## Untitled

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## 1 Blatt 08

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- 1.2 Aufgabe 17
- 1.2.1 c)

3

21.1

20.0

Nun gilt es, den Informationsgewinn auch für die anderen Attribute anzeigen zu lassen.

```
[1]: # Libraries we use for data analysis and visualisation
     import matplotlib.pyplot as plt
     import seaborn as sns
     import pandas as pd
     import numpy as np
     # libraries used for the notebook
     from IPython.display import SVG
     from graphviz import Source
     from IPython.display import display
     from ipywidgets import interactive
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.datasets import make_moons
[2]: df = pd.read_csv("data.csv", sep = ";") # read data from csv file
[3]: df.head()
[3]:
                                                                Fußball
        Temperatur
                    Wettervorhersage
                                      Luftfeuchtigkeit
                                                          Wind
              29.4
     0
                                   2
                                                     85 False
                                                                  False
     1
              26.7
                                   2
                                                     90
                                                         True
                                                                  False
     2
              28.3
                                                     78 False
                                                                   True
                                   1
```

```
[4]: # sono molto contento
def entropy(p):
    m, counts = np.unique(p, return_counts = True) # possible values
```

0

0

96 False

80 False

True

True

```
H = 0

for i in range(len(m)):
    H += counts[i]/len(p) * np.log2(counts[i]/len(p))

return - H
```

```
[5]: entropy(df["Fußball"]) # to test result of a)
```

[5]: 0.9402859586706309

```
def info_gained(y, x):
    mask = x == True
    n = mask.sum()
    n_ges = len(y)
    n_spec = n_ges - n

H_y = entropy(y)

H_yx = n/n_ges * entropy(y[mask]) + n_spec/n_ges * entropy(y[-mask])

IG_yx = H_y - H_yx

return IG_yx
```

```
[7]: def find_max_cut(y, x):
    states = np.unique(x)

    IG_states = np.array([info_gained(y, x > i) for i in states])

    plt.figure()
    plt.plot(states[:], IG_states[:], ".")

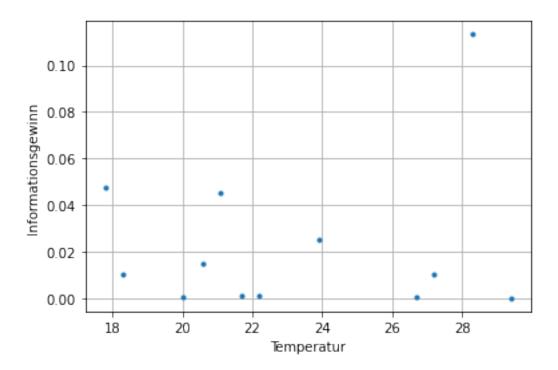
    plt.xlabel(x.name)
    plt.ylabel("Informationsgewinn")

    plt.grid()

    return IG_states.max()
```

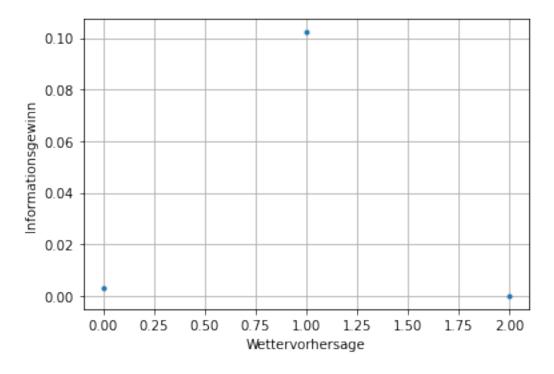
```
[8]: IG_max_T = find_max_cut(df["Fußball"], df["Temperatur"])
print("Maximum Information gained = ", IG_max_T)
```

Maximum Information gained = 0.11340086418110318



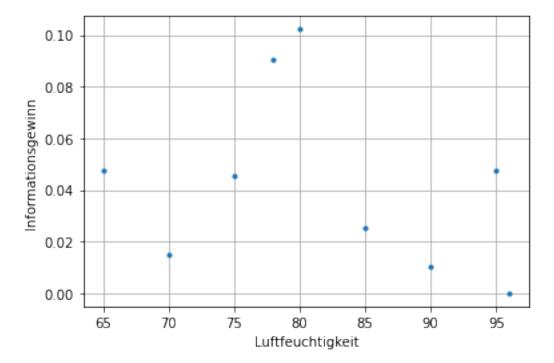
```
[9]: IG_max_WV = find_max_cut(df["Fußball"], df["Wettervorhersage"])
print("Maximum Information gained = ", IG_max_WV)
```

Maximum Information gained = 0.10224356360985054



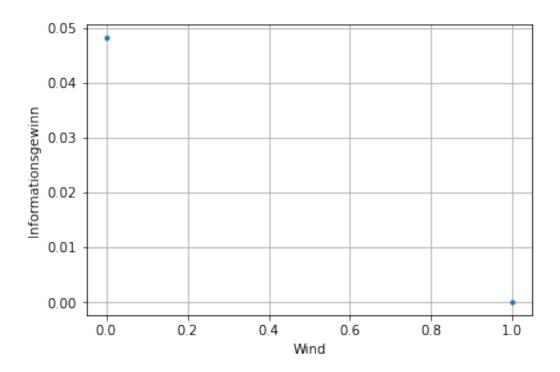
```
[10]: IG_max_L = find_max_cut(df["Fußball"], df["Luftfeuchtigkeit"])
print("Maximum Information gained = ", IG_max_L)
```

Maximum Information gained = 0.10224356360985054



```
[12]: IG_max_W= find_max_cut(df["Fußball"], df["Wind"])
print("Maximum Information gained = ", IG_max_W)
```

Maximum Information gained = 0.04812703040826927



## 1.2.2 d)

Es zeigt sich also, dass durch einen cut bei der Temperatur die meiste Information gewonnen werden kann, gleich darauf folgen auf derselben Höhe die Luftfeuchtigkeit und Wettervorhersage.

[]: