

Let's import the dataset and the necessary Python libraries that we need for this task:

```
In [1]: import numpy as np
import pandas as pd
data = pd.read_csv(r"C:\Users\SHREE\Downloads\Python CODES\Health Insurance Premium Prediction with Machine Learning\Health Insurance Premium Prediction.csv")
data.head()
```

Out[1]:

	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

Before moving forward, let's have a look at whether this dataset contains any null values or not:

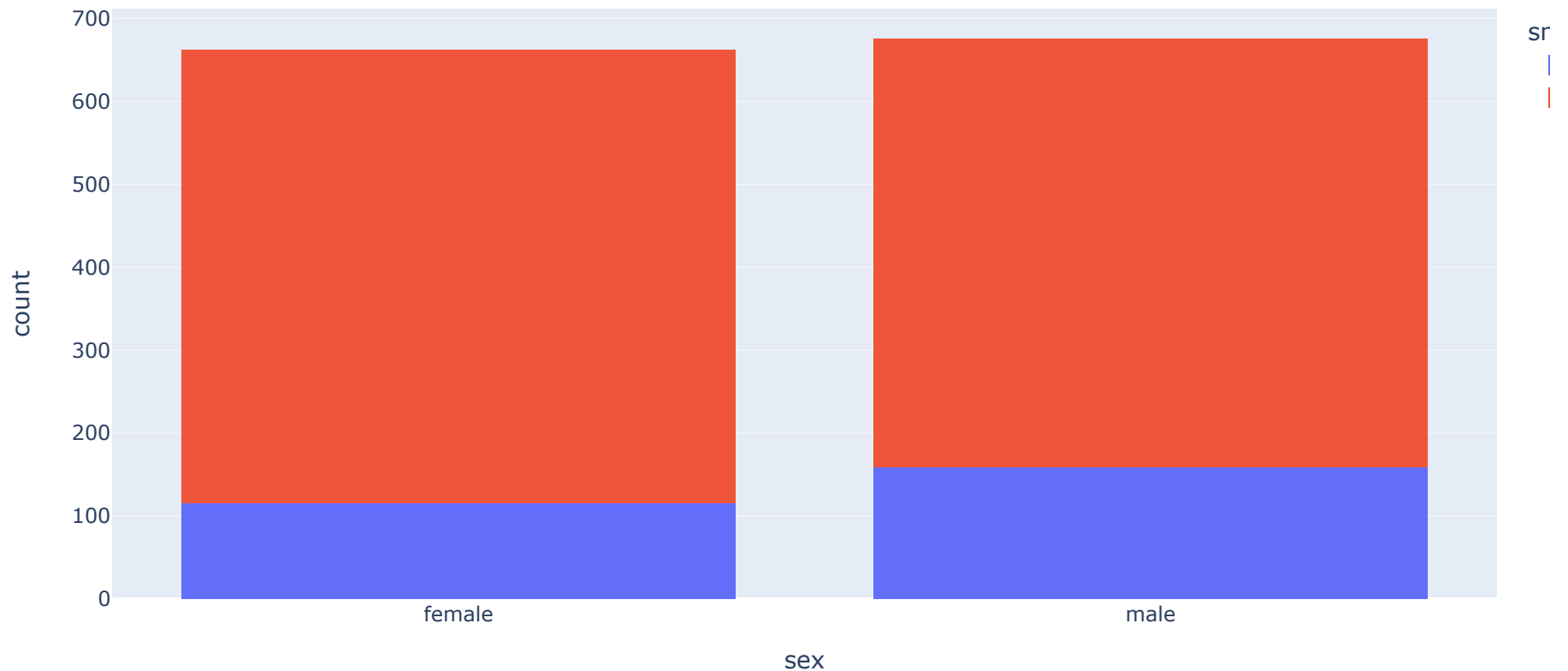
```
In [2]: data.isnull().sum()
```

```
Out[2]: age      0
sex        0
bmi        0
children    0
smoker      0
region      0
charges     0
dtype: int64
```

The dataset is therefore ready to be used. After getting the first impressions of this data, I noticed the “smoker” column, which indicates whether the person smokes or not. This is an important feature of this dataset because a person who smokes is more likely to have major health problems compared to a person who does not smoke. So let's look at the distribution of people who smoke and who do not:

```
In [3]: import plotly.express as px
data = data
figure = px.histogram(data, x = "sex", color = "smoker", title= "Number of Smokers")
figure.show()
```

Number of Smokers



According to the above visualisation, 547 females, 517 males don't smoke, and 115 females, 159 males do smoke. It is important to use this feature while training a machine learning model, so now I will replace the values of the "sex" and "smoker" columns with 0 and 1 as both these columns

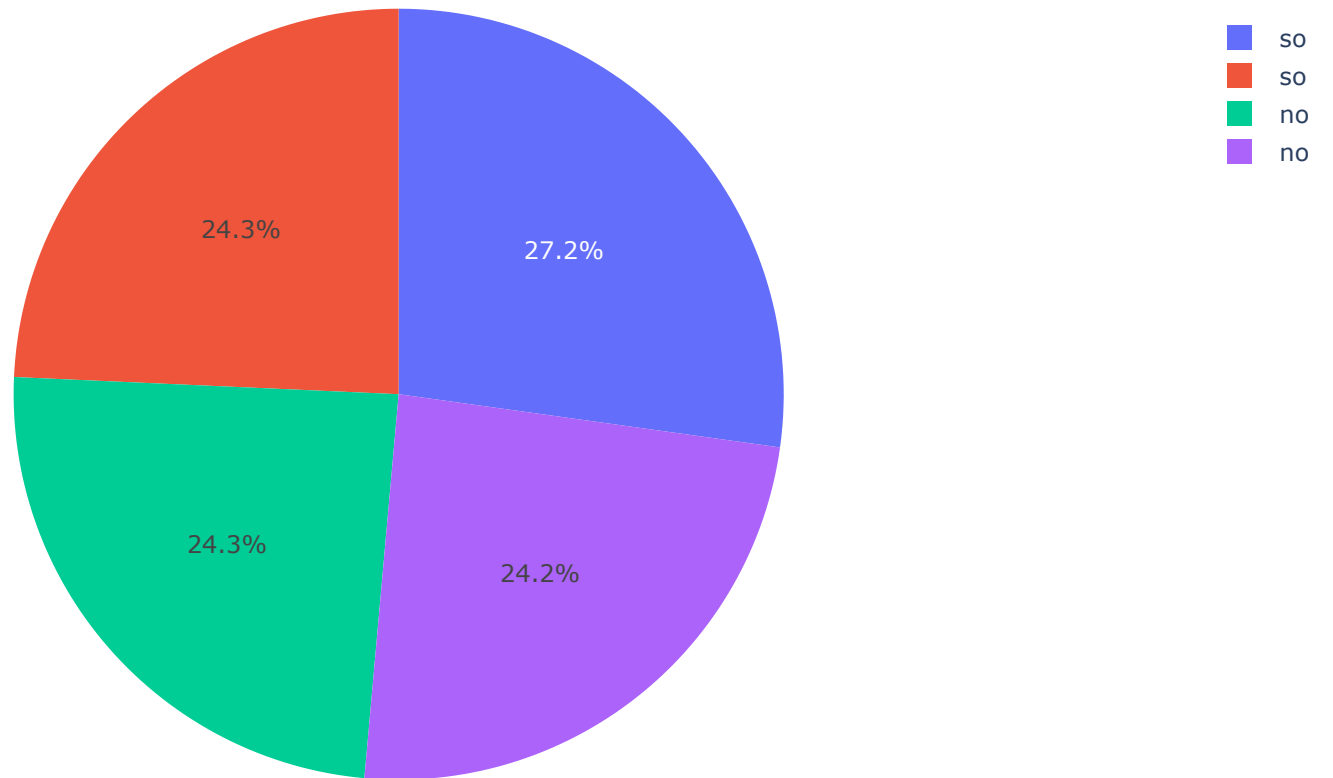
contain string values:

```
In [4]: data["sex"] = data["sex"].map({"female": 0, "male": 1})
data["smoker"] = data["smoker"].map({"no": 0, "yes": 1})
print(data.head())
```

	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

Now let's have a look at the distribution of the regions where people are living according to the dataset:

```
In [5]: import plotly.express as px
pie = data["region"].value_counts()
regions = pie.index
population = pie.values
fig = px.pie(data, values=population, names=regions)
fig.show()
```



Now let's have a look at the correlation between the features of this dataset:

In [6]: `print(data.corr())`

	age	sex	bmi	children	smoker	charges
age	1.000000	-0.020856	0.109272	0.042469	-0.025019	0.299008
sex	-0.020856	1.000000	0.046371	0.017163	0.076185	0.057292
bmi	0.109272	0.046371	1.000000	0.012759	0.003750	0.198341
children	0.042469	0.017163	0.012759	1.000000	0.007673	0.067998
smoker	-0.025019	0.076185	0.003750	0.007673	1.000000	0.787251
charges	0.299008	0.057292	0.198341	0.067998	0.787251	1.000000

Health Insurance Premium Prediction Model

Now let's move on to training a machine learning model for the task of predicting health insurance premiums. First, I'll split the data into training and test sets:

```
In [7]: x = np.array(data[["age", "sex", "bmi", "smoker"]])
        y = np.array(data["charges"])

        from sklearn.model_selection import train_test_split
        xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2, random_state=42)
```

After using different machine learning algorithms, I found the random forest algorithm as the best performing algorithm for this task. So here I will train the model by using the random forest regression algorithm:

```
In [8]: from sklearn.ensemble import RandomForestRegressor
        forest = RandomForestRegressor()
        forest.fit(xtrain, ytrain)
```

```
Out[8]: RandomForestRegressor
        RandomForestRegressor()
```

Now let's have a look at the predicted values of the model:

```
In [9]: ypred = forest.predict(xtest)
data = pd.DataFrame(data={"Predicted Premium Amount": ypred})
print(data.head())
```

	Predicted Premium Amount
0	10195.399276
1	5593.074187
2	28390.352115
3	9643.372430
4	34670.883579

myr