

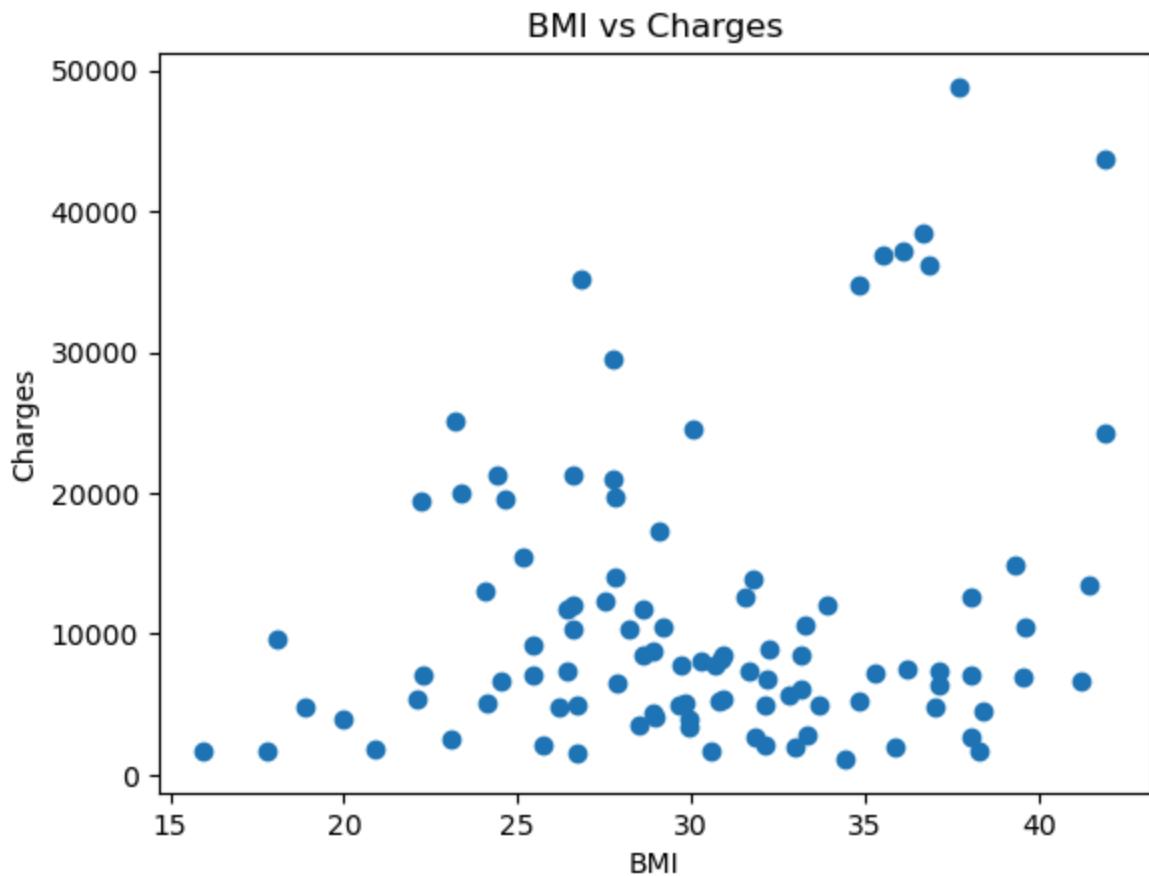
```
In [1]: ## =====  
## EET-4501 - Applied Machine Learning  
## Assignment 2  
## =====
```

```
In [2]: ## Part 1: Data Loading and Exploration  
  
import pandas as pd  
  
inputData = pd.read_csv("Ass2_dataset.csv")  
  
inputData.head()  
  
inputData.tail(10)  
  
inputData.shape  
  
inputData['age'].describe()  
  
trainingData = inputData.iloc[150:250]  
  
features = trainingData[['bmi', 'charges']]  
  
labels = trainingData.iloc[:, -1]  
  
trainingData2F = pd.concat([features, labels], axis=1)  
trainingData2F.head()
```

```
Out[2]:
```

	bmi	charges	insuranceclaim
150	24.130	5125.21570	0
151	29.700	7789.63500	1
152	37.145	6334.34355	0
153	23.370	19964.74630	1
154	25.460	7077.18940	0

```
In [3]: ## Part 2: Data Visualization  
  
import matplotlib.pyplot as plt  
  
plt.figure()  
plt.scatter(trainingData2F['bmi'], trainingData2F['charges'])  
plt.xlabel('BMI')  
plt.ylabel('Charges')  
plt.title('BMI vs Charges')  
plt.show()
```



```
In [4]: ## Part 3: Feature Engineering & Scaling

from sklearn.preprocessing import MinMaxScaler

minmax_scaler = MinMaxScaler()
features_normalized = minmax_scaler.fit_transform(features)

features_normalized

from sklearn.preprocessing import StandardScaler

standard_scaler = StandardScaler()
features_standardized = standard_scaler.fit_transform(features)

features_standardized
```

```
Out[4]: array([[-1.07931651, -0.59993178],  
   [-0.10972722, -0.33845565],  
   [ 1.18624984, -0.4812725 ],  
   [-1.21161235,  0.85636415],  
   [-0.84779878, -0.40837239],  
   [ 1.59967436, -0.42098178],  
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   [-0.89740972,  0.41999482],  
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   [-0.64935501,  0.99218077],  
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   [ 1.16100919, -0.62884131],  
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   [ 0.32545648, -0.43793382],  
   [ 0.30978987, -0.61978472],  
   [ 0.21579019,  0.12945509],  
   [-0.7189844 , -0.62361693],  
   [-0.7990582 , -0.89311993],  
   [-0.64935501,  0.07908535],  
   [ 0.71364034, -0.99127423],  
   [ 0.04519818, -0.94200066],  
   [ 0.42990057, -0.54845938],  
   [-0.30120805, -0.26709156],  
   [-2.13768327, -0.15645044],  
   [ 1.5666004 ,  0.35947777],  
   [ 0.30978987, -0.89380469],  
   [ 0.33067869, -0.23231965],  
   [-1.09585349,  0.17406816],  
   [ 1.00086159,  2.54128039],  
   [-1.39787097, -0.40151114],  
   [-0.25246747, -0.67721216],
```

```
[-0.6841697 ,  0.04954212],  
[-0.45091124,  0.95639908],  
[ 0.25582709,  0.25932308],  
[ 1.89734001, -0.45420984],  
[ 0.46471527, -0.90858459],  
[ 0.09480912, -0.30184286],  
[-0.3186154 , -0.75572433],  
[-0.62672546, -0.61194689],  
[ 0.09916096, -0.26677782],  
[ 1.17841654, -0.37946316],  
[-0.64935501, -0.08663744],  
[-1.25861219, -0.8591567 ],  
[-0.07143105, -0.76992721],  
[-1.23946411,  1.35852947],  
[ 0.5865667 , -0.61099612],  
[ 0.50823364, -0.06610306],  
[ 0.08175361, -0.58734008],  
[ 0.77804753,  2.31023986],  
[-0.99053903,  0.81228089],  
[ 0.61789993,  0.07347453],  
[ 1.34552708, -0.83896424],  
[ 2.01570998,  1.27467886],  
[ 0.22710497, -0.38079746],  
[-0.84779878, -0.19756915],  
[ 1.02088004, -0.37241   ],  
[-0.43524463,  0.27113257],  
[-2.18120164, -0.9333429 ],  
[-0.49268888,  0.10749442],  
[-1.01316858, -0.44438815],  
[-1.41179685,  0.80528615],  
[-0.62672546, -0.94433595],  
[ 1.40297133, -0.66489902],  
[-0.21939351,  0.60002578],  
[ 1.34552708, -0.40096487],  
[ 1.10356494,  2.67648566],  
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[-0.61454031,  2.34758262],  
[ 0.86508427, -0.39662768],  
[-0.45091124,  1.79439161],  
[-0.0540237 ,  1.29912865],  
[ 1.34552708,  0.13839512],  
[ 0.96256542, -0.90791104],  
[-1.64157385, -0.9231064 ],  
[-0.23593049, -0.706376  ]])
```

```
In [5]: ## Part 4: Sampling & Cross-Validation
```

```
from sklearn.model_selection import train_test_split  
  
X = features  
y = labels  
  
X_train, X_valandtest, y_train, y_valandtest = train_test_split(  
    X, y, test_size=0.30, stratify=y, random_state=42  
)
```

```

X_val, X_test, y_val, y_test = train_test_split(
    X_valandtest, y_valandtest, test_size=0.50, stratify=y_valandtest, random_state
)

from sklearn.model_selection import KFold, cross_val_score
from sklearn.tree import DecisionTreeClassifier

model = DecisionTreeClassifier(random_state=42)
kf = KFold(n_splits=5, shuffle=True, random_state=42)

cv_scores = cross_val_score(model, X_train, y_train, cv=kf)

cv_scores

print("Accuracy per fold:", cv_scores)
print("Mean accuracy:", cv_scores.mean())

model.fit(X_train, y_train)
importances = model.feature_importances_

feature_importance_df = pd.DataFrame({
    'Feature': X.columns,
    'Importance': importances
})

feature_importance_df

```

*## Stratified sampling ensures that each data split maintains the same class as the
Feature importance helps find which input variables contribute most to the model*

Accuracy per fold: [0.57142857 0.5 0.57142857 0.64285714 0.42857143]
Mean accuracy: 0.5428571428571428

Out[5]:

	Feature	Importance
0	bmi	0.371575
1	charges	0.628425

In [6]:

```

## Part 5: Data Export

trainingData2F.to_csv("Ass2Output_Lieu.csv", index=False)

```