## logistische Regression

```
In [1]: # Load the Pandas libraries, matplotlib and numpy
import pandas as pd
import matplotlib.pyplot as plt
import numpy as np

# Read data
data = pd.read_csv("Umfragedaten_v1_an.csv")

# Preview the first 5 lines of the loaded data
data.head()
```

Out[1]:

	ID	GESCHL	GEBJAHR	BERUFSTAETIG	ARBEITSSTD	ARZTBES	RAUCH	GRC
0	1359	WEIBLICH	1967.0	NICHT ERWERBSTAETIG	NaN	1.0	JA	162.
1	2455	WEIBLICH	1964.0	HAUPTBERUFL.HALBTAGS	30.0	1.0	NEIN	165.
2	200	MAENNLICH	1980.0	NICHT ERWERBSTAETIG	NaN	3.0	NEIN	166.
3	1280	MAENNLICH	1968.0	HAUPTBERUFL.GANZTAGS	50.0	0.0	NEIN	180.
4	2384	WEIBLICH	1995.0	NICHT ERWERBSTAETIG	NaN	1.0	NEIN	161.

```
In [2]: #load libraries for regression
from sklearn.linear_model import LogisticRegression
from sklearn import metrics
import seaborn as sn
```

Data often have NaN or other bad values, which means data need to be cleaned first.

```
In [4]: #need to dummy code first
x = data['NETTO'].values.reshape(-1,1)
dummies = pd.get_dummies(data, columns=['RAUCH'])
y = dummies['RAUCH_JA'].values.reshape(-1,1)
dummies.head()
```

Out[4]:

	ID	GESCHL	GEBJAHR	BERUFSTAETIG	ARBEITSSTD	ARZTBES	GRO	GEW	;
C	1359	WEIBLICH	1967.0	NICHT ERWERBSTAETIG	NaN	1.0	162.0	79.0	
1	2455	WEIBLICH	1964.0	HAUPTBERUFL.HALBTAGS	30.0	1.0	165.0	59.0	
2	200	MAENNLICH	1980.0	NICHT ERWERBSTAETIG	NaN	3.0	166.0	86.0	[i
3	1280	MAENNLICH	1968.0	HAUPTBERUFL.GANZTAGS	50.0	0.0	180.0	95.0	l
4	2384	WEIBLICH	1995.0	NICHT ERWERBSTAETIG	NaN	1.0	161.0	46.0	Ŀ

0.0

10000

20000

```
In [5]: #not sure why pandas creates 2D arrays, but need 1D otherwise logistic regression
        x=x[:,0]
        y=y[:,0]
        #need to clean data, there seem to be some NaN in x
        print(np.any(np.isnan(x)))
        print(np.any(np.isnan(y)))
        True
        False
In [6]: | #print(np.where(np.isnan(x) == True))
        print(x.ndim)
        print(len(x))
        index = np.where(~np.isnan(x)) #index of good data entries
        #print(len(index))
        xnew = x[index] #only keep good entries
        ynew = y[index]
        print(xnew)
        1
        3471
                  780. 4000. ..., 1500. 1300. 1100.]
          475.
In [7]: #show a scatter plot
        plt.plot(xnew,ynew,'o')
        plt.show()
        #1=raucher, 0=nichtraucher
         1.0
         0.8
         0.6
         0.4
         0.2
```

Now we train the model by splitting it into a training set and then the test set is used to validate it.

40000

50000

60000

30000

```
In [8]: #train the model
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(xnew.reshape(-1,1), ynew, test
```

```
In [9]: # logistic regression using the default parameters)
logreg = LogisticRegression()

# fit the model with data
logreg.fit(X_train,y_train)

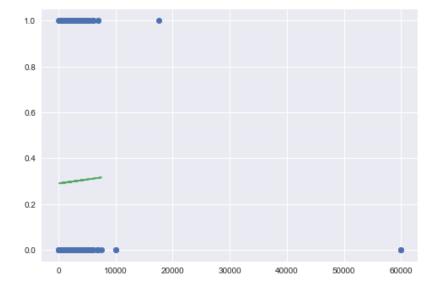
#
y_pred=logreg.predict(X_test)
y_pred2 = logreg.predict_proba(X_test)[:,1]

print(logreg.coef_, logreg.intercept_)

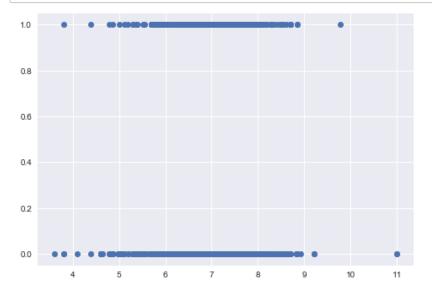
(array([[ 1.63918095e-05]]), array([-0.89435454]))
In [10]: #confusion matrix
```

In [11]: #plotting
 #show a scatter plot
 plt.plot(xnew,ynew,'o')
 plt.plot(X\_test,y\_pred2)
 plt.show()

#this also shows a bad fit.



```
In [12]: #let's try changing the model by taking the (natural) log of the income
    plt.plot(np.log(xnew),ynew,'o')
    plt.show()
    #one can see that there is actually not a very clear separation between non-smoker
    #We fit it anyway.
```



```
In [13]: xnew2 = np.log(xnew)
X_train, X_test, y_train, y_test = train_test_split(xnew2.reshape(-1,1), ynew, tes

# logistic regression using the default parameters)
logreg = LogisticRegression(C=9)

# fit the model with data
logreg.fit(X_train,y_train)

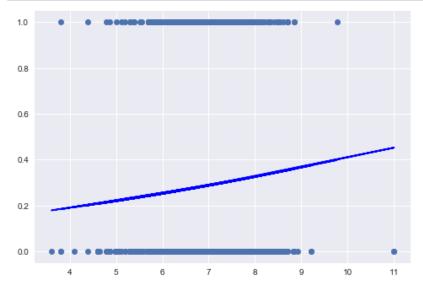
#
y_pred=logreg.predict(X_test)
y_pred2 = logreg.predict_proba(X_test)[:,1]

print(logreg.coef_, logreg.intercept_)
```

(array([[ 0.1815504]]), array([-2.17652385]))

```
In [14]: #create the model manually to verify the prediction of python works (yes, it's the
    pi = np.exp(-2.17+.18*xnew2) / (1+np.exp(-2.17+.18*xnew2))

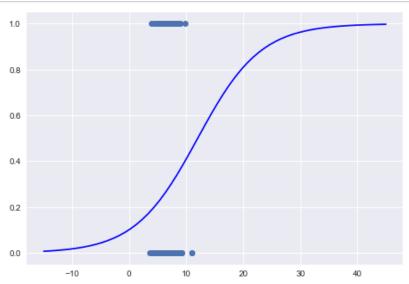
#plotting
    #show a scatter plot
    plt.plot(xnew2,ynew,'o')
    plt.plot(X_test,y_pred2)
    plt.plot(xnew2,pi,color='b')
    plt.show()
```



```
In [15]: #plot a larger x-axis to see the full model
    xlarge = np.linspace(-15,45,60).reshape((-1,1))
    ylarge = logreg.predict_proba(xlarge)[:,1]

    plt.plot(xnew2,ynew,'o')
    plt.plot(xlarge,ylarge,color='b')
    plt.show()

#yes, the model is a logit model, but the data are not clearly separated, which me
```



```
In [16]: #wahrscheinlichkeit bei 2000 gehalt [%]
print(logreg.predict_proba(np.log(2000))[:,1])*100.
```

[ 31.07582063]

In [ ]: