	Importing Libraries
In [1]:	<pre>import pandas as pd import numpy as np import matplotlib.pyplot as plt</pre>
	import seaborn as sns from sklearn.model_selection import train_test_split Data Collection and Analysis
In [2]:	# Loading data from CSV file to Pandas Dataframe df = pd.read_csv('insurance.csv')
In [3]:	# First 5 rows of the Dataframe df.head()
Out[3]:	age sex bmi children smoker region charges insuranceclaim 0 19 0 27.900 0 1 3 16884.92400 1
	1 18 1 33.770 1 0 2 1725.55230 1 2 28 1 33.000 3 0 2 4449.46200 0 3 33 1 22.705 0 0 1 21984.47061 0
	4 32 1 28.880 0 0 1 3866.85520 1 # Number of rows and columns
	df.shape (1338, 8)
	<pre># Getting Information about Dataset df.info()</pre>
	<pre><class 'pandas.core.frame.dataframe'=""> RangeIndex: 1338 entries, 0 to 1337 Data columns (total 8 columns): # Column Non-Null Count Dtype</class></pre>
	0 age
	4 smoker 1338 non-null int64 5 region 1338 non-null int64 6 charges 1338 non-null float64 7 insuranceclaim 1338 non-null int64
In [6]:	dtypes: float64(2), int64(6) memory usage: 83.8 KB # checking for Missing values
Out[6]:	<pre>df.isnull().sum() age 0 sex 0 bmi 0</pre>
	bmi 0 children 0 smoker 0 region 0 charges 0
In [7]:	insuranceclaim 0 dtype: int64 # Statical Information of dataset
Out[7]:	df.describe() age sex bmi children smoker region charges insuranceclaim
	count 1338.000000 1338.000000 1338.000000 1338.000000 1338.000000 1338.000000 1338.000000 mean 39.207025 0.505232 30.663397 1.094918 0.204783 1.515695 13270.422265 0.585202 std 14.049960 0.500160 6.098187 1.205493 0.403694 1.104885 12110.011237 0.492871
	min 18.000000 0.000000 15.960000 0.000000 0.000000 1121.873900 0.000000 25% 27.000000 0.000000 26.296250 0.000000 1.000000 4740.287150 0.000000 50% 39.00000 1.000000 1.000000 2.000000 9382.033000 1.000000
	75% 51.000000 1.000000 34.693750 2.000000 0.000000 2.000000 16639.912515 1.000000 max 64.000000 1.000000 53.130000 5.000000 1.000000 63770.428010 1.000000
In [19]:	<pre># Checking for Age distribution plt.figure(figsize=(5,6)) sns.displot(df['age']) plt.title('Age Distribution')</pre>
	plt.show() <figure 0="" 360x432="" axes="" size="" with=""> Age Distribution 200 -</figure>
	175 -
	150 - 125 - tg 100 -
	75 -
	50 - 25 - 0
In [20]:	20 30 40 50 60 age # Checking for BMI distribution
	<pre>plt.figure(figsize=(5,6)) sns.displot(df['bmi']) plt.title('BMI Distribution') plt.show()</pre>
	<pre><figure 0="" 360x432="" axes="" size="" with=""> BMI Distribution 140 -</figure></pre>
	120 -
	80 - 80 - 60 - 60 - 60 - 60 - 60 - 60 -
	40 -
	20 15 20 25 30 35 40 45 50 bmi
In [21]:	# Checking for Charges distribution plt.figure(figsize=(5,6))
	<pre>sns.displot(df['charges']) plt.title('Charges Distribution') plt.show() </pre> <pre><figure 0="" 360x432="" axes="" size="" with=""></figure></pre>
	Charges Distribution
	150 -
	125 - 8 100 - 75 -
	50 - 25 - 10 10 10 10 10 10 10 10 10 10 10 10 10
	0 10000 20000 30000 40000 50000 60000 charges
In [8]:	<pre>Splitting Feature & Target x = df.drop(['insuranceclaim'], axis=1) y = df['insuranceclaim']</pre>
In [9]:	# Feature X
Out[9]:	age sex bmi children smoker region charges 0 19 0 27.900 0 1 3 16884.92400
	1 18 1 33.770 1 0 2 1725.55230 2 28 1 33.000 3 0 2 4449.46200 3 33 1 22.705 0 0 1 21984.47061
	4 32 1 28.880 0 0 1 3866.85520 1333 50 1 30.970 3 0 1 10600.54830
	1334 18 0 31.920 0 0 0 2205.98080 1335 18 0 36.850 0 0 2 1629.83350 1336 21 0 25.800 0 0 3 2007.94500
:	1337 61 0 29.070 0 1 1 29141.36030 1338 rows × 7 columns
In [10]:	<pre># Target y 0 1</pre>
Out[10]:	1 1 2 0 3 0 4 1
	1333 0 1334 1 1335 1 1336 0
	1337 1 Name: insuranceclaim, Length: 1338, dtype: int64 Splitting data into Training & Tetsing Data
	<pre>X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=42) # Printing shape</pre>
	<pre>print("X_train :", X_train shape) print("X_test :", X_test shape) print("y_train :", y_train shape) print("y_test :", y_test shape)</pre>
	<pre>X_train : (936, 7) X_test : (402, 7) y_train : (936,) y_test : (402,)</pre>
In [13]:	Model Training # Loading the Decision Tree Classification model
[±0];	<pre>from sklearn.tree import DecisionTreeClassifier dt = DecisionTreeClassifier()</pre>
	<pre># Fitting the model dt.fit(X_train,y_train) DesignerTranClassifier()</pre>
000[14].	DecisionTreeClassifier() Model Evaluation
In [15]:	<pre># Checking Accuracy on Training Data acc1 = dt.score(X_train,y_train) print("Accuracy on Training data :",acc1)</pre>
In [16]:	Accuracy on Training data : 1.0 # Prediction on Testing Data
	<pre>y_pred = dt.predict(X_test) print(y_pred) [0 1 1 1 1 1 0 1 1 1 0 0 0 1 1 1 1 0 1 0</pre>
	1 1 0 0 0 1 1 1 1 0 1 1 1 0 1 0 0 1 1 1 1 0 1 0 1 1 1 0 0 0 0 1 1 0 0 0 0 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
	0 0 0 1 1 1 0 1 1 0 0 0 0 0 0 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1
In [17]:	<pre># Checking Accuracy on Testing Data acc2 = dt.score(X_test,y_test) print("Accuracy on Tetsing data :",acc2)</pre>
In [18]:	Accuracy on Tetsing data : 0.9751243781094527 # I used Logistic Regression model on Same dataset and I got the Accuracy:0.815920398