```
In [1]:
```

```
#Importing required libraries
import pandas as pd
import numpy as np
```

### In [2]:

```
#Importing the dataset
data_preprocessed = pd.read_csv("Absenteeism_preprocessed.csv")
data_preprocessed.head()
```

# Out[2]:

	Reason Type 1	Reason Type 2		Reason Type 4	Month	Weekdays	Transportation Expense	Distance to Work	Age	Daily Work Load Average	Body Mass Index	Education	Children	Pets	F
0	0	0	0	1	7	1	289	36	33	239.554	30	0	2	1	
1	0	0	0	0	7	1	118	13	50	239.554	31	0	1	0	
2	0	0	0	1	7	2	179	51	38	239.554	31	0	0	0	
3	1	0	0	0	7	3	279	5	39	239.554	24	0	2	0	
4	0	0	0	1	7	3	289	36	33	239.554	30	0	2	1	
4															F

### In [3]:

```
#Creating Targets
data_preprocessed['Absenteeism Time in Hours'].median()
```

# Out[3]:

3.0

# In [4]:

```
#creating targets
targets = np.where(data_preprocessed['Absenteeism Time in Hours']>3,1,0)
data_preprocessed['Excessive Absenteeism'] = targets
data_with_targets = data_preprocessed.drop(['Absenteeism Time in Hours','Distance to Work','Daily
Work Load Average','Weekdays'],axis=1)
data_with_targets.head()
```

### Out[4]:

	Reason Type 1	Reason Type 2	Reason Type 3	Reason Type 4	Month	Transportation Expense	Age	Body Mass Index	Education	Children	Pets	Excessive Absenteeism
0	0	0	0	1	7	289	33	30	0	2	1	1
1	0	0	0	0	7	118	50	31	0	1	0	0
2	0	0	0	1	7	179	38	31	0	0	0	0
3	1	0	0	0	7	279	39	24	0	2	0	1
4	0	0	0	1	7	289	33	30	0	2	1	0

# targets.sum()/targets.shape()

### In [5]:

```
targets
```

# Out[5]:

```
0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 1,
       0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0,
       0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 0,
       1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 1,
       0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1,
       1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1,
       0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0,
       0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0,
       0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1,
       1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1,
                                                    1,
                                                        1, 0, 0, 0, 0, 0,
       1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0,
       1, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0, 1,
       1, 1, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1,
       1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1,
       0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1,
       1, 1, 1, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 1,
       1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1,
       1, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0,
       1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1,
                                                     1,
                                                        0, 0, 0, 1, 1, 1,
       0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0,
       0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1,
       0, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 0,
       1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1,
       1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0])
In [6]:
targets.sum()
Out[6]:
319
In [7]:
targets.shape[0]
Out[7]:
700
In [8]:
targets.sum()/targets.shape[0]
Out[81:
0.45571428571428574
In [9]:
unscaled inputs = data with targets.iloc[:,:-1]
In [10]:
#Standardizing the data
from sklearn.preprocessing import StandardScaler
absenteeism scaler = StandardScaler()
absenteeism scaler.fit(unscaled inputs)
scaled inputs = absenteeism scaler.transform(unscaled inputs)
scaled inputs
Out[10]:
array([[-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003, 0.88046927, 0.26848661],
```

0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1,

```
[-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003, -0.01928035, -0.58968976],
[-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003, -0.91902997, -0.58968976],
...,
[1.73205081, -0.09298136, -0.31448545, ..., 2.23224237, -0.91902997, -0.58968976],
[-0.57735027, -0.09298136, -0.31448545, ..., 2.23224237, -0.91902997, -0.58968976],
[-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003, -0.01928035, 0.26848661]])
```

### In [11]:

```
from sklearn.model_selection import train_test_split
train_test_split(scaled_inputs, targets)
```

### Out[11]:

```
[array([[ 1.73205081, -0.09298136, -0.31448545, ..., -0.44798003,
         -0.91902997, -0.58968976],
        [-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
       [ 1.73205081, -0.09298136, -0.31448545, ..., 2.23224237,
        -0.91902997, -0.58968976],
       [1.73205081, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.01928035, 1.12666297],
        [-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
       [-0.57735027, -0.09298136, -0.31448545, ..., 2.23224237,
        -0.91902997, -0.58968976]]),
array([[-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
        [-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
         -0.01928035, 1.12666297],
       [1.73205081, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
        . . . ,
       [-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
        [ 1.73205081, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.91902997, -0.58968976],
       [-0.57735027, -0.09298136, -0.31448545, ..., -0.44798003,
        -0.01928035, 1.12666297]]),
array([1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1,
        1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0,
        1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1,
       0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0,
       0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0,
       0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0,
       1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0,
       0, 0, 1, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 0,
       1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1,
       1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 0,
       0, 1, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0,
       1, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1,
       1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0,
       0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 1,
       1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1,
       0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1,
       0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0,
       1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1,
       0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1,
       1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0,
       0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 1,\ 1,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,
       0, 0, 1, 1, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 1, 0,
       1, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0]),
array([0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0,
        0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1,
       0,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 1,\ 1,\ 1,\ 0,\ 1,\ 0,
       0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1,
       1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 1,
       1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 1, 0, 0, 0,
        0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1,
```

```
0, 0, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0])]
In [12]:
#splitting the data into train and test
x train,x test,y train,y test = train test split(scaled inputs,targets,train size=0.8,random state=
In [13]:
print(x_train.shape,y_train.shape)
(560, 11) (560,)
In [22]:
#performing logistic regression
from sklearn.linear_model import LogisticRegression
reg = LogisticRegression()
reg.fit(x_train,y_train)
reg.score(x_train,y_train)
Out[22]:
0.7839285714285714
In [15]:
#finding the intercept
reg.intercept
Out[15]:
array([-0.22375632])
In [16]:
#finding the coefficient
reg.coef
Out[16]:
array([[ 2.0695214 , 0.33581057, 1.56203942, 1.31277993, 0.18469755, 0.69126151, -0.19828303, 0.32435951, -0.12665599, 0.37017705,
         -0.32535864]])
In [17]:
feature names = unscaled inputs.columns.values
In [18]:
summary table = pd.DataFrame(columns=['Features'], data=feature names)
summary table
Out[18]:
              Features
 0
          Reason Type 1
          Reason Type 2
 1
 2
          Reason Type 3
 3
          Reason Type 4
 4
                Month
```

5 Transportation Expense

```
6 Features
7 Body Mass Index
8 Education
9 Children
10 Pets
```

### In [19]:

```
summary_table['Cofficients'] = np.transpose(reg.coef_)
summary_table
```

# Out[19]:

#### Features Cofficients 0 Reason Type 1 2.069521 1 Reason Type 2 0.335811 2 Reason Type 3 1.562039 3 Reason Type 4 1.312780 4 Month 0.184698 0.691262 5 Transportation Expense -0.198283 6 Body Mass Index 7 0.324360 8 Education -0.126656 9 Children 0.370177 10 Pets -0.325359

### In [20]:

```
summary_table.index = summary_table.index+1
summary_table
```

# Out[20]:

#### Features Cofficients Reason Type 1 2.069521 2 Reason Type 2 0.335811 3 Reason Type 3 1.562039 4 Reason Type 4 1.312780 5 0.184698 Month Transportation Expense 0.691262 7 Age -0.198283 8 **Body Mass Index** 0.324360 9 Education -0.126656 10 Children 0.370177 11 Pets -0.325359

### In [21]:

```
summary_table.loc[0] = ['Intercept',reg.intercept_[0]]
summary_table = summary_table.sort_index()
summary_table
```

# Out[21]:

### Features Cofficients

```
0
                 Fretetroreps
                          Coff2@27f16
 1
                             2.069521
            Reason Type 1
 2
            Reason Type 2
                             0.335811
            Reason Type 3
                             1.562039
 3
 4
            Reason Type 4
                             1.312780
                             0.184698
 5
                   Month
 6
   Transportation Expense
                             0.691262
 7
                     Age
                            -0.198283
 8
          Body Mass Index
                             0.324360
 9
                Education
                            -0.126656
10
                  Children
                             0.370177
11
                     Pets
                            -0.325359
In [23]:
reg.score(x_test,y_test)
```

```
reg.score(x_test,y_test)
```

#### Out[23]:

0.7357142857142858

### In [30]:

```
predicted_proba = reg.predict_proba(x_test)[:,:-1]
predicted_proba
```

### Out[30]:

```
array([[0.70804629],
        [0.57178351],
        [0.39895299],
        [0.78714111],
        [0.06691516],
        [0.31109284],
       [0.28582378],
        [0.08093986],
        [0.79926793],
        [0.7496687],
        [0.46772305],
       [0.18449682],
       [0.04086632],
        [0.75596678],
        [0.23709315],
        [0.54140798],
        [0.53354387],
       [0.52039975],
        [0.40507469],
        [0.02759587],
        [0.70148836],
        [0.78714111],
       [0.40545618],
        [0.40545618],
        [0.17391667],
        [0.75362875],
        [0.48817233],
        [0.88053255],
       [0.13059235],
        [0.78714111],
        [0.62247425],
        [0.30437845],
        [0.28895164],
       [0.44157966],
        [0.78714111],
        [0.43705325],
        [0.79067389],
```

[0.18986981], [0.58710261], [0.36998588],

```
[0.79584274],
[0.57848411],
[0.77268351],
[0.89513347],
[0.14276767],
[0.34529847],
[0.29370238],
[0.70804629],
[0.78792994],
[0.80427711],
[0.43450126],
[0.92225898],
[0.31109284],
[0.73029003],
[0.14019617],
[0.54578776],
[0.06949094],
[0.78211875],
[0.91468256],
[0.91047636],
[0.29440565],
[0.3225053],
[0.71278024],
[0.13921991],
[0.81010673],
[0.72936044],
[0.98595531],
[0.79067389],
[0.18707135],
[0.71499343],
[0.79067389],
[0.93124347],
[0.07140162],
[0.52206954],
[0.36134577],
[0.78714111],
[0.14962662],
[0.26408704],
[0.22931828],
[0.32006017],
[0.76328838],
[0.98845557],
[0.78211875],
[0.22297761],
[0.55457206],
[0.90100263],
[0.26777377],
[0.40545618],
[0.94848341],
[0.30437845],
[0.85602571],
[0.86122209],
[0.77817196],
[0.72936044],
[0.7496687],
[0.12573789],
[0.69702716],
[0.21427718],
[0.76029427],
[0.81248944],
[0.33737954],
[0.30437845],
[0.27918665],
[0.24273428],
[0.53520877],
[0.4957175],
[0.71882631],
[0.12573789],
[0.20662001],
[0.86740544],
[0.98959909],
[0.07487408],
[0.2494824],
[0.65886996],
[0.45006774],
[0.42034788],
[0.14588413],
```

```
[0.15471885],
       [0.45730036],
       [0.69032997],
       [0.73964286],
       [0.862864],
       [0.14806763],
       [0.57848411],
       [0.78714111],
       [0.65838291],
       [0.79584274],
       [0.90442096],
       [0.22848616],
       [0.70148836],
       [0.35778052],
       [0.79584274],
       [0.69702716],
       [0.65838291],
       [0.74404968],
       [0.45006774],
       [0.52039975],
       [0.70412694],
       [0.76029427],
       [0.52206954]])
In [31]:
import pickle
with open('model','wb') as file:
   pickle.dump(reg,file)
with open('scaler','wb') as file:
    pickle.dump(absenteeism scaler,file)
In [ ]:
```