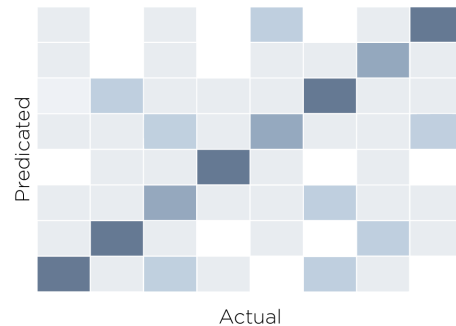


Confusion Matrix



What is our GOAL for this MODULE?

The goal of this module is to check for the accuracy of the prediction model in linear and multi linear regression using the confusion matrix.

What did we ACHIEVE in the class TODAY?

We checked for the accuracy of the prediction model in linear regression and multi linear regression using the confusion matrix and heatmap

Which CONCEPTS/CODING BLOCKS did we cover today?

- Usage of sklearn library
- Training and testing of the prediction model
- Using the heatmap and creating a confusion matrix

How did we DO the activities?

1. We learned about the confusion matrix.

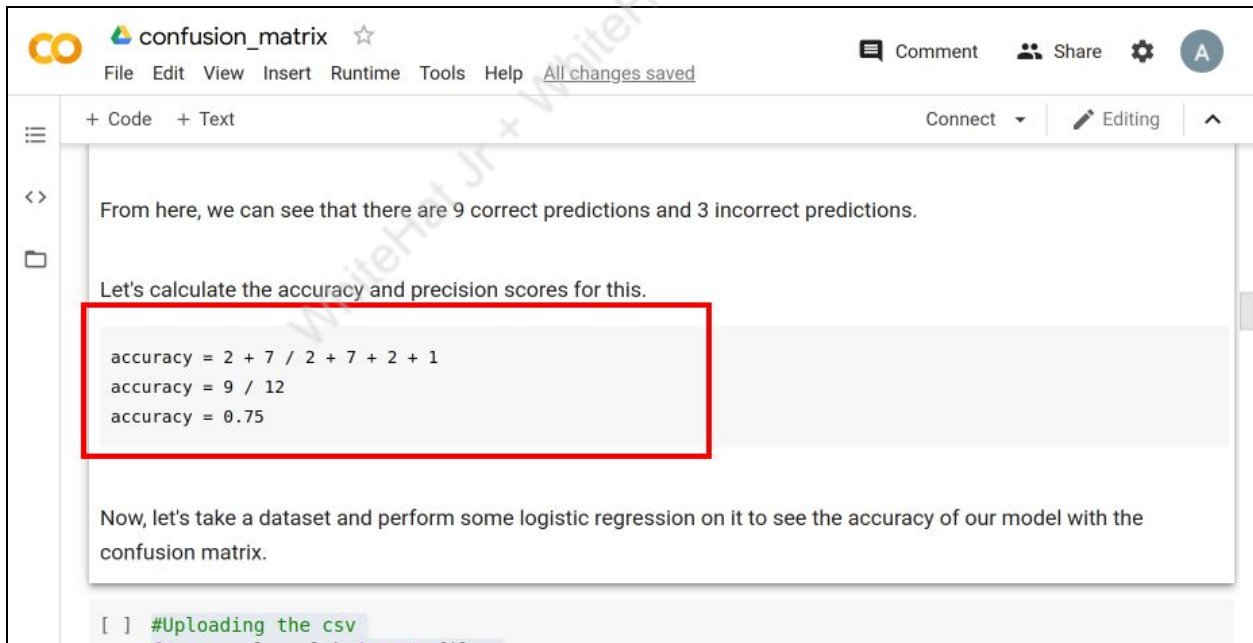
| | | Actual Values | |
|-------------------------|-----|----------------------|----------------|
| | | Yes | No |
| Predicted Values | Yes | True Positive | False Positive |
| | No | False Negative | True Negative |

2. We saw how our prediction matrix looks like.

| Predicted Class | Actual Class |
|-----------------|--------------|
| No Cancer | No Cancer |
| Has Cancer | Has Cancer |
| No Cancer | No Cancer |
| No Cancer | No Cancer |
| No Cancer | Has Cancer |
| Has Cancer | Has Cancer |
| No Cancer | No Cancer |
| Has Cancer | No Cancer |
| No Cancer | No Cancer |
| No Cancer | No Cancer |
| Has Cancer | Has Cancer |
| No Cancer | No Cancer |

| | | Actual Class | |
|-----------------|------------|--------------|-----------|
| | | Has Cancer | No Cancer |
| Predicted Class | Has Cancer | 2 | 2 |
| | No Cancer | 1 | 7 |

3. We calculated the accuracy of the prediction model.



From here, we can see that there are 9 correct predictions and 3 incorrect predictions.

Let's calculate the accuracy and precision scores for this.

```
accuracy = 2 + 7 / 2 + 7 + 2 + 1
accuracy = 9 / 12
accuracy = 0.75
```

Now, let's take a dataset and perform some logistic regression on it to see the accuracy of our model with the confusion matrix.

```
[ ] #Uploading the csv
```

4. We uploaded the heart attack rate csv file in the colab notebook and using the pandas library we read it and print it's content.



```
[ ] #Uploading the csv
from google.colab import files
data_to_load = files.upload()
```

Choose Files No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.
Saving heart.csv to heart.csv

Let's see how the age of the person increases the list of a heart attack, by using single variable logistic regression.

```
[ ] import pandas as pd
df = pd.read_csv("heart.csv")
print(df.head())
```

| | age | sex | cp | trestbps | chol | fbs | ... | exang | oldpeak | slope | ca | thal | target |
|---|-----|-----|----|----------|------|-----|-----|-------|---------|-------|----|------|--------|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | ... | 0 | 2.3 | 0 | 0 | 1 | 1 |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | ... | 0 | 3.5 | 0 | 0 | 2 | 1 |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | ... | 0 | 1.4 | 2 | 0 | 2 | 1 |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | ... | 0 | 0.8 | 2 | 0 | 2 | 1 |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | ... | 1 | 0.6 | 2 | 0 | 2 | 1 |

[5 rows x 14 columns]

```
[ ] from sklearn.model_selection import train_test_split
```

5. Then we split the data into 75% and 25% to train and test the prediction model.



Let's see how the age of the person increases the list of a heart attack, by using single variable logistic regression.

```
[ ] import pandas as pd
df = pd.read_csv("heart.csv")
print(df.head())
```

| | age | sex | cp | trestbps | chol | fbs | ... | exang | oldpeak | slope | ca | thal | target |
|---|-----|-----|----|----------|------|-----|-----|-------|---------|-------|----|------|--------|
| 0 | 63 | 1 | 3 | 145 | 233 | 1 | ... | 0 | 2.3 | 0 | 0 | 1 | 1 |
| 1 | 37 | 1 | 2 | 130 | 250 | 0 | ... | 0 | 3.5 | 0 | 0 | 2 | 1 |
| 2 | 41 | 0 | 1 | 130 | 204 | 0 | ... | 0 | 1.4 | 2 | 0 | 2 | 1 |
| 3 | 56 | 1 | 1 | 120 | 236 | 0 | ... | 0 | 0.8 | 2 | 0 | 2 | 1 |
| 4 | 57 | 0 | 0 | 120 | 354 | 0 | ... | 1 | 0.6 | 2 | 0 | 2 | 1 |

[5 rows x 14 columns]

```
[ ] from sklearn.model_selection import train_test_split
age = df["age"]
heart_attack = df["target"]
age_train, age_test, heart_attack_train, heart_attack_test = train_test_split(age, heart_attack, test_size = 0.25, random_state = 0)
```

```
from sklearn.linear_model import LogisticRegression
import numpy as np
X = np.reshape(age_train.ravel(), (len(age_train), 1))
Y = np.reshape(heart_attack_train.ravel(), (len(heart_attack_train), 1))
classifier = LogisticRegression(random_state = 0)
classifier.fit(X, Y)
```

Training the data:

```

confusion_matrix
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+ Code + Text
heart_attack = 0 if target == 0 else 1

age_train, age_test, heart_attack_train, heart_attack_test = train_test_split(age, heart_attack, test_size = 0.25, random_state = 0)

from sklearn.linear_model import LogisticRegression
import numpy as np

X = np.reshape(age_train.ravel(), (len(age_train), 1))
Y = np.reshape(heart_attack_train.ravel(), (len(heart_attack_train), 1))

classifier = LogisticRegression(random_state = 0)
classifier.fit(X, Y)

/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py:760: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ) to match the expected shape.
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
intercept_scaling=1, l1_ratio=None, max_iter=100,
multi_class='auto', n_jobs=None, penalty='l2',
random_state=0, solver='lbfgs', tol=0.0001, verbose=0,
warm_start=False)

[ ] X_test = np.reshape(age_test.ravel(), (len(age_test), 1))
Y_test = np.reshape(heart_attack_test.ravel(), (len(heart_attack_test), 1))

heart_attack_prediction = classifier.predict(X_test)

predicted_values = []

```

- Then we test the model using the remaining data and substitute the value of 0 as No and 1 as Yes and create 2 arrays for the values in heart attack predictions and actual_values and create a labels array with values Yes and No for the confusion matrix.

```

confusion_matrix
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+ Code + Text
heart_attack = 0 if target == 0 else 1

age_train, age_test, heart_attack_train, heart_attack_test = train_test_split(age, heart_attack, test_size = 0.25, random_state = 0)

from sklearn.linear_model import LogisticRegression
import numpy as np

X = np.reshape(age_train.ravel(), (len(age_train), 1))
Y = np.reshape(heart_attack_train.ravel(), (len(heart_attack_train), 1))

classifier = LogisticRegression(random_state = 0)
classifier.fit(X, Y)

/usr/local/lib/python3.6/dist-packages/sklearn/utils/validation.py:760: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ) to match the expected shape.
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
intercept_scaling=1, l1_ratio=None, max_iter=100,
multi_class='auto', n_jobs=None, penalty='l2',
random_state=0, solver='lbfgs', tol=0.0001, verbose=0,
warm_start=False)

[ ] X_test = np.reshape(age_test.ravel(), (len(age_test), 1))
Y_test = np.reshape(heart_attack_test.ravel(), (len(heart_attack_test), 1))

heart_attack_prediction = classifier.predict(X_test)

predicted_values = []
for i in heart_attack_prediction:
    if i == 0:
        predicted_values.append("No")
    else:
        predicted_values.append("Yes")

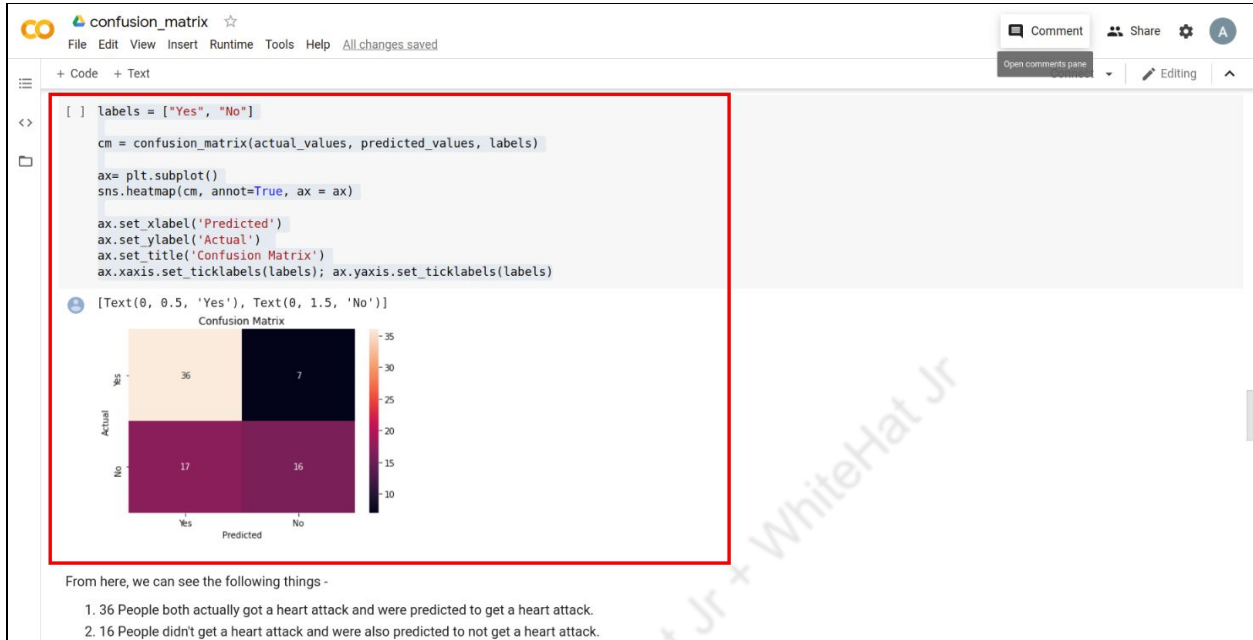
actual_values = []
for i in Y_test.ravel():
    if i == 0:
        actual_values.append("No")
    else:
        actual_values.append("Yes")

[ ] labels = ["Yes", "No"]

cm = confusion_matrix(actual_values, predicted_values, labels)

```

7. We plotted the heatmap for the data and calculated its accuracy.



The screenshot shows a Jupyter Notebook interface with the following code and output:

```
[ ] labels = ["Yes", "No"]

cm = confusion_matrix(actual_values, predicted_values, labels)

ax= plt.subplot()
sns.heatmap(cm, annot=True, ax = ax)

ax.set_xlabel('Predicted')
ax.set_ylabel('Actual')
ax.set_title('Confusion Matrix')
ax.xaxis.set_ticklabels(labels); ax.yaxis.set_ticklabels(labels)

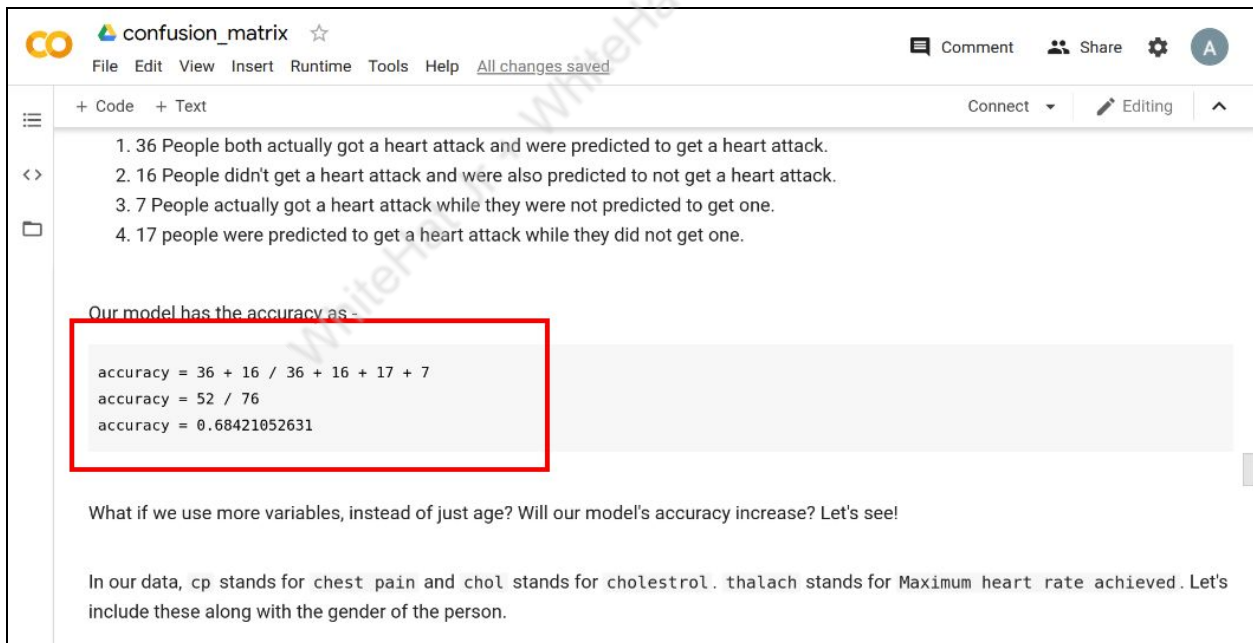
[Text(0, 0.5, 'Yes'), Text(0, 1.5, 'No')]
```

The output is a Confusion Matrix heatmap titled "Confusion Matrix". The x-axis is labeled "Predicted" with ticks for "Yes" and "No". The y-axis is labeled "Actual" with ticks for "Yes" and "No". The heatmap cells contain the following counts:

| Actual \ Predicted | Yes | No |
|--------------------|-----|----|
| Yes | 36 | 7 |
| No | 17 | 16 |

From here, we can see the following things -

1. 36 People both actually got a heart attack and were predicted to get a heart attack.
2. 16 People didn't get a heart attack and were also predicted to not get a heart attack.



The screenshot shows a Jupyter Notebook interface with the following text and code:

1. 36 People both actually got a heart attack and were predicted to get a heart attack.
2. 16 People didn't get a heart attack and were also predicted to not get a heart attack.
3. 7 People actually got a heart attack while they were not predicted to get one.
4. 17 people were predicted to get a heart attack while they did not get one.

Our model has the accuracy as -

```
accuracy = 36 + 16 / 36 + 16 + 17 + 7
accuracy = 52 / 76
accuracy = 0.68421052631
```

What if we use more variables, instead of just age? Will our model's accuracy increase? Let's see!

In our data, cp stands for chest pain and chol stands for cholesterol. thalach stands for Maximum heart rate achieved. Let's include these along with the gender of the person.

8. We also checked how multiple factors(variables) affect the accuracy of a prediction model.
9. We took multiple factors and created two different data frames and then split that data into 75% and 25%.

confusion_matrix

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+ Code + Text

What if we use more variables, instead of just age? Will our model's accuracy increase? Let's see!

In our data, cp stands for chest pain and chol stands for cholesterol. thalach stands for Maximum heart rate achieved. Let's include these along with the gender of the person.

Student side code starts here

```
[
    factors = df[["age", "sex", "cp", "chol", "thalach"]]
    heart_attack = df["target"]

    factors_train, factors_test, heart_attack_train, heart_attack_test = train_test_split(factors, heart_attack, test_size = 0.25, random_state = 0)
```

Since all of age, sex, cp and chol have different measurement units, let's make them scalar to analyse them well.

```
[ ] from sklearn.preprocessing import StandardScaler
    sc_x = StandardScaler()

    factors_train = sc_x.fit_transform(factors_train)
    factors_test = sc_x.transform(factors_test)
```

```
[ ] classifier2 = LogisticRegression(random_state = 0)
    classifier2.fit(factors_train, heart_attack_train)
```

```
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                    intercept_scaling=1, l1_ratio=None, max_iter=100,
                    multi_class='auto', n_jobs=None, penalty='l2',
                    random_state=0, solver='lbfgs', tol=0.0001, verbose=0,
                    warm_start=False)
```

10. Then we made all the factors scalar as all had different measurements.

confusion_matrix

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+ Code + Text

```
[ ] heart_attack = df["target"]

    factors_train, factors_test, heart_attack_train, heart_attack_test = train_test_split(factors, heart_attack, t
```

Since all of age, sex, cp and chol have different measurement units, let's make them scalar to analyse them well.

```
[ ] from sklearn.preprocessing import StandardScaler
    sc_x = StandardScaler()

    factors_train = sc_x.fit_transform(factors_train)
    factors_test = sc_x.transform(factors_test)
```

```
[ ] classifier2 = LogisticRegression(random_state = 0)
    classifier2.fit(factors_train, heart_attack_train)
```

```
LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
                    intercept_scaling=1, l1_ratio=None, max_iter=100,
                    multi_class='auto', n_jobs=None, penalty='l2',
                    random_state=0, solver='lbfgs', tol=0.0001, verbose=0,
                    warm_start=False)
```

11. We trained our model on the remaining data.



```

[ ] heart_attack = df["target"]

factors_train, factors_test, heart_attack_train, heart_attack_test = train_test_split(factors, heart_attack, t

Since all of age, sex, cp and chol have different measurement units, let's make them scaler to analyse them well.

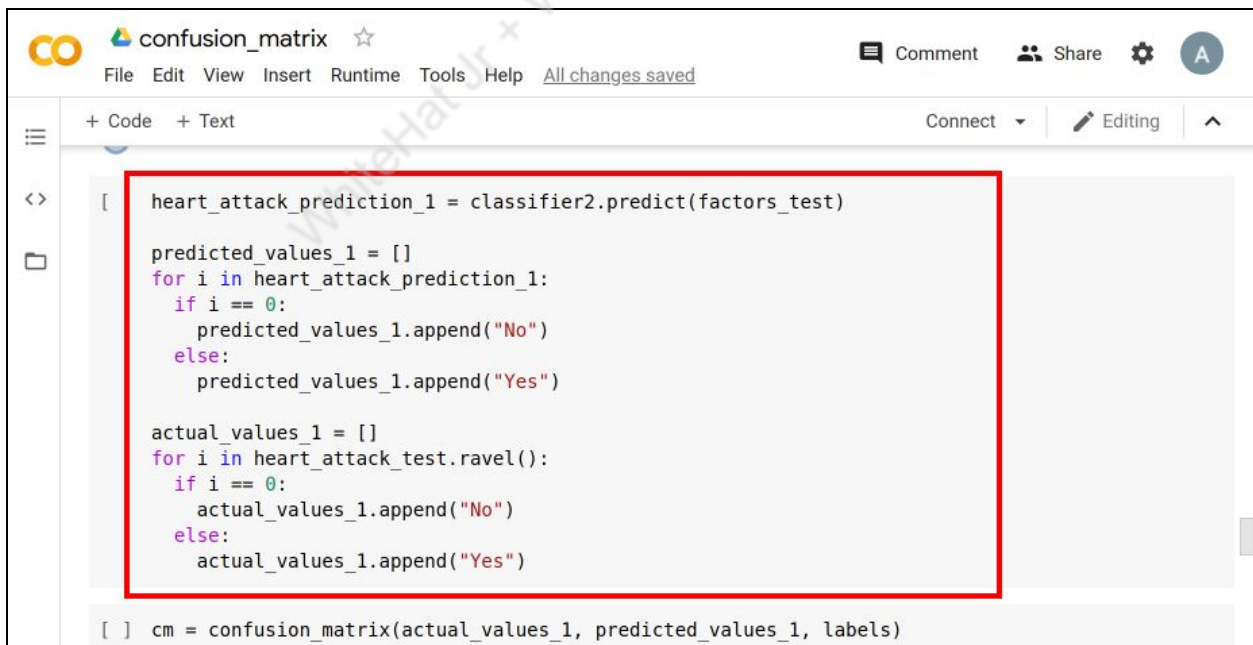
[ ] from sklearn.preprocessing import StandardScaler
sc_x = StandardScaler()

factors_train = sc_x.fit_transform(factors_train)
factors_test = sc_x.transform(factors_test)

[ ] classifier2 = LogisticRegression(random_state = 0)
classifier2.fit(factors_train, heart_attack_train)

LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True,
intercept_scaling=1, l1_ratio=None, max_iter=100,
multi_class='auto', n_jobs=None, penalty='l2',
random_state=0, solver='lbfgs', tol=0.0001, verbose=0,
warm_start=False)
  
```

12. We then substituted the values of 0 by No and 1 by Yes and created it's list for the heat map.



```

[ ] heart_attack_prediction_1 = classifier2.predict(factors_test)

predicted_values_1 = []
for i in heart_attack_prediction_1:
    if i == 0:
        predicted_values_1.append("No")
    else:
        predicted_values_1.append("Yes")

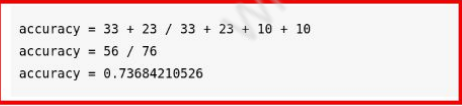
actual_values_1 = []
for i in heart_attack_test.ravel():
    if i == 0:
        actual_values_1.append("No")
    else:
        actual_values_1.append("Yes")

[ ] cm = confusion_matrix(actual_values_1, predicted_values_1, labels)
  
```

13. We plotted the heatmap.



14. Calculated the accuracy of the model.



```

accuracy = 33 + 23 / 33 + 23 + 10 + 10
accuracy = 56 / 76
accuracy = 0.73684210526
  
```

Our model has the accuracy as -

3. 10 People actually got a heart attack while they were not predicted to get one.
 4. 10 people were predicted to get a heart attack while they did not get one.

With the new model that we just built, we have a higher accuracy to detect if a person will get a heart attack or not.

You can try this out with a combination of different set of variables, or add more variables to this to see if that improves the accuracy of the model?

We concluded that more number of variables affects the accuracy of the model.

What's NEXT?

In the next class, we will learn about clustering.

EXTEND YOUR KNOWLEDGE:

Learn more about the confusion matrix from the following doc:

<https://www.geeksforgeeks.org/confusion-matrix-machine-learning/>

WhiteHat Jr + WhiteHat Jr + WhiteHat Jr