problem statement :which model is suitable best for insurance dataset

```
In [ ]: import pandas as pd
    import numpy as np
    import matplotlib.pyplot as plt
    import seaborn as sns
```

data collection

Read the data

In [2]: df=pd.read_csv(r"C:\Users\rakesh\Downloads\insurance.csv")
df

Out[2]:

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

1338 rows × 7 columns

DATA CLEANING AND PREPROCESSING

In [3]: df.head()

Out[3]:

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

In [4]: df.tail()

Out[4]:

		age	sex	bmi	children	smoker	region	charges
1	1333	50	male	30.97	3	no	northwest	10600.5483
1	1334	18	female	31.92	0	no	northeast	2205.9808
1	335	18	female	36.85	0	no	southeast	1629.8335
1	1336	21	female	25.80	0	no	southwest	2007.9450
1	1337	61	female	29.07	0	yes	northwest	29141.3603

In [5]: df.shape

Out[5]: (1338, 7)

```
In [6]: df.describe()
```

Out[6]:

	age	bmi	children	charges
count	1338.000000	1338.000000	1338.000000	1338.000000
mean	39.207025	30.663397	1.094918	13270.422265
std	14.049960	6.098187	1.205493	12110.011237
min	18.000000	15.960000	0.000000	1121.873900
25%	27.000000	26.296250	0.000000	4740.287150
50%	39.000000	30.400000	1.000000	9382.033000
75%	51.000000	34.693750	2.000000	16639.912515
max	64.000000	53.130000	5.000000	63770.428010

to find null values

```
In [7]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):

	(, •
#	Column	Non-N	Null Count	Dtype
0	age	1338	non-null	int64
1	sex	1338	non-null	object
2	bmi	1338	non-null	float64
3	children	1338	non-null	int64
4	smoker	1338	non-null	object
5	region	1338	non-null	object
6	charges	1338	non-null	float64
dtype	es: float6	4(2),	int64(2),	object(3)

memory usage: 73.3+ KB

To find Duplicate Values

```
In [72]: df.duplicated().sum()
Out[72]: 1
```

To find Unique Values

```
In [73]: df['age'].unique()
    df['children'].unique()
    df['bmi'].unique()
```

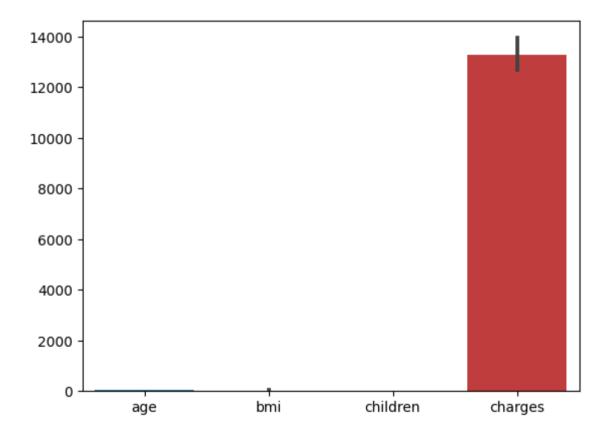
```
Traceback (most recent call last)
KeyError
File ~\AppData\Local\Programs\Pvthon\Pvthon311\Lib\site-packages\pandas\core\indexes\base.pv:3652, in Index.get loc
(self, key)
   3651 trv:
-> 3652
            return self. engine.get loc(casted key)
   3653 except KeyError as err:
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\ libs\index.pyx:147, in pandas. libs.index.I
ndexEngine.get loc()
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\ libs\index.pyx:176, in pandas. libs.index.I
ndexEngine.get loc()
File pandas\ libs\hashtable class helper.pxi:7080, in pandas. libs.hashtable.PyObjectHashTable.get item()
File pandas\ libs\hashtable class helper.pxi:7088, in pandas. libs.hashtable.PyObjectHashTable.get item()
KeyError: 'bni'
The above exception was the direct cause of the following exception:
                                          Traceback (most recent call last)
KevError
Cell In[73], line 3
      1 df['age'].unique()
      2 df['children'].unique()
----> 3 df['bni'].unique()
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\frame.py:3761, in DataFrame. getitem
(self, key)
   3759 if self.columns.nlevels > 1:
            return self. getitem multilevel(key)
-> 3761 indexer = self.columns.get_loc(key)
   3762 if is integer(indexer):
            indexer = [indexer]
   3763
File ~\AppData\Local\Programs\Python\Python311\Lib\site-packages\pandas\core\indexes\base.py:3654, in Index.get loc
(self, key)
   3652
            return self. engine.get loc(casted key)
   3653 except KeyError as err:
            raise KeyError(key) from err
-> 3654
```

```
3655 except TypeError:
3656  # If we have a listlike key, _check_indexing_error will raise
3657  # InvalidIndexError. Otherwise we fall through and re-raise
3658  # the TypeError.
3659  self._check_indexing_error(key)
```

KeyError: 'bni'

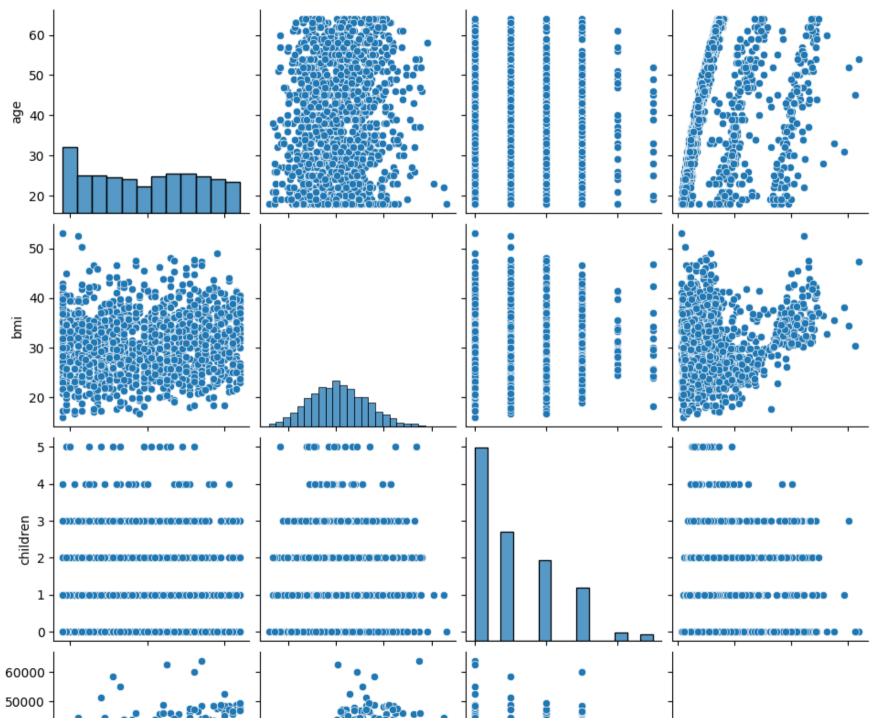


Out[9]: <Axes: >



```
In [10]: sns.pairplot(df)
```

Out[10]: <seaborn.axisgrid.PairGrid at 0x1547f574850>



children

charges

```
In [11]: df.columns
```

bmi

Out[11]: Index(['age', 'sex', 'bmi', 'children', 'smoker', 'region', 'charges'], dtype='object')

age

```
In [12]: t={"smoker":{"yes":1,"no":0}}
df=df.replace(t)
df
```

Out[12]:

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

1338 rows × 7 columns

```
In [13]: h={"sex":{"male":1,"female":0}}
    df=df.replace(h)
    df
```

Out[13]:

	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

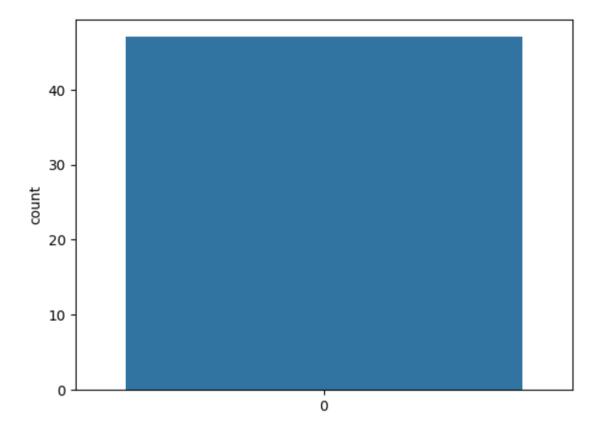
1338 rows × 7 columns

```
In [14]: |ho=df[['age', 'sex', 'bmi', 'children', 'smoker', 'charges']]
          plt.figure(figsize=(4,4))
          sns.heatmap(ho.corr(),annot=True)
Out[14]: <Axes: >
                                                           - 1.0
                          -0.021 0.11 0.042-0.025 0.3
                                                           - 0.8
                               0.046 0.017 0.076 0.057
                sex -0.021
                                                            0.6
               bmi - 0.11 0.046
                                     0.0130.0038 0.2
                                 1
           children -0.042 0.017 0.013
                                           0.00770.068
                                                            0.4
            smoker --0.0250.0760.00380.0077
                                                 0.79
                                                            0.2
                     0.3 0.057 0.2 0.068 0.79
           charges -
                                                   1
                                       children
                                             smoker
In [15]: x=df[['age', 'sex', 'bmi', 'children', 'smoker']]
         y=df['charges']
In [16]: x=df[['age', 'sex', 'bmi', 'children', 'smoker']]
         y=df['charges']
```

Data Visualize: Visualization the unique counts

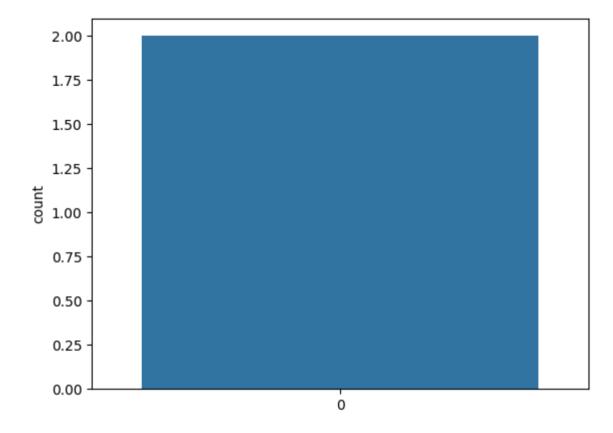
```
In [17]: sns.countplot(df['age'].unique())
```

Out[17]: <Axes: ylabel='count'>



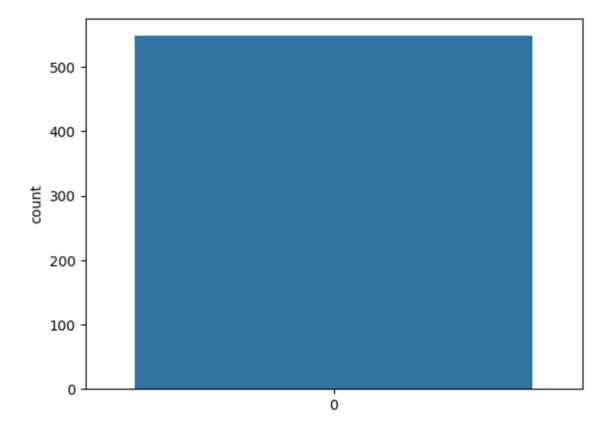
```
In [18]: sns.countplot(df['sex'].unique())
```

Out[18]: <Axes: ylabel='count'>

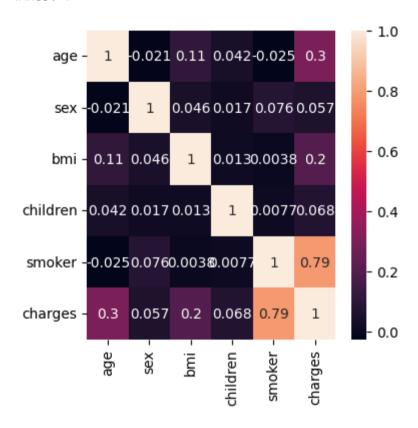


```
In [19]: sns.countplot(df['bmi'].unique())
```

Out[19]: <Axes: ylabel='count'>



```
In [20]: Insuranced=df[['age','sex','bmi','children','smoker','charges']]
    plt.figure(figsize=(4,4))
    sns.heatmap(Insuranced.corr(),annot=True)
Out[20]: <Axes: >
```



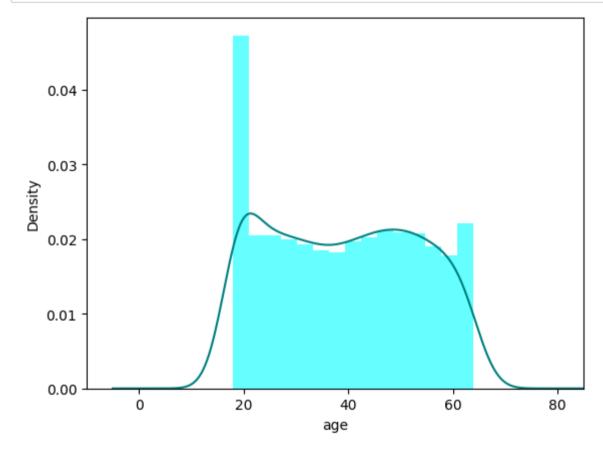
to find Mean and median

```
In [21]: print(df["age"].mean(skipna=True))
    print(df["age"].median(skipna=True))

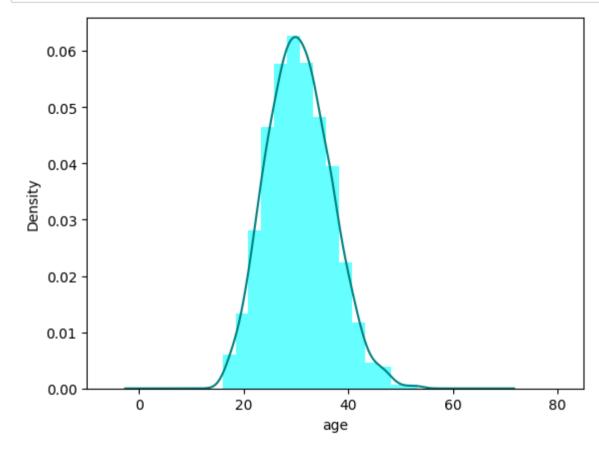
39.20702541106129
39.0
```

data visualization in histogram

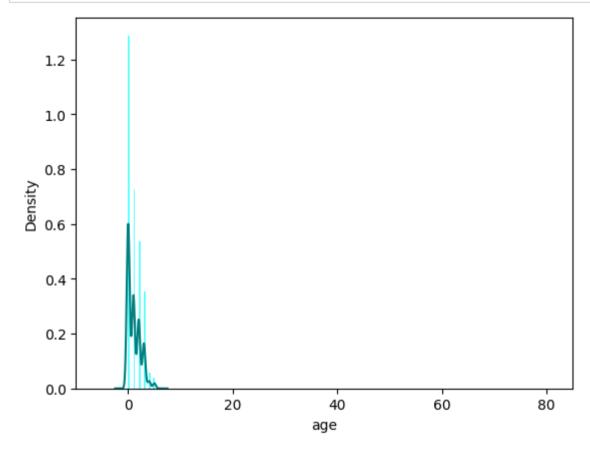
```
In [25]: ax=df["age"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
    df["age"].plot(kind='density',color='teal')
    ax.set(xlabel='age')
    plt.xlim(-10,85)
    plt.show()
```



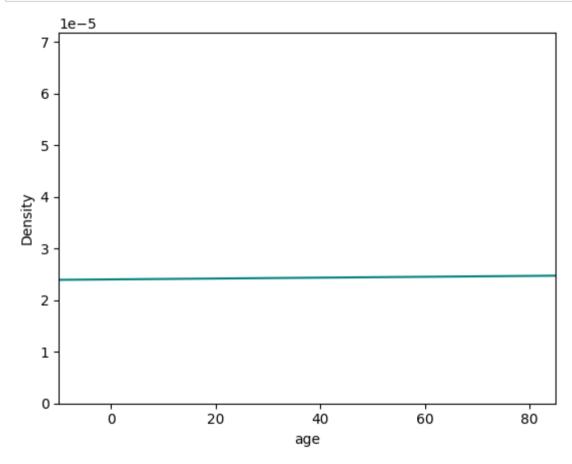
```
In [26]: ax=df["bmi"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
    df["bmi"].plot(kind='density',color='teal')
    ax.set(xlabel='age')
    plt.xlim(-10,85)
    plt.show()
```



```
In [27]: ax=df["children"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
    df["children"].plot(kind='density',color='teal')
    ax.set(xlabel='age')
    plt.xlim(-10,85)
    plt.show()
```



```
In [28]: ax=df["charges"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6)
    df["charges"].plot(kind='density',color='teal')
    ax.set(xlabel='age')
    plt.xlim(-10,85)
    plt.show()
```



To Check The Null Values

```
In [29]: df.replace(np.nan,'0',inplace=True)
```

Feature Scaling: To split the data into train and test data

```
In [30]: from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=100)

In [74]: from sklearn.linear_model import LinearRegression
    regr=LinearRegression()
    regr.fit(X_train,y_train)
    print(regr.score(x_test,y_test))

-4082835.427787215

In [32]: score=regr.score(X_test,y_test)
    print(score)

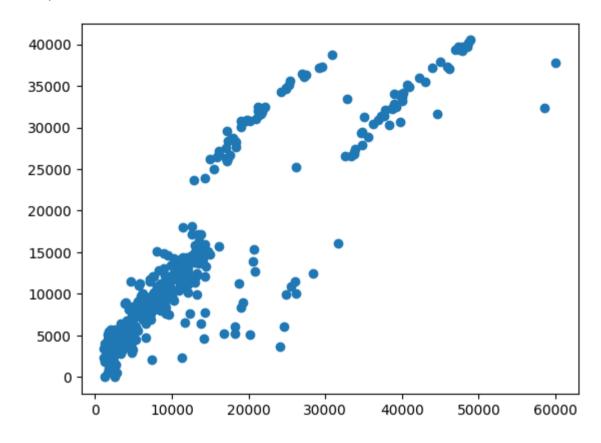
0.780095696440481
```

In the Linear Regression is not suitable for this model because of accuracy is very less

```
In [33]: predictions=regr.predict(X_test)
```

```
In [34]: plt.scatter(y_test,predictions)
```

Out[34]: <matplotlib.collections.PathCollection at 0x15407232ad0>



```
In [35]: x=np.array(df['smoker']).reshape(-1,1)
y=np.array(df['charges']).reshape(-1,1)
df.dropna(inplace=True)
```

```
In [36]: X_train,X_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
         regr.fit(X_train,y_train)
         regr.fit(X_train,y_train)
Out[36]:
          ▼ LinearRegression
          LinearRegression()
In [37]: y_pred=regr.predict(X_test)
         plt.scatter(X_test,y_test,color='y')
         plt.plot(X_test,y_pred,color='b')
         plt.show()
          60000
           50000
           40000
           30000
           20000
           10000
```

Logistic Regresssion

```
In [75]: 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=1)
    from sklearn.linear_model import LogisticRegression
    from sklearn.preprocessing import StandardScaler
    lr=LogisticRegression(max_iter=10000)

In [76]: lr.fit(x_train,y_train)

    C:\Users\SASIDHAR ROYAL\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 the shape of y to (n _samples, ), for example using ravel().
    y = column_or_1d(y, warn=True)

Out[76]:    LogisticRegression
    LogisticRegression(max_iter=10000)
```

0.8930348258706468

print(score)

In [40]: | score=lr.score(x test,y test)

In [41]: pip install statsmodels

Requirement already satisfied: statsmodels in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\si te-packages (0.14.0)

Requirement already satisfied: numpy>=1.18 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\si te-packages (from statsmodels) (1.24.3)

Requirement already satisfied: scipy!=1.9.2,>=1.4 in c:\users\sasidhar royal\appdata\local\programs\python\python311 \lib\site-packages (from statsmodels) (1.10.1)

Requirement already satisfied: pandas>=1.0 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\si te-packages (from statsmodels) (2.0.2)

Requirement already satisfied: patsy>=0.5.2 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\s ite-packages (from statsmodels) (0.5.3)

Requirement already satisfied: packaging>=21.3 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\site-packages (from statsmodels) (23.1)

Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\sasidhar royal\appdata\local\programs\python\pyt

Requirement already satisfied: pytz>=2020.1 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\s ite-packages (from pandas>=1.0->statsmodels) (2023.3)

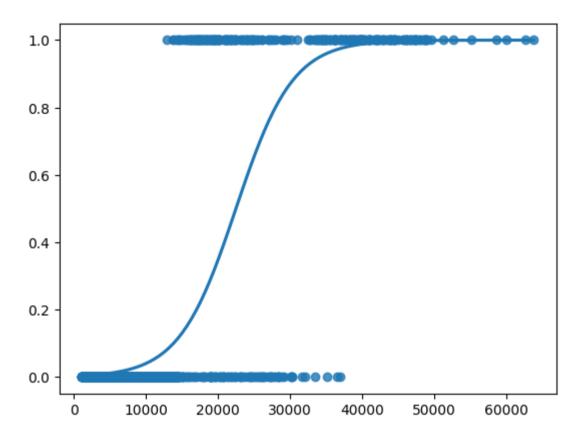
Requirement already satisfied: tzdata>=2022.1 in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib \site-packages (from pandas>=1.0->statsmodels) (2023.3)

Requirement already satisfied: six in c:\users\sasidhar royal\appdata\local\programs\python\python311\lib\site-packa ges (from patsy>=0.5.2->statsmodels) (1.16.0)

Note: you may need to restart the kernel to use updated packages.

```
In [42]: sns.regplot(x=x,y=y,data=df,logistic=True,ci=None)
```

Out[42]: <Axes: >



Decision Tree

```
In [43]: from sklearn.tree import DecisionTreeClassifier
         clf=DecisionTreeClassifier(random_state=0)
         clf.fit(x_train,y_train)
Out[43]:
                  DecisionTreeClassifier
         DecisionTreeClassifier(random state=0)
        score=clf.score(x test,y test)
In [44]:
         print(score)
         0.8880597014925373
         Random Forest
In [45]: #Random forest classifier
         from sklearn.ensemble import RandomForestClassifier
         rfc=RandomForestClassifier()
         rfc.fit(X train,y train)
         C:\Users\SASIDHAR ROYAL\AppData\Local\Temp\ipykernel 15444\2470359396.py:4: DataConversionWarning: A column-vector y
         was passed when a 1d array was expected. Please change the shape of y to (n samples,), for example using ravel().
           rfc.fit(X train,y train)
Out[45]:
          ▼ RandomForestClassifier
          RandomForestClassifier()
```

In [46]: params={'max depth':[2,3,5,10,20], 'min samples leaf':[5,10,20,50,100,200], 'n estimators':[10,25,30,50,100,200]}

```
In [47]: from sklearn.model selection import GridSearchCV
         grid search=GridSearchCV(estimator=rfc,param grid=params,cv=2,scoring="accuracy")
        grid search.fit(X train, y train)
In [48]:
         C:\Users\SASIDHAR ROYAL\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ validat
         ion.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the sh
         ape of y to (n samples,), for example using ravel().
           estimator.fit(X train, y train, **fit params)
         C:\Users\SASIDHAR ROYAL\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ validat
         ion.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the sh
         ape of v to (n samples,), for example using ravel().
           estimator.fit(X train, y train, **fit params)
         C:\Users\SASIDHAR ROYAL\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ validat
         ion.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the sh
         ape of y to (n samples,), for example using ravel().
           estimator.fit(X train, v train, **fit params)
         C:\Users\SASIDHAR ROYAL\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ validat
         ion.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the sh
         ape of y to (n samples,), for example using ravel().
           estimator.fit(X train, y train, **fit params)
         C:\Users\SASIDHAR ROYAL\AppData\Local\Programs\Python\Python311\Lib\site-packages\sklearn\model selection\ validat
         ion.py:686: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the sh
         ape of y to (n samples,), for example using ravel().
In [49]:
         grid search.best score
Out[49]: 0.7938034188034188
In [50]: rf best=grid search.best estimator
         rf best
Out[50]:
                                    RandomForestClassifier
          RandomForestClassifier(max depth=2, min samples leaf=5, n estimators=10)
```

```
In [52]: from sklearn.tree import plot_tree
    plt.figure(figsize=(80,40))
    plot_tree(rf_best.estimators_[4],class_names=['1','0'],filled=True);
```

x[0] <= 0.5 gini = 0.33 samples = 576 value = [741, 195]class = 1

gini = 0.317 samples = 460 value = [601, 148] class = 1 gini = 0.376 samples = 116 value = [140, 47] class = 1

```
In [53]: score=rfc.score(x_test,y_test)
print(score)
```

0.7985074626865671

Conclusion

for the above different typees of models we get accuracy based on the accuracy we can predict the which model is better for this dataset for this dataset. Logistic regression and Decision tree getting more accuracy among all the models, but logistic regression is more accuracy than Decision tree. so, the given data set is best fit for Logistic Regression

In []: