# Bartosz Mikulski

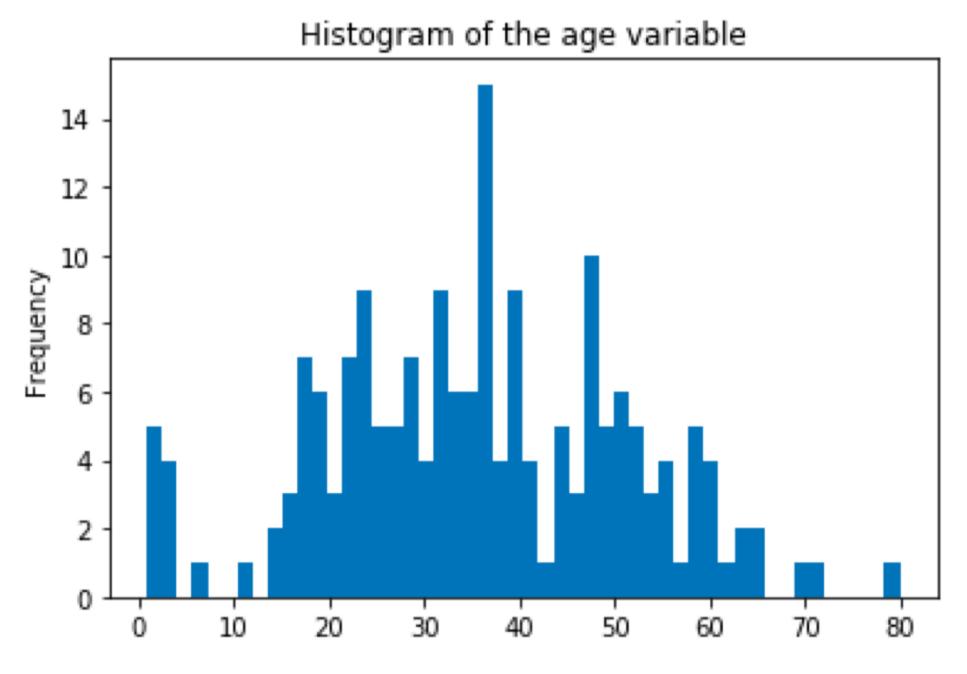
Data Craft - making data pipelines trustworthy

# **Outlier detection with Scikit Learn**

In this example, we are going to use the Titanic dataset. I remove the rows containing missing values because dealing with them is not the topic of this blog post.

First, we are going to find the outliers in the age column. To decide which method of finding outliers we should use, we must plot the histogram of the variable and look at its distribution.

```
import seaborn as sns
import pandas as pd
titanic = sns.load_dataset('titanic')
titanic = titanic.copy()
titanic = titanic.dropna()
titanic['age'].plot.hist(
bins = 50,
title = "Histogram of the age variable"
)
```



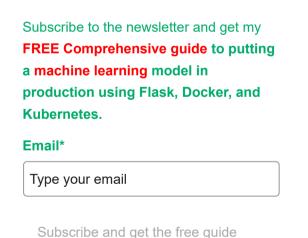
It looks a little bit like Gaussian distribution so we will use z-score. Z-score is the difference between the value and the sample mean expressed as the number of standard deviations. If the z-score is smaller than 2.5 or larger than 2.5, the value is in the 5% of smallest or largest values (2.5% of values at both ends of the distribution).

```
from scipy.stats import zscore
titanic["age_zscore"] = zscore(titanic["age"])
titanic["is_outlier"] = titanic["age_zscore"].apply(
    lambda x: x <= -2.5 or x >= 2.5
)
titanic[titanic["is_outlier"]]
```

	survived	pclass	sex	age	sibsp	parch	fare	embarked	class	who	adult_male	deck	embark_town	alive	alone	age_zscore	is_01
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We have found the one person who is older than the other Titanic passengers

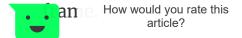




# DBSCAN — Density-Based Spatial Clustering of Applications with Noise

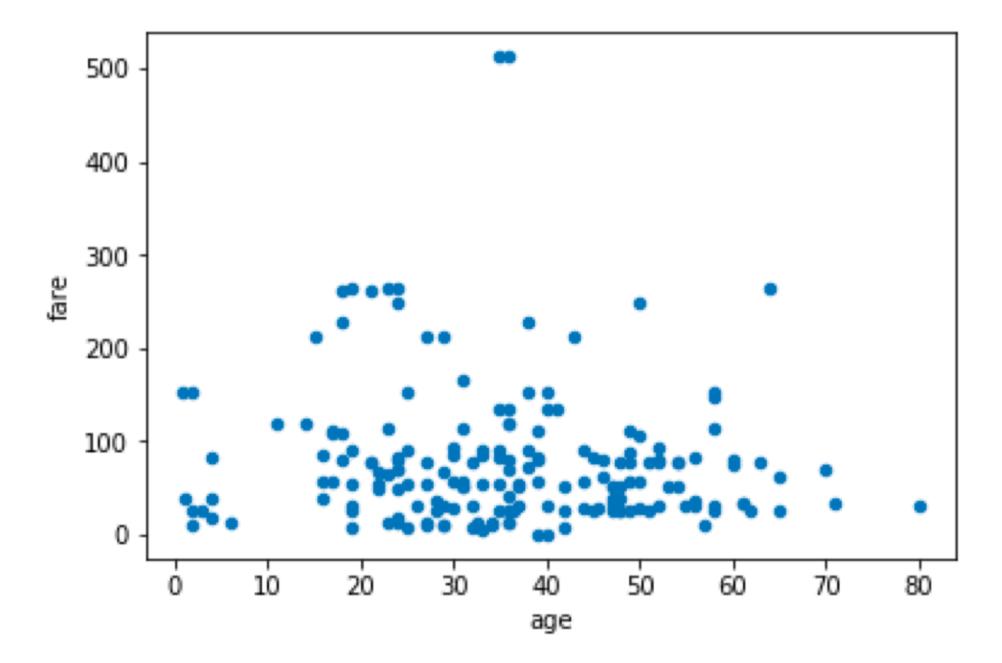
Another way to find the outliers is DBSCAN which uses clustering. In this method, we calculate the distance between points (the Euclidean distance or some other distance) and look for points which are far away from others.

As an example, we will select the age and fare from the Titanic dataset and look for the outliers in the data



First, we are going to plot the data frame and check if there are some values which may be considered to be outliers.

```
ageAndFare = titanic[["age", "fare"]]
ageAndFare.plot.scatter(x = "age", y = "fare")
```



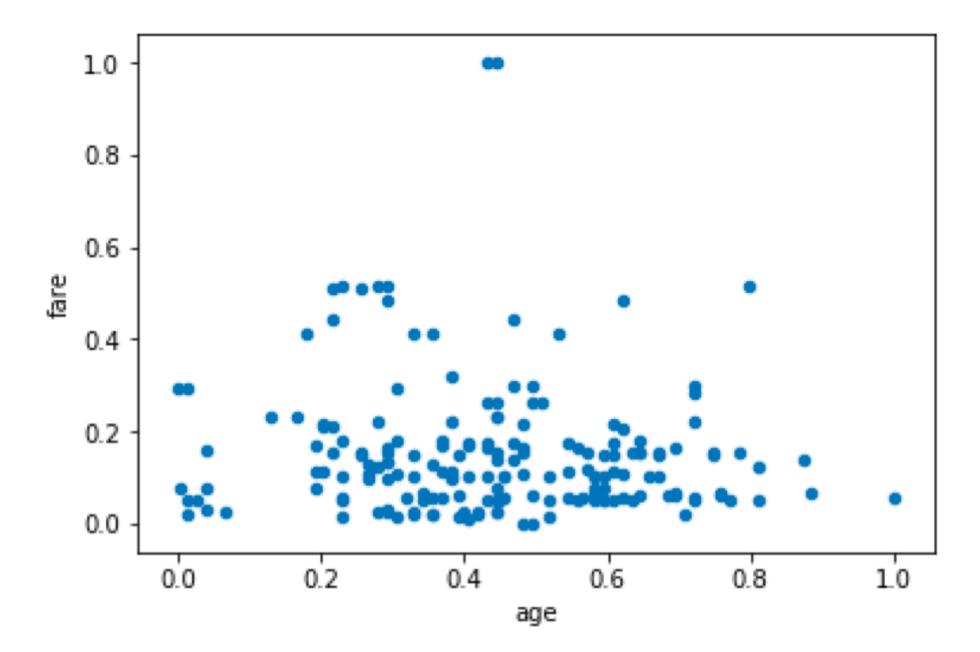
We see two points with a large value of the "fare" attribute.

We also see that both variables have different scales. Because of that, the distance would be dominated by the fare variable. We don't want that so we must normalize variables.

```
from sklearn.preprocessing import MinMaxScaler
scaler = MinMaxScaler()
ageAndFare = scaler.fit_transform(ageAndFare)
ageAndFare = pd.DataFrame(ageAndFare, columns = ["age", "fare"])
ageAndFare.plot.scatter(x = "age", y = "fare")
```



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DBSCAN is going to assign points to clusters and return the labels of clusters. If it cannot assign the value to any cluster (because it is an outlier), it returns -1. In this example, it may also return a cluster which contains only two points, but for the sake of demonstration I want -1 so I set the minimal number of samples in a cluster to 3.

```
from sklearn.cluster import DBSCAN

outlier_detection = DBSCAN(
    eps = 0.5,
    metric="euclidean",
    min_samples = 3,
    n_jobs = -1)

clusters = outlier_detection.fit_predict(ageAndFare)

clusters
```



How would you rate this article?

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### Cluster identifiers

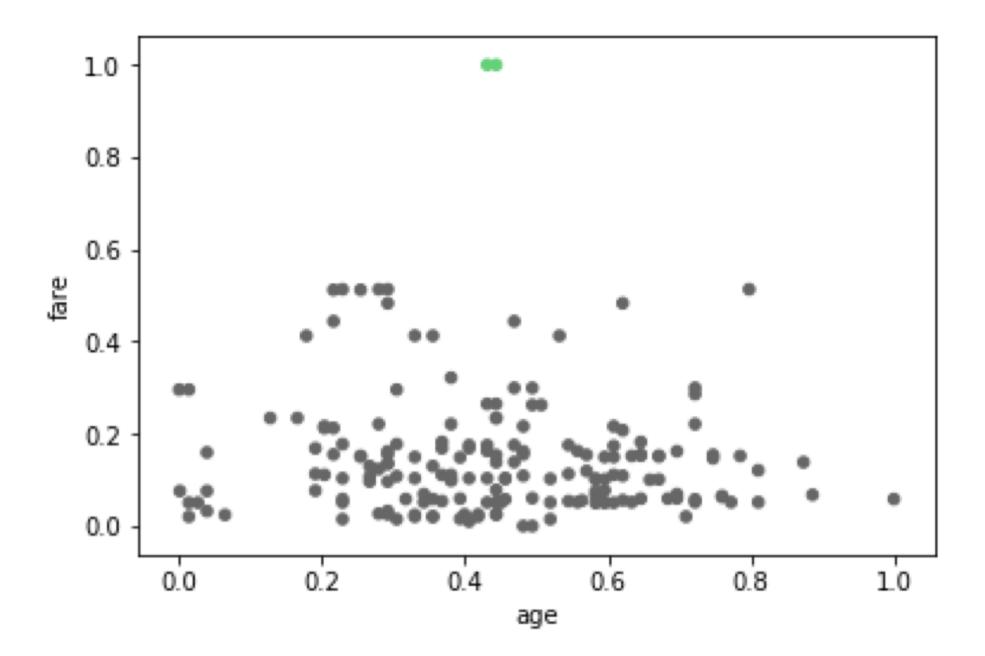
As expected we have found two outliers. Now we should verify whether the points marked as outliers are the expected ones. We can either:

- merge the output to the data frame and print the output, or
- we can plot the dataset and use the cluster identifier to color the points.

```
from matplotlib import cm
cmap = cm.get_cmap('Accent')
ageAndFare.plot.scatter(
    x = "age",
    y = "fare",
    c = clusters,
    cmap = cmap,
    colorbar = False
)
```



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# Clusters returned by DBSCAN

# Done!;)

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# Bartosz Mikulski (Follow)

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