- Principal component analysis to select the components that account for the variation in the dataset
- Feature importances using ensembled or tree classifiers

Statistical Tests to Select the Best *k* Features Using the SelectKBest Module

The following list is a selection of statistical tests to use with **SelectKBest**. The choice depends if the dataset target variable is numerical or categorical:

- ANOVA F-value, **f_classif** (classification)
- Chi-squared stats of non-negative features, chi2 (classification)
- F-value, **f_regression** (regression)
- Mutual information for a continuous target, mutual_info_regression (regression)

Let's see an example using chi-squared test to select the best variables.

```
# import packages
from sklearn import datasets
from sklearn.feature_selection import SelectKBest
from sklearn.feature_selection import chi2
# load dataset
data = datasets.load_iris()
# separate features and target
X = data.data
y = data.target
# display first 5 rows
X[0:5,:]
# feature engineering. Let's see the best 3 features by setting k = 3
kBest_chi = SelectKBest(score_func=chi2, k=3)
fit_test = kBest_chi.fit(X, y)
```

From the test scores, the top 3 important features in the dataset are ranked from feature 3 to 4 to 1 and to 2 in order. The data scientist can choose to drop the second column and observe the effect on the model performance.

We can transform the dataset to subset only the important features.

The result drops the second column of the dataset.

Recursive Feature Elimination (RFE)

RFE is used together with a learning model to recursively select the desired number of top performing features.

Let's use RFE with **LinearRegression**.

```
# import packages
from sklearn.feature_selection import RFE
from sklearn.linear_model import LinearRegression
from sklearn import datasets
# load dataset
data = datasets.load_boston()
# separate features and target
X = data.data
y = data.target
```