_66	False False False False
feat_28 feat_29 feat_65	False False False
feat_28 feat_29 feat_65 feat_66	False False False False
feat_28 feat_29 feat_65 feat_66 feat_67	False False False False False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68	False False False False False False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69	False False False False False False False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70	False False False False False False False False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72	False False False False False False False False False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73	False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74	False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75	False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76	False
feat_28 feat_29 feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80 feat_81	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80 feat_81 feat_82	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80 feat_81 feat_82 feat_83	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80 feat_81 feat_82 feat_83 feat_84	False
feat_28 feat_29  feat_65 feat_66 feat_67 feat_68 feat_69 feat_70 feat_71 feat_72 feat_73 feat_74 feat_75 feat_76 feat_77 feat_78 feat_79 feat_80 feat_81 feat_82 feat_83	False

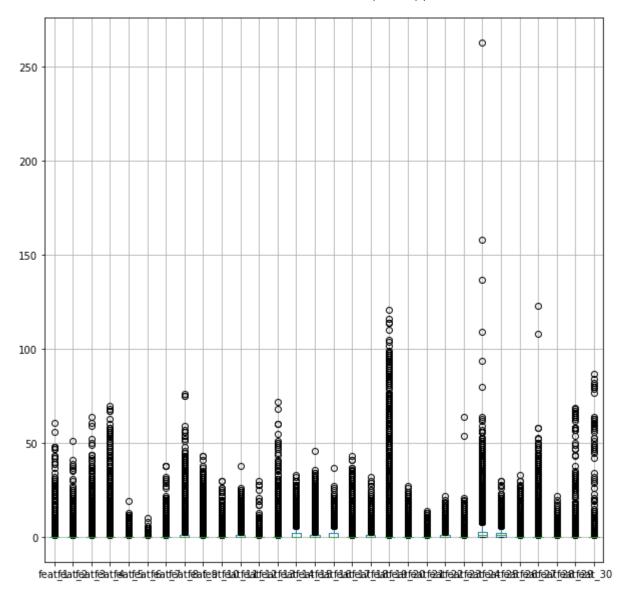
```
feat 86
           False
feat_87
           False
feat_88
           False
feat_89
           False
feat_90
           False
feat_91
           False
feat_92
           False
feat_93
           False
           False
target
Length: 95, dtype: bool
```

```
# We only looked at some of the features here, make sure to do a quick check for all
feat = data.loc[:, 'feat_1':'feat_30']
feat.hist(figsize=(30,20));
```



If you look at all the histograms, you can tell that a lot of the data are zero-inflated, so most of the variables contain mostly zeros and then some higher values here and there. No normality, but for most machine learning techniques this is not an issue.

```
feat.boxplot(figsize=(10,10));
```



Because there are so many zeroes, most values above zero will seem to be outliers. The safe decision for this data is to not delete any outliers and see what happens. With many 0s, sparse data is available and high values may be super informative. Moreover, without having any intuitive meaning for each of the features, we don't know if a value of ~260 is actually an outlier.

# Is there any missing data?

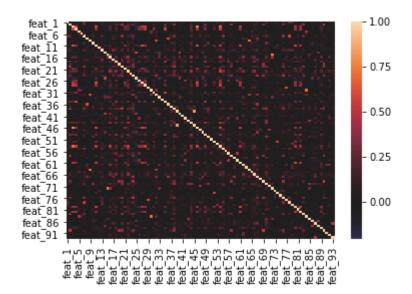
feat.isna().any().any()

False

# Feature engineering and selection with PCA

Have a look at the correlation structure of your features using a heatmap.

```
feat = data.loc[:, 'feat_1':'feat_93']
import seaborn as sns
sns.heatmap(feat.corr(), center=0);
```



Use PCA to select a number of features in a way that you still keep 80% of your explained variance.

```
from sklearn.decomposition import PCA
pca_1 = PCA(n_components=20)
pca_2 = PCA(n_components=40)
pca_3 = PCA(n_components=60)

principalComponents = pca_1.fit_transform(feat)
principalComponents = pca_2.fit_transform(feat)
principalComponents = pca_3.fit_transform(feat)

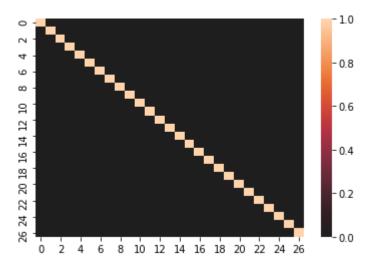
print(np.sum(pca_1.explained_variance_ratio_))
print(np.sum(pca_2.explained_variance_ratio_))
print(np.sum(pca_3.explained_variance_ratio_))

pca = PCA(n_components=27)
principalComponents = pca.fit_transform(feat)
print(np.sum(pca.explained_variance_ratio_))
```

```
0.7275643792253975
```

- 0.8885505230944062
- 0.9598045557819248
- 0.8003287639023293

```
import seaborn as sns
sns.heatmap(pd.DataFrame(principalComponents).corr(), center=0);
```



# Create a train-test split with a test size of 40%

This is a relatively big training set, so you can assign 40% to the test set. Set the random\_state to 42.

```
y = data['target']
X = data.loc[:, 'feat_1':'feat_93']

# Load and split the data

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_stat)
```

## Create a baseline model

Create your baseline model in a pipeline setting. In the pipeline:

- Your first step will be to scale your features down to the number of features that ensure you keep just 80% of your explained variance (which we saw before)
- Your second step will be to build a basic logistic regression model

Make sure to fit the model using the training set and test the result by obtaining the accuracy using the test set. Set the random\_state to 123.

# Create a pipeline consisting of a linear SVM, a simple decision tree, and a simple random forest classifier

Repeat the above, but now create three different pipelines:

- One for a standard linear SVM
- One for a default decision tree
- One for a random forest classifier

```
# This cell may take several minutes to run
from sklearn import svm
from sklearn.pipeline import Pipeline
```

```
from sklearn.ensemble import RandomForestClassifier
from sklearn import tree
## KEEP IT FOR NOW
# Construct some pipelines
pipe_svm = Pipeline([('pca', PCA(n_components=27)),
                     ('clf', svm.SVC(random state=123))])
pipe tree = Pipeline([('pca', PCA(n components=27)),
                      ('clf', tree.DecisionTreeClassifier(random state=123))])
pipe rf = Pipeline([('pca', PCA(n components=27)),
                    ('clf', RandomForestClassifier(random state=123))])
# List of pipelines and pipeline names
pipelines = [pipe svm, pipe tree, pipe rf]
pipeline_names = ['Support Vector Machine','Decision Tree','Random Forest']
# Loop to fit each of the three pipelines
for pipe in pipelines:
    print(pipe)
    pipe.fit(X_train, y_train)
# Compare accuracies
for index, val in enumerate(pipelines):
    print('%s pipeline test accuracy: %.3f' % (pipeline_names[index], val.score(X_te
Pipeline(memory=None,
     steps=[('pca', PCA(copy=True, iterated_power='auto', n_components=27,
random state=None,
  svd solver='auto', tol=0.0, whiten=False)), ('clf', SVC(C=1.0, cache size=200,
class_weight=None, coef0=0.0,
  decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
  max iter=-1, probability=False, random state=123, shrinking=True,
  tol=0.001, verbose=False))])
Pipeline(memory=None,
     steps=[('pca', PCA(copy=True, iterated_power='auto', n_components=27,
random state=None,
  svd_solver='auto', tol=0.0, whiten=False)), ('clf',
DecisionTreeClassifier(class_weight=None, criterion='gini', max_depth=None,
            max features=None, max leaf nodes=None,
            min_impurity_decrease=0.0, min_impurity_split=None,
            min samples leaf=1, min samples split=2,
            min weight fraction leaf=0.0, presort=False, random state=123,
            splitter='best'))])
Pipeline(memory=None,
     steps=[('pca', PCA(copy=True, iterated_power='auto', n_components=27,
```

## Pipeline with grid search

Construct two pipelines with grid search:

- one for random forests try to have around 40 different models
- one for the AdaBoost algorithm

## Random Forest pipeline with grid search

```
# imports
from sklearn import svm
from sklearn.model selection import GridSearchCV
from sklearn.pipeline import Pipeline
# 🔯 This cell may take a long time to run!
# Construct pipeline
pipe_rf = Pipeline([('pca', PCA(n_components=27)),
                    ('clf', RandomForestClassifier(random state = 123))])
# Set grid search params
param_grid_forest = [
 {'clf_n_estimators': [120],
   'clf criterion': ['entropy', 'gini'],
   'clf max depth': [4, 5, 6],
   'clf min samples leaf':[0.05,0.1, 0.2],
   'clf__min_samples_split':[0.05 ,0.1, 0.2]
 }
1
# Construct grid search
gs_rf = GridSearchCV(estimator=pipe_rf,
                     param grid=param grid forest,
                     scoring='accuracy',
```

```
cv=3, verbose=2, return train score = True)
# Fit using grid search
gs_rf.fit(X_train, y_train)
# Best accuracy
print('Best accuracy: %.3f' % gs_rf.best_score_)
# Best params
print('\nBest params:\n', gs_rf.best_params_)
Fitting 3 folds for each of 54 candidates, totalling 162 fits
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.05,
clf__min_samples_split=0.05, clf__n_estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.05, clf__n_estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.05, clf__n_estimators=120
[Parallel(n jobs=1)]: Done 1 out of 1 | elapsed: 7.4s remaining:
                                                                          0.0s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.05, clf__n_estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.05, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.05, clf__n_estimators=120, total= 7.3s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.1, clf__n_estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.1, clf n estimators=120, total= 7.1s
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.05,
clf__min_samples_split=0.1, clf__n_estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.1, clf n estimators=120, total= 7.2s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.1, clf__n_estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.1, clf__n_estimators=120, total= 7.2s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.2, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf__min_samples_split=0.2, clf__n_estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.2, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
```

```
clf min samples split=0.2, clf n estimators=120, total=
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.05,
clf__min_samples_split=0.2, clf__n_estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.05,
clf min samples split=0.2, clf n estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf__min_samples_split=0.05, clf__n_estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.05, clf n estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.05, clf n estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.05, clf n estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.05, clf n estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf__min_samples_split=0.05, clf__n_estimators=120, total=
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.1, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.1, clf n estimators=120, total= 4.7s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.1, clf n estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.1, clf n estimators=120, total=
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf__min_samples_split=0.1, clf__n_estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.1, clf n estimators=120, total=
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.2, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf min samples split=0.2, clf n estimators=120, total=
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf__min_samples_split=0.2, clf__n_estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf__min_samples_split=0.2, clf__n_estimators=120, total= 4.7s
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.1,
clf min samples split=0.2, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.1,
clf__min_samples_split=0.2, clf__n_estimators=120, total= 4.8s
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.2,
clf min samples split=0.05, clf n estimators=120
[CV] clf__criterion=entropy, clf__max_depth=4, clf__min_samples_leaf=0.2,
clf__min_samples_split=0.05, clf__n_estimators=120, total=
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.2,
clf min samples split=0.05, clf n estimators=120
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.2,
clf min samples split=0.05, clf n estimators=120, total=
```

```
[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.2,
clf min samples split=0.05, clf n estimators=120
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clf min samples split=0.1, clf n estimators=120
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clf min samples split=0.2, clf n estimators=120
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[CV] clf criterion=entropy, clf max depth=4, clf min samples leaf=0.2,
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clf min samples split=0.2, clf n estimators=120
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```

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  [Parallel(n jobs=1)]: Done 162 out of 162 | elapsed: 11.1min finished
  Best accuracy: 0.623
  Best params:
   {'clf criterion': 'entropy', 'clf max depth': 6, 'clf min samples leaf':
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Use your grid search object along with .cv results to get the full result overview
  gs rf.cv results
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```

```
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        0.0121848 , 0.0034191 , 0.00338819, 0.00362946])}
```

```
'clf learning rate': [1.0, 0.5, 0.1]
}
# Construct grid search
gs ab = GridSearchCV(estimator=pipe ab,
           param grid=adaboost param grid,
           scoring='accuracy',
           cv=3, verbose=2, return train score = True)
# Fit using grid search
gs ab.fit(X train, y train)
# Best accuracy
print('Best accuracy: %.3f' % gs_ab.best_score_)
# Best params
print('\nBest params:\n', gs_ab.best_params_)
Fitting 3 folds for each of 9 candidates, totalling 27 fits
[CV] clf__learning_rate=1.0, clf__n_estimators=30 .....
[CV] ..... clf learning rate=1.0, clf n estimators=30, total=
[CV] clf__learning_rate=1.0, clf__n_estimators=30 .....
[Parallel(n jobs=1)]: Done 1 out of 1 | elapsed: 3.8s remaining:
                                                                    0.0s
[CV] ..... clf__learning_rate=1.0, clf__n_estimators=30, total=
[CV] clf__learning_rate=1.0, clf__n_estimators=30 .....
[CV] ..... clf__learning_rate=1.0, clf__n_estimators=30, total=
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[CV] ..... clf learning rate=0.5, clf n estimators=30, total=
[CV] clf__learning_rate=0.5, clf__n_estimators=30 ......
[CV] ..... clf__learning_rate=0.5, clf__n_estimators=30, total=
```

```
[CV] clf learning rate=0.5, clf n estimators=50 .....
 [CV] ..... clf learning rate=0.5, clf n estimators=50, total= 5.9s
 [CV] clf__learning_rate=0.5, clf__n_estimators=50 .....
 [CV] ..... clf_learning_rate=0.5, clf__n_estimators=50, total=
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 [CV] .... clf learning rate=0.5, clf n estimators=50, total=
 [CV] clf__learning_rate=0.5, clf__n_estimators=70 .....
 [CV] ..... clf learning rate=0.5, clf n estimators=70, total=
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 [CV] ..... clf__learning_rate=0.5, clf__n_estimators=70, total=
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 [CV] ..... clf learning rate=0.5, clf n estimators=70, total= 8.1s
 [CV] clf learning rate=0.1, clf n estimators=30 .....
 [CV] ..... clf learning rate=0.1, clf n estimators=30, total=
 [CV] clf__learning_rate=0.1, clf__n_estimators=30 .....
 [CV] ..... clf learning rate=0.1, clf n estimators=30, total=
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 [CV] clf learning rate=0.1, clf n estimators=50 .....
 [CV] ..... clf_learning_rate=0.1, clf__n_estimators=50, total= 5.9s
 [CV] clf learning rate=0.1, clf n estimators=50 .....
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  [CV] .... clf learning rate=0.1, clf n estimators=70, total=
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 [CV] ..... clf__learning_rate=0.1, clf__n_estimators=70, total=
 [Parallel(n jobs=1)]: Done 27 out of 27 | elapsed: 2.7min finished
 Best accuracy: 0.665
 Best params:
  {'clf__learning_rate': 0.5, 'clf__n_estimators': 70}
Use your grid search object along with .cv results to get the full result overview:
 gs ab.cv results
```

## Level-up (Optional): SVM pipeline with grid search

As extra level-up work, construct a pipeline with grid search for support vector machines.

 Make sure your grid isn't too big. You'll see it takes quite a while to fit SVMs with nonlinear kernel functions!

```
# Mar This cell may take a very long time to run!
# Construct pipeline
pipe_svm = Pipeline([('pca', PCA(n_components=27)),
                     ('clf', svm.SVC(random_state=123))])
# Set grid search params
param_grid_svm = [
  {'clf_C': [0.1, 1, 10] , 'clf_kernel': ['linear']},
  {'clf_C': [1, 10], 'clf_gamma': [0.001, 0.01], 'clf_kernel': ['rbf']},
 ]
# Construct grid search
gs svm = GridSearchCV(estimator=pipe svm,
                      param_grid=param_grid_svm,
                      scoring='accuracy',
                      cv=3, verbose=2, return train score = True)
# Fit using grid search
gs_svm.fit(X_train, y_train)
# Best accuracy
print('Best accuracy: %.3f' % gs_svm.best_score_)
# Best params
print('\nBest params:\n', gs_svm.best_params_)
```

Use your grid search object along with .cv\_results to get the full result overview:

```
gs_svm.cv_results_
```

## Note

Note that this solution is only one of many options. The results in the Random Forest and AdaBoost models show that there is a lot of improvement possible by tuning the hyperparameters further, so make sure to explore this yourself!

# **Summary**

Great! You've gotten a lot of practice in using PCA in pipelines. What algorithm would you choose and why?

#### **Releases**

No releases published

### **Packages**

No packages published

#### Contributors 2



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#### Languages

Jupyter Notebook 100.0%