## linear regreesion

## March 20, 2024

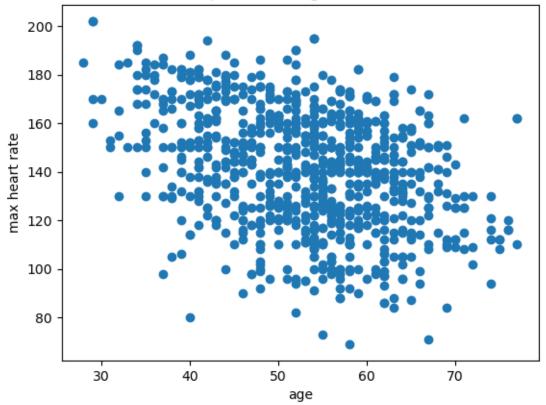
```
[8]: import numpy as np
      import pandas as pd
      import warnings
      warnings.filterwarnings('ignore')
 [9]: #importing data
      data=pd.read_csv("C:\\Users\\user\\Downloads\\Dataset Heart Disease.csv")
      data
 [9]:
           age
                 chest pain type
                                   resting bps
                                                 cholesterol
                                                               max heart rate
      0
            40
                                2
                                            140
                                                          289
                                                                           172
      1
            49
                                3
                                                          180
                                                                           156
                                            160
      2
                                2
            37
                                                          283
                                                                            98
                                            130
      3
            48
                                                                           108
                                4
                                            138
                                                          214
      4
            54
                                3
                                            150
                                                          195
                                                                           122
      893
            42
                                3
                                            130
                                                          180
                                                                           150
      894
                                                                           170
            45
                                2
                                            128
                                                          308
      895
            57
                                1
                                            165
                                                          289
                                                                           124
      896
                                3
                                                                           131
            64
                                            125
                                                          309
      897
            41
                                3
                                            112
                                                          250
                                                                           179
      [898 rows x 5 columns]
     calling arrays
[10]: x=np.array(data["age"]).reshape(-1,1)
[11]: y=np.array(data["max heart rate"])
     checking for missing data
[12]: data.isna().sum()
                          0
[12]: age
      chest pain type
                          0
      resting bps
                          0
      cholesterol
                          0
      max heart rate
```

dtype: int64

graph

```
[13]: import matplotlib.pyplot as plt
   plt.scatter(x,y)
   plt.xlabel("age")
   plt.ylabel("max heart rate")
   plt.title("Relationship between age and heart disease")
   plt.show()
```

## Relationship between age and heart disease



splitting data

#standardizing data from sklearn.preprocessing import StandardScaler scaler = StandardScaler()  $x_{train}$  scaled = scaler.fit\_transform( $x_{train}$ )  $x_{test}$  scaled = scaler.transform( $x_{test}$ )

```
[15]: #building the model
      from sklearn.linear_model import LinearRegression
      model = LinearRegression()
      model
[15]: LinearRegression()
[16]: #model fitting
      model.fit(x_train,y_train)
[16]: LinearRegression()
[17]: #making predictions
      y_pred=model.predict(x_test)
[18]: model.coef_
[18]: array([-0.96626539])
[19]: #finding intercept(regression line intercepts with the y axis)
      model.intercept_
[19]: 193.4557211223224
[20]: #model evaluation
      #model accuracy on train values
      model.score(x_train,y_train)
[20]: 0.1411214865128494
[21]: #model accuracy on the test values
      model.score(x_test,y_test)
[21]: 0.09956280781317806
[22]: from sklearn.metrics import mean_squared_error,r2_score,mean_absolute_error
      print(mean_squared_error(y_test,y_pred))
     573.7417929335393
[23]: print(mean_absolute_error(y_test,y_pred))
     19.89992588788073
[24]: print(r2_score(y_test,y_pred))
     0.09956280781317806
     MODEL OPTIMIZATION
```

```
[25]: from sklearn.model_selection import GridSearchCV
      model = LinearRegression()
[26]: #performing GridSearchCV to find best hyperparameters
      param_grid = {'fit_intercept':[True,False],'positive':[True,False],'positive':
       →[True,False]}
      param_grid
[26]: {'fit_intercept': [True, False], 'positive': [True, False]}
[27]: #fitting data
      data= GridSearchCV(model,param_grid, cv=5)
      data.fit(x_train,y_train)
[27]: GridSearchCV(cv=5, estimator=LinearRegression(),
                   param_grid={'fit_intercept': [True, False],
                               'positive': [True, False]})
[28]: #getting the best parameter from the GridSearch
      best_params = data.best_params_['fit_intercept']
      best_params
[28]: True
[29]: best_params = data.best_params_['positive']
      best_params
[29]: False
[30]: #training the model using the best parameters
      model=LinearRegression(fit_intercept=best_params)
      model.fit(x train,y train)
[30]: LinearRegression(fit_intercept=False)
[31]: #making predictions
      y_pred = model.predict(x_test)
[32]: #evaluating model performance
      print(mean_absolute_error(y_test,y_pred))
     32.93329029906092
[33]: print(r2_score(y_test,y_pred))
     -1.5683088109567058
[34]: #finding accuracy on train values
      model.score(x_train,y_train)
```

[34]: -1.813959890359337

[35]: #finding accuracy on test values model.score(x\_test,y\_test)

[35]: -1.5683088109567058