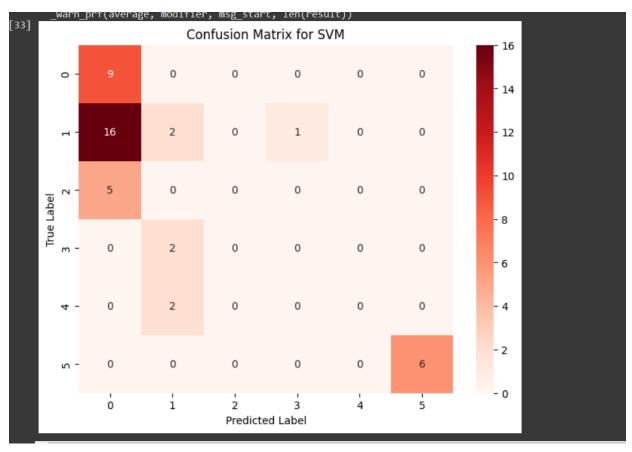
```
from google.colab import drive
       drive.mount('/content/drive')

→ Mounted at /content/drive

[33] # Importing necessary libraries
       import pandas as pd
       import matplotlib.pyplot as plt
       import seaborn as sns
       from sklearn.model_selection import train_test_split
       from sklearn.svm import SVC
        from sklearn.metrics import classification_report, confusion_matrix
       from sklearn.preprocessing import MinMaxScaler
       url = 'https://drive.google.com/uc?id={}'.format('1cau0ol63D2PxRQ9jf-PTqWrpaW6vZSAu')
       glass_data = pd.read_csv(url)
       # Normalizing the features
       scaler = MinMaxScaler()
       X = scaler.fit_transform(glass_data.drop(columns=['Type']))
       y = glass data['Type']
       # Splitting the dataset into training and testing sets
       X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
```

```
# SVM Classifier
# Creating and training the model
clf = SVC(kernel='linear', random_state=0)
clf.fit(X_train, y_train)
# Making predictions
y_pred = clf.predict(X_test)
# Evaluating the model
svm_accuracy = clf.score(X_test, y_test)
print(f"SVM Accuracy: {svm_accuracy:}")
print(confusion_matrix(y_test, y_pred))
print(classification_report(y_test, y_pred, zero_division=1))
# Visualization
svm_cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(8, 6))
sns.heatmap(svm_cm, annot=True, fmt='d', cmap='Reds')
plt.title('Confusion Matrix for SVM')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```

F	SVM A	ccu	rac	y:	0.3953488372093023				
_	[[9	0	0	์ 0	0	0]			
	[16	2	0	1	0	ø <u>j</u>			
	Ĩ 5	0	0	0	0	ø j			
	[0	2	0	0	0	øj			
	[0	2	0	0	0	øj			
	ĪØ	0	0	0	0	6]]			
	-				pre	cision	recall	f1-score	support
				1		0.30	1.00	0.46	9
				2		0.33	0.11	0.16	19
	3					0.00	0.00	0.00	5
	5					0.00	0.00	0.00	2
				6		0.00	0.00	0.00	2
	7					1.00	1.00	1.00	6
	a	ccu	rac	y				0.40	43
	ma	cro	av	g		0.27	0.35	0.27	43
	weigh	ted	av	g'g		0.35	0.40	0.31	43

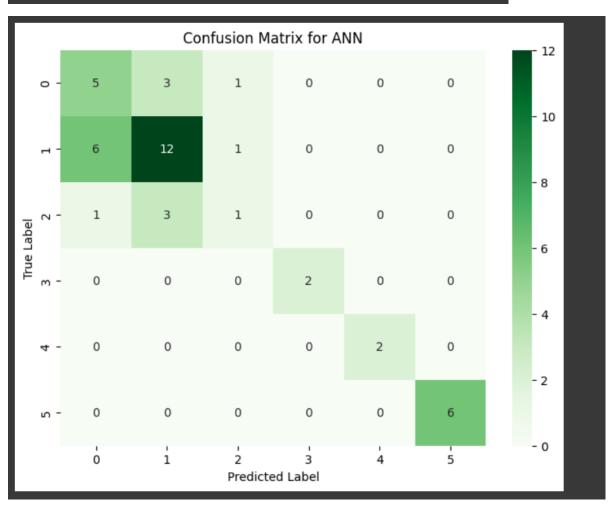


```
# Importing necessary libraries
    import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns
    from sklearn.model_selection import train_test_split
    from sklearn.neural_network import MLPClassifier
    from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
    from sklearn.preprocessing import MinMaxScaler
    url = 'https://drive.google.com/uc?id={}'.format('1cau0o163D2PxRQ9jf-PTqWrpaW6vZSAu')
    glass_data = pd.read_csv(url)
    scaler = MinMaxScaler()
    X = scaler.fit_transform(glass_data.drop(columns=['Type']))
    y = glass_data['Type']
    # Splitting the dataset into training and testing sets
    X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
    # Neural Network Classifier (ANN)
    # Creating and training the model
    mlp = MLPClassifier(hidden_layer_sizes=(100, 50), max_iter=2000, random_state=0)
    mlp.fit(X train, y train)
    # Making predictions
    predict test = mlp.predict(X test)
```

```
# Evaluating the model
ann_accuracy = accuracy_score(y_test, predict_test)
print(f"ANN Accuracy: {ann_accuracy:}")
print(confusion_matrix(y_test, predict_test))
print(classification_report(y_test, predict_test))

# Visualization
mlp_cm = confusion_matrix(y_test, predict_test)
plt.figure(figsize=(8, 6))
sns.heatmap(mlp_cm, annot=True, fmt='d', cmap='Greens')
plt.title('Confusion Matrix for ANN')
plt.xlabel('Predicted Label')
plt.ylabel('True Label')
plt.show()
```

同	ANN Accuracy:				y:	0.6511627906976745						
	[[5	3	1	0	0	0]					
	Ī	6	12	1	0	0	0]					
	Ì	1	3	1	0	0	øj					
	Ì	0	0	0	2	0	øj					
	ĺ	0	0	0	0	2						
	Ī	0	0	0	0	0	6]]					
	_						cision	recall	f1-score	support		
		1					0.42	0.56	0.48	9		
			2				0.67	0.63	0.65	19		
					3		0.33	0.20	0.25	5		
				5			1.00	1.00	1.00	2		
		6					1.00	1.00	1.00	2		
		7					1.00	1.00	1.00	6		
	accuracy								0.65	43		
	macro avg						0.74	0.73	0.73	43		
	weighted avg						0.65	0.65	0.65	43		
		_			_							



Conclusion:

The SVM model achieves an accuracy of, around 0.40.

On the hand the ANN model demonstrates an accuracy of 0.65.

When it comes to the Glass dataset the ANN model outperforms the SVM model by a margin.

- 1. The ANN model has the ability to capture linear patterns that may exist within the dataset something that cannot be adequately handled by the linear kernel used in the SVM model.
- 2. In terms of structure and parameters the ANN model is more complex compared to its counterpart, SVM. This complexity allows for modeling of relationships within the data resulting in improved performance.
- 3. Neural networks like ANN are better suited for datasets like Glass due to their ability to uncover hidden patterns and handle linear relationships that might not be easily captured by linear models such as SVM.

To summarize, based on our observations of accuracy and other metrics we can conclude that the ANN model is more suitable, for handling the Glass dataset.

YouTube Video Link: https://youtu.be/QuixvEYUW34

Github Repo Link: https://github.com/Krypton0626/Bigdata/tree/main/ICP%205