Problem statement: To predict How Best the data fits

1. Data collection

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn import preprocessing,svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [34]:

df=pd.read_csv(r"C:\Users\jyothi reddy\Downloads\insurance.csv")
df
```

Out[34]:

	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

#Exploratory data anlysis

In [35]:

df.head()

Out[35]:

	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 [36]:

df.tail()

Out[36]:

	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 [37]:
                                                                                                                                        M
df.shape
Out[37]:
(1338, 7)
In [38]:
                                                                                                                                        M
df.describe
Out[38]:
<bound method NDFrame.describe of</pre>
                                                sex
                                                        bmi children smoker
                                                                                 region
                                                                                             charges
                                       age
       19 female 27.900
                                      yes southwest
                                                      16884.92400
                   33.770
                                                        1725.55230
1
       18
            male
                                           southeast
                                       no
                   33.000
                                                        4449.46200
       28
            male
                                       no
                                           southeast
3
       33
            male
                   22.705
                                  0
                                           northwest
                                                      21984.47061
                                       no
4
            male 28.880
                                  0
                                                        3866.85520
       32
                                       no northwest
                   30.970
                                 3
                                                      10600.54830
1333
      50
            male
                                       no northwest
       18
          female
                   31.920
                                  0
                                                        2205.98080
1334
                                       no northeast
                   36.850
                                  0
                                                        1629.83350
1335
                                           southeast
       18
          female
                                       no
                   25.800
                                                        2007.94500
1336
       21
          female
                                  0
                                       no southwest
          female 29.070
                                  0
                                      yes northwest 29141.36030
      61
1337
[1338 rows x 7 columns]>
In [39]:
                                                                                                                                        M
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):
    Column
              Non-Null Count Dtype
0
               1338 non-null
                              int64
    age
               1338 non-null
                              object
1
     sex
2
               1338 non-null
                               float64
     bmi
3
     children 1338 non-null
                               int64
     smoker
               1338 non-null
                              object
              1338 non-null
     region
                               object
              1338 non-null
                               float64
    charges
dtypes: float64(2), int64(2), object(3)
memory usage: 73.3+ KB
In [40]:
                                                                                                                                        M
df.isnull().any()
Out[40]:
           False
age
           False
sex
           False
bmi
children
           False
smoker
           False
region
           False
           False
charges
dtype: bool
In [41]:
                                                                                                                                        M
df.isna().sum()
Out[41]:
           0
age
sex
bmi
children
smoker
region
charges
dtype: int64
```

```
In [42]:
                                                                                                                                                      M
df['region'].value_counts()
Out[42]:
region
southeast
              364
              325
southwest
northwest
              325
northeast
              324
Name: count, dtype: int64
In [43]:
                                                                                                                                                      H
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
Out[43]:
      age
                 bmi children smoker
                                         region
            1 27.900
       19
   0
                            0
                                  yes southwest 16884.92400
    1
       18
             0 33.770
                            1
                                                 1725.55230
                                   no southeast
    2
       28
             0 33.000
                            3
                                                 4449.46200
                                   no southeast
   3
       33
             0 22.705
                            0
                                   no northwest 21984.47061
   4
       32
             0 28.880
                            0
                                                 3866.85520
                                   no northwest
                                   no northwest 10600.54830
 1333
       50
            0 30.970
                            3
 1334
       18
             1 31.920
                            0
                                   no northeast
                                                 2205.98080
 1335
       18
             1 36.850
                            0
                                   no southeast
                                                 1629.83350
                            0
 1336
      21
            1 25.800
                                   no southwest
                                                 2007.94500
                                                                                                                                                      M
In [44]:
convert={"smoker":{"yes":1, "no":0}}
df=df.replace(convert)
df
Out[44]:
                 bmi children smoker
                                         region
                                                    charges
      age sex
    0
       19
             1 27.900
                                    1 southwest 16884.92400
       18
             0 33.770
                                    0 southeast
                                                 1725.55230
    2
       28
             0 33 000
                            3
                                                4449 46200
                                    0 southeast
   3
       33
             0 22.705
                                    0 northwest 21984.47061
    4
       32
             0 28.880
                            0
                                    0 northwest
                                                 3866.85520
 1333
       50
            0 30.970
                                    0 northwest 10600.54830
                            0
 1334
       18
             1 31.920
                                    0 northeast
                                                 2205.98080
                            0
       18
                                                 1629.83350
 1335
             1 36.850
                                    0 southeast
       21
            1 25.800
                                    0 southwest
                                                 2007.94500
```

```
In [45]:

convert={"region":{"southwest":1,"southeast":2,"northwest":3,"northeast":4}}
df=df.replace(convert)
df
```

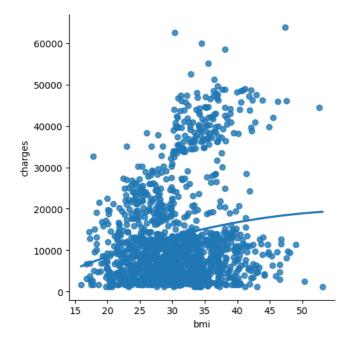
Out[45]:

	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

```
In [46]:
sns.lmplot(x='bmi',y='charges',order=2,data=df,ci=None)
plt.show()
```



```
In [47]:

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

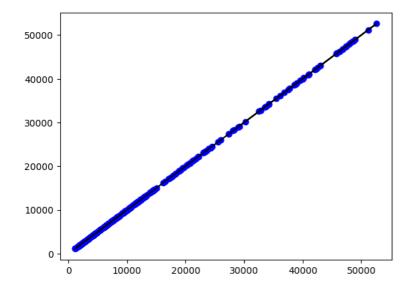
```
In [48]:

x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,random_state=0)
lr=LinearRegression()
lr.fit(x_train,y_train)
print(lr.score(x_test,y_test))
```

1.0

```
In [49]:

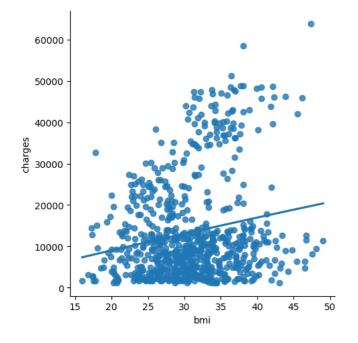
y_pred=lr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



working with subset of data

```
In [50]:

df700=df[:][:700]
sns.lmplot(x='bmi',y='charges',order=2,ci=None,data=df700)
plt.show()
```



```
In [51]:
df700.fillna(method='ffill',inplace=True)
```

```
In [52]:

x=np.array(df700["bmi"]).reshape(-1,1)
y=np.array(df700['charges']).reshape(-1,1)
```

```
M
In [53]:
df700.dropna(inplace=True)
                                                                                                                                                  M
In [54]:
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))
-0.1630229146000015
In [55]:
                                                                                                                                                  M
y_pred=lr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
 50000
 40000
 30000
 20000
 10000
```

Evaluation of model

20

25

0

```
In [56]:

from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score

In [57]:

Ir=LinearRegression()
Ir.fit(x_train,y_train)
y_pred=lr.predict(x_test)
```

50

40

45

-0.1630229146000015

print(r2)

r2=r2_score(y_test,y_pred)

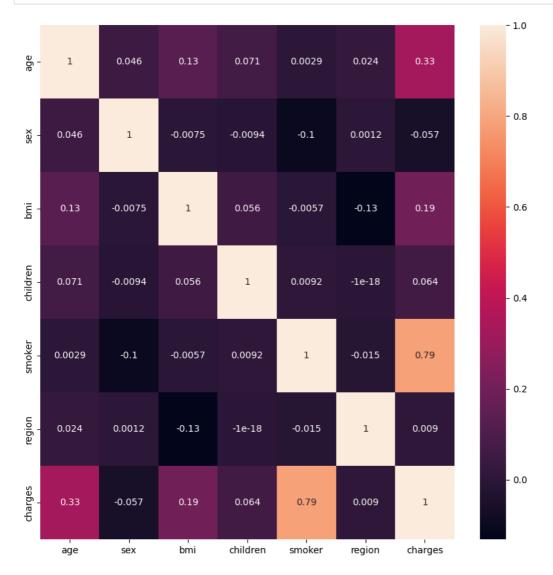
Ridge Regression

```
In [58]:

from sklearn.linear_model import Lasso,Ridge
from sklearn.preprocessing import StandardScaler
```

```
In [59]:
```

plt.figure(figsize=(10,10))
sns.heatmap(df700.corr(),annot=True)
plt.show()



```
In [60]:
```

features=df.columns[0:1]
target=df.columns[-1]

```
In [61]:
```

```
x=df[features].values
y=df[target].values
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30,random_state=1)
print("The dimension of X_train is {}".format(x_train.shape))
print("The dimension of X_test is {}".format(x_test.shape))
```

The dimension of X_{train} is (936, 1) The dimension of X_{test} is (402, 1)

```
In [62]:
                                                                                                                                                  M
lr = LinearRegression()
#Fit model
lr.fit(x_train, y_train)
#predict
actual = y_test
train_score_lr = lr.score(x_train, y_train)
test_score_lr = lr.score(x_test, y_test)
print("\nLinear Regression Model:\n")
print("The train score for lr model is {}".format(train_score_lr))
print("The test score for lr model is {}".format(test_score_lr))
Linear Regression Model:
The train score for lr model is 0.0910963973805714
The test score for lr model is 0.08490473916580776
In [63]:
ridgeReg = Ridge(alpha=10)
ridgeReg.fit(x_train,y_train)
#train and test scorefor ridge regression
train_score_ridge = ridgeReg.score(x_train, y_train)
test_score_ridge = ridgeReg.score(x_test, y_test)
print("\nRidge Model:\n")
print("The train score for ridge model is {}".format(train_score_ridge))
print("The test score for ridge model is {}".format(test_score_ridge))
Ridge Model:
The train score for ridge model is 0.09109639711159634
The test score for ridge model is 0.08490538609860176
In [64]:
                                                                                                                                                  M
plt.figure(figsize=(10,10))
Out[64]:
<Figure size 1000x1000 with 0 Axes>
<Figure size 1000x1000 with 0 Axes>
In [66]:
                                                                                                                                                  M
 plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker="*",markersize=5,color='red',label=r'Ridge;$\alpha=100$')
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker="o",markersize=7,color='green',label='LinearRegression')
plt.xticks(rotation=90)
plt.legend()
plt.show()
                                                              Ridge; \alpha = 100
 262.412
                                                              LinearRegression
 262.410
 262.408
 262,406
 262.404
 262.402
 262.400
 262.398
```

Lasso Regression

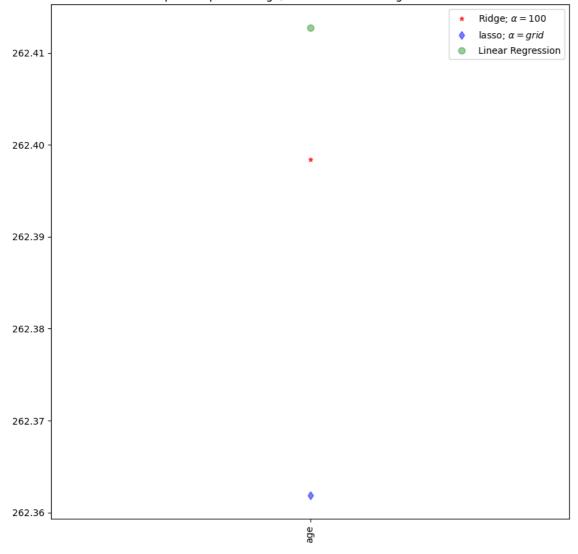
```
In [67]:
                                                                                                                                             M
lasso= Lasso(alpha=10)
lasso.fit(x_train,y_train)
#train and test scorefor ridge regression
train_score_ls = lasso.score(x_train, y_train)
test_score_ls= lasso.score(x_test, y_test)
print("\nRidge Model:\n")
print("The train score for lasso model is {}".format(train_score_ls))
print("The test score for lasso model is {}".format(test_score_ls))
Ridge Model:
The train score for lasso model is 0.09109639395809055
The test score for lasso model is 0.08490704421828055
                                                                                                                                             M
In [68]:
plt.figure(figsize=(10,10))
Out[68]:
<Figure size 1000x1000 with 0 Axes>
<Figure size 1000x1000 with 0 Axes>
                                                                                                                                            H
pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "bar")
plt.show()
 250
 200
 150
 100
  50
    0
                                       age
In [70]:
                                                                                                                                            H
from sklearn.linear_model import LassoCV
In [71]:
                                                                                                                                             Ы
#using the linear cv model
from sklearn.linear_model import RidgeCV
#cross validation
ridge_cv=RidgeCV(alphas =[0.0001,0.001,0.01,0.1,1,10]).fit(x_train,y_train)
print(ridge_cv.score(x_train,y_train))
print(ridge_cv.score(x_test,y_test))
```

0.09109639711159612

0.08490538609884779

```
M
In [72]:
#using the linear cv model
from sklearn.linear_model import LassoCV
#cross validation
lasso_cv=LassoCV(alphas =[0.0001,0.001,0.01,0.1,1,10]).fit(x_train,y_train)
print(lasso_cv.score(x_train,y_train))
print(lasso_cv.score(x_test,y_test))
0.09109639395809055
0.08490704421828055
In [73]:
plt.figure(figsize = (10, 10))
#add plot for ridge regression
plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker='*',markersize=5,color='red',label=r'Ridge; $\alpha=100$')
#add plot for lasso regression
plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue',label=r'lasso; $\alpha = grid$')
#add plot for linear model
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green',label='Linear Regression')
plt.xticks(rotation = 90)
plt.legend()
plt.title("Comparison plot of Ridge, Lasso and Linear regression model")
plt.show()
```

Comparison plot of Ridge, Lasso and Linear regression model



ElasticNet Regression

```
In [74]:
                                                                                                                                         M
from sklearn.linear_model import ElasticNet
                                                                                                                                         H
In [75]:
el=ElasticNet()
el.fit(x_train,y_train)
print(el.coef_)
print(el.intercept_)
[261.74450967]
3115.083177426244
In [76]:
                                                                                                                                         H
y_pred_elastic=el.predict(x_train)
In [77]:
                                                                                                                                         M
mean_squared_error=np.mean((y_pred_elastic-y_train)**2)
print(mean_squared_error)
135077142.70714515
In [78]:
el=ElasticNet()
el.fit(x_train,y_train)
print(el.score(x_train,y_train))
```

0.09109580670592365

Logistic Regression

```
In [79]:

import numpy as np
import pandas as pd
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler

In [80]:

df=pd.read_csv(r"C:\Users\jyothi reddy\Downloads\insurance.csv")
df
```

Out[80]:

	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

In [81]:

df.shape

Out[81]:

(1338, 7)

```
M
In [82]:
pd.set_option('display.max_rows',10000000000)
pd.set_option('display.max_columns',100000000000)
pd.set_option('display.width',95)
                                                                                                                                                M
In [83]:
print('This Dataset has %d rows and %d columns'%(df.shape))
This Dataset has 1338 rows and 7 columns
In [84]:
                                                                                                                                                М
df.head()
Out[84]:
                bmi children smoker
                                       region
   age
          sex
                                                 charges
    19
       female
                                 yes southwest 16884.92400
    18
              33,770
                                     southeast
                                               1725.55230
                                               4449.46200
    28
              33.000
                          3
                                    southeast
         male
                                 no
                                 no northwest 21984.47061
    33
                          0
                          0
    32
         male 28 880
                                 no northwest
                                              3866 85520
In [85]:
                                                                                                                                                M
df.describe
42
       41
             male
                    21.780
                                    1
                                              southeast
                                                           6272.477200
                                          no
43
                                                           6313.759000
       37
           female
                    30.800
                                    2
                                              southeast
                                          no
44
                    37.050
                                                           6079.671500
       38
             male
                                    1
                                              northeast
                                          no
45
       55
                    37.300
                                    0
             male
                                          no
                                              southwest
                                                          20630.283510
46
       18
           female
                    38.665
                                    2
                                          no
                                              northeast
                                                           3393.356350
47
       28
                    34,770
                                    0
                                                           3556,922300
           female
                                          no
                                              northwest
48
                                    0
       60
                    24.530
                                                          12629.896700
           female
                                          no
                                              southeast
49
       36
             male
                    35.200
                                    1
                                         yes
                                              southeast
                                                          38709.176000
50
                                    0
       18
           female
                    35.625
                                          no
                                              northeast
                                                           2211.130750
51
                                    2
       21
           female
                    33.630
                                          no
                                              northwest
                                                           3579.828700
52
                                    1
       48
             male
                    28,000
                                         yes
                                              southwest
                                                          23568.272000
53
       36
             male
                    34.430
                                    0
                                         yes
                                               southeast
                                                          37742.575700
54
       40
           female
                    28.690
                                    3
                                          no
                                              northwest
                                                           8059.679100
55
       58
             male
                    36.955
                                    2
                                         yes
                                              northwest
                                                          47496,494450
56
       58
           female
                    31.825
                                    2
                                          no
                                              northeast
                                                          13607.368750
57
       18
             male
                    31.680
                                    2
                                         yes
                                               southeast
                                                          34303.167200
58
       53
           female
                    22.880
                                    1
                                         yes
                                               southeast
                                                          23244.790200
59
       34
           female
                    37.335
                                    2
                                              northwest
                                                           5989.523650
60
       43
             male
                    27.360
                                    3
                                              northeast
                                                           8606.217400
61
       25
             male
                    33.660
                                               southeast
                                                           4504.662400
In [86]:
                                                                                                                                                M
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):
               Non-Null Count Dtype
#
    Column
0
               1338 non-null
                                 int64
     age
               1338 non-null
1
     sex
                                 object
                                 float64
 2
     bmi
               1338 non-null
 3
     children
               1338 non-null
                                 int64
 4
     smoker
               1338 non-null
                                 object
     region
               1338 non-null
                                 object
     charges
               1338 non-null
                                 float64
dtypes: float64(2), int64(2), object(3)
memory usage: 73.3+ KB
In [87]:
                                                                                                                                                M
df.isnull().sum()
Out[87]:
            0
age
sex
            0
bmi
            0
children
            0
smoker
            a
region
            0
charges
dtype: int64
```

```
M
In [88]:
convert={"smoker":{"yes":1,"no":0}}
df=df.replace(convert)
df
       28
             male 33 000
                                3
    2
                                        0 southeast
                                                      4449 462000
    3
       33
             male 22.705
                                0
                                        0
                                           northwest 21984.470610
    4
       32
                   28.880
                                0
                                            northwest
                                                      3866.855200
                                0
    5
       31 female
                  25.740
                                           southeast
                                                      3756.621600
                                        0
       46
           female
                  33.440
                                        0
                                            southeast
                                                      8240.589600
    7
       37
           female
                  27 740
                                3
                                        0
                                            northwest
                                                      7281.505600
    8
       37
             male 29.830
                                2
                                            northeast
                                                      6406.410700
                                        0
    9
       60
                   25.840
                                                     28923.136920
           female
                                           northwest
   10
       25
             male
                  26.220
                                0
                                        0
                                            northeast
                                                      2721.320800
       62 female 26.290
                                0
                                           southeast 27808.725100
   11
                                        1
   12
       23
             male
                   34.400
                                0
                                        0 southwest
                                                      1826.843000
                                0
   13
       56 female 39 820
                                        Ω
                                           southeast 11090 717800
                                0
                                        1 southeast 39611.757700
   14
       27
             male 42.130
                                                                                                                                                              M
In [89]:
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
Out[89]:
       age
           sex
                   bmi children smoker
                                            region
                                                        charges
    0
       19
                27.900
                              0
                                                   16884.924000
              1
                                        southwest
    1
       18
             0 33.770
                                                    1725.552300
    2
       28
             0 33.000
                              3
                                      0
                                         southeast
                                                    4449.462000
    3
       33
             0
                              0
                                                   21984.470610
                22.705
                                      0
                                         northwest
    4
       32
             0 28.880
                              0
                                      0
                                         northwest
                                                    3866.855200
    5
                              0
       31
              1 25.740
                                      0
                                         southeast
                                                    3756.621600
       46
              1 33.440
                                      0
                                                    8240.589600
                                         southeast
    7
       37
              1 27.740
                              3
                                      0 northwest
                                                    7281.505600
                              2
    8
       37
             0 29.830
                                                    6406.410700
                                      0
                                        northeast
    9
       60
             1 25.840
                                      0 northwest 28923.136920
In [90]:
                                                                                                                                                              M
convert={"region":{"southeast":1,"southwest":2,"northeast":3,"northwest":4}}
df=df.replace(convert)
df
    1
       18
             0 33.770
                                                 1725.552300
    2
       28
             0 33.000
                              3
                                      0
                                                 4449.462000
    3
       33
             0 22.705
                              0
                                      0
                                                21984.470610
    4
       32
             0 28.880
                              0
                                      0
                                             4
                                                 3866.855200
    5
                              0
       31
             1 25,740
                                      0
                                             1
                                                 3756.621600
    6
       46
                33.440
                                                 8240.589600
    7
       37
              1 27.740
                              3
                                      0
                                             4
                                                 7281.505600
    8
       37
             0 29.830
                                      0
                                             3
                                                 6406.410700
    9
       60
                25.840
                              0
                                      0
                                             4
   10
       25
             0 26.220
                              n
                                      0
                                             3
                                                 2721.320800
   11
       62
              1 26.290
                                              1 27808.725100
   12
       23
             0 34.400
                              0
                                      0
                                                 1826.843000
                                                11090 717800
       56
                              n
                                      0
                                             1
   13
             1 39 820
In [91]:
                                                                                                                                                              M
features_matrix=df.iloc[:,0:4]
```

```
In [92]:
                                                                                                                                          M
target_vector=df.iloc[:,-3]
In [93]:
                                                                                                                                          M
print('The Feature Matrix has %d Rows and %d columns(s)'%(features_matrix.shape))
print('The Target Matrix has %d Rows and %d columns(s)'%(np.array(target_vector).reshape(-1,1).shape))
The Feature Matrix has 1338 Rows and 4 columns(s)
The Target Matrix has 1338 Rows and 1 columns(s)
In [94]:
                                                                                                                                          М
import matplotlib.pyplot as plt
import seaborn as sns
In [95]:
                                                                                                                                          M
sns.barplot(x='smoker', y='age', data=df, color="mediumturquoise")
plt.show()
    40
    35
    30
    25
 age
20
    15
    10
     5
                       0
                                                        1
                                     smoker
In [96]:
                                                                                                                                          M
features_matrix_standardized=StandardScaler().fit_transform(features_matrix)
In [97]:
                                                                                                                                          Ы
algorithm=LogisticRegression(max_iter=10000)
In [98]:
                                                                                                                                          M
Logistic_Regression_Model=algorithm.fit(features_matrix_standardized,target_vector)
In [99]:
                                                                                                                                          M
observation=[[1,0,0.99539,-0.0588]]
In [101]:
                                                                                                                                          M
\verb|predictions=Logistic_Regression_Model.predict(observation)|\\
print('The model predicted the observation to belong to class %s'%(predictions))
The model predicted the observation to belong to class [0]
In [102]:
                                                                                                                                          M
print('The algoritham was trained to predict one of the two classes:%s'%(algorithm.classes_))
The algoritham was trained to predict one of the two classes:[0 1]
```

localhost:8888/notebooks/Documents/insurance.ipynb

```
In [104]:
"The Model says the probability of the observation we passed belonging to class[0] %s" " "%(algorithm.predict_proba(observation)[0][0]))
 The Model says the probability of the observation we passed belonging to class[0] 0.8057075871331396
In [107]:
Model says the probability of the observation we passed belonging to class['g'] Is %s" " "%(algorithm.predict_proba(observation)[0][1]))
 The Model says the probability of the observation we passed belonging to class['g'] Is 0.19429241286686041
In [108]:
                                                                                                                                                  M
x=np.array(df['age']).reshape(-1,1)
y=np.array(df['smoker']).reshape(-1,1)
                                                                                                                                                  Ы
In [109]:
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.05)
lo=LogisticRegression()
lo.fit(x_train,y_train)
print(lo.score(x_test,y_test))
0.7611940298507462
C:\Users\jyothi reddy\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\utils\validation.py:1143: DataConve
rsionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for
example using ravel().
  y = column_or_1d(y, warn=True)
Decision Tree
In [110]:
                                                                                                                                                  Ы
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
In [111]:
                                                                                                                                                  M
df=pd.read_csv(r"C:\Users\jyothi reddy\Downloads\insurance.csv")
df
            male 28.880
                                                  3866.855200
       32
                              0
                                     no northwest
   5
       31
                 25.740
                              0
                                         southeast
                                                  3756.621600
       46 female 33,440
                                     no
                                        southeast
                                                  8240.589600
       37
           female 27.740
                                                  7281.505600
                                     no
                                        northwest
                                         northeast
   8
       37
            male 29.830
                              2
                                                  6406.410700
   9
       60 female 25.840
                              0
                                     no
                                        northwest
                                                 28923.136920
   10
                                                  2721.320800
       25
            male
                 26.220
                                         northeast
                                     no
   11
       62 female 26,290
                              0
                                         southeast
                                                 27808.725100
                              0
                                                  1826.843000
  12
       23
            male
                 34.400
                                        southwest
                                     no
       56 female
                                                  11090.717800
                 39.820
  14
       27
            male 42 130
                              n
                                         southeast
                                                 39611 757700
  15
       19
                              1
                                                  1837.237000
            male 24.600
                                     no southwest
  16
       52 female 30.780
                                         northeast
                                                 10797.336200
In [112]:
                                                                                                                                                  M
df.shape
Out[112]:
(1338, 7)
```

```
In [113]:
                                                                                                                                                       M
df.isnull().any()
Out[113]:
             False
age
             False
sex
             False
bmi
children
             False
smoker
             False
region
             False
charges
             False
dtype: bool
In [114]:
                                                                                                                                                       M
df['region'].value_counts()
Out[114]:
region
              364
southeast
              325
southwest
northwest
              325
              324
northeast
Name: count, dtype: int64
In [115]:
                                                                                                                                                       M
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
df
   ΖU
       บบ
             T 30.005
                                   no normeast 13228.848950
  21
       30
             1 32.400
                            1
                                                 4149.736000
                                   no southwest
  22
       18
             0 34.100
                                       southeast
                                                  1137.011000
                                   no
  23
       34
             1 31.920
                                        northeast 37701.876800
       37
                            2
  24
             0 28.025
                                                  6203.901750
                                   no northwest
  25
       59
             1 27.720
                                   no southeast 14001.133800
  26
       63
             1 23.085
                            0
                                   no
                                       northeast 14451.835150
  27
       55
             1 32.775
                                   no northwest 12268.632250
  28
       23
             0 17.385
                                                  2775.192150
       31
             0 36.300
                            2
  29
                                   yes
                                       southwest 38711.000000
                            0
  30
       22
             0 35.600
                                   yes southwest 35585.576000
  31
       18
             1 26.315
                            n
                                   no northeast
                                                 2198.189850
                            5
  32
       19
             1 28.600
                                   no southwest
                                                4687.797000
In [116]:
                                                                                                                                                       M
convert={"smoker":{"yes":1,"no":0}}
df=df.replace(convert)
df
   19
       30
             0 35.300
                            U
                                    1 southwest 36837.467000
  20
       60
             1 36.005
                                    0 northeast 13228.846950
  21
       30
             1 32.400
                                                  4149.736000
                            0
  22
       18
             0 34.100
                                    southeast
                                                  1137.011000
  23
       34
             1 31.920
                                       northeast 37701.876800
  24
       37
             0 28.025
                                       northwest
                                                  6203.901750
       59
                                    0 southeast 14001.133800
  25
             1 27.720
                            3
                                       northeast 14451.835150
   26
       63
             1 23.085
  27
       55
             1 32.775
                                    0 northwest 12268.632250
  28
       23
             0 17.385
                            1
                                                  2775.192150
                                    0 northwest
   29
       31
             0 36.300
                                    1 southwest 38711.000000
                            0
  30
       22
             0 35.600
                                    1 southwest 35585.576000
  31
       18
             1 26.315
                                    0 northeast
                                                 2198.189850
```

```
In [117]:
                                                                                                                                           M
x=["bmi","children"]
y=["Yes","No"]
all_inputs=df[x]
all_classes=df["sex"]
In [118]:
                                                                                                                                           M
(x_train,x_test,y_train,y_test)=train_test_split(all_inputs,all_classes,test_size=0.03)
In [119]:
                                                                                                                                           M
clf=DecisionTreeClassifier(random_state=0)
                                                                                                                                           M
In [120]:
clf.fit(x_train,y_train)
Out[120]:
         DecisionTreeClassifier
DecisionTreeClassifier(random_state=0)
In [121]:
                                                                                                                                           M
score=clf.score(x_test,y_test)
print(score)
0.5853658536585366
Random Forest
In [122]:
                                                                                                                                           M
```

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt ,seaborn as sns
In [123]:
                                                                                                                                                       M
df=pd.read_csv(r"C:\Users\jyothi reddy\Downloads\insurance.csv")
df
                                     yes southwest 51194.559140
  34
       28
            male 36.400
                               1
  35
       19
            male 20.425
                               0
                                          northwest
                                                    1625.433750
  36
       62 female 32.965
                               3
                                          northwest 15612.193350
                                                   2302.300000
  37
            male 20.800
                               0
       26
                                      no southwest
   38
       35
            male 36.670
                                          northeast 39774.276300
                                     yes
                                         southwest 48173.361000
  39
       60
            male 39.900
                               0
                                     yes
   40
       24 female 26.600
                                                    3046.062000
                                          northeast
                                      no
       31 female
                  36.630
                               2
                                          southeast
                                                    4949.758700
  42
       41
            male 21.780
                               1
                                      no
                                          southeast
                                                    6272.477200
       37 female 30.800
                                                    6313.759000
   43
                                      no southeast
   44
       38
            male 37.050
                                          northeast
                                                    6079.671500
            male 37.300
                               0
                                      no southwest 20630,283510
  45
       55
   46
       18 female 38.665
                                                   3393.356350
In [124]:
                                                                                                                                                       H
df.shape
```

Out[124]:

(1338, 7)

```
In [125]:
                                                                                                                                              M
df['region'].value_counts()
Out[125]:
region
southeast
             364
             325
southwest
             325
northwest
northeast
             324
Name: count, dtype: int64
In [126]:
                                                                                                                                              M
df['bmi'].value_counts()
25.800
32.775
27.645
           7
32.110
           7
38.060
25.460
30.590
           7
27.360
24.320
           7
34.800
           7
27.500
           6
19.950
           6
           6
29.920
30.115
           6
26.600
           6
30.200
           6
35,530
           6
33.630
           6
28.595
           6
27
In [127]:
                                                                                                                                              Ы
m={"sex":{"female":1,"male":0}}
df=df.replace(m)
print(df)
      age
           sex
                    bmi
                         children smoker
                                              region
                                                           charges
0
       19
             1
                27.900
                                0
                                     yes
                                          southwest
                                                      16884.924000
1
       18
             0
                33.770
                                          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
                                          southeast
                                                       3756.621600
6
       46
             1
                33.440
                                1
                                      no
                                           southeast
                                                       8240.589600
7
       37
             1
                27.740
                                3
                                          northwest
                                                       7281.505600
                                      no
8
       37
             0
                29.830
                                2
                                          northeast
                                                       6406.410700
                                      no
9
       60
             1
                25.840
                                0
                                          northwest
                                                      28923.136920
10
       25
             0
                26.220
                                0
                                      no
                                          northeast
                                                       2721.320800
11
                                                      27808.725100
       62
             1
                26.290
                                     yes
                                          southeast
                                                       1826.843000
12
       23
             0
                34.400
                                0
                                           southwest
                                      no
                39.820
                                0
13
       56
             1
                                          southeast
                                                      11090.717800
                                      no
14
       27
             0
                42.130
                                0
                                     yes
                                          southeast
                                                      39611.757700
                24.600
                                                       1837.237000
15
       19
             0
                                          southwest
                                1
                                      no
       52
                30.780
                                          northeast
                                                      10797.336200
16
             1
                                1
                                      no
             0
                                                      2395.171550
17
       23
                23.845
                                      no
                                          northeast
In [128]:
                                                                                                                                              M
n={"smoker":{"yes":1,"no":0}}
df=df.replace(n)
print(df)
                    bmi
                         children
                                   smoker
                                               region
                                                            charges
      age
           sex
0
                27.900
                                                       16884.924000
       19
             1
                                0
                                            southwest
                33.770
1
       18
                                        0
                                           southeast
                                                        1725.552300
                                1
2
       28
             0
                33.000
                                3
                                        0
                                            southeast
                                                        4449.462000
3
             0
                22.705
                                0
                                           northwest
                                                       21984.470610
       33
                                        0
4
       32
             0
                28.880
                                0
                                        0
                                                        3866.855200
                                           northwest
5
       31
             1
                25.740
                                0
                                        0
                                            southeast
                                                        3756.621600
6
       46
             1
                33.440
                                1
                                        0
                                            southeast
                                                        8240.589600
7
                27.740
                                3
       37
                                                        7281.505600
             1
                                        0
                                           northwest
8
       37
             0
                29.830
                                2
                                        0
                                           northeast
                                                        6406.410700
9
       60
                25.840
                                0
                                                       28923,136920
             1
                                        0
                                           northwest
10
             0
                                0
       25
                26.220
                                        0
                                                        2721,320800
                                            northeast
11
       62
                                0
                                                       27808.725100
             1
                26.290
                                        1
                                            southeast
12
       23
             0
                34.400
                                0
                                        0
                                            southwest
                                                        1826.843000
                39.820
                                        0
13
       56
             1
                                0
                                            southeast
                                                       11090.717800
14
       27
             a
                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
             0
                23.845
                                0
                                        0
                                            northeast
                                                        2395.171550
```

```
In [129]:
                                                                                                                 M
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
Out[129]:
▼ RandomForestClassifier
RandomForestClassifier()
In [130]:
                                                                                                                 M
rf=RandomForestClassifier()
'n_estimators':[10,25,30,50,100,200]}
In [132]:
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rf,param_grid=params,cv=2,scoring="accuracy")
grid_search.fit(x_train,y_train)
Out[132]:
           GridSearchCV
 ▶ estimator: RandomForestClassifier
      ▶ RandomForestClassifier
In [133]:
                                                                                                                 M
grid_search.best_score_
Out[133]:
0.5096254446536932
In [134]:
                                                                                                                 Ы
rf_best=grid_search.best_estimator_
print(rf_best)
RandomForestClassifier(max_depth=2, min_samples_leaf=100, n_estimators=10)
In [135]:
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[4],class_names=['1','0'],filled=True);
                                          x[0] \le 28.39
                                             gini = 0.5
                                          samples = 820
                                        value = [649, 648]
                                              class = 1
                 x[0] \le 26.46
                                                                      x[1] <= 0.5
                  gini = 0.496
                                                                      gini = 0.498
                                                                    samples = 507
                samples = 313
              value = [227, 273]
                                                                 value = [422, 375]
                    class = 0
                                                                       class = 1
                               gini = 0.488
     gini = 0.498
                                                         gini = 0.495
                                                                                    gini = 0.5
   samples = 213
                             samples = 100
                                                       samples = 221
                                                                                 samples = 286
 value = [159, 180]
                            value = [68, 93]
                                                     value = [189, 155]
                                                                              value = [233, 220]
                                                           class = 1
       class = 0
                                 class = 0
                                                                                     class = 1
```

```
In [136]:
                                                                                                                     M
from sklearn.tree import plot_tree
plt.figure(figsize=(70,30))
\verb|plot_tree(rf_best.estimators_[6], class_names=["1","0"], filled=True);|
                                            x[0] \le 33.517
                                              gini = 0.499
                                            samples = 833
                                          value = [684, 613]
                                                class = 1
                 x[0] \le 25.917
                                                                         x[1] <= 0.5
                                                                        gini = 0.477
                     gini = 0.5
                                                                       samples = 255
                  samples = 578
                value = [445, 459]
                                                                     value = [239, 154]
                     class = 0
                                                                          class = 1
                                                           gini = 0.437
      gini = 0.498
                                                                                      gini = 0.493
                                  gini = 0.5
    samples = 193
                               samples = 385
                                                          samples = 101
                                                                                    samples = 154
  value = [141, 162]
                             value = [304, 297]
                                                        value = [109, 52]
                                                                                  value = [130, 102]
        class = 0
                                  class = 1
                                                             class = 1
                                                                                        class = 1
In [137]:
                                                                                                                     M
rf_best.feature_importances_
Out[137]:
array([0.8019108, 0.1980892])
In [140]:
                                                                                                                     M
rf=RandomForestClassifier(random_state=0)
In [141]:
                                                                                                                     M
rf.fit(x_train,y_train)
Out[141]:
        RandomForestClassifier
RandomForestClassifier(random_state=0)
In [142]:
                                                                                                                     H
score=rf.score(x_test,y_test)
print(score)
0.5609756097560976
In [ ]:
                                                                                                                     Ы
```