

## 20230412\_Wirtschaftsinformatik\_MV2

April 12, 2023

```
[40]: import numpy as np
```

```
[41]: # mittelwert
```

```
[42]: geschwindigkeit = [99,86,87,88,111,86,103,87,94,78,77,85,86]
```

```
[43]: # (99+86+87+88+111+86+103+87+94+78+77+85+86) / 13 = 89.77
```

```
[44]: mittelwert = np.mean(geschwindigkeit)
```

```
[45]: print(mittelwert)
```

```
89.76923076923077
```

```
[46]: # standard abweichung
```

```
[47]: std = np.std(geschwindigkeit)
```

```
[48]: print(std)
```

```
9.258292301032677
```

```
[69]: print('Die Standard Abweichung STD ist ', std, '.')
```

```
Die Standard Abweichung STD ist  9.258292301032677 .
```

```
[49]: # percentile
```

```
[50]: # percentile 60 bedeutet der Wert, welche 60% der Daten "links" lässt
```

```
[51]: percentile_60 = np.percentile(geschwindigkeit, 60)
```

```
[52]: print(percentile_60)
```

```
87.2
```

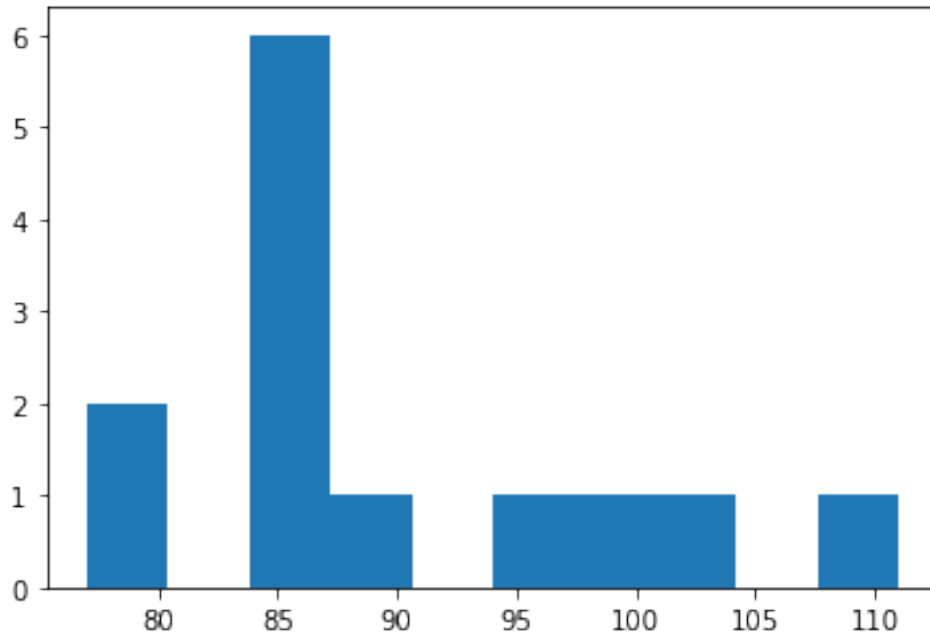
```
[53]: # 60% der Daten sind kleiner als 87.2
```

```
[54]: # daten distribution (Verteilung)
```

```
[55]: # Verteilungen werden idR mit Hilfe von einem Histogramm dargestellt
```

```
[56]: import matplotlib.pyplot as plt
```

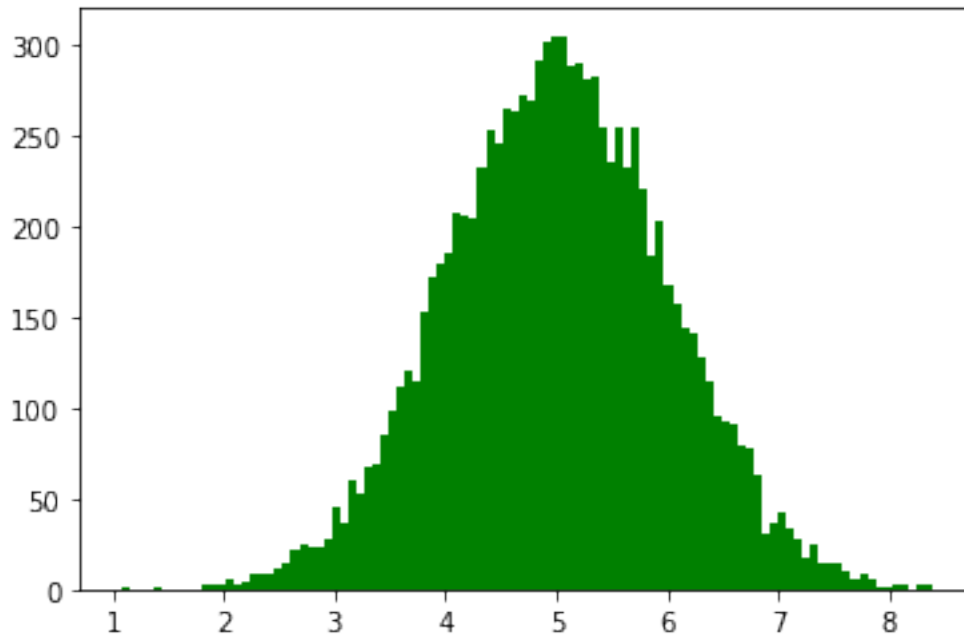
```
[71]: plt.hist(geschwindigkeit)  
plt.show()
```



```
[78]: # beispiel
```

```
x = np.random.normal(5, 1, 10000)
```

```
plt.hist(x, bins = 100, color = 'g')  
plt.show()
```



```
[79]: import pandas as pd
url = 'https://raw.githubusercontent.com/ProfH4/Lecture_Notes/main/
↳2023%20Sommersemester/410636%20Wirtschaftsinformatik/20230412_data.csv'
df = pd.read_csv(url,index_col=0,parse_dates=[0])
```

```
[81]: print(df.head())
```

	Model	Volume	Weight	CO2
Car				
Toyoty	Aygo	1000	790	99
Mitsubishi	Space Star	1200	1160	95
Skoda	Citigo	1000	929	95
Fiat	500	900	865	90
Mini	Cooper	1500	1140	105

```
[82]: # normierung der Daten

# z = (x-mittelwert)/standard_abweichung

import pandas as pd
```

```
[83]: !pip install sklearn
```

```
Requirement already satisfied: sklearn in
/Users/h4/anaconda3/lib/python3.9/site-packages (0.0)
Requirement already satisfied: scikit-learn in
```

```
/Users/h4/anaconda3/lib/python3.9/site-packages (from sklearn) (1.0.2)
Requirement already satisfied: numpy>=1.14.6 in
/Users/h4/anaconda3/lib/python3.9/site-packages (from scikit-learn->sklearn)
(1.23.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in
/Users/h4/anaconda3/lib/python3.9/site-packages (from scikit-learn->sklearn)
(2.2.0)
Requirement already satisfied: scipy>=1.1.0 in
/Users/h4/anaconda3/lib/python3.9/site-packages (from scikit-learn->sklearn)
(1.9.1)
Requirement already satisfied: joblib>=0.11 in
/Users/h4/anaconda3/lib/python3.9/site-packages (from scikit-learn->sklearn)
(1.1.0)
```

```
[84]: import sklearn
```

```
[86]: from sklearn.preprocessing import StandardScaler
scale = StandardScaler()
df_nummerisch = df[['Weight', 'Volume', 'CO2']]
scaled_df = scale.fit_transform(df_nummerisch)

print(scaled_df)
```

```
[[-2.10389253 -1.59336644 -0.41192538]
 [-0.55407235 -1.07190106 -0.95612037]
 [-1.52166278 -1.59336644 -0.95612037]
 [-1.78973979 -1.85409913 -1.63636411]
 [-0.63784641 -0.28970299  0.40436711]
 [-1.52166278 -1.59336644  0.40436711]
 [-0.76769621 -0.55043568 -1.63636411]
 [ 0.3046118  -0.28970299 -1.36426661]
 [-0.7551301  -0.28970299 -0.54797412]
 [-0.59595938 -0.0289703  -0.41192538]
 [-1.30803892 -1.33263375 -0.41192538]
 [-1.26615189 -0.81116837 -0.13982788]
 [-0.7551301  -1.59336644 -0.41192538]
 [-0.16871166 -0.0289703  -1.09216911]
 [ 0.14125238 -0.0289703  -0.68402287]
 [ 0.15800719 -0.0289703  -0.68402287]
 [ 0.3046118  -0.0289703  -0.41192538]
 [-0.05142797  1.53542584  0.26831836]
 [-0.72580918 -0.0289703  0.26831836]
 [ 0.14962979  1.01396046  0.40436711]
 [ 1.2219378  -0.0289703  -1.09216911]
 [ 0.5685001   1.01396046 -0.41192538]
 [ 0.3046118   1.27469315 -0.41192538]
 [ 0.51404696 -0.0289703  -0.41192538]
 [ 0.51404696  1.01396046 -0.41192538]
```

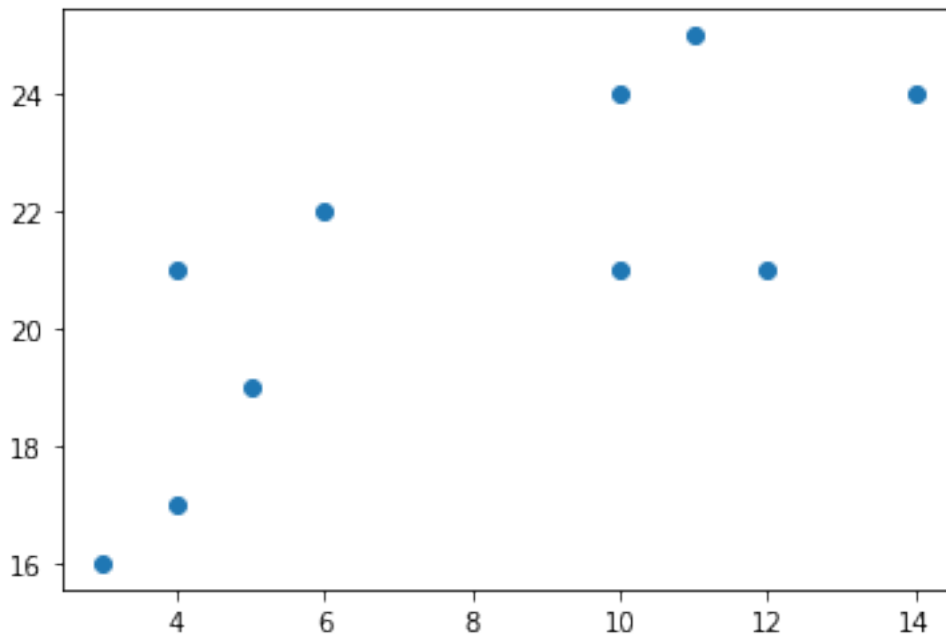
```
[ 0.72348212 -0.28970299 -0.00377913]
[ 0.8281997   1.01396046  0.26831836]
[ 1.81254495  1.01396046  1.62880584]
[ 0.96642691 -0.0289703   0.9485621 ]
[ 1.72877089  1.01396046  1.62880584]
[ 1.30990057  1.27469315  1.76485459]
[ 1.90050772  1.01396046  2.03695208]
[-0.23991961 -0.0289703   0.26831836]
[ 0.40932938 -0.0289703   0.81251335]
[ 0.47215993 -0.0289703   0.9485621 ]
[ 0.4302729   2.31762392  2.44509833]]
```

```
[87]: # Gruppen finden in Daten - Clustering
```

```
[89]: # K-Means Clustering (K sind die Anzahl Gruppen oder Zentren)
```

```
[91]: Gehalt = [4, 5, 10, 4, 3, 11, 14, 6, 10, 12]
Ausgaben_Webseite = [21, 19, 24, 17, 16, 25, 24, 22, 21, 21]

plt.scatter(Gehalt, Ausgaben_Webseite)
plt.show()
```

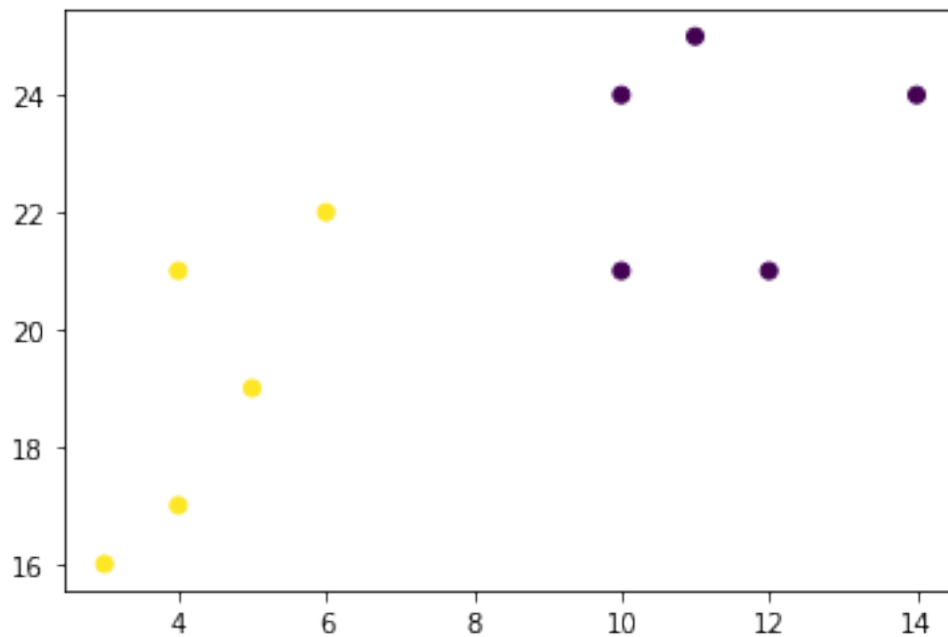


```
[92]: data = list(zip(Gehalt, Ausgaben_Webseite))
```

```
[93]: data
```

```
[93]: [(4, 21),  
      (5, 19),  
      (10, 24),  
      (4, 17),  
      (3, 16),  
      (11, 25),  
      (14, 24),  
      (6, 22),  
      (10, 21),  
      (12, 21)]
```

```
[94]: from sklearn.cluster import KMeans  
  
kmeans = KMeans(n_clusters = 2)  
kmeans.fit(data)  
  
plt.scatter(Gehalt, Ausgaben_Webseite, c = kmeans.labels_)  
plt.show()
```



```
[95]: # Übung für die Prüfungsvorbereitung:  
      # KMeans Cluster umsetzen für dataset Autos (siehe oben)  
      # für Weight und Volume.
```

```
[ ]:
```