```
In [1]:
          import numpy as np
          import pandas as pd
          import seaborn as sns
          import matplotlib.pyplot as plt
In [2]:
          from sklearn.linear_model import LogisticRegression
In [3]:
          df=pd.read_csv("C6_bmi.csv")
          df
Out[3]:
              Gender Height Weight Index
           0
                Male
                         174
                                  96
                                         4
           1
                Male
                         189
                                  87
                                         2
              Female
                         185
                                 110
                                         4
              Female
                         195
                                 104
                                         3
                                         3
                Male
                         149
                                  61
                                  •••
         495 Female
                         150
                                 153
                                         5
         496
              Female
                         184
                                 121
                                         4
         497
              Female
                         141
                                 136
                                         5
         498
                                         5
                Male
                         150
                                  95
         499
                                         5
                Male
                         173
                                 131
        500 rows × 4 columns
In [4]:
          df=df.dropna()
          df
Out[4]:
              Gender Height Weight Index
           0
                Male
                         174
                                  96
                                         4
           1
                Male
                         189
                                  87
                                         2
           2 Female
                         185
                                 110
                                         4
              Female
                         195
                                 104
                                         3
                Male
                         149
                                  61
                                         3
         495 Female
                         150
                                 153
                                         5
         496
              Female
                         184
                                 121
                                         4
```

	Gender	Height	Weight	Index
498	Male	150	95	5
499	Male	173	131	5

500 rows × 4 columns

```
In [5]:
          df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 500 entries, 0 to 499
         Data columns (total 4 columns):
              Column Non-Null Count Dtype
          0
              Gender
                      500 non-null
                                       object
              Height
                      500 non-null
                                       int64
          1
          2
              Weight
                       500 non-null
                                       int64
              Index
                       500 non-null
                                       int64
         dtypes: int64(3), object(1)
         memory usage: 19.5+ KB
 In [6]:
          df.columns
 Out[6]: Index(['Gender', 'Height', 'Weight', 'Index'], dtype='object')
 In [7]:
          feature_matrix=df[['Height', 'Weight']]
          target_vector=df[ 'Index']
 In [8]:
          feature matrix.shape
 Out[8]: (500, 2)
 In [9]:
          target vector.shape
 Out[9]: (500,)
In [10]:
          from sklearn.preprocessing import StandardScaler
In [11]:
          fs=StandardScaler().fit_transform(feature_matrix)
In [12]:
          logr=LogisticRegression()
          logr.fit(fs,target_vector)
Out[12]: LogisticRegression()
In [13]:
          observation=[[1,2]]
```

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In [14]:
          prediction=logr.predict(observation)
          print(prediction)
         [5]
In [15]:
          logr.classes_
Out[15]: array([0, 1, 2, 3, 4, 5], dtype=int64)
In [16]:
          logr.predict_proba(observation)[0][0]
         5.5956697582538237e-11
Out[16]:
In [17]:
          logr.predict proba(observation)
Out[17]: array([[5.59566976e-11, 6.05990036e-10, 1.19071465e-07, 4.99471797e-05,
                  2.03791363e-02, 9.79570797e-01]])
In [18]:
          df['Index'].value counts()
               198
Out[18]:
         5
               130
                69
                68
                22
                13
         Name: Index, dtype: int64
In [19]:
          x=df[['Height', 'Weight']]
          y=df['Index']
In [20]:
          #g1={ 'TenYearCHD':{'True':1, 'False':2}}
          #df=df.replace(g1)
          #df
In [21]:
          from sklearn.model_selection import train_test_split
In [22]:
          x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
In [23]:
          from sklearn.ensemble import RandomForestClassifier
In [24]:
          rfc=RandomForestClassifier()
          rfc.fit(x_train,y_train)
Out[24]: RandomForestClassifier()
```

```
In [25]:
         parameters={ 'max_depth':[1,2,3,4,5],
                     'min_samples_leaf':[5,10,15,20,25],
                     'n estimators':[10,20,30,40,50]
         }
In [26]:
         from sklearn.model selection import GridSearchCV
         grid search =GridSearchCV(estimator=rfc,param grid=parameters,cv=2,scoring="accuracy")
         grid_search.fit(x_train,y_train)
Out[26]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),
                     param_grid={'max_depth': [1, 2, 3, 4, 5],
                                'min_samples_leaf': [5, 10, 15, 20, 25],
                                'n estimators': [10, 20, 30, 40, 50]},
                     scoring='accuracy')
In [27]:
         grid search.best score
Out[27]: 0.7714285714285715
In [28]:
         rfc best=grid search.best estimator
In [35]:
         from sklearn.tree import plot tree
         plt.figure(figsize=(80,40))
         plot tree(rfc best.estimators [5],feature names=x.columns,class names=['a','b','c','d'
Out[35]: [Text(2547.391304347826, 1993.2, 'Weight <= 129.0\ngini = 0.723\nsamples = 222\nvalue =
         [4, 16, 41, 45, 111, 133]\nclass = f'),
         Text(1407.1304347826087, 1630.8000000000000, 'Weight <= 65.5\ngini = 0.75\nsamples = 14
        9\nvalue = [4, 16, 41, 45, 91, 36]\nclass = e'),
         Text(582.2608695652174, 1268.4, 'Weight <= 61.5\ngini = 0.635\nsamples = 31\nvalue =
         [4, 16, 22, 4, 0, 0] \setminus class = c'),
         Text(388.17391304347825, 906.0, 'Height <= 171.5\ngini = 0.624\nsamples = 23\nvalue =
         [4, 14, 14, 1, 0, 0] \setminus class = b'),
         Text(194.08695652173913, 543.599999999999, 'gini = 0.436\nsamples = 10\nvalue = [2, 1,
         \nvalue = [2, 13, 3, 0, 0, 0] \setminus (ass = b'),
         Text(388.17391304347825, 181.1999999999982, 'gini = 0.298\nsamples = 8\nvalue = [2, 9,
        0, 0, 0, 0]\nclass = b'),
         Text(776.3478260869565, 181.1999999999982, 'gini = 0.49\nsamples = 5\nvalue = [0, 4,
         3, 0, 0, 0] \cap (ass = b'),
         Text(776.3478260869565, 906.0, 'gini = 0.544\nsamples = 8\nvalue = [0, 2, 8, 3, 0, 0]\n
         class = c'),
         Text(2232.0, 1268.4, 'Weight <= 105.5\ngini = 0.668\nsamples = 118\nvalue = [0, 0, 19,
         41, 91, 36]\nclass = e'),
         Text(1746.782608695652, 906.0, 'Weight <= 84.5\ngini = 0.709\nsamples = 75\nvalue = [0,
         0, 19, 40, 40, 15]\nclass = d'),
         Text(1358.608695652174, 543.599999999999, 'Height <= 179.5\ngini = 0.656\nsamples = 36
         \nvalue = [0, 0, 16, 12, 23, 1]\nclass = e'),
         Text(1164.5217391304348, 181.19999999999982, 'gini = 0.53\nsamples = 26\nvalue = [0, 0,
         2, 12, 23, 1]\nclass = e'),
         Text(1552.695652173913, 181.199999999999, 'gini = 0.0\nsamples = 10\nvalue = [0, 0, 1
         4, 0, 0, 0]\nclass = c'),
         9\nvalue = [0, 0, 3, 28, 17, 14]\nclass = d'),
```

 $Text(1940.8695652173913, 181.1999999999982, 'gini = 0.691 \nsamples = 33 \nvalue = [0,]$ 0, 3, 20, 15, 13]\nclass = d'), Text(2329.0434782608695, 181.1999999999982, 'gini = 0.43\nsamples = 6\nvalue = [0, 0, 0] $0, 8, 2, 1] \setminus class = d'),$ Text(2717.217391304348, 906.0, 'Height <= 165.0\ngini = 0.429\nsamples = 43\nvalue = $[0, 0, 0, 1, 51, 21] \setminus e = e'$ Text(2523.1304347826085, 543.599999999999, 'gini = 0.0\nsamples = 15\nvalue = [0, 0, 0, 0, 0, 21]\nclass = f'), Text(2911.304347826087, 543.599999999999, 'Weight <= 115.5\ngini = 0.038\nsamples = 28 \nvalue = [0, 0, 0, 1, 51, 0]\nclass = e'), Text(2717.217391304348, 181.199999999999, 'gini = 0.1\nsamples = 11\nvalue = [0, 0, 0, 1, 18, 0]\nclass = e'), Text(3105.391304347826, 181.199999999999, 'gini = 0.0\nsamples = 17\nvalue = [0, 0, $0, 0, 33, 0 \cap e^{-1}$ Text(3687.6521739130435, 1630.80000000000002, 'Weight <= 138.5\ngini = 0.283\nsamples = 73\nvalue = [0, 0, 0, 0, 20, 97]\nclass = f'), Text(3299.478260869565, 1268.4, 'Height <= 181.0\ngini = 0.436\nsamples = 19\nvalue = $[0, 0, 0, 0, 9, 19] \setminus class = f'),$ $Text(3105.391304347826, 906.0, 'gini = 0.0 \nsamples = 13 \nvalue = [0, 0, 0, 0, 0, 19] \n$ class = f'), Text(3493.565217391304, 906.0, 'gini = 0.0\nsamples = 6\nvalue = [0, 0, 0, 0, 9, 0]\ncl ass = e'), Text(4075.8260869565215, 1268.4, 'Height <= 191.5\ngini = 0.217\nsamples = 54\nvalue = $[0, 0, 0, 0, 11, 78] \setminus class = f'),$ Text(3881.7391304347825, 906.0, 'gini = 0.0\nsamples = 47\nvalue = [0, 0, 0, 0, 76] \nclass = f'), Text(4269.913043478261, 906.0, 'gini = 0.26\nsamples = 7\nvalue = [0, 0, 0, 0, 11, 2]\n class = e')

