# Problem Statement: "Evaluating Which model is best-fit for Insurance data

```
import numpy as np,pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [98]: df=pd.read\_csv(r"C:\Users\Y.Saranya\Downloads\insurance.csv")
df

#### Out[98]:

	age	sex	bmi	children	smoker	region	charges
0	19	female	27.900	0	yes	southwest	16884.92400
1	18	ma <b>l</b> e	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

# **Data cleaning and Preprocessing**

```
In [99]: 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
            1
                          1338 non-null
                                           object
                sex
                                           float64
            2
                          1338 non-null
                bmi
            3
                children 1338 non-null
                                           int64
            4
                smoker
                          1338 non-null
                                           object
            5
                region
                          1338 non-null
                                           object
            6
                                           float64
                charges
                          1338 non-null
           dtypes: float64(2), int64(2), object(3)
           memory usage: 73.3+ KB
In [100]: df.shape
Out[100]: (1338, 7)
In [101]: df.describe
Out[101]: <bound method NDFrame.describe of</pre>
                                                                      bmi children smoker
                                                     age
                                                             sex
           region
                       charges
           0
                      female 27.900
                                              0
                                                         southwest
                                                                    16884.92400
                  19
                                                    yes
           1
                  18
                        male
                              33.770
                                              1
                                                     no
                                                         southeast
                                                                      1725.55230
                                                                      4449.46200
           2
                  28
                        male 33.000
                                               3
                                                         southeast
                                                     no
           3
                  33
                        male 22.705
                                              0
                                                     no
                                                         northwest
                                                                    21984.47061
           4
                  32
                        male 28.880
                                                         northwest
                                                                      3866.85520
                                               0
                                                     no
                 . . .
                         . . .
                                  . . .
                                                    . . .
                                                                . . .
                                             . . .
           1333
                  50
                        male
                               30.970
                                               3
                                                     no
                                                         northwest
                                                                    10600.54830
           1334
                  18
                      female 31.920
                                                         northeast
                                                                      2205.98080
                                               0
                                                     no
                      female
                               36.850
           1335
                  18
                                               0
                                                     no
                                                         southeast
                                                                      1629.83350
           1336
                  21
                      female 25.800
                                               0
                                                         southwest
                                                                      2007.94500
                                                     no
                                                         northwest 29141.36030
           1337
                  61
                      female 29.070
                                                    yes
           [1338 rows x 7 columns]>
In [102]: df.isna().any()
Out[102]: age
                       False
           sex
                       False
           bmi
                       False
           children
                       False
           smoker
                       False
           region
                       False
           charges
                       False
           dtype: bool
```

```
Insurance - Jupyter Notebook
In [103]:
           #To finding null values
           df.isnull().sum()
Out[103]: age
                        0
           sex
                        0
           bmi
                        0
           children
                        0
           smoker
                        0
           region
                        0
           charges
                        0
           dtype: int64
In [104]: df.head
Out[104]: <bound method NDFrame.head of</pre>
                                                                    bmi children smoker
                                                  age
                                                           sex
           region
                        charges
                       female
                   19
                               27.900
                                                0
                                                           southwest
                                                                       16884.92400
           0
                                                      yes
           1
                   18
                         male
                               33.770
                                                1
                                                       no
                                                           southeast
                                                                        1725.55230
                         male 33.000
                                                           southeast
                                                                        4449.46200
           2
                   28
                                                3
                                                       no
                         male 22.705
           3
                   33
                                                0
                                                           northwest 21984.47061
                                                       no
           4
                   32
                         male
                               28.880
                                                0
                                                       no
                                                           northwest
                                                                        3866.85520
                          . . .
                  . . .
                                   . . .
                                                      . . .
                                                                  . . .
           . . .
                   50
                         male
                                30.970
                                                                       10600.54830
           1333
                                                3
                                                       no
                                                           northwest
           1334
                   18
                       female 31.920
                                                0
                                                           northeast
                                                                        2205.98080
                                                       no
           1335
                   18
                       female
                               36.850
                                                           southeast
                                                                        1629.83350
                                                0
                                                       no
                       female 25.800
           1336
                   21
                                                0
                                                       no
                                                           southwest
                                                                        2007.94500
           1337
                   61
                       female 29.070
                                                0
                                                           northwest
                                                                       29141.36030
                                                      yes
           [1338 rows x 7 columns]>
In [105]: df.tail
Out[105]: <bound method NDFrame.tail of
                                                  age
                                                                    bmi
                                                                         children smoker
                                                           sex
           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
                                                           southeast
                                                                        4449.46200
                                                       no
           3
                   33
                               22.705
                         male
                                                0
                                                       no
                                                           northwest
                                                                       21984.47061
           4
                   32
                         male
                                28.880
                                                0
                                                       no
                                                           northwest
                                                                        3866.85520
                  . . .
                                               . . .
                                                      . . .
                                                                  . . .
                         male
                               30.970
                                                3
                                                                       10600.54830
           1333
                   50
                                                       no
                                                           northwest
                               31.920
           1334
                   18
                       female
                                                0
                                                       no
                                                           northeast
                                                                        2205.98080
           1335
                   18
                       female
                               36.850
                                                0
                                                           southeast
                                                                        1629.83350
                                                       no
           1336
                   21
                       female
                                25.800
                                                0
                                                       no
                                                           southwest
                                                                        2007.94500
```

yes

northwest

29141.36030

localhost:8888/notebooks/Desktop/Insurance.ipynb#Random-forest

1337

61

[1338 rows x 7 columns]>

female 29.070

```
In [106]: convert={"sex":{"female":0,"male":1}}
    df=df.replace(convert)
    df
```

# Out[106]:

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

1338 rows × 7 columns

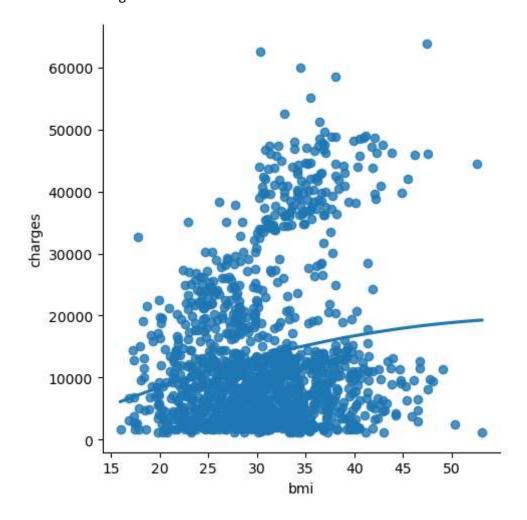
```
In [107]: convert={'smoker':{"yes":1,"no":0}}
    df=df.replace(convert)
    df
```

# Out[107]:

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

In [108]: sb.lmplot(x="bmi",y="charges",data=df,order=2,ci=None)

Out[108]: <seaborn.axisgrid.FacetGrid at 0x1d34af76110>



In [109]: df.drop("charges",axis=1)

# Out[109]:

	age	sex	bmi	children	smoker	region
0	19	0	27.900	0	1	southwest
1	18	1	33.770	1	0	southeast
2	28	1	33.000	3	0	southeast
3	33	1	22.705	0	0	northwest
4	32	1	28.880	0	0	northwest
1333	50	1	30.970	3	0	northwest
1334	18	0	31.920	0	0	northeast
1335	18	0	36.850	0	0	southeast
1336	21	0	25.800	0	0	southwest
1337	61	0	29.070	0	1	northwest

1338 rows × 6 columns

```
In [110]: sex={"sex":{"female":0,"male":1}}
    df=df.replace(sex)
    df
```

# Out[110]:

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

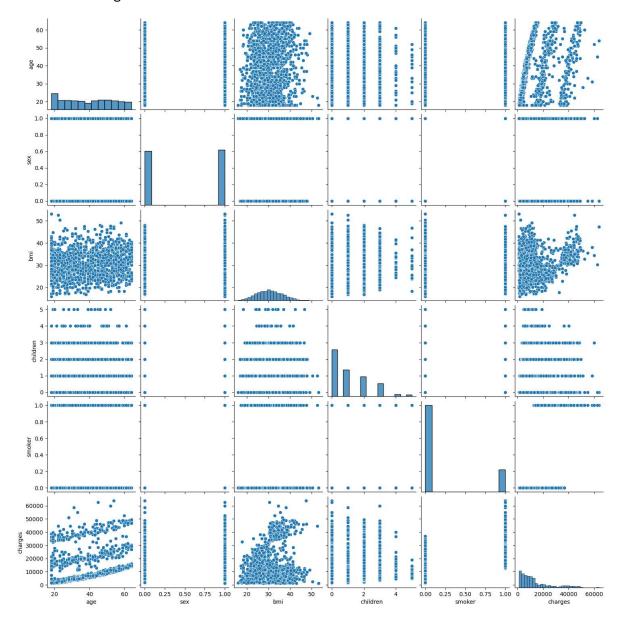
```
In [111]: smoker={"smoker":{"yes":1,"no":0}}
df=df.replace(smoker)
df
```

# Out[111]:

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

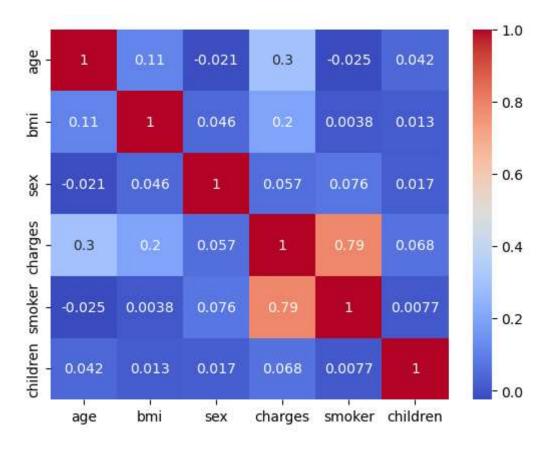
In [112]: #Data visualization
sns.pairplot(df)

Out[112]: <seaborn.axisgrid.PairGrid at 0x1d34fa0ff10>



```
In [113]: #heat map for data
    columns=df[['age','bmi','sex','charges','smoker','children']]
    subset=columns.corr()
    sns.heatmap(subset,annot=True,cmap='coolwarm')
```

Out[113]: <Axes: >



```
In [114]: #feature scaling or trainning our model
    from sklearn.model_selection import train_test_split
        X=df[['age','bmi','sex','charges','children']]
        y=df['smoker']
        x_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.30,random_state
```

# **Data modelling**

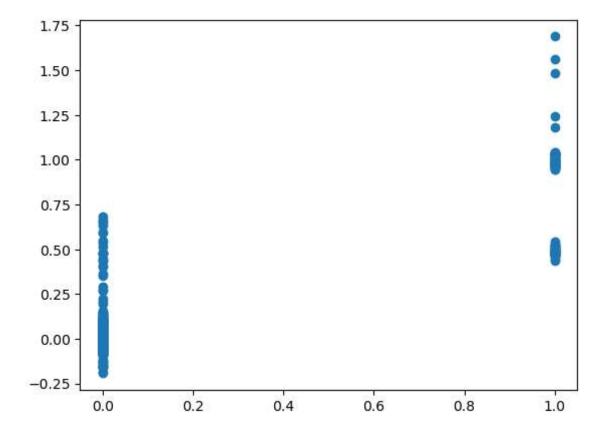
```
In [115]: #Now we are calculating our data fit for linear regression model
    from sklearn.linear_model import LinearRegression
    lr=LinearRegression()
    lr.fit(x_train,y_train)
```

#### 0.42484839188702184

#### Out[116]:

	coefficient
age	-0.007772
bmi	-0.010035
sex	0.019906
charges	0.000030
children	-0.017475

#### Out[117]: <matplotlib.collections.PathCollection at 0x1d34c9ef310>



Out[118]: 'In above linear regression model our insurance data is not fitted accuratly.'\n so now we are on logistic regression model'

```
In [119]: #importing libraries& dropping null values
          x=np.array(df['charges']).reshape(-1,1)
          y=np.array(df['smoker']).reshape(-1,1)
          df.dropna(inplace=True)
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=
          from sklearn.linear_model import LogisticRegression
          lg=LogisticRegression()
In [120]: | x=df[features].values
          y=df[target].values
          NameError
                                                     Traceback (most recent call last)
          Cell In[120], line 1
          ----> 1 x=df[features].values
                2 y=df[target].values
          NameError: name 'features' is not defined
In [121]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.5)
In [122]: | a=LinearRegression()
          a.fit(x train,y train)
Out[122]:
          ▼ LinearRegression
           LinearRegression()
In [123]: print(a.score(x test,y test))
          0.6632708523988279
In [124]: | a=LinearRegression()
          a.fit(x_train,y_train)
          train_score_a=a.score(x_train,y_train)
          test score a=a.score(x test,y test)
          print("\nLinearModel\n")
          print("The train score for lr model is {}".format(train_score_a))
          print("The train score for lr model is {}".format(test score a))
          LinearModel
          The train score for 1r model is 0.5683171218495688
          The train score for lr model is 0.6632708523988279
```

# **Decision tree**

```
In [125]: import numpy as np
    import pandas as pd
    from sklearn.tree import DecisionTreeClassifier
    from sklearn.model_selection import train_test_split
```

In [126]: df=pd.read\_csv(r"C:\Users\Y.Saranya\Downloads\insurance.csv")
 df

# Out[126]:

	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	ma <b>l</b> e	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 [127]: df["region"].value\_counts()

#### Out[127]: southeast

southeast 364 southwest 325 northwest 325 northeast 324

Name: region, dtype: int64

```
In [128]: convert={"sex":{"female":0,"male":1}}
    df=df.replace(convert)
    df
```

# Out[128]:

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

1338 rows × 7 columns

```
In [129]: convert={"smoker":{"yes":0,"no":1}}
    df=df.replace(smoker)
    df
```

# Out[129]:

		age	sex	bmi	children	smoker	region	charges
	0	19	0	27.900	0	1	southwest	16884.92400
	1	18	1	33.770	1	0	southeast	1725.55230
	2	28	1	33.000	3	0	southeast	4449.46200
	3	33	1	22.705	0	0	northwest	21984.47061
	4	32	1	28.880	0	0	northwest	3866.85520
1	333	50	1	30.970	3	0	northwest	10600.54830
1	334	18	0	31.920	0	0	northeast	2205.98080
1	335	18	0	36.850	0	0	southeast	1629.83350
1	336	21	0	25.800	0	0	southwest	2007.94500
1	337	61	0	29.070	0	1	northwest	29141.36030

```
In [130]: x=["age","sex","children","bmi","charges"]
y=["0","1"]
all_inputs=df[x]
all_classes=df["smoker"]

In [131]: x_train,X_test,y_train,y_test=train_test_split(all_inputs,all_classes,test_siz x_train.shape,x_test.shape

Out[131]: ((669, 5), (669, 1))
```

```
In [132]: s=DecisionTreeClassifier(random_state=20)
s.fit(x_train,y_train)
score=s.score(x_test,y_test)
print(score)
```

C:\Users\Y.Saranya\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarni
ng: X does not have valid feature names, but DecisionTreeClassifier was fitte
d with feature names
 warnings.warn(

```
ValueError
                                          Traceback (most recent call last)
Cell In[132], line 3
      1 s=DecisionTreeClassifier(random state=20)
      2 s.fit(x train,y train)
----> 3 score=s.score(x_test,y_test)
      4 print(score)
File ~\anaconda3\lib\site-packages\sklearn\base.py:649, in ClassifierMixin.sc
ore(self, X, y, sample_weight)
    624 """
    625 Return the mean accuracy on the given test data and labels.
    626
   (…)
            Mean accuracy of ``self.predict(X)`` wrt. `y`.
    646 """
    647 from .metrics import accuracy score
--> 649 return accuracy_score(y, self.predict(X), sample_weight=sample_weigh
t)
File ~\anaconda3\lib\site-packages\sklearn\tree\_classes.py:426, in BaseDecis
ionTree.predict(self, X, check_input)
    403 """Predict class or regression value for X.
   405 For a classification model, the predicted class for each sample in X
is
   (\ldots)
            The predicted classes, or the predict values.
    423
   424 """
    425 check is fitted(self)
--> 426 X = self._validate_X_predict(X, check input)
    427 proba = self.tree .predict(X)
    428 n samples = X.shape[0]
File ~\anaconda3\lib\site-packages\sklearn\tree\ classes.py:392, in BaseDecis
ionTree. validate X predict(self, X, check input)
    390 """Validate the training data on predict (probabilities)."""
    391 if check input:
--> 392
            X = self. validate data(X, dtype=DTYPE, accept sparse="csr", rese
t=False)
            if issparse(X) and (
    393
    394
                X.indices.dtype != np.intc or X.indptr.dtype != np.intc
    395
            ):
    396
                raise ValueError("No support for np.int64 index based sparse
matrices")
File ~\anaconda3\lib\site-packages\sklearn\base.py:569, in BaseEstimator. val
idate data(self, X, y, reset, validate separately, **check params)
    566
            out = X, y
    568 if not no val X and check params.get("ensure 2d", True):
            self._check_n_features(X, reset=reset)
    571 return out
File ~\anaconda3\lib\site-packages\sklearn\base.py:370, in BaseEstimator. che
ck n features(self, X, reset)
    367
            return
    369 if n_features != self.n_features_in_:
```

--> 370

raise ValueError(

```
371
                           f"X has {n_features} features, but {self.__class__.__name__}}
                           f"is expecting {self.n features in } features as input."
               372
               373
                       )
           ValueError: X has 1 features, but DecisionTreeClassifier is expecting 5 featu
           res as input.
In [135]:
          # random forest
           import numpy as np
           import pandas as pd
           import matplotlib.pyplot as plt,seaborn as sb
In [136]: | df=pd.read_csv(r"C:\Users\Y.Saranya\Downloads\insurance.csv")
In [137]: df.describe
Out[137]: <bound method NDFrame.describe of
                                                                          children smoker
                                                    age
                                                             sex
                                                                     bmi
           region
                       charges
                      female 27.900
           0
                  19
                                              0
                                                   yes
                                                         southwest 16884.92400
           1
                  18
                        male 33.770
                                              1
                                                    no
                                                         southeast
                                                                     1725.55230
           2
                  28
                        male
                              33.000
                                              3
                                                         southeast
                                                                     4449.46200
                                                    no
           3
                  33
                        male 22.705
                                                         northwest 21984.47061
           4
                  32
                        male 28.880
                                              0
                                                        northwest
                                                                     3866.85520
                                                    no
           1333
                  50
                        male 30.970
                                              3
                                                         northwest
                                                                    10600.54830
                                                    no
                                                         northeast
           1334
                  18
                      female 31.920
                                              0
                                                    no
                                                                     2205.98080
           1335
                  18
                      female 36.850
                                                    no
                                                         southeast
                                                                     1629.83350
           1336
                      female
                              25.800
                  21
                                                    no
                                                         southwest
                                                                     2007.94500
                      female 29.070
           1337
                  61
                                                   yes
                                                        northwest 29141.36030
           [1338 rows x 7 columns]>
In [138]: df.info
Out[138]: <bound method DataFrame.info of</pre>
                                                           sex
                                                                   bmi
                                                                        children smoker
                                                  age
           region
                       charges
           0
                  19
                      female 27.900
                                              0
                                                   yes
                                                         southwest
                                                                    16884.92400
           1
                  18
                              33.770
                                              1
                        male
                                                    no
                                                         southeast
                                                                     1725.55230
           2
                  28
                        male
                              33.000
                                              3
                                                         southeast
                                                                     4449.46200
                                                    no
           3
                        male 22.705
                  33
                                              0
                                                    no
                                                         northwest
                                                                    21984.47061
           4
                  32
                        male 28.880
                                              0
                                                         northwest
                                                                     3866.85520
                                                    no
           . . .
                 . . .
                         . . .
                                  . . .
                                            . . .
                                                    . . .
                                                               . . .
                        male 30.970
                                                         northwest 10600.54830
           1333
                  50
                                              3
                                                    no
           1334
                     female 31.920
                                                        northeast
                                                                     2205.98080
                  18
                                              0
                                                    no
           1335
                  18
                      female 36.850
                                                         southeast
                                                                     1629.83350
                                              0
                                                    no
                      female 25.800
                                                         southwest
           1336
                  21
                                              0
                                                                     2007.94500
                                                    no
           1337
                      female 29.070
                                                        northwest 29141.36030
                                                   yes
           [1338 rows x 7 columns]>
```

```
In [139]: df.isna().any()
Out[139]: age
                          False
                          False
            sex
            bmi
                          False
            children
                          False
            smoker
                          False
            region
                          False
            charges
                          False
            dtype: bool
In [140]: df.shape
Out[140]: (1338, 7)
            convert={"sex":{"female":0,"male":1}}
In [141]:
            df=df.replace(convert)
            df
Out[141]:
                                   children smoker
                                                       region
                                                                  charges
                   age
                       sex
                         0 27.900
                0
                    19
                                          0
                                                     southwest
                                                              16884.92400
                                                yes
                1
                    18
                          1 33.770
                                          1
                                                     southeast
                                                                1725.55230
                                                 no
                2
                    28
                         1 33.000
                                          3
                                                     southeast
                                                                4449.46200
                                                 no
                3
                    33
                         1 22.705
                                          0
                                                              21984.47061
                                                     northwest
                4
                    32
                            28.880
                                          0
                         1
                                                 no
                                                     northwest
                                                                3866.85520
               ...
                    ...
                                         ...
                                                 ...
             1333
                    50
                            30.970
                                          3
                                                     northwest
                                                               10600.54830
                         1
                                                 no
             1334
                    18
                                          0
                         0 31.920
                                                     northeast
                                                                2205.98080
                                                 no
             1335
                    18
                         0 36.850
                                          0
                                                     southeast
                                                                1629.83350
             1336
                    21
                         0 25.800
                                          0
                                                 no
                                                     southwest
                                                                2007.94500
             1337
                         0 29.070
                                          0
                                                     northwest 29141.36030
                    61
                                                yes
            1338 rows × 7 columns
In [142]: df["region"].value_counts()
Out[142]: southeast
                           364
                           325
            southwest
            northwest
                           325
            northeast
                           324
            Name: region, dtype: int64
```

#### Out[143]:

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

```
In [144]: x=df.drop("smoker",axis=1)
    y=df["smoker"]
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.5)
    from sklearn.ensemble import RandomForestClassifier
```

- In [145]: #importing libraries& fittingdata
   from sklearn.ensemble import RandomForestClassifier
   rfc=RandomForestClassifier()
   rfc.fit(x\_train,y\_train)
- Out[145]: 

  \* RandomForestClassifier

  RandomForestClassifier()
- In [146]: params={"max\_depth":[1,23,4,56,85],"min\_samples\_leaf":[4,6,8,10,12],"n\_estimat
- In [147]: #for finding optimal parameter values we are importing GridSearchCv
  from sklearn.model\_selection import GridSearchCV
  grid\_search=GridSearchCV(estimator=rfc,param\_grid=params,cv=2,scoring="accurac")

```
In [148]: grid_search.fit(x_train,y_train)
Out[148]:
                        GridSearchCV
            ▶ estimator: RandomForestClassifier
                 ▶ RandomForestClassifier
In [149]:
          grid_search.best_score_
          rf_best=grid_search.best_estimator_
          rf_best
Out[149]:
                                     RandomForestClassifier
           RandomForestClassifier(max_depth=23, |min_samples_leaf=4, n_estimators=42)
In [150]: from sklearn.tree import plot tree
          plt.figure(figsize=(80,40))
          plot_tree(rf_best.estimators_[4],feature_names=X.columns,class_names=['1','0']
          IndexError
                                                     Traceback (most recent call las
          t)
          Cell In[150], line 3
                1 from sklearn.tree import plot tree
                2 plt.figure(figsize=(80,40))
          ----> 3 plot tree(rf best.estimators [4], feature names=X.columns, class nam
          es=['1','0'],filled=True)
          File ~\anaconda3\lib\site-packages\sklearn\tree\_export.py:194, in plot tr
          ee(decision_tree, max_depth, feature_names, class_names, label, filled, im
          purity, node ids, proportion, rounded, precision, ax, fontsize)
              179 check is fitted(decision tree)
              181 exporter = MPLTreeExporter(
              182
                      max depth=max depth,
              183
                      feature_names=feature_names,
             (…)
              192
                      fontsize=fontsize,
In [151]: |#accurate score for random forest
          score=rfc.score(x_test,y_test)
          print(score)
          0.9611360239162929
          """In above all three models we more accuracy in LINEAR REGRESSION
In [152]:
           with respect to other two models"""
Out[152]: 'In above all three models we more accuracy in LINEAR REGRESSION\n with respc
          et to other two models'
```