

medical-premium-insurance-project

November 16, 2023

```
[1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.metrics import accuracy_score
```

```
[2]: #IMPORTING THE DATA
```

```
[3]: dt = pd.read_csv("C:/Users/Adithya Das/Desktop/Medicalpremium.csv")
```

```
[4]: #DATA ANALYSIS
```

```
[5]: dt.shape
```

```
[5]: (986, 11)
```

```
[6]: dt.head()
```

```
[6]:   Age  Diabetes  BloodPressureProblems  AnyTransplants  AnyChronicDiseases  \
0    45         0                      0                0                  0
1    60         1                      0                0                  0
2    36         1                      1                0                  0
3    52         1                      1                0                  1
4    38         0                      0                0                  1

   Height  Weight  KnownAllergies  HistoryOfCancerInFamily  \
0     155     57                0                        0
1     180     73                0                        0
2     158     59                0                        0
3     183     93                0                        0
4     166     88                0                        0

   NumberOfMajorSurgeries  PremiumPrice
0                        0          25000
1                        0          29000
```

2	1	23000
3	2	28000
4	1	23000

```
[7]: dt.tail()
```

```
[7]:
```

	Age	Diabetes	BloodPressureProblems	AnyTransplants	AnyChronicDiseases	\
981	18	0	0	0	0	
982	64	1	1	0	0	
983	56	0	1	0	0	
984	47	1	1	0	0	
985	21	0	0	0	0	

	Height	Weight	KnownAllergies	HistoryOfCancerInFamily	\
981	169	67	0	0	
982	153	70	0	0	
983	155	71	0	0	
984	158	73	1	0	
985	158	75	1	0	

	NumberOfMajorSurgeries	PremiumPrice
981	0	15000
982	3	28000
983	1	29000
984	1	39000
985	1	15000

```
[8]: dt.describe()
```

```
[8]:
```

	Age	Diabetes	BloodPressureProblems	AnyTransplants	\
count	986.000000	986.000000	986.000000	986.000000	
mean	41.745436	0.419878	0.468560	0.055781	
std	13.963371	0.493789	0.499264	0.229615	
min	18.000000	0.000000	0.000000	0.000000	
25%	30.000000	0.000000	0.000000	0.000000	
50%	42.000000	0.000000	0.000000	0.000000	
75%	53.000000	1.000000	1.000000	0.000000	
max	66.000000	1.000000	1.000000	1.000000	

	AnyChronicDiseases	Height	Weight	KnownAllergies	\
count	986.000000	986.000000	986.000000	986.000000	
mean	0.180527	168.182556	76.950304	0.215010	
std	0.384821	10.098155	14.265096	0.411038	
min	0.000000	145.000000	51.000000	0.000000	
25%	0.000000	161.000000	67.000000	0.000000	
50%	0.000000	168.000000	75.000000	0.000000	
75%	0.000000	176.000000	87.000000	0.000000	

max	1.000000	188.000000	132.000000	1.000000
-----	----------	------------	------------	----------

	HistoryOfCancerInFamily	NumberOfMajorSurgeries	PremiumPrice
count	986.000000	986.000000	986.000000
mean	0.117647	0.667343	24336.713996
std	0.322353	0.749205	6248.184382
min	0.000000	0.000000	15000.000000
25%	0.000000	0.000000	21000.000000
50%	0.000000	1.000000	23000.000000
75%	0.000000	1.000000	28000.000000
max	1.000000	3.000000	40000.000000

```
[9]: #CHECK FOR NULL VALUE
```

```
[10]: dt.isnull().sum()
```

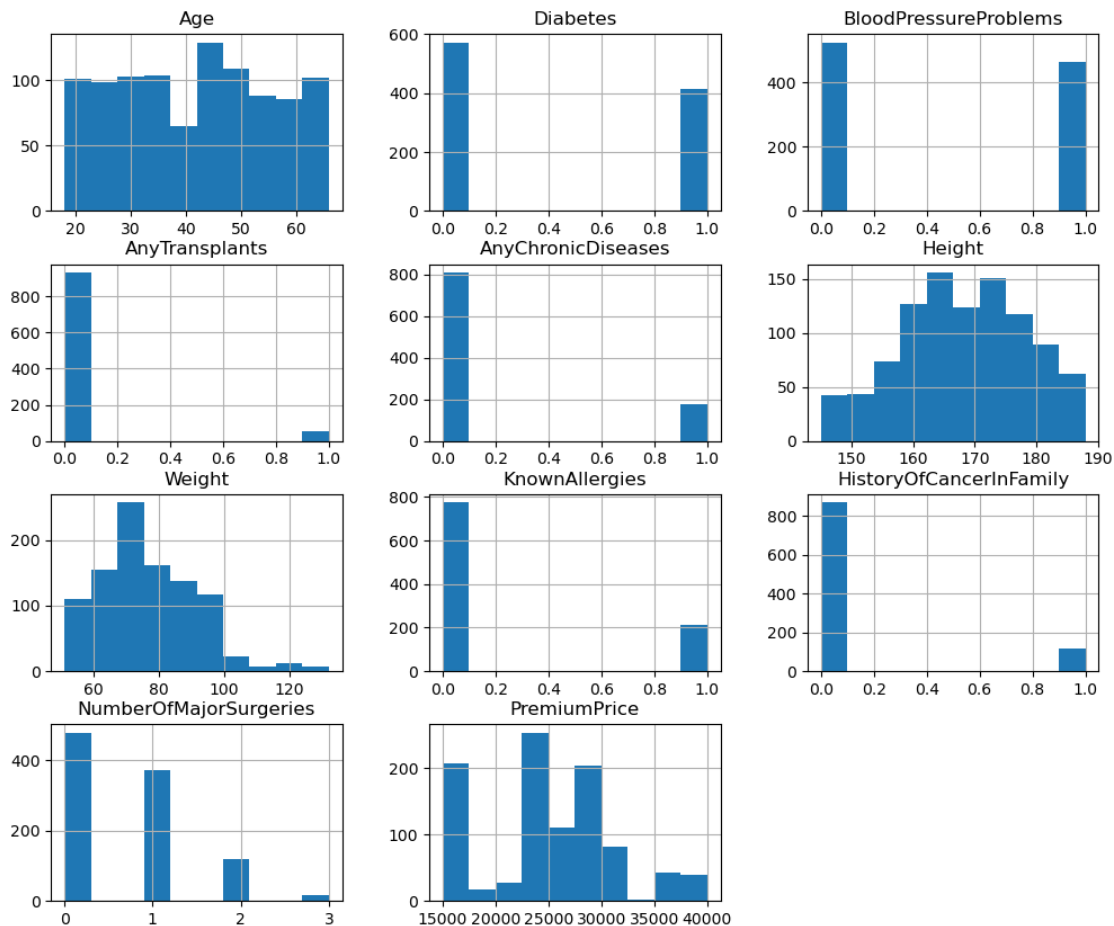
```
[10]: Age                                0
      Diabetes                            0
      BloodPressureProblems              0
      AnyTransplants                     0
      AnyChronicDiseases                  0
      Height                             0
      Weight                              0
      KnownAllergies                      0
      HistoryOfCancerInFamily             0
      NumberOfMajorSurgeries              0
      PremiumPrice                        0
      dtype: int64
```

```
[11]: dt.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 986 entries, 0 to 985
Data columns (total 11 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age                                    986 non-null    int64
1   Diabetes                              986 non-null    int64
2   BloodPressureProblems                 986 non-null    int64
3   AnyTransplants                        986 non-null    int64
4   AnyChronicDiseases                    986 non-null    int64
5   Height                                986 non-null    int64
6   Weight                                986 non-null    int64
7   KnownAllergies                        986 non-null    int64
8   HistoryOfCancerInFamily                986 non-null    int64
9   NumberOfMajorSurgeries                 986 non-null    int64
10  PremiumPrice                           986 non-null    int64
```

```
dtypes: int64(11)
memory usage: 84.9 KB
```

```
[12]: dt.hist(figsize=(12, 10))
plt.show()
```



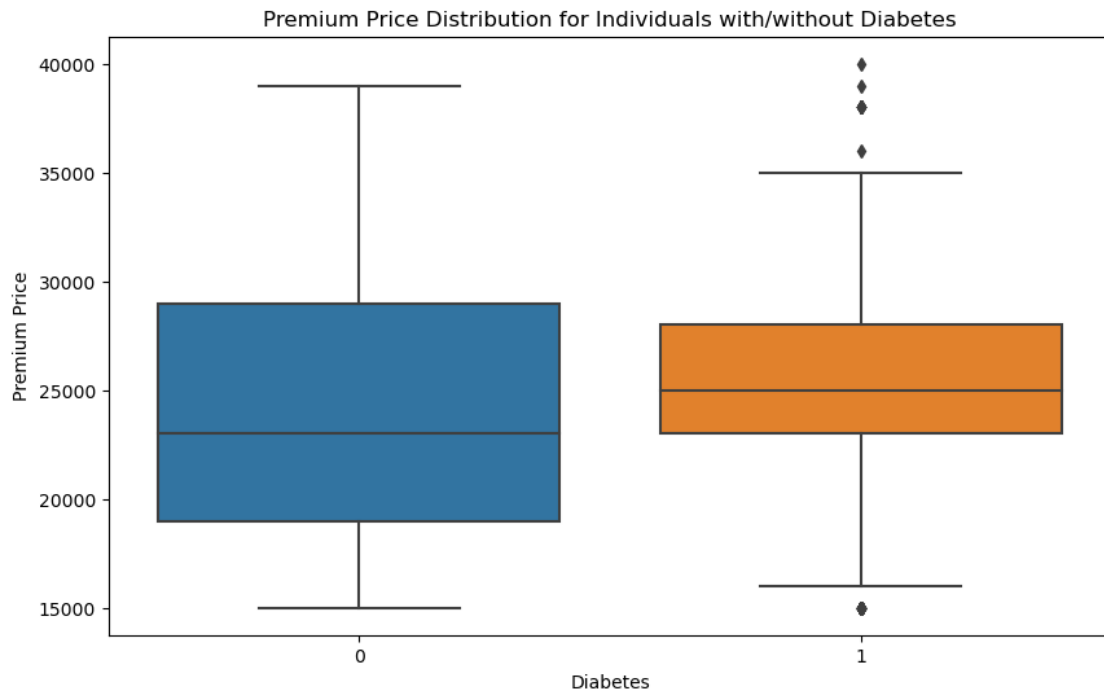
```
[13]: #CATEGORICAL DATA ANALYSIS
```

```
[14]: dt.groupby('Diabetes')['PremiumPrice'].mean()
dt.groupby('BloodPressureProblems')['PremiumPrice'].mean()
```

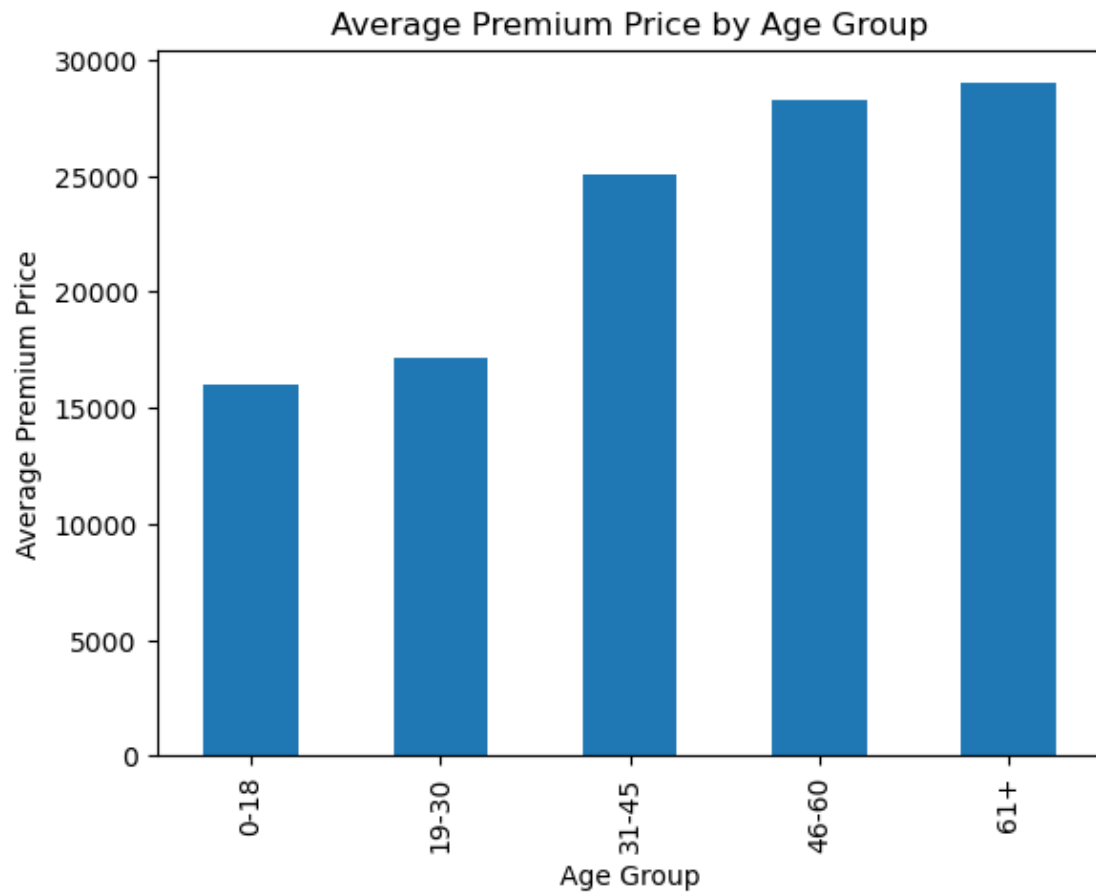
```
[14]: BloodPressureProblems
0    23356.870229
1    25448.051948
Name: PremiumPrice, dtype: float64
```

```
[15]: plt.figure(figsize=(10, 6))
sns.boxplot(x='Diabetes', y='PremiumPrice', data=dt)
```

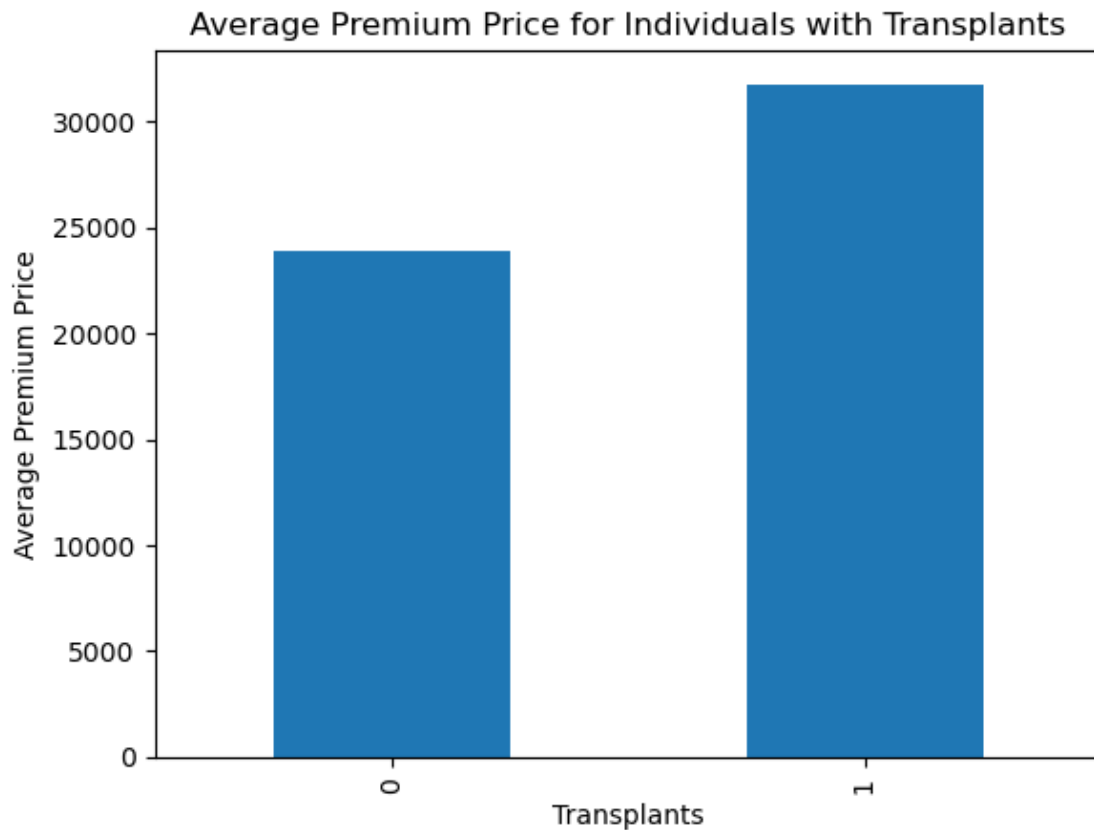
```
plt.xlabel('Diabetes')
plt.ylabel('Premium Price')
plt.title('Premium Price Distribution for Individuals with/without Diabetes')
plt.show()
```



```
[16]: age_bins = [0, 18, 30, 45, 60, 100]
age_labels = ['0-18', '19-30', '31-45', '46-60', '61+']
dt['AgeGroup'] = pd.cut(dt['Age'], bins=age_bins, labels=age_labels)
dt.groupby('AgeGroup')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('Age Group')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price by Age Group')
plt.show()
```

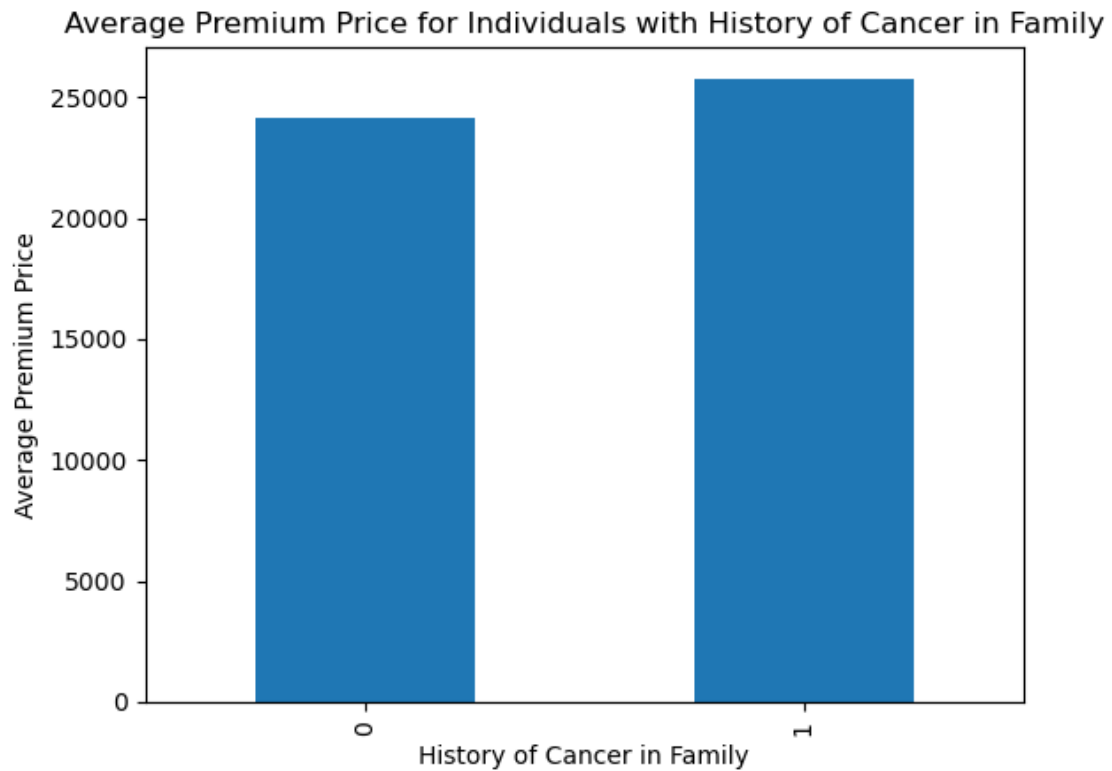


```
[17]: dt.groupby('AnyTransplants')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('Transplants')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price for Individuals with Transplants')
plt.show()
```



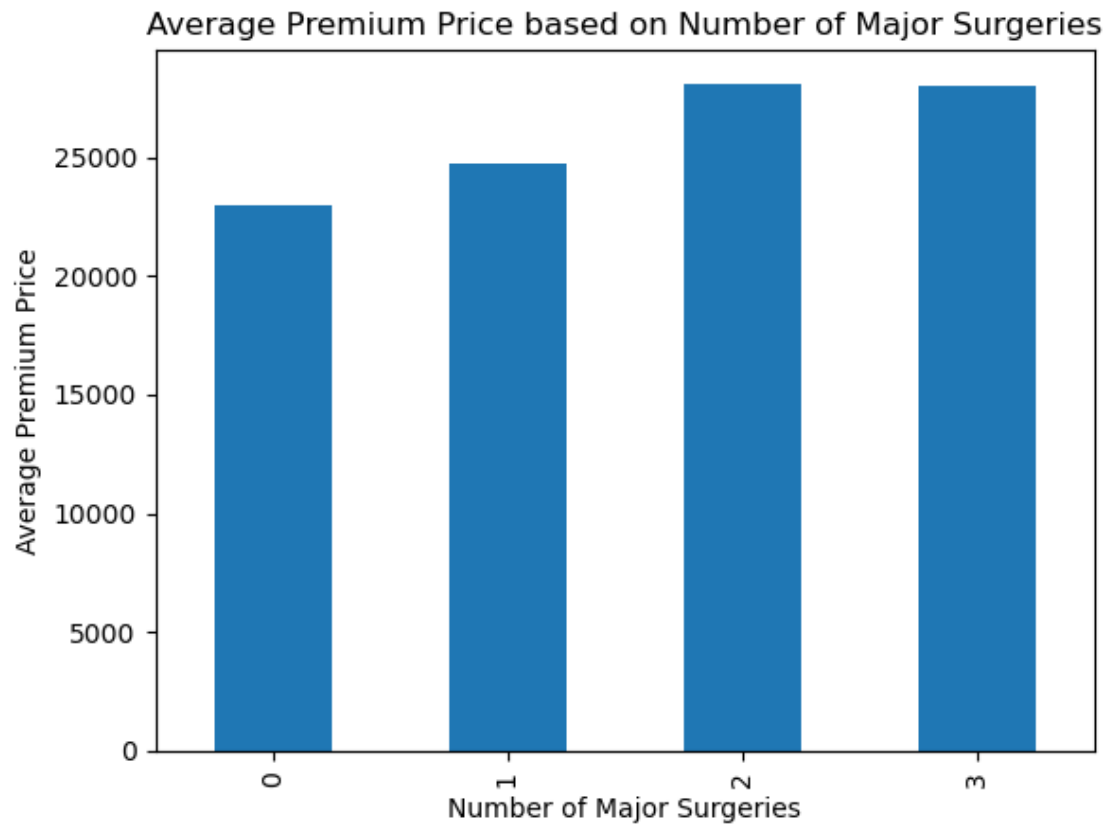
```
[18]: #HISTORY ANALYSIS
```

```
[19]: dt.groupby('HistoryOfCancerInFamily')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('History of Cancer in Family')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price for Individuals with History of Cancer in_
↪Family')
plt.show()
```



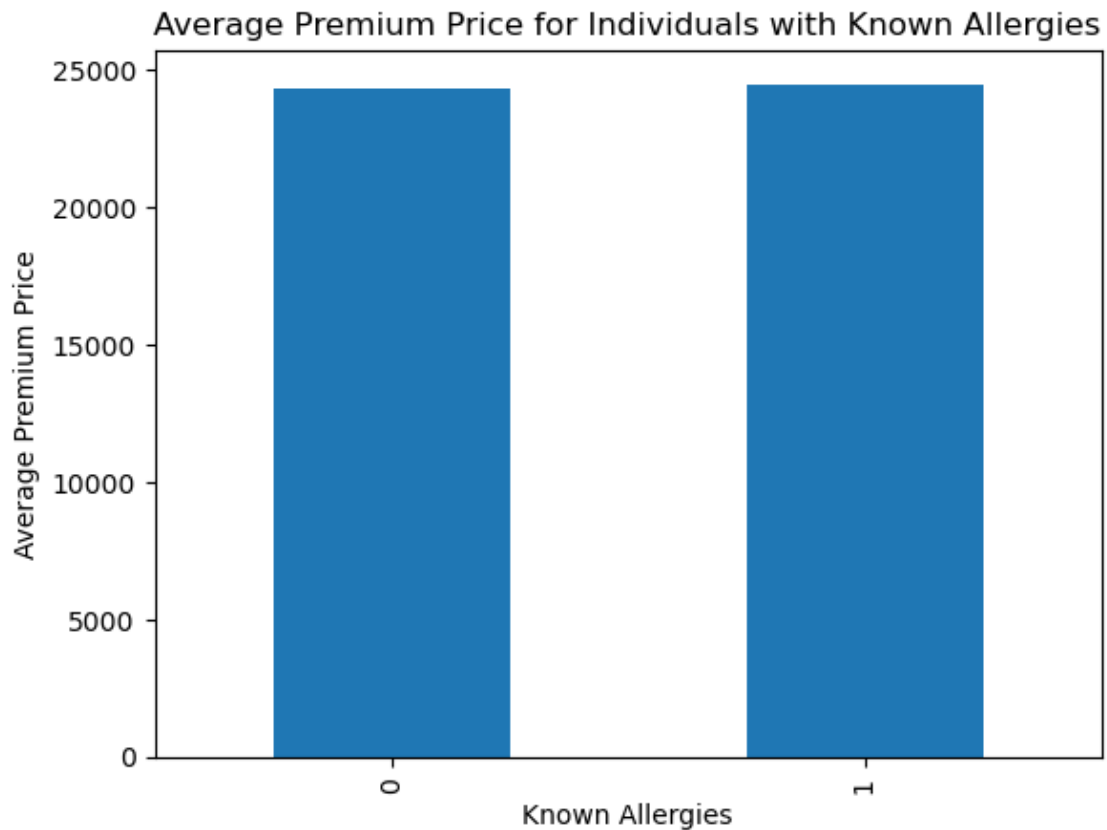
[20]: *#NO. OF SURGERY ANALYSIS*

```
[21]: dt.groupby('NumberOfMajorSurgeries')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('Number of Major Surgeries')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price based on Number of Major Surgeries')
plt.show()
```

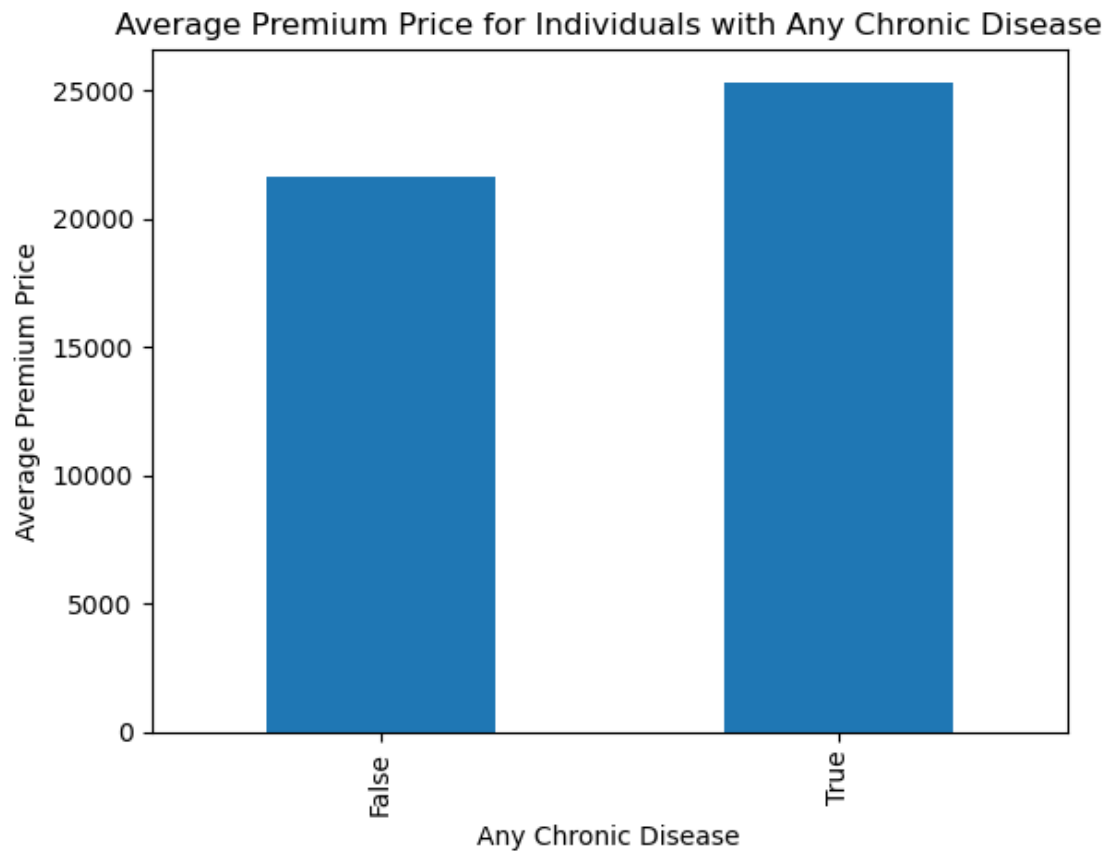



```
[22]: #KNOWN ALLERGIES ANALYSIS
```

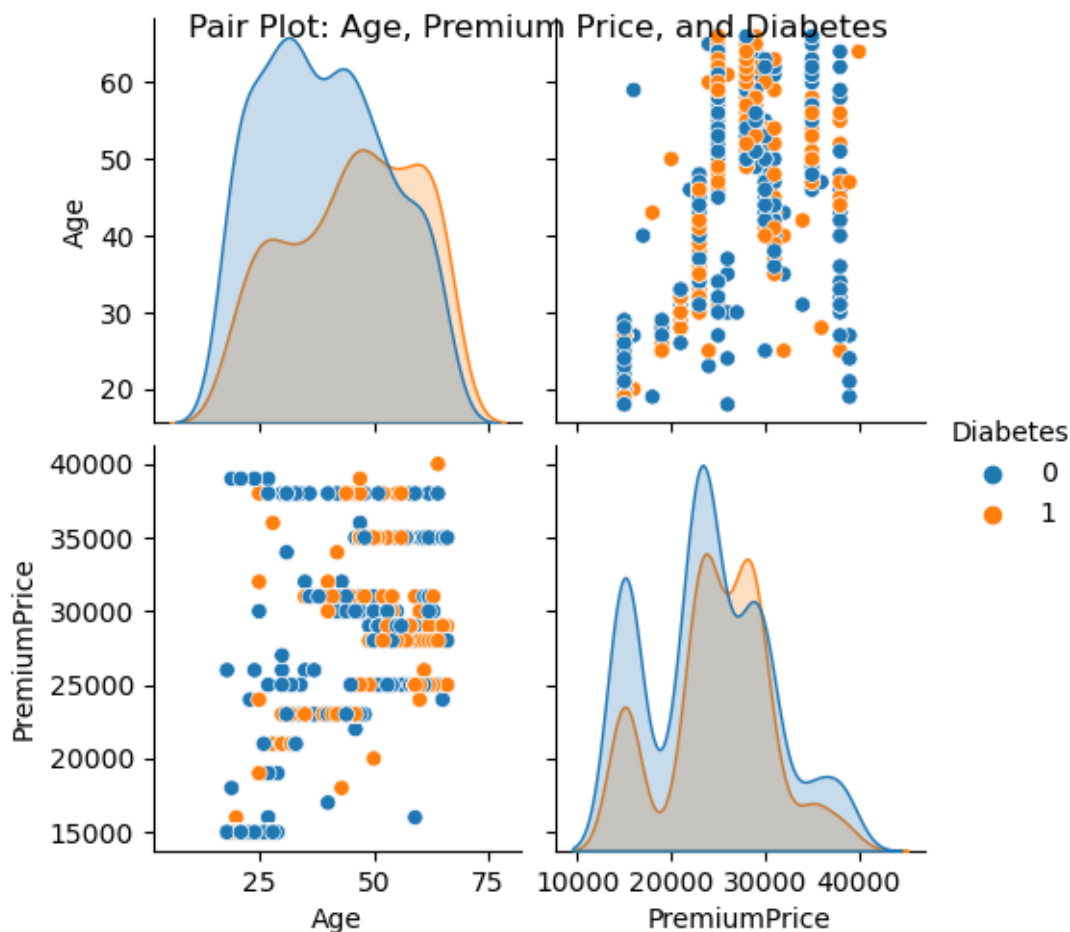
```
[23]: dt.groupby('KnownAllergies')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('Known Allergies')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price for Individuals with Known Allergies')
plt.show()
```



```
[24]: chronic_diseases = ['Diabetes', 'BloodPressureProblems', 'AnyChronicDiseases']
dt['AnyChronicDisease'] = dt[chronic_diseases].any(axis=1)
dt.groupby('AnyChronicDisease')['PremiumPrice'].mean().plot(kind='bar')
plt.xlabel('Any Chronic Disease')
plt.ylabel('Average Premium Price')
plt.title('Average Premium Price for Individuals with Any Chronic Disease')
plt.show()
```



```
[25]: sns.pairplot(dt, vars=['Age', 'PremiumPrice'], hue='Diabetes', diag_kind='kde')  
plt.suptitle('Pair Plot: Age, Premium Price, and Diabetes')  
plt.show()
```



[26]: #CORRELATION

[27]: dt.corr()

C:\Users\Adithya Das\AppData\Local\Temp\ipykernel_8652\1921767823.py:1:
FutureWarning: The default value of numeric_only in DataFrame.corr is
deprecated. In a future version, it will default to False. Select only valid
columns or specify the value of numeric_only to silence this warning.
dt.corr()

[27]:

	Age	Diabetes	BloodPressureProblems	\
Age	1.000000	0.210908	0.244888	
Diabetes	0.210908	1.000000	0.127727	
BloodPressureProblems	0.244888	0.127727	1.000000	
AnyTransplants	-0.008549	-0.036652	-0.024538	
AnyChronicDiseases	0.051072	-0.089428	0.045424	
Height	0.039879	-0.003783	-0.037926	
Weight	-0.018590	-0.024563	-0.061016	

KnownAllergies	-0.024416	-0.080102	-0.011550
HistoryOfCancerInFamily	-0.027623	-0.055527	0.048239
NumberOfMajorSurgeries	0.429181	0.122722	0.251568
PremiumPrice	0.697540	0.076209	0.167097
AnyChronicDisease	0.300208	0.510450	0.563387

	AnyTransplants	AnyChronicDiseases	Height \
Age	-0.008549	0.051072	0.039879
Diabetes	-0.036652	-0.089428	-0.003783
BloodPressureProblems	-0.024538	0.045424	-0.037926
AnyTransplants	1.000000	0.035285	-0.031543
AnyChronicDiseases	0.035285	1.000000	0.047419
Height	-0.031543	0.047419	1.000000
Weight	0.002087	-0.033318	0.066946
KnownAllergies	0.001876	-0.027418	-0.010200
HistoryOfCancerInFamily	-0.020171	0.008666	0.010549
NumberOfMajorSurgeries	-0.004154	0.014835	0.037289
PremiumPrice	0.289056	0.208610	0.026910
AnyChronicDisease	0.015615	0.281615	0.009486

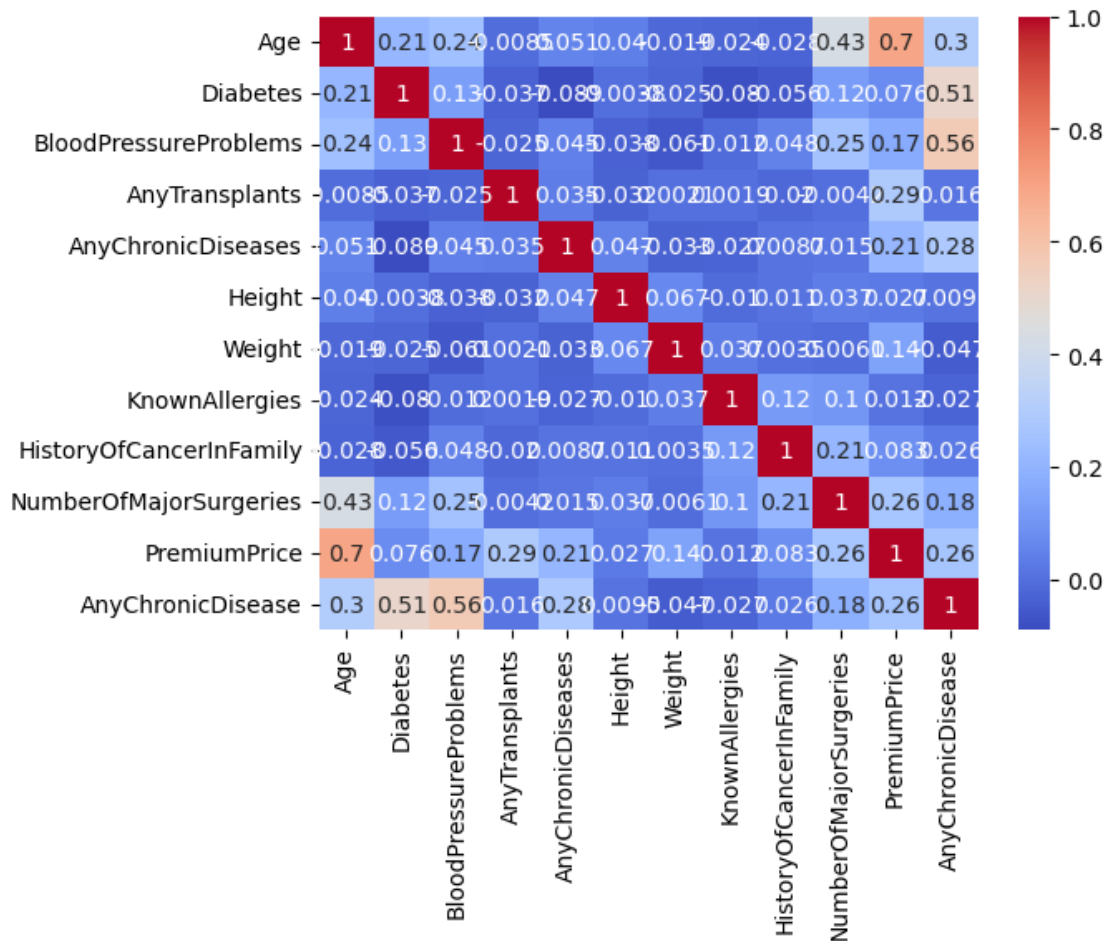
	Weight	KnownAllergies	HistoryOfCancerInFamily \
Age	-0.018590	-0.024416	-0.027623
Diabetes	-0.024563	-0.080102	-0.055527
BloodPressureProblems	-0.061016	-0.011550	0.048239
AnyTransplants	0.002087	0.001876	-0.020171
AnyChronicDiseases	-0.033318	-0.027418	0.008666
Height	0.066946	-0.010200	0.010549
Weight	1.000000	0.037492	0.003481
KnownAllergies	0.037492	1.000000	0.115383
HistoryOfCancerInFamily	0.003481	0.115383	1.000000
NumberOfMajorSurgeries	-0.006108	0.103923	0.212657
PremiumPrice	0.141507	0.012103	0.083139
AnyChronicDisease	-0.047237	-0.027320	0.026442

	NumberOfMajorSurgeries	PremiumPrice \
Age	0.429181	0.697540
Diabetes	0.122722	0.076209
BloodPressureProblems	0.251568	0.167097
AnyTransplants	-0.004154	0.289056
AnyChronicDiseases	0.014835	0.208610
Height	0.037289	0.026910
Weight	-0.006108	0.141507
KnownAllergies	0.103923	0.012103
HistoryOfCancerInFamily	0.212657	0.083139
NumberOfMajorSurgeries	1.000000	0.264250
PremiumPrice	0.264250	1.000000
AnyChronicDisease	0.184738	0.260210

	AnyChronicDisease
Age	0.300208
Diabetes	0.510450
BloodPressureProblems	0.563387
AnyTransplants	0.015615
AnyChronicDiseases	0.281615
Height	0.009486
Weight	-0.047237
KnownAllergies	-0.027320
HistoryOfCancerInFamily	0.026442
NumberOfMajorSurgeries	0.184738
PremiumPrice	0.260210
AnyChronicDisease	1.000000

```
[28]: sns.heatmap(dt.corr(), annot=True, cmap='coolwarm')
plt.show()
```

```
C:\Users\Adithya Das\AppData\Local\Temp\ipykernel_8652\1692465306.py:1:
FutureWarning: The default value of numeric_only in DataFrame.corr is
deprecated. In a future version, it will default to False. Select only valid
columns or specify the value of numeric_only to silence this warning.
sns.heatmap(dt.corr(), annot=True, cmap='coolwarm')
```



```
[29]: from sklearn.cluster import KMeans

# Prepare features
X = dt[['Age', 'Weight', 'PremiumPrice']]

# Use KMeans clustering
kmeans = KMeans(n_clusters=3, random_state=42)
dt['Cluster'] = kmeans.fit_predict(X)

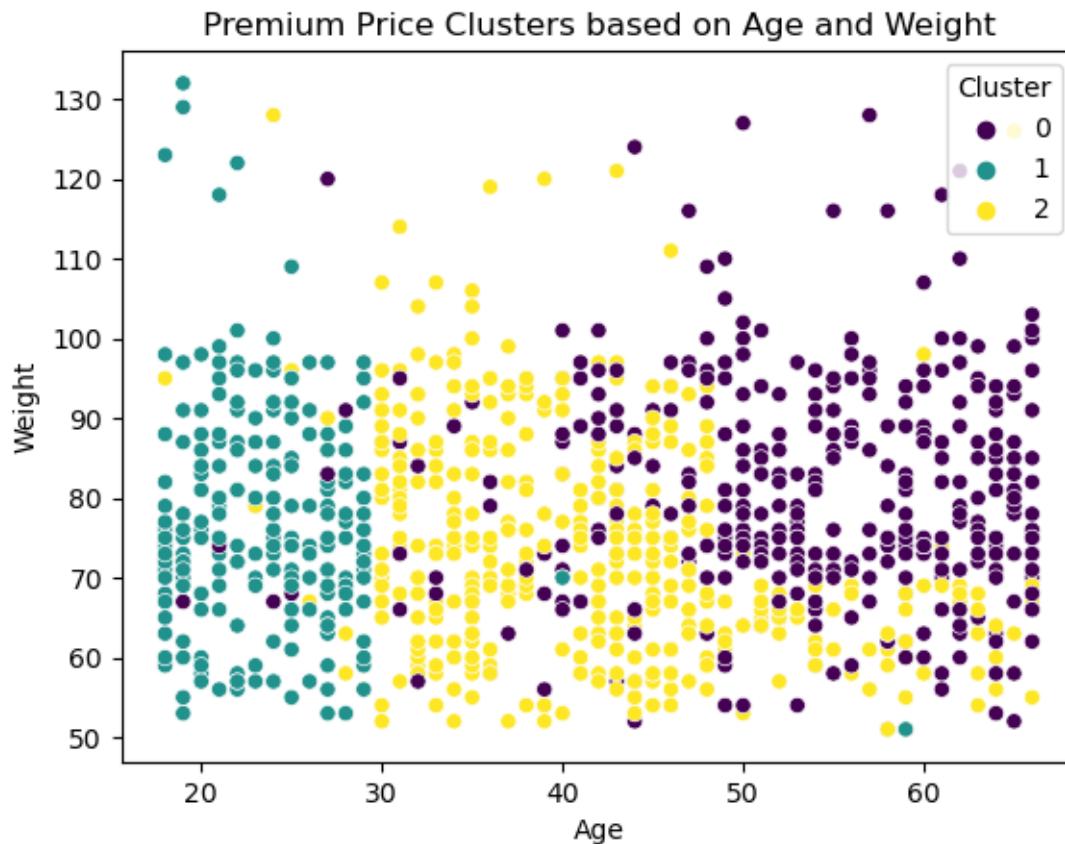
# Visualize clusters
sns.scatterplot(data=dt, x='Age', y='Weight', hue='Cluster', palette='viridis')
plt.xlabel('Age')
plt.ylabel('Weight')
plt.title('Premium Price Clusters based on Age and Weight')
plt.show()
```

C:\Users\Adithya Das\anaconda3\Lib\site-packages\sklearn\cluster_kmeans.py:870:
FutureWarning: The default value of `n_init` will change from 10 to 'auto' in

1.4. Set the value of `n_init` explicitly to suppress the warning

```
warnings.warn(  
C:\Users\Adithya Das\anaconda3\Lib\site-  
packages\sklearn\cluster\_kmeans.py:1382: UserWarning: KMeans is known to have a  
memory leak on Windows with MKL, when there are less chunks than available  
threads. You can avoid it by setting the environment variable OMP_NUM_THREADS=4.  
warnings.warn(  

```



```
[30]: from sklearn.linear_model import LinearRegression  
from sklearn.metrics import mean_squared_error  
  
# Prepare features and target variable  
features = ['Age', 'Diabetes', 'BloodPressureProblems', 'AnyChronicDiseases',  
            ↪ 'Height', 'Weight', 'KnownAllergies', 'HistoryOfCancerInFamily',  
            ↪ 'NumberOfMajorSurgeries']  
X = dt[features]  
y = dt['PremiumPrice']  
  
# Split data into training and testing sets
```



```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,  
↪random_state=42)  
  
# Build and train the regression model  
model = LinearRegression()  
model.fit(X_train, y_train)  
  
# Make predictions  
predictions = model.predict(X_test)  
  
# Calculate and print mean squared error  
mse = mean_squared_error(y_test, predictions)  
print(f'Mean Squared Error: {mse}')
```

Mean Squared Error: 16775137.514773391

[]:

[]:

[]: