## problem Statement:To predict How Best the data fits

## 1)Data Collection

df

Out[2]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	male	33.770	1	no	southeast	1725.55230
2	28	male	33.000	3	no	southeast	4449.46200
3	33	male	22.705	0	no	northwest	21984.47061
4	32	male	28.880	0	no	northwest	3866.85520
1333	50	male	30.970	3	no	northwest	10600.54830
1334	18	female	31.920	0	no	northeast	2205.98080
1335	18	female	36.850	0	no	southeast	1629.83350
1336	21	female	25.800	0	no	southwest	2007.94500
1337	61	female	29.070	0	yes	northwest	29141.36030

1338 rows × 7 columns

### 2) Data Cleaning and Preprocessing

In [3]: ► df.head()

Out[3]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	male	33.770	1	no	southeast	1725.55230
2	28	male	33.000	3	no	southeast	4449.46200
3	33	male	22.705	0	no	northwest	21984.47061
4	32	male	28.880	0	no	northwest	3866.85520

In [4]: ► df.tail()

Out[4]:

	age	sex	bmi	children	smoker	region	charges
1333	50	male	30.97	3	no	northwest	10600.5483
1334	18	female	31.92	0	no	northeast	2205.9808
1335	18	female	36.85	0	no	southeast	1629.8335
1336	21	female	25.80	0	no	southwest	2007.9450
1337	61	female	29.07	0	yes	northwest	29141.3603

In [5]: ► df.shape

Out[5]: (1338, 7)

In [6]: ► df.describe

Out[6]:	<bound< th=""><th>met</th><th>hod NDFr</th><th>ame.describ</th><th>e of</th><th>age</th><th>sex</th><th>bmi children</th><th>smo</th></bound<>	met	hod NDFr	ame.describ	e of	age	sex	bmi children	smo
	ker	re	gion	charges					
	0	19	female	27.900	0	yes	southwest	16884.92400	
	1	18	male	33.770	1	no	southeast	1725.55230	
	2	28	male	33.000	3	no	southeast	4449.46200	
	3	33	male	22.705	0	no	northwest	21984.47061	
	4	32	male	28.880	0	no	northwest	3866.85520	
	• • •			• • •	• • •			• • •	
	1333	50	male	30.970	3	no	northwest	10600.54830	
	1334	18	female	31.920	0	no	northeast	2205.98080	
	1335	18	female	36.850	0	no	southeast	1629.83350	
	1336	21	female	25.800	0	no	southwest	2007.94500	
	1337	61	female	29.070	0	yes	northwest	29141.36030	

[1338 rows x 7 columns]>

```
df.info()
In [7]:
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 1338 entries, 0 to 1337
             Data columns (total 7 columns):
              #
                   Column
                             Non-Null Count Dtype
              0
                  age
                             1338 non-null
                                              int64
              1
                             1338 non-null
                  sex
                                              object
              2
                  bmi
                             1338 non-null
                                              float64
              3
                  children 1338 non-null
                                              int64
              4
                                              object
                  smoker
                             1338 non-null
              5
                  region
                             1338 non-null
                                              object
              6
                   charges
                             1338 non-null
                                              float64
             dtypes: float64(2), int64(2), object(3)
             memory usage: 73.3+ KB

    df.isnull().any()

In [8]:
    Out[8]: age
                          False
             sex
                          False
             bmi
                          False
             children
                          False
             smoker
                          False
             region
                          False
             charges
                          False
             dtype: bool
          df.isna().sum()
In [9]:
    Out[9]: age
                          0
             sex
                          0
             bmi
                          0
                          0
             children
             smoker
                          0
             region
                          0
             charges
             dtype: int64

    | df['region'].value_counts()

In [10]:
   Out[10]: region
             southeast
                           364
                           325
             southwest
             northwest
                           325
             northeast
                           324
             Name: count, dtype: int64
```

#### Out[11]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	yes	southwest	16884.92400
1	18	0	33.770	1	no	southeast	1725.55230
2	28	0	33.000	3	no	southeast	4449.46200
3	33	0	22.705	0	no	northwest	21984.47061
4	32	0	28.880	0	no	northwest	3866.85520
1333	50	0	30.970	3	no	northwest	10600.54830
1334	18	1	31.920	0	no	northeast	2205.98080
1335	18	1	36.850	0	no	southeast	1629.83350
1336	21	1	25.800	0	no	southwest	2007.94500
1337	61	1	29.070	0	yes	northwest	29141.36030

#### Out[12]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	1	southwest	16884.92400
1	18	0	33.770	1	0	southeast	1725.55230
2	28	0	33.000	3	0	southeast	4449.46200
3	33	0	22.705	0	0	northwest	21984.47061
4	32	0	28.880	0	0	northwest	3866.85520
1333	50	0	30.970	3	0	northwest	10600.54830
1334	18	1	31.920	0	0	northeast	2205.98080
1335	18	1	36.850	0	0	southeast	1629.83350
1336	21	1	25.800	0	0	southwest	2007.94500
1337	61	1	29.070	0	1	northwest	29141.36030

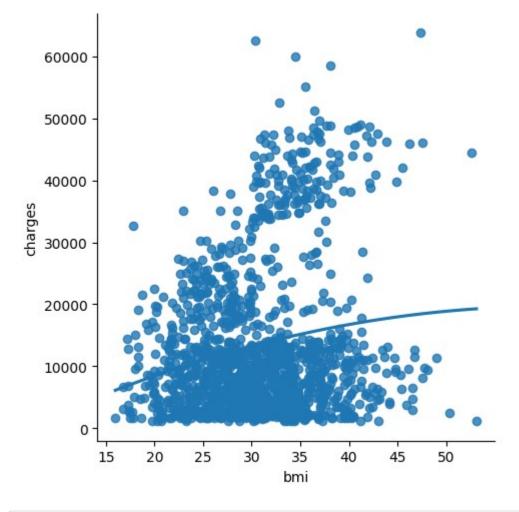
1338 rows × 7 columns

#### Out[13]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	1	1	16884.92400
1	18	0	33.770	1	0	2	1725.55230
2	28	0	33.000	3	0	2	4449.46200
3	33	0	22.705	0	0	3	21984.47061
4	32	0	28.880	0	0	3	3866.85520
1333	50	0	30.970	3	0	3	10600.54830
1334	18	1	31.920	0	0	4	2205.98080
1335	18	1	36.850	0	0	2	1629.83350
1336	21	1	25.800	0	0	1	2007.94500
1337	61	1	29.070	0	1	3	29141.36030

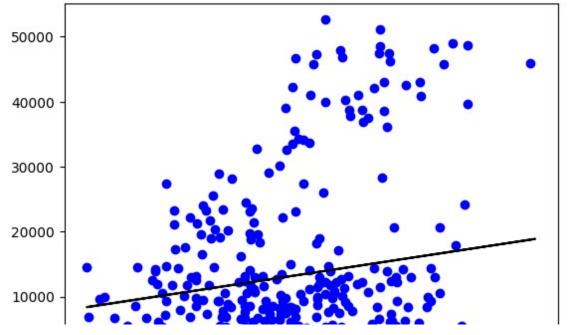
1338 rows × 7 columns

## 3) Data Visualization

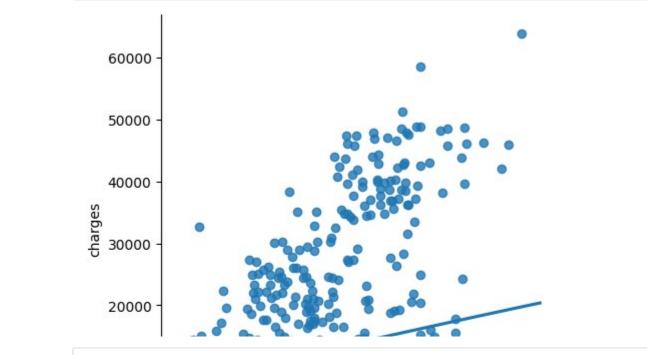


0.050136213258239914





## **Working With Subset Of Data**



In [19]: ► df700.fillna(method='ffill',inplace=True)

```
x=np.array(df['bmi']).reshape(-1,1)
In [20]:
            y=np.array(df['charges']).reshape(-1,1)
In [21]:

    df700.dropna(inplace=True)

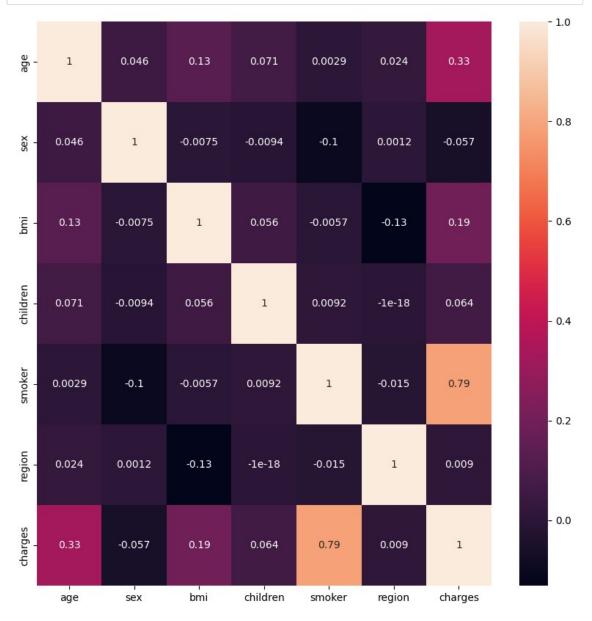
In [22]:
         | x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25)
            lr=LinearRegression()
            lr.fit(x_train,y_train)
            print(lr.score(x_test,y_test))
            0.03834819458441918
In [23]:
         plt.scatter(x_test,y_test,color='b')
            plt.plot(x_test,y_pred,color='k')
            plt.show()
             60000
             50000
              40000
             30000
             20000
              10000
```

## **Evaluation of model**

0.03834819458441918

## **Ridge Regression**

```
In [26]:  plt.figure(figsize= (10,10))
    sns.heatmap(df700.corr(),annot = True)
    plt.show()
```



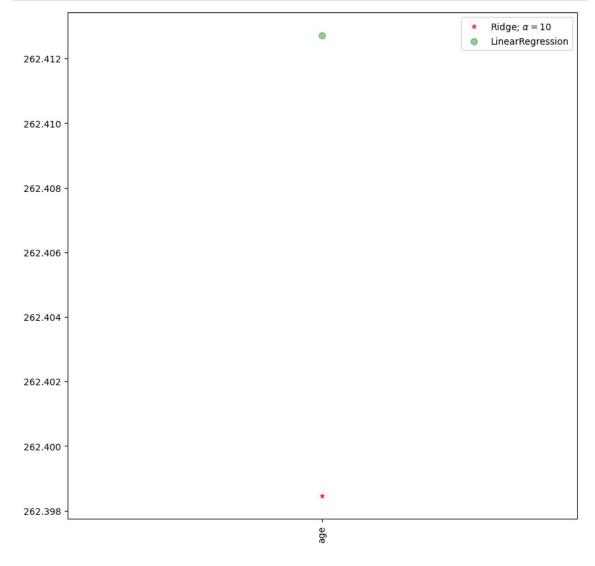
The dimension of x\_train is (936, 1)The dimension of x\_test is (402, 1)

#### Linear Regression Model :

The train score for lr model is 0.0910963973805714 The test score for lr model is 0.08490473916580776

#### Ridge Model:

The train score for ridge model is 0.09109639711159634 The test score for ridge model is 0.08490538609860176



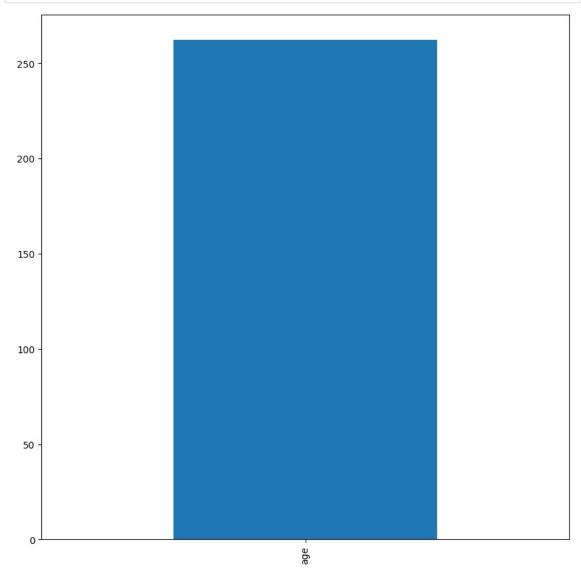
## **Lasso Regression**

```
In [31]: N lasso=Lasso(alpha=10)
lasso.fit(x_train,y_train)
train_score_ls =lasso.score(x_train,y_train)
test_score_ls =lasso.score(x_test,y_test)
print("\nRidge Model:\n")
print("The train score for ls model is {}".format(train_score_ls))
print("The test score for ls model is {}".format(test_score_ls))
```

#### Ridge Model:

The train score for ls model is 0.09109639395809055 The test score for ls model is 0.08490704421828055

#### 



In [33]: ▶ from sklearn.linear\_model import LassoCV

# In [34]: #using the Linear CV model from sklearn.linear\_model import RidgeCV #Lasso Cross validation ridge\_cv = RidgeCV(alphas = [0.0001,0.001,0.01,1,10]).fit(x\_train,y\_train) #score print(ridge\_cv.score(x\_train,y\_train)) print(ridge\_cv.score(x\_test,y\_test))

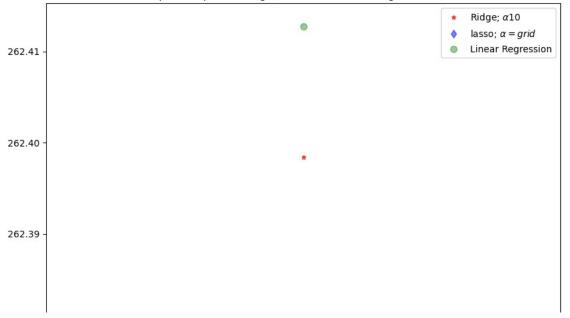
- 0.09109639711159612
- 0.08490538609884779

## In [47]: #using the linear CV model from sklearn.linear\_model import LassoCV #Ridge Cross validation lasso\_cv=LassoCV(alphas = [0.0001,0.001,0.01,1,10]).fit(x\_train,y\_train) #score print("The train score for ls model is {}".format(ridge\_cv.score(x\_train,y\_print("The test score for ls model is {}".format(ridge\_cv.score(x\_test,y\_test)).".

The train score for ls model is 0.09109639711159612 The test score for ls model is 0.08490538609884779

```
In [48]: In plt.figure(figsize = (10, 10))
#add plot for ridge regression
plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markerd plot for lasso regression
plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6
#add plot for linear model
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=6
#rotate axis
plt.xticks(rotation = 90)
plt.legend()
plt.title("Comparison plot of Ridge, Lasso and Linear regression model")
plt.show()
```





## **Elastic Net Regression**

```
In [49]:
            el=ElasticNet()
            el.fit(x,y)
            print(el.coef_)
            print(el.intercept_)
            [257.0684655]
            3191.532406056678
In [50]:

    | y_pred_elastic=el.predict(x_train)
In [51]:

    | mean_squared_error=np.mean((y_pred_elastic-y_train)**2)

            print("Mean Squared Error on test", mean_squared_error)
            Mean Squared Error on test 135093713.48214507

▶ el=ElasticNet()

In [52]:
            el.fit(x_train,y_train)
            print(el.score(x_train,y_train))
            0.09109580670592365
```

## **Logistic Regression**

```
In [80]: ▶ import numpy as np
    import pandas as pd
    from sklearn.linear_model import LogisticRegression
    from sklearn.preprocessing import StandardScaler
```

Out[81]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.924000
1	18	male	33.770	1	no	southeast	1725.552300
2	28	male	33.000	3	no	southeast	4449.462000
3	33	male	22.705	0	no	northwest	21984.470610
4	32	male	28.880	0	no	northwest	3866.855200
5	31	female	25.740	0	no	southeast	3756.621600
6	46	female	33.440	1	no	southeast	8240.589600
7	37	female	27.740	3	no	northwest	7281.505600
8	37	male	29.830	2	no	northeast	6406.410700
9	60	female	25.840	0	no	northwest	28923.136920
10	25	male	26.220	0	no	northeast	2721.320800

Out[82]: (1338, 7)

In [83]: pd.set\_option('display.max\_rows',10000000000)
 pd.set\_option('display.max\_columns',10000000000)
 pd.set\_option('display.width',95)

In [84]: ▶ print('This Dataset has %d rows and %d columns'%(df.shape))

This Dataset has 1338 rows and 7 columns

In [85]: ► df.head()

Out[85]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	male	33.770	1	no	southeast	1725.55230
2	28	male	33.000	3	no	southeast	4449.46200
3	33	male	22.705	0	no	northwest	21984.47061
4	32	male	28.880	0	no	northwest	3866.85520

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```
▶ df.describe
In [86]:
    Out[86]: <bound method NDFrame.describe of
                                                         age
                                                                  sex
                                                                          bmi
                                                                                children smo
              ker
                       region
                                     charges
              0
                          female
                                                   0
                      19
                                  27.900
                                                        yes
                                                              southwest
                                                                         16884.924000
              1
                                                   1
                     18
                            male
                                  33.770
                                                              southeast
                                                                          1725.552300
                                                         no
              2
                      28
                            male
                                  33.000
                                                   3
                                                              southeast
                                                                          4449.462000
                                                         no
              3
                      33
                            male
                                  22.705
                                                   0
                                                         no
                                                             northwest
                                                                         21984.470610
              4
                      32
                                  28.880
                                                   0
                            male
                                                         no
                                                             northwest
                                                                           3866.855200
              5
                      31
                          female
                                  25.740
                                                   0
                                                              southeast
                                                                           3756.621600
                                                         no
              6
                          female
                                                   1
                      46
                                  33.440
                                                         no
                                                              southeast
                                                                           8240.589600
              7
                      37
                                                   3
                          female
                                  27.740
                                                             northwest
                                                                          7281.505600
                                                         no
              8
                      37
                            male
                                  29.830
                                                   2
                                                         no
                                                             northeast
                                                                           6406.410700
              9
                      60
                          female
                                  25.840
                                                   0
                                                             northwest
                                                                         28923.136920
                                                         no
              10
                      25
                            male
                                  26.220
                                                   0
                                                             northeast
                                                         no
                                                                           2721.320800
              11
                      62
                          female
                                  26.290
                                                   0
                                                        yes
                                                             southeast
                                                                         27808.725100
              12
                      23
                                                   0
                            male
                                  34.400
                                                         no
                                                              southwest
                                                                          1826.843000
              13
                      56
                          female
                                  39.820
                                                   0
                                                              southeast
                                                                         11090.717800
                                                         no
                      27
                                  42.130
                                                   0
              14
                            male
                                                        yes
                                                              southeast
                                                                         39611.757700
              15
                      19
                            male
                                  24.600
                                                   1
                                                              southwest
                                                                          1837.237000
                                                         no
              16
                      52
                          female
                                  30.780
                                                   1
                                                                         10797.336200
                                                         no
                                                             northeast
In [87]:
           df.info()
              <class 'pandas.core.frame.DataFrame'>
              RangeIndex: 1338 entries, 0 to 1337
              Data columns (total 7 columns):
                   Column
                              Non-Null Count Dtype
              - - -
                                               ----
               0
                                               int64
                   age
                              1338 non-null
               1
                   sex
                              1338 non-null
                                               object
               2
                                               float64
                   bmi
                              1338 non-null
               3
                   children
                              1338 non-null
                                               int64
               4
                   smoker
                              1338 non-null
                                               object
               5
                              1338 non-null
                   region
                                               object
               6
                   charges
                              1338 non-null
                                               float64
              dtypes: float64(2), int64(2), object(3)
              memory usage: 73.3+ KB
In [88]:
           df.isnull().sum()
    Out[88]:
                           0
              age
                           0
              sex
                           0
              bmi
              children
                           0
              smoker
                           0
              region
                           0
              charges
                           0
              dtype: int64
```

#### Out[89]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	1	southwest	16884.924000
1	18	male	33.770	1	0	southeast	1725.552300
2	28	male	33.000	3	0	southeast	4449.462000
3	33	male	22.705	0	0	northwest	21984.470610
4	32	male	28.880	0	0	northwest	3866.855200
5	31	female	25.740	0	0	southeast	3756.621600
6	46	female	33.440	1	0	southeast	8240.589600
7	37	female	27.740	3	0	northwest	7281.505600
8	37	male	29.830	2	0	northeast	6406.410700
9	60	female	25.840	0	0	northwest	28923.136920
10	25	male	26.220	0	0	northeast	2721.320800

```
In [90]: N convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
df
```

#### Out[90]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	1	southwest	16884.924000
1	18	0	33.770	1	0	southeast	1725.552300
2	28	0	33.000	3	0	southeast	4449.462000
3	33	0	22.705	0	0	northwest	21984.470610
4	32	0	28.880	0	0	northwest	3866.855200
5	31	1	25.740	0	0	southeast	3756.621600
6	46	1	33.440	1	0	southeast	8240.589600
7	37	1	27.740	3	0	northwest	7281.505600
8	37	0	29.830	2	0	northeast	6406.410700
9	60	1	25.840	0	0	northwest	28923.136920
10	25	0	26.220	0	0	northeast	2721.320800

#### Out[91]:

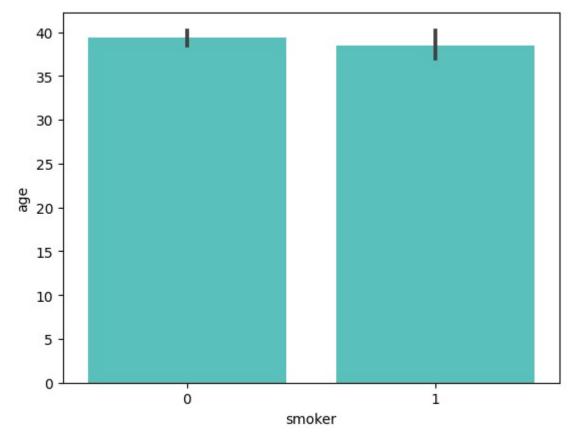
	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	1	2	16884.924000
1	18	0	33.770	1	0	1	1725.552300
2	28	0	33.000	3	0	1	4449.462000
3	33	0	22.705	0	0	4	21984.470610
4	32	0	28.880	0	0	4	3866.855200
5	31	1	25.740	0	0	1	3756.621600
6	46	1	33.440	1	0	1	8240.589600
7	37	1	27.740	3	0	4	7281.505600
8	37	0	29.830	2	0	3	6406.410700
9	60	1	25.840	0	0	4	28923.136920
10	25	0	26.220	0	0	3	2721.320800

```
In [94]: print('The Feature Matrix has %d Rows and %d columns(s)'%(features_matrix.print('The Target Matrix has %d Rows and %d columns(s)'%(np.array(target_variations))
```

The Feature Matrix has 1338 Rows and 4 columns(s)
The Target Matrix has 1338 Rows and 1 columns(s)

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```
In [95]: | import matplotlib.pyplot as plt
import seaborn as sns
sns.barplot(x='smoker', y='age', data=df, color="mediumturquoise")
plt.show()
```



The algoritham was trained to predict one of the two classes:[0 1]

```
In [142]: ▶ print(" " "The Model says the probability of the observation we passed below print()
```

The Model says the probability of the observation we passed belonging to class[0] 0.8057075871331396

```
In [148]: ▶ print(" " "The Model says the probability of the observation we passed beloprint()
```

The Model says the probability of the observation we passed belonging to class['g'] is 0.19429241286686041

```
In [146]:  x=np.array(df['age']).reshape(-1,1)
y=np.array(df['smoker']).reshape(-1,1)
```

#### 0.7910447761194029

C:\Users\chinta pavani\AppData\Local\Programs\Python\Python311\Lib\site-p
ackages\sklearn\utils\validation.py:1143: DataConversionWarning: A column
-vector y was passed when a 1d array was expected. Please change the shap
e of y to (n\_samples, ), for example using ravel().
 y = column\_or\_1d(y, warn=True)

### **Decision Tree**

```
import numpy as np
import pandas as pd
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
```

```
    df=pd.read_csv(r"C:\Users\chinta pavani\Documents\insurance.csv")

In [104]:
                df
    Out[104]:
                                      bmi children
                                                     smoker
                       age
                               sex
                                                                region
                                                                             charges
                                    27.900
                     0
                         19
                                                  0
                                                             southwest
                                                                        16884.924000
                            female
                     1
                         18
                                    33.770
                                                  1
                                                              southeast
                                                                         1725.552300
                              male
                                                         no
                     2
                         28
                              male
                                    33.000
                                                  3
                                                              southeast
                                                                         4449.462000
                                                         no
                     3
                         33
                                                  0
                              male
                                    22.705
                                                              northwest
                                                                        21984.470610
                     4
                                    28.880
                                                  0
                         32
                              male
                                                              northwest
                                                                         3866.855200
                                                         no
                     5
                                    25.740
                                                  0
                         31
                            female
                                                              southeast
                                                                         3756.621600
                     6
                         46
                            female
                                    33.440
                                                  1
                                                         no
                                                              southeast
                                                                         8240.589600
                    7
                         37
                            female
                                   27.740
                                                  3
                                                              northwest
                                                                         7281.505600
                                                         no
                     8
                                    29.830
                                                  2
                                                              northeast
                                                                         6406.410700
                         37
                              male
                                    25.840
                    9
                                                  0
                         60
                            female
                                                              northwest
                                                                        28923.136920
                                                         no
                    10
                                   26.220
                                                  0
                         25
                              male
                                                              northeast
                                                                         2721.320800
                                                          no
In [105]:

    df.shape

    Out[105]: (1338, 7)
In [106]:

    df.isnull().any()

    Out[106]: age
                               False
                               False
                 sex
                 bmi
                               False
                               False
                 children
                 smoker
                               False
                 region
                               False
                               False
                 charges
                 dtype: bool
In [107]:

    | df['region'].value_counts()

    Out[107]: region
                 southeast
                                364
                 southwest
                                325
                 northwest
                                325
                                324
```

northeast

Name: count, dtype: int64

#### Out[108]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	yes	southwest	16884.924000
1	18	0	33.770	1	no	southeast	1725.552300
2	28	0	33.000	3	no	southeast	4449.462000
3	33	0	22.705	0	no	northwest	21984.470610
4	32	0	28.880	0	no	northwest	3866.855200
5	31	1	25.740	0	no	southeast	3756.621600
6	46	1	33.440	1	no	southeast	8240.589600
7	37	1	27.740	3	no	northwest	7281.505600
8	37	0	29.830	2	no	northeast	6406.410700
9	60	1	25.840	0	no	northwest	28923.136920
10	25	0	26.220	0	no	northeast	2721.320800

#### Out[109]:

	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	1	southwest	16884.924000
1	18	0	33.770	1	0	southeast	1725.552300
2	28	0	33.000	3	0	southeast	4449.462000
3	33	0	22.705	0	0	northwest	21984.470610
4	32	0	28.880	0	0	northwest	3866.855200
5	31	1	25.740	0	0	southeast	3756.621600
6	46	1	33.440	1	0	southeast	8240.589600
7	37	1	27.740	3	0	northwest	7281.505600
8	37	0	29.830	2	0	northeast	6406.410700
9	60	1	25.840	0	0	northwest	28923.136920
10	25	0	26.220	0	0	northeast	2721.320800

```
In [111]: ► (x_train,x_test,y_train,y_test)=train_test_split(all_inputs,all_classes,te
```

```
In [112]:
In [113]:
          clf.fit(x_train,y_train)
   Out[113]:
                     DecisionTreeClassifier
             DecisionTreeClassifier(random_state=0)
In [114]:
          ▶ DecisionTreeClassifier(random_state=0)
            score=clf.score(x_test,y_test)
            print(score)
            0.4878048780487805
         Random Forest
In [115]:

    import pandas as pd

            import numpy as np
            import matplotlib.pyplot as plt ,seaborn as sns
In [116]:

    df=pd.read_csv(r"C:\Users\chinta pavani\Documents\insurance.csv")

   Out[116]:
```

age	sex	bmi	children	smoker	region	charges
19	female	27.900	0	yes	southwest	16884.924000
18	male	33.770	1	no	southeast	1725.552300
28	male	33.000	3	no	southeast	4449.462000
33	male	22.705	0	no	northwest	21984.470610
32	male	28.880	0	no	northwest	3866.855200
31	female	25.740	0	no	southeast	3756.621600
46	female	33.440	1	no	southeast	8240.589600
37	female	27.740	3	no	northwest	7281.505600
37	male	29.830	2	no	northeast	6406.410700
60	female	25.840	0	no	northwest	28923.136920
25	male	26.220	0	no	northeast	2721.320800
	19 18 28 33 32 31 46 37 37 60	19 female 18 male 28 male 33 male 32 male 31 female 46 female 37 female 37 male 60 female	19 female 27.900 18 male 33.770 28 male 33.000 33 male 22.705 32 male 28.880 31 female 25.740 46 female 33.440 37 female 27.740 37 male 29.830 60 female 25.840	19 female 27.900 0 18 male 33.770 1 28 male 33.000 3 33 male 22.705 0 32 male 28.880 0 31 female 25.740 0 46 female 33.440 1 37 female 27.740 3 37 male 29.830 2 60 female 25.840 0	19 female 27.900 0 yes 18 male 33.770 1 no 28 male 33.000 3 no 33 male 22.705 0 no 32 male 28.880 0 no 31 female 25.740 0 no 46 female 33.440 1 no 37 female 27.740 3 no 37 male 29.830 2 no 60 female 25.840 0 no	19 female 27.900 0 yes southwest 18 male 33.770 1 no southeast 28 male 33.000 3 no southeast 33 male 22.705 0 no northwest 32 male 28.880 0 no northwest 31 female 25.740 0 no southeast 46 female 33.440 1 no southeast 37 female 27.740 3 no northwest 37 male 29.830 2 no northwest 60 female 25.840 0 no northwest

In [117]: ► df.shape

Out[117]: (1338, 7)

```
In [118]:

    | df['region'].value_counts()

    Out[118]: region
               southeast
                              364
               southwest
                              325
               northwest
                              325
                              324
               northeast
               Name: count, dtype: int64

    df['bmi'].value_counts()

In [119]:
    Out[119]: bmi
                          13
               32.300
               28.310
                            9
               30.495
                            8
               30.875
                            8
                            8
               31.350
               30.800
                            8
               34.100
                            8
               28.880
                            8
               33.330
                            7
                            7
               35.200
                            7
               25.800
               32.775
                            7
                            7
               27.645
                            7
               32.110
                            7
               38.060
                            7
               25.460
                            7
               30.590
                            7
               27.360
               24 222
In [121]:
            M | m={"sex":{"female":1,"male":0}}
               df=df.replace(m)
               print(df)
                                          children smoker
                                    bmi
                                                                region
                      age
                            sex
                                                                               charges
               0
                       19
                              1
                                 27.900
                                                  0
                                                       yes
                                                             southwest
                                                                         16884.924000
               1
                       18
                                 33.770
                                                  1
                                                             southeast
                                                                          1725.552300
                              0
                                                        no
               2
                                                  3
                       28
                              0
                                 33.000
                                                             southeast
                                                                          4449.462000
                                                        no
               3
                                                  0
                                                             northwest
                                                                         21984.470610
                       33
                              0
                                 22.705
                                                        no
               4
                       32
                                                  0
                              0
                                 28.880
                                                             northwest
                                                                          3866.855200
                                                        no
               5
                       31
                                                  0
                                                             southeast
                              1
                                 25.740
                                                        no
                                                                          3756.621600
               6
                                                  1
                       46
                              1
                                 33.440
                                                             southeast
                                                                          8240.589600
                                                        no
               7
                       37
                              1
                                 27.740
                                                  3
                                                             northwest
                                                                          7281.505600
                                                        no
               8
                       37
                                 29.830
                                                  2
                                                        no
                                                             northeast
                                                                          6406.410700
               9
                                                  0
                       60
                                 25.840
                                                             northwest
                                                                         28923.136920
                              1
                                                        no
                                                  0
               10
                       25
                              0
                                 26.220
                                                             northeast
                                                                          2721.320800
                                                        no
               11
                       62
                              1
                                 26.290
                                                  0
                                                       yes
                                                             southeast
                                                                         27808.725100
                                                  0
               12
                       23
                              0
                                 34.400
                                                             southwest
                                                                          1826.843000
                                                        no
               13
                       56
                                 39.820
                                                  0
                                                             southeast
                                                                         11090.717800
                              1
                                                        no
               14
                       27
                                 42.130
                                                  0
                                                       yes
                                                             southeast
                                                                         39611.757700
                                                  1
               15
                       19
                              0
                                 24.600
                                                        no
                                                             southwest
                                                                          1837.237000
                       52
                                 30.780
                                                  1
                                                                         10797.336200
               16
                              1
                                                             northeast
                                                        no
                                 23.845
               17
                       23
                              0
                                                  0
                                                                          2395.171550
                                                        no
                                                             northeast
```

```
    | n={"smoker":{"yes":1,"no":0}}

In [122]:
             df=df.replace(n)
             print(df)
                                     children
                                               smoker
                   age
                        sex
                                bmi
                                                          region
                                                                      charges
             0
                    19
                             27.900
                                            0
                                                       southwest 16884.924000
                          1
                                                    1
                                                      southeast
             1
                    18
                          0
                             33.770
                                            1
                                                                   1725.552300
             2
                                            3
                    28
                          0
                             33.000
                                                    0 southeast
                                                                  4449.462000
             3
                    33
                          0 22.705
                                            0
                                                    0 northwest 21984.470610
             4
                    32
                          0
                             28.880
                                            0
                                                       northwest
                                                                   3866.855200
             5
                    31
                                            0
                                                       southeast
                          1 25.740
                                                                   3756.621600
             6
                    46
                          1 33.440
                                            1
                                                    0 southeast
                                                                   8240.589600
             7
                    37
                          1 27.740
                                            3
                                                    0 northwest
                                                                  7281.505600
             8
                                            2
                    37
                          0 29.830
                                                    0 northeast
                                                                  6406.410700
             9
                    60
                          1
                             25.840
                                            0
                                                       northwest 28923.136920
                                            0
             10
                    25
                          0 26.220
                                                    0 northeast
                                                                   2721.320800
             11
                    62
                          1 26.290
                                            0
                                                    1 southeast 27808.725100
             12
                    23
                          0 34.400
                                            0
                                                    0 southwest
                                                                 1826.843000
                    56
             13
                          1 39.820
                                            0
                                                    0 southeast 11090.717800
                    27
             14
                          0 42.130
                                            0
                                                    1 southeast 39611.757700
             15
                    19
                          0 24.600
                                            1
                                                    0 southwest
                                                                 1837.237000
             16
                    52
                          1
                             30.780
                                            1
                                                    0 northeast 10797.336200
             17
                    23
                             23.845
                                            0
                                                       northeast
                                                                 2395.171550
                                                       In [123]:
           rfc=RandomForestClassifier()
             rfc.fit(x_train,y_train)
   Out[123]:
              ▼ RandomForestClassifier
              RandomForestClassifier()
In [124]:

    | rf=RandomForestClassifier()
             params={'max_depth':[2,3,5,20],
              'min_samples_leaf':[5,10,20,50,100,200],
              'n_estimators':[10,25,30,50,100,200]}

    ★ from sklearn.model selection import GridSearchCV

In [126]:
             grid_search=GridSearchCV(estimator=rf,param_grid=params,cv=2,scoring="accul

             grid_search.fit(x_train,y_train)
   Out[126]:
                           GridSearchCV
               ▶ estimator: RandomForestClassifier
                    ▶ RandomForestClassifier
In [127]:

■ grid_search.best_score_
   Out[127]: 0.5304492666780802
```

```
    | rf_best=grid_search.best_estimator_
In [128]:
             print(rf_best)
             RandomForestClassifier(max_depth=3, min_samples_leaf=200, n_estimators=1
In [129]:
          plt.figure(figsize=(80,40))
             plot_tree(rf_best.estimators_[4],class_names=['1','0'],filled=True);
                                       x[1] <= 1.5
                                        gini = 0.5
                                     samples = 828
                                  value = [658, 639]
                                         class = 1
                      gini = 0.499
                                                          gini = 0.5
                    samples = 559
                                                      samples = 269
                                                    value = [208, 221]
                  value = [450, 418]
                        class = 1
                                                          class = 0
In [131]:
          plt.figure(figsize=(70,30))
             plot_tree(rf_best.estimators_[6],class_names=["1","0"],filled=True);
                                                 x[1] <= 1.5
                                                  gini = 0.5
                                                samples = 830
                                               value = [662, 635]
                                                  class = 1
                                   x[0] \le 32.47
                                                               gini = 0.497
                                    gini = 0.5
                                                              samples = 266
                                  samples = 564
                                                             value = [214, 182]
                                 value = [448, 453]
                                                                class = 1
                                     class = 0
                      gini = 0.495
                                                 gini = 0.485
                     samples = 364
                                                samples = 200
                    value = [267, 325]
                                               value = [181, 128]
                       class = 0
                                                  class = 1
In [132]:

    | rf_best.feature_importances_
   Out[132]: array([0.53541719, 0.46458281])
In [133]:

    | rf=RandomForestClassifier(random_state=0)
```