# Problem statement:To predict the best fit data

## 1. Data Collection

```
In [3]:
```

```
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 [4]:

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

#### Out[4]:

	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 [5]:

df.head()

## Out[5]:

	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 [6]:
```

```
df.tail()
```

## Out[6]:

	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	ves	northwest	29141.3603

## In [7]:

df.shape

## Out[7]:

(1338, 7)

## In [8]:

df.describe

## Out[8]:

<boun< th=""><th>d met</th><th>hod NDFr</th><th>ame.descr</th><th>ibe of</th><th>age</th><th>sex</th><th>bmi children smo</th><th>ker region</th><th>charges</th></boun<>	d met	hod NDFr	ame.descr	ibe of	age	sex	bmi children smo	ker 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 x 7 columns]>

[1338 rows x 7 columns]>

## In [9]:

df.info

## Out[9]:

<box< td=""><td>d met</td><td>hod Data</td><td>Frame.info</td><td>of</td><td>age</td><td>sex</td><td>bmi children</td><td>smoker</td><td>region</td><td>charges</td></box<>	d met	hod Data	Frame.info	of	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			

```
In [10]:
df.isnull().any()
Out[10]:
            False
age
sex
            False
            False
bmi
            False
children
            False
smoker
region
            False
charges
            False
dtype: bool
In [11]:
df.isna().sum()
Out[11]:
age
sex
            0
bmi
            0
children
            0
smoker
            0
region
charges
dtype: int64
In [12]:
df['region'].value_counts()
Out[12]:
region
southeast
             364
southwest
             325
northwest
             325
northeast
             324
Name: count, dtype: int64
In [13]:
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
df
```

## Out[13]:

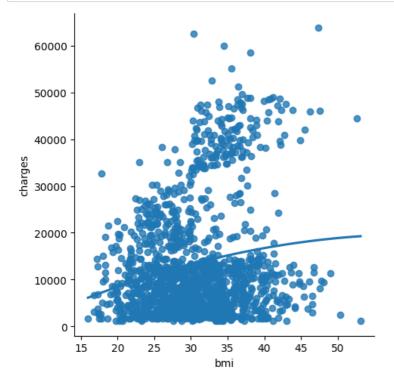
	age	sex	bmi	children	smoker	region	charges
0	19	1	27.900	0	yes	southwest	16884.92400
1	18	0	33.770	1	no	southeast	1725.55230
2	28	0	33.000	3	no	southeast	4449.46200
3	33	0	22.705	0	no	northwest	21984.47061
4	32	0	28.880	0	no	northwest	3866.85520
1333	50	0	30.970	3	no	northwest	10600.54830
1334	18	1	31.920	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

# 3. Data Visualization

1338 rows × 7 columns

#### In [15]:

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



## In [16]:

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

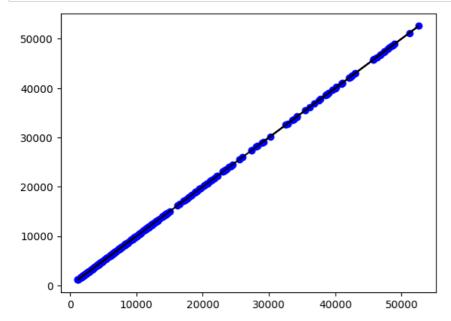
## In [17]:

```
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 [18]:
```

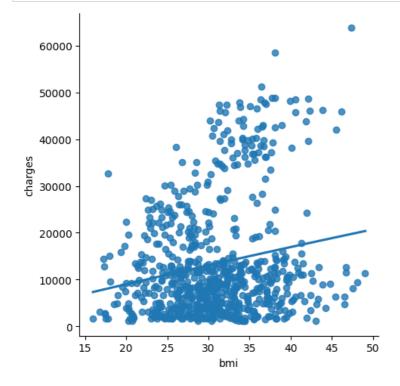
```
y_pred=lr.predict(x_test)
plt.scatter(x_test,y_test,color='b')
plt.plot(x_test,y_pred,color='k')
plt.show()
```



## working with subset of data

```
In [20]:
```

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



## In [21]:

```
df700.fillna(method='ffill',inplace=True)
```

```
In [22]:
```

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

#### In [23]:

```
df700.dropna(inplace=True)
```

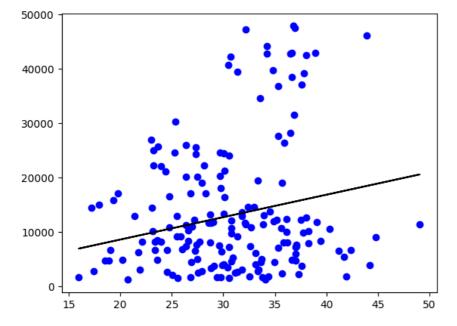
#### In [24]:

```
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))
```

#### 0.02505872132466691

#### In [26]:

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



## **Evaluation of Model**

#### In [27]:

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

## In [28]:

```
lr=LinearRegression()
lr.fit(x_train,y_train)
y_pred=lr.predict(x_test)
r2=r2_score(y_test,y_pred)
print(r2)
```

0.02505872132466691

## **Ridge Regression**

```
In [29]:
```

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

```
In [32]:
plt.figure(figsize=(10,10))
sns.heatmap(df700.corr(),annot=True)
plt.show()
______
ValueError
                                          Traceback (most recent call last)
Cell In[32], line 2
      1 plt.figure(figsize=(10,10))
   -> 2 sns.heatmap(df700.corr(),annot=True)
      3 plt.show()
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\frame.py:10059, in DataFrame.c
orr(self, method, min_periods, numeric_only)
  10057 cols = data.columns
  10058 idx = cols.copy()
> 10059 mat = data.to_numpy(dtype=float, na_value=np.nan, copy=False)
  10061 if method == "pearson":
            correl = libalgos.nancorr(mat, minp=min_periods)
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\frame.py:1838, in DataFrame.to
_numpy(self, dtype, copy, na_value)
   1836 if dtype is not None:
   1837
           dtype = np.dtype(dtype)
-> 1838 result = self._mgr.as_array(dtype=dtype, copy=copy, na_value=na_value)
   1839 if result.dtype is not dtype:
   1840
            result = np.array(result, dtype=dtype, copy=False)
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\internals\managers.py:1732, in
BlockManager.as_array(self, dtype, copy, na_value)
   1730
                arr.flags.writeable = False
   1731 else:
-> 1732
            arr = self._interleave(dtype=dtype, na_value=na_value)
   1733
            # The underlying data was copied within _interleave, so no need
            # to further copy if copy=True or setting na_value
   1736 if na_value is not lib.no_default:
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\internals\managers.py:1794, in
BlockManager._interleave(self, dtype, na_value)
   1792
            else:
   1793
                arr = blk.get_values(dtype)
-> 1794
            result[rl.indexer] = arr
   1795
            itemmask[rl.indexer] = 1
   1797 if not itemmask.all():
ValueError: could not convert string to float: 'yes'
<Figure size 1000x1000 with 0 Axes>
In [33]:
features=df.columns[0:1]
target=df.columns[-1]
In [34]:
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 [35]:
```

```
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 [36]:

```
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 0.08490538609860199

#### In [37]:

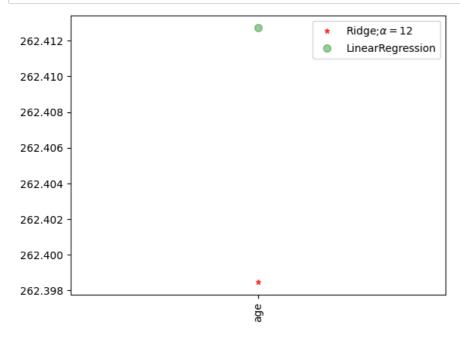
```
plt.figure(figsize=(10,10))
```

### Out[37]:

<Figure size 1000x1000 with 0 Axes>
<Figure size 1000x1000 with 0 Axes>

#### In [39]:

```
plt.plot(features,ridgeReg.coef_,alpha=0.7,linestyle='none',marker="*",markersize=5,color='red',label=r'Ridge;$\alpha=12$
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()
```



## **Lasso Regression**

```
In [40]:
```

```
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 0.09109639395809044 The test score for lasso model is 0.08490704421828055

#### In [41]:

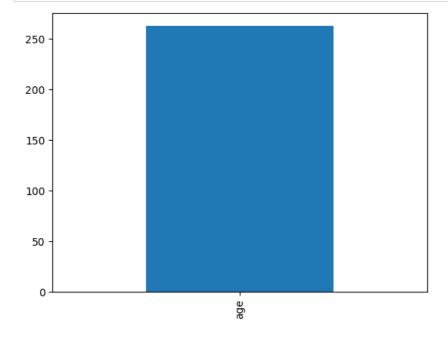
```
plt.figure(figsize=(10,10))
```

#### Out[41]:

```
<Figure size 1000x1000 with 0 Axes>
<Figure size 1000x1000 with 0 Axes>
```

#### In [42]:

```
pd.Series(lasso.coef_, features).sort_values(ascending = True).plot(kind = "bar")
plt.show()
```



## In [43]:

```
from sklearn.linear_model import LassoCV
```

## In [44]:

```
#using the linear cv model
from sklearn.linear_model import RidgeCV
#cross validation
ridge_cv=RidgeCV(alphas =[0.0001,0.001,0.01,1,1,10]).fit(x_train,y_train)
#score
print(ridge_cv.score(x_train,y_train))
print(ridge_cv.score(x_test,y_test))
```

0.09109639711159612

0.08490538609885023

#### In [45]:

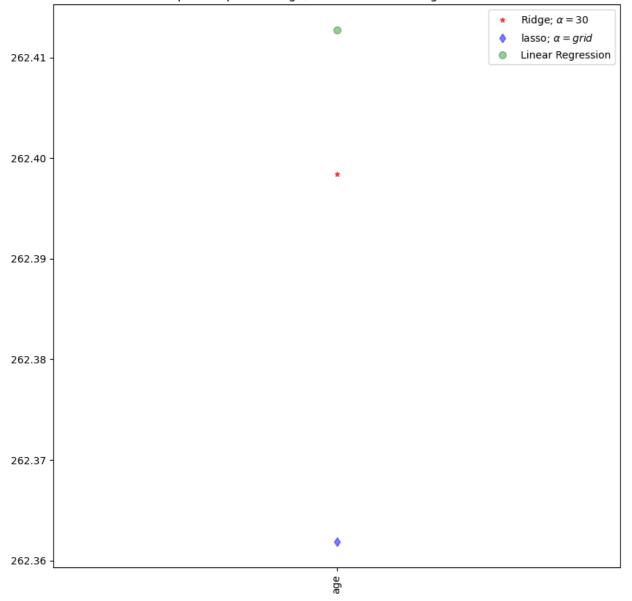
```
#using the Linear cv model
from sklearn.linear_model import LassoCV
#cross validation
lasso_cv=LassoCV(alphas =[0.0001,0.001,0.01,1,1,10]).fit(x_train,y_train)
#score
print(lasso_cv.score(x_train,y_train))
print(lasso_cv.score(x_test,y_test))
```

0.09109639395809044 0.08490704421828055

#### In [46]:

```
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=30:
#add plot for lasso regression
plt.plot(lasso_cv.coef_,alpha=0.5,linestyle='none',marker='d',markersize=6,color='blue',label=r'lasso; $\alpha=\text{grid}$')
#add plot for linear model
plt.plot(features,lr.coef_,alpha=0.4,linestyle='none',marker='o',markersize=7,color='green',label='Linear Regression')
#rotate axis
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



In [47]:

## **Elastic Net Regression**

```
from sklearn.linear_model import ElasticNet

In [48]:
el=ElasticNet()
el.fit(x_train,y_train)
print(el.coef_)
print(el.intercept_)

[261.74450967]
3115.0831774262424

In [49]:
y_pred_elastic=el.predict(x_train)

In [50]:
```

135077142.70714515

print(mean\_squared\_error)

```
In [51]:
```

```
el=ElasticNet()
el.fit(x_train,y_train)
print(el.score(x_train,y_train))
```

0.09109580670592365

## **Logistic Regression**

```
In [52]:
import numpy as np
import pandas as pd
from sklearn.linear_model import LogisticRegression
from sklearn.preprocessing import StandardScaler
```

mean\_squared\_error=np.mean((y\_pred\_elastic-y\_train)\*\*2)

```
In [53]:
```

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

Out[53]:

	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 [54]:
```

df.shape

#### Out[54]:

(1338, 7)

#### In [55]:

```
pd.set_option('display.max_rows',10000000000)
pd.set_option('display.max_columns',10000000000)
pd.set_option('display.width',95)
```

#### In [56]:

```
print('This Dataset has %d rows and %d columns'%(df.shape))
```

This Dataset has 1338 rows and 7 columns

#### In [57]:

df.head()

#### Out[57]:

	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 [58]:

df.tail()

#### Out[58]:

	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	ves	northwest	29141 3603

## In [59]:

```
df.describe
83
       48
          temale 41.230
                                       no northwest
                                                      11033.661700
84
       37
          female
                  34.800
                                 2
                                      yes
                                            southwest
                                                      39836.519000
85
       45
            male
                  22.895
                                           northwest 21098.554050
                                      yes
                                      yes northwest 43578.939400
86
       57
          female
                  31.160
                                 0
87
       56
           female
                  27.200
                                 0
                                           southwest
                                                      11073.176000
                                       no
       46
          female 27.740
88
                                 0
                                       no northwest
                                                       8026,666600
89
       55
          female 26.980
                                 0
                                       no northwest 11082.577200
90
       21
          female
                  39.490
                                 0
                                       no southeast
                                                       2026.974100
                  24.795
                                           northwest 10942.132050
91
       53
          female
                                 1
                                       no
92
       59
            male
                  29.830
                                      yes northeast 30184.936700
93
       35
            male
                  34.770
                                 2
                                       no northwest
                                                      5729.005300
94
       64
          female
                  31.300
                                 2
                                           southwest
                                                      47291.055000
                                      yes
95
          female
       28
                  37,620
                                 1
                                           southeast
                                                      3766.883800
                                       no
96
       54
          female
                  30.800
                                 3
                                           southwest 12105.320000
97
       55
                                 0
            male
                  38.280
                                       no
                                           southeast
                                                      10226.284200
98
       56
            male
                  19.950
                                 0
                                      yes
                                           northeast
                                                      22412.648500
99
       38
            male 19.300
                                           southwest 15820.699000
                                      yes
100
                                 0
       41
                  31.600
                                                       6186.127000
          female
                                       no
                                           southwest
101
       30
             male
                  25.460
                                 0
                                       no
                                           northeast
                                                       3645.089400
102
       18
          female 30.115
                                           northeast 21344.846700
                                       no
```

```
In [60]:
```

```
df.info
                                                                                                                                      Out[60]:
                                                                                                                                      <bound method DataFrame.info of</pre>
                                                             bmi children smoker
                                           age
                                                                                         region
                                                                                                       charges
                                                   sex
        19
            female 27.900
                                            yes
                                                 southwest
                                                             16884.924000
1
        18
              male
                     33.770
                                      1
                                                 southeast
                                                               1725.552300
                                             no
2
        28
              male
                     33.000
                                      3
                                             no
                                                 southeast
                                                               4449.462000
                     22.705
3
        33
              male
                                      0
                                                 northwest
                                                              21984.470610
                                             no
4
                     28.880
                                      a
        32
              male
                                             no
                                                 northwest
                                                               3866.855200
5
        31
            female
                     25.740
                                      0
                                                               3756.621600
                                             no
                                                 southeast
6
        46
            female
                     33,440
                                                               8240.589600
                                      1
                                             no
                                                 southeast
7
        37
            female
                     27.740
                                      3
                                                 northwest
                                                               7281.505600
8
        37
                     29.830
                                      2
                                                               6406.410700
              male
                                                 northeast
                                             no
9
        60
            female
                     25.840
                                      0
                                             no
                                                 northwest
                                                             28923.136920
10
        25
              male
                     26.220
                                      0
                                                 northeast
                                                               2721.320800
                                             no
                     26.290
                                      0
11
        62
            female
                                            yes
                                                 southeast
                                                              27808.725100
12
        23
              male
                     34.400
                                      0
                                                 southwest
                                                               1826.843000
                                             no
13
        56
                     39.820
                                      a
            female
                                                 southeast
                                                             11090.717800
                                             nο
14
        27
              male
                     42.130
                                      0
                                                 southeast
                                                              39611.757700
                                            yes
15
        19
                     24.600
                                      1
                                                 southwest
                                                              1837.237000
              male
                                             no
16
        52
            female
                     30.780
                                      1
                                             no
                                                 northeast
                                                              10797.336200
In [61]:
df.isnull().sum()
Out[61]:
age
             0
sex
bmi
             0
children
             0
smoker
             0
             0
region
charges
             a
dtype: int64
In [62]:
convert={"smoker":{"yes":1,"no":0}}
df=df.replace(convert)
df
   35
       19
            male 20.425
                               0
                                          northwest
                                                    1625.433750
                  32.965
                                          northwest 15612.193350
   36
       62
          female
       26
                  20.800
                                          southwest
                                                    2302.300000
   37
            male
                                       0
   38
       35
            male
                  36.670
                                          northeast 39774.276300
   39
       60
             male
                  39.900
                                          southwest 48173.361000
   40
       24
           female
                  26.600
                                          northeast
                                                    3046.062000
   41
       31
          female
                 36.630
                                       0
                                          southeast
                                                    4949.758700
   42
       41
             male
                  21.780
                                       0
                                          southeast
                                                    6272.477200
   43
       37
           female
                  30.800
                               2
                                       0
                                          southeast
                                                    6313.759000
       38
                  37.050
                                                    6079.671500
             male
                                          northeast
                 37.300
                               0
                                                   20630.283510
   45
       55
            male
                                       0
                                          southwest
                               2
   46
        18
           female
                  38.665
                                       0
                                          northeast
                                                    3393.356350
   47
       28
          female 34.770
                               0
                                          northwest
                                                    3556.922300
```

```
In [63]:
```

```
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
df
```

## Out[63]:

	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

#### In [64]:

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

## Out[64]:

	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

## In [65]:

```
features_matrix=df.iloc[:,0:4]
```

## In [66]:

```
target_vector=df.iloc[:,-3]
```

#### In [67]:

```
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))
```

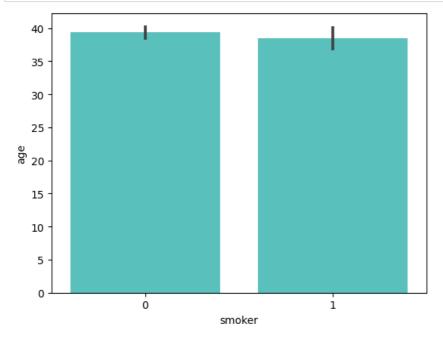
The Feature Matrix has 1338 Rows and 4 columns(s)
The Target Matrix has 1338 Rows and 1 columns(s)

## In [68]:

```
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [69]:
```

```
sns.barplot(x='smoker', y='age', data=df, color="mediumturquoise")
plt.show()
```



#### In [70]:

features\_matrix\_standardized=StandardScaler().fit\_transform(features\_matrix)

#### In [71]:

algorithm=LogisticRegression(max\_iter=10000)

## In [72]:

 $Logistic\_Regression\_Model = algorithm.fit (features\_matrix\_standardized, target\_vector)$ 

#### In [73]:

observation=[[1,0,0.99539,-0.0588]]

#### In [74]:

```
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 [75]:

```
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]

#### In [76]:

```
print(" " "The Model says the probability of the observation we passed belonging to class['0'] Is %s" " "%(algorithm.pred
print()
```

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

#### In [77]:

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

```
In [78]:
```

```
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))
```

#### 0.746268656716418

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

## **Decision tree**

## In [79]:

```
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 [81]:

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

#### 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

#### In [82]:

```
df.shape
```

#### Out[82]:

(1338, 7)

## In [83]:

```
df.isnull().any()
```

## Out[83]:

age	False
sex	False
bmi	False
children	False
smoker	False
region	False
charges	False
dtype: bool	

In [84]:

```
df['region'].value_counts()
Out[84]:
region
southeast
               364
southwest
               325
northwest
               325
               324
northeast
Name: count, dtype: int64
In [85]:
convert={"sex":{"female":1,"male":0}}
df=df.replace(convert)
df
   51
        21
              1 33.630
                              2
                                         northwest
                                                     3579.828700
                                      no
                              1
   52
        48
              0 28.000
                                     yes
                                          southwest 23568.272000
   53
        36
              0 34.430
                              0
                                          southeast
                                                    37742.575700
        40
              1 28.690
                              3
                                                     8059.679100
                                          northwest
                                      no
        58
              0 36.955
                              2
                                          northwest 47496.494450
   55
                                     ves
                              2
        58
              1 31.825
                                                    13607.368750
   56
                                      no
                                          northeast
   57
        18
              0 31.680
                              2
                                          southeast
                                                    34303.167200
                                     yes
   58
        53
              1 22.880
                                                    23244.790200
                                     yes
                                          southeast
        34
              1 37.335
                              2
                                                     5989.523650
   59
                                          northwest
                                      no
              0 27 360
                              3
                                                     8606 217400
        43
                                          northeast
   60
                                      nο
   61
        25
              0 33.660
                              4
                                      no
                                          southeast
                                                     4504.662400
   62
        64
              0 24.700
                                          northwest 30166.618170
                                      no
   63
        28
              1 25.935
                                         northwest
                                                     4133.641650
                                      no
In [86]:
convert={"smoker":{"yes":1,"no":0}}
df=df.replace(convert)
df
Out[86]:
                        children
            sex
                   bmi
                                 smoker
                                             region
                                                         charges
       age
                27 900
                                          southwest
    0
        19
              1
                              0
                                                    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
                                         northwest 21984.470610
              0 28.880
                                                     3866.855200
        32
                              0
                                          northwest
                                       0
                              0
    5
        31
              1 25.740
                                       0
                                          southeast
                                                     3756.621600
    6
        46
              1 33.440
                                       0
                                          southeast
                                                     8240.589600
        37
              1 27.740
                                                     7281.505600
                                       0
                                          northwest
                              2
    8
        37
              0 29 830
                                       0
                                                     6406 410700
                                          northeast
    9
        60
              1 25.840
                              Λ
                                       0 northwest 28923.136920
In [87]:
x=["bmi","children"]
y=["Yes","No"]
all inputs=df[x]
all_classes=df["sex"]
In [88]:
(x\_train, x\_test, y\_train, y\_test) = train\_test\_split(all\_inputs, all\_classes, test\_size = 0.03)
```

```
In [89]:
clf=DecisionTreeClassifier(random_state=0)
In [90]:
clf.fit(x_train,y_train)
Out[90]:
          DecisionTreeClassifier
DecisionTreeClassifier(random_state=0)
In [91]:
score=clf.score(x_test,y_test)
print(score)
0.5121951219512195
Random Forest
In [92]:
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt ,seaborn as sns
In [93]:
df=pd.read_csv(r"C:\Users\Svijayalakshmi\Downloads\insurance.csv")
df
Out[93]:
             sex
                    bmi
                        children smoker
                                           region
                                                      charges
          female
                 27.900
                                                  16884.924000
       19
                                        southwest
                                    yes
                                                   1725.552300
       18
                 33.770
            male
                                         southeast
                                     no
                                                   4449.462000
   2
       28
            male
                 33.000
                              3
                                     no
                                         southeast
   3
       33
            male
                 22.705
                              0
                                         northwest
                                                 21984.470610
       32
                                                   3866.855200
            male 28.880
                                         northwest
                                     no
       31 female 25.740
                              0
                                                   3756.621600
                                         southeast
                                     no
       46
          female 33.440
                                                   8240.589600
                                     no
                                         southeast
       37
          female 27.740
                              3
                                         northwest
                                                   7281.505600
       37
            male 29.830
                                     no
                                         northeast
                                                   6406.410700
       60 female 25.840
                                     no northwest 28923.136920
In [94]:
df.shape
Out[94]:
(1338, 7)
In [95]:
df['region'].value_counts()
Out[95]:
region
              364
southeast
southwest
              325
              325
northwest
northeast
              324
Name: count, dtype: int64
```

```
In [96]:
df['bmi'].value_counts()
                                                                                                                                 Out[96]:
                                                                                                                                 bmi
32.300
           13
28.310
           9
30.495
            8
30.875
            8
            8
31,350
30.800
            8
34,100
            8
28.880
            8
33.330
            7
35.200
            7
25.800
            7
            7
32.775
27.645
            7
            7
32,110
38.060
            7
25.460
            7
            7
30.590
In [97]:
m={"sex":{"female":1,"male":0}}
df=df.replace(m)
print(df)
            sex
                    bmi
                         children smoker
      age
                                               region
                                                             charges
0
       19
                 27.900
                                                       16884.924000
                                 0
                                            southwest
              1
                                      ves
1
       18
              0
                 33.770
                                 1
                                            southeast
                                                         1725.552300
2
       28
              0
                 33.000
                                 3
                                            southeast
                                                         4449.462000
                                        no
3
       33
              a
                 22.705
                                 0
                                        no
                                            northwest
                                                        21984.470610
4
                 28.880
                                                         3866.855200
       32
              0
                                 0
                                            northwest
                                       no
5
                                 0
                                            southeast
       31
              1
                 25.740
                                        no
                                                         3756.621600
6
       46
              1
                 33.440
                                 1
                                            southeast
                                                         8240.589600
                                        no
7
       37
              1
                 27,740
                                 3
                                            northwest
                                                         7281,505600
                                        no
8
       37
              0
                 29.830
                                 2
                                        no
                                            northeast
                                                         6406.410700
9
       60
              1
                 25.840
                                 0
                                            northwest
                                                        28923.136920
                                        no
10
       25
              0
                 26.220
                                 0
                                       no
                                            northeast
                                                         2721.320800
                 26.290
                                 0
                                                        27808.725100
11
       62
              1
                                      yes
                                            southeast
                 34,400
                                 0
12
       23
              0
                                       no
                                            southwest
                                                        1826.843000
13
       56
              1
                 39.820
                                 0
                                            southeast
                                                        11090.717800
                                        no
       27
                 42.130
                                 0
14
              0
                                            southeast
                                                       39611.757700
                                      yes
15
       19
              0
                 24.600
                                 1
                                            southwest
                                                         1837.237000
16
       52
              1
                 30.780
                                 1
                                            northeast
                                                       10797.336200
                                        no
17
       23
              0
                 23.845
                                 0
                                        no
                                            northeast
                                                        2395.171550
In [98]:
n={"smoker":{"yes":1,"no":0}}
df=df.replace(n)
print(df)
            sex
                    bmi
                         children
                                    smoker
                                                region
                                                              charges
      age
0
       19
              1
                 27.900
                                 0
                                          1
                                             southwest
                                                         16884.924000
                 33.770
1
       18
              0
                                             southeast
                                                         1725.552300
                                 1
              a
                 33.000
2
       28
                                 3
                                          a
                                             southeast
                                                          4449.462000
3
       33
              0
                 22.705
                                 0
                                          0
                                             northwest
                                                         21984.470610
4
       32
              a
                 28.880
                                 a
                                          a
                                             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
                                             northeast
                                                          6406.410700
                                 a
9
       60
              1
                 25.840
                                          0
                                             northwest
                                                         28923.136920
10
       25
              0
                 26.220
                                 0
                                          0
                                             northeast
                                                          2721.320800
                 26.290
                                 0
                                             southeast 27808.725100
11
       62
              1
                                          1
12
       23
              0
                 34.400
                                 0
                                          0
                                             southwest
                                                          1826.843000
13
       56
              1
                 39.820
                                 0
                                          0
                                             southeast
                                                         11090.717800
       27
14
              0
                 42.130
                                 0
                                          1
                                             southeast
                                                         39611.757700
15
       19
              0
                 24.600
                                 1
                                             southwest
                                                         1837.237000
                 30.780
       52
                                 1
                                          a
                                             northeast 10797.336200
16
              1
17
       23
              0
                 23.845
                                 0
                                          0
                                             northeast
                                                         2395.171550
```

#### In [99]:

```
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
```

#### Out[99]:

```
r RandomForestClassifier
RandomForestClassifier()
```

#### In [100]:

```
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]}
```

## In [101]:

```
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[101]:

```
► GridSearchCV

► estimator: RandomForestClassifier

► RandomForestClassifier
```

#### In [102]:

```
grid_search.best_score_
```

#### Out[102]:

0.5204398029256787

#### In [103]:

```
rf_best=grid_search.best_estimator_
print(rf_best)
```

 $Random Forest Classifier (\verb|max_depth=3|, \verb|min_samples_leaf=200|, \verb|n_estimators=10|)$ 

#### In [104]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[4],class_names=['1','0'],filled=True)
Out[104]:
[Text(0.4, 0.833333333333334, 'x[1] <= 0.5\ngini = 0.5\nsamples = 830\nvalue = [634, 663]\nclass = 0'),
 Text(0.2, 0.5, 'gini = 0.498\nsamples = 356\nvalue = [265, 301]\nclass = 0'),
Text(0.6, 0.5, 'x[0] <= 29.595\ngini = 0.5\nsamples = 474\nvalue = [369, 362]\nclass = 1'),
 Text(0.4, 0.1666666666666666, 'gini = 0.497\nsamples = 209\nvalue = [151, 176]\nclass = 0'),
Text(0.8, 0.166666666666666, 'gini = 0.497\nsamples = 265\nvalue = [218, 186]\nclass = 1')]
```

x[1] <= 0.5gini = 0.5samples = 830value = [634, 663]class = 0

qini = 0.498samples = 356value = [265, 301]class = 0

 $x[0] \le 29.595$ gini = 0.5samples = 474value = [369, 362]class = 1

qini = 0.497samples = 209value = [151, 176]class = 0

gini = 0.497samples = 265value = [218, 186]class = 1

#### In [105]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(70,30))
plot_tree(rf_best.estimators_[6],class_names=["1","0"],filled=True)
```

## Out[105]:

```
[\text{Text}(0.4, 0.833333333333333333, 'x[1] <= 0.5 \text{ ngini} = 0.5 \text{ nsamples} = 815 \text{ nvalue} = [658, 639] \text{ nclass} = 1'),
 Text(0.2, 0.5, 'gini = 0.499\nsamples = 356\nvalue = [279, 306]\nclass = 0'),
Text(0.6, 0.5, 'x[0] <= 29.922\ngini = 0.498\nsamples = 459\nvalue = [379, 333]\nclass = 1'),
 Text(0.4, 0.1666666666666666, 'gini = 0.5\nsamples = 212\nvalue = [156, 166]\nclass = 0'),
Text(0.8, 0.166666666666666, 'gini = 0.49\nsamples = 247\nvalue = [223, 167]\nclass = 1')]
```

 $x[1] \le 0.5$ gini = 0.5samples = 815value = [658, 639]class = 1

gini = 0.499samples = 356value = [279, 306]class = 0

 $x[0] \le 29.922$ gini = 0.498samples = 459value = [379, 333]class = 1

gini = 0.5samples = 212value = [156, 166]class = 0

gini = 0.49samples = 247value = [223, 167]class = 1

```
In [106]:
rf_best.feature_importances_
Out[106]:
array([0.53685391, 0.46314609])
In [107]:
rf=RandomForestClassifier(random_state=0)
In [108]:
rf.fit(x_train,y_train)
Out[108]:
         RandomForestClassifier
RandomForestClassifier(random_state=0)
In [109]:
score=rf.score(x_test,y_test)
print(score)
0.4146341463414634
In [ ]:
```