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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()
```

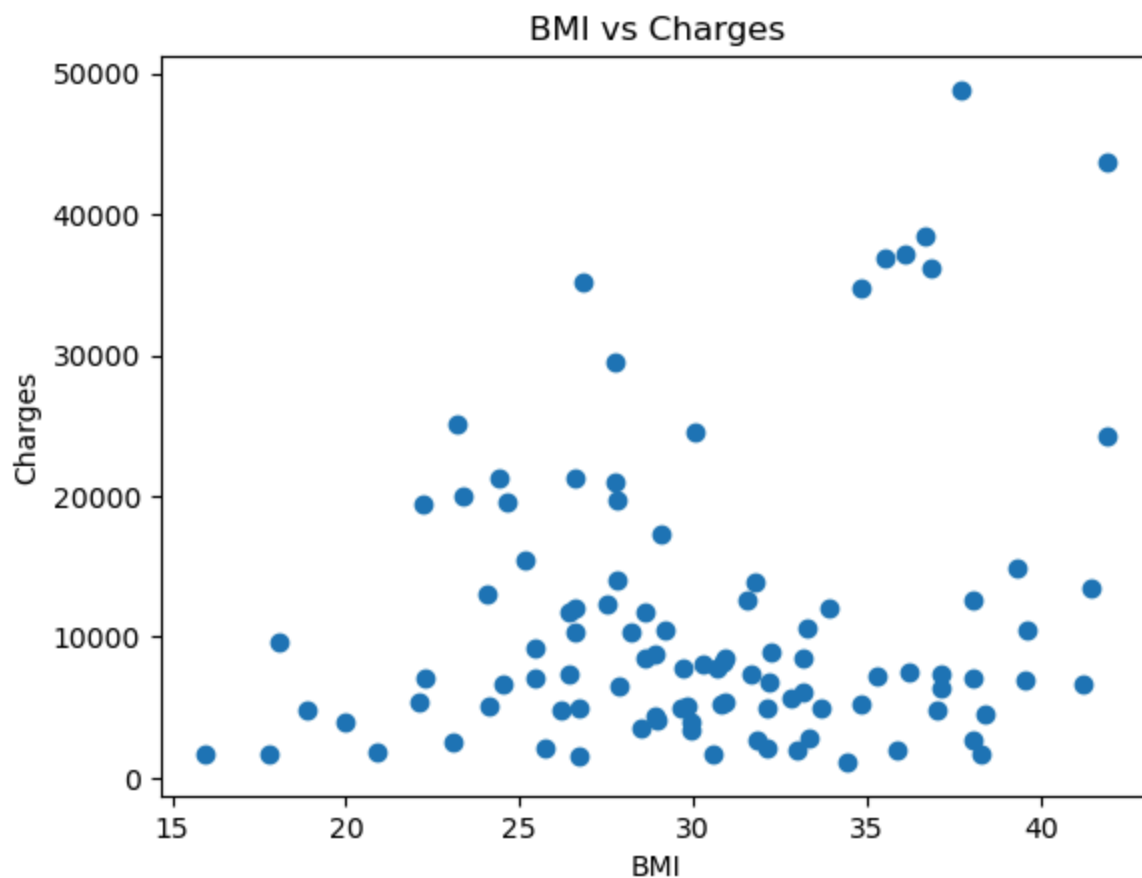
```
Out[2]:
```

	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 ],
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 [ -0.84779878, -0.40837239],
 [  1.59967436, -0.42098178],
 [ -1.0288352 ,  0.97991078],
 [ -0.89740972,  0.41999482],
 [  0.90512117,  2.5232585 ],
 [ -0.43524463,  0.83522924],
 [ -0.64935501,  0.99218077],
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 [ -0.09231987, -0.59729012],
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 [ -0.248986 , -0.2370177 ],
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 [ -0.7990582 , -0.89311993],
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 [  0.71364034, -0.99127423],
 [  0.04519818, -0.94200066],
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 [ -0.30120805, -0.26709156],
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 [  0.33067869, -0.23231965],
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 [  1.00086159,  2.54128039],
 [ -1.39787097, -0.40151114],
 [ -0.25246747, -0.67721216],
```

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[-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],
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[ 0.5865667 , -0.61099612],
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[-2.18120164, -0.9333429 ],
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[-1.41179685,  0.80528615],
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[-0.61454031,  2.34758262],
[ 0.86508427, -0.39662768],
[-0.45091124,  1.79439161],
[-0.0540237 ,  1.29912865],
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[ 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
)

```

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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)
```