

Faculty of Computers and Artificial Intelligence

Computer Science Department

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## CS 395 Selected Topics in CS-1

### Research Project

Report Submitted for Fulfillment of the Requirements and ILO's  
for Selected Topics in CS-1 course for Fall 2021

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# I. NUMERICAL DATASET

## 1. Project Introduction

### a. Dataset Name

(What is the dataset used?)

### b. Number of classes and their labels

(Specify number of classes and their labels.)

There are 2 Classes

Labels:

1 - Spruce/Fir

2 - Lodgepole Pine

### c. Dataset Samples Numbers

(The total number of samples in dataset)

500K Samples

### d. Training, Validation and Testing

(The number of samples used in training, validation and testing.)

Training = 75% (375K)

Testing = 25% (125K)

## 2.Implementation Details

### **a. Extracted Features**

(How many features were extracted, their names, the dimension of resulted features)

### **b. Cross-validation**

(Is cross-validation is used in any of implemented models? If yes, specify the number of fold and ratio of training/validation)

### **c. Artificial Neural Network (ANN)**

#### **⌘ Hyper-parameters**

(Specify all the hyper-parameters (initial learning rate, optimizer, regularization, batch size, no. of epochs...) with their specified value in implementation)

### **d. Support Vector Machine (SVM)**

#### **⌘ Hyper-parameters**

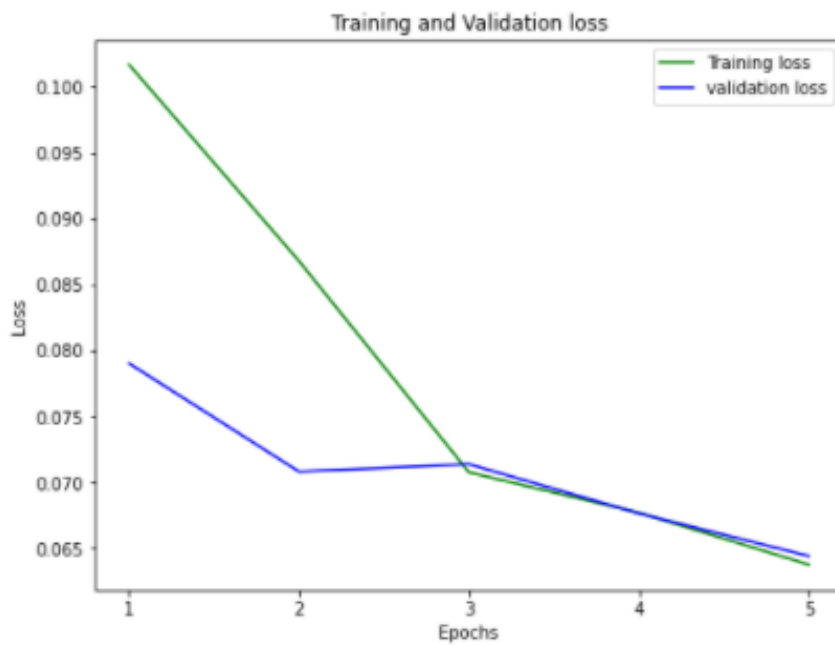
(Specify all the hyper-parameters (optimizer, regularization, ...) with their specified value in implementation)

### 3. Models Results

For each model you should show all these results for your model on testing data (loss curve, accuracy, confusion matrix, ROC curve)

#### a. ANN Results

```
plt.figure(figsize=(8,6))
loss_train = history.history['loss']
loss_val = history.history['val_loss']
epochs = range(1,6)
plt.plot(epochs, loss_train, 'g', label='Training loss')
plt.plot(epochs, loss_val, 'b', label='validation loss')
plt.title('Training and Validation loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.xticks(np.arange(1,6))
plt.legend()
plt.show()
```



Accuracy of the NN

FM

```
In [40]: nn_model.evaluate(X_test_nn, y_test_nn)
```

```
3907/3907 [=====] - 45s 12ms/step - loss: 0.0644 - acc: 0.9724
```

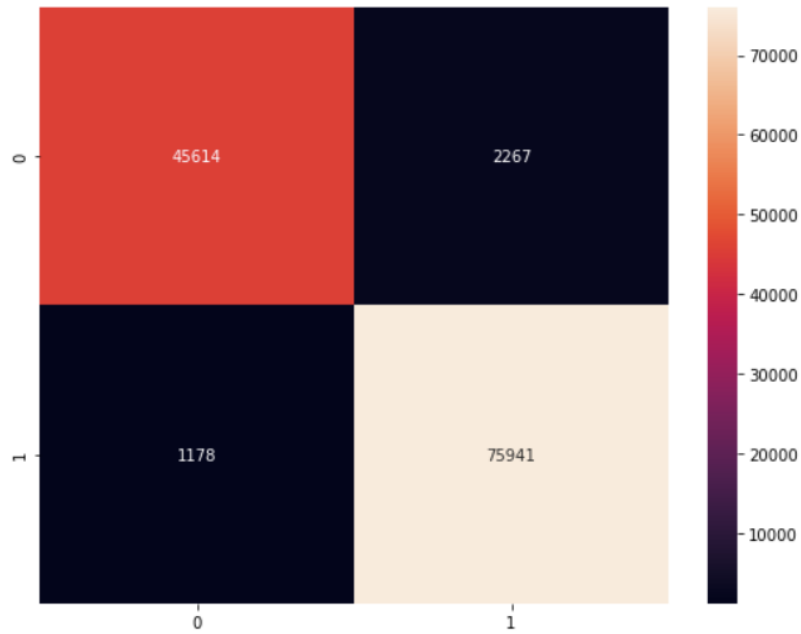
```
Out[40]: [0.06441473215818405, 0.9724400043487549]
```

### Confusion Matrix

FM

```
In [68]: plt.figure(figsize=(9,7))  
sns.heatmap(cm, annot=True, fmt='d')
```

Out[68]: <AxesSubplot:>



## b.SVM Results

### ROC Curve

```
In [42]: y_pred_svm = svm_model.decision_function(X_test)
```

```
In [47]: y_test_repl = y_test.replace(to_replace=[1,2], value=[0,1])
```

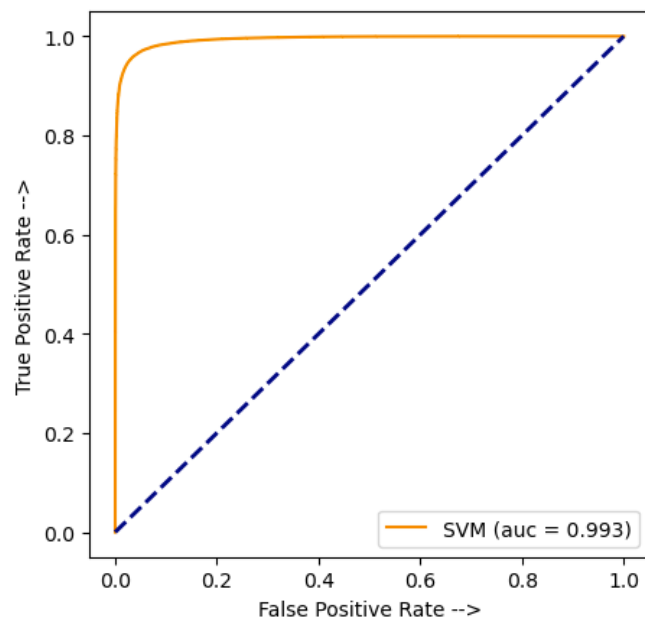
```
In [138]: svm_fpr, svm_tpr, _ = roc_curve(y_test_repl, y_pred_svm)
auc_svm = auc(svm_fpr, svm_tpr)

plt.figure(figsize=(5, 5), dpi=100)
plt.plot(svm_fpr, svm_tpr, color="darkorange", label='SVM (auc = %0.3f)' % auc_svm)
plt.plot([0, 1], [0, 1], color="navy", lw=2, linestyle="--")

plt.xlabel('False Positive Rate -->')
plt.ylabel('True Positive Rate -->')

plt.legend()

plt.show()
```



### Accuracy of the model

```
In [31]: svm_model = SVC()
train(svm_model, "svm")
```

SVM model score on Training data: 96.91733333333333%  
SVM model score on Testing data: 96.188%

## II. IMAGE DATASET

### 1. Project Introduction

#### a. Dataset Name

(What is the dataset used?)

Bayern

#### b. Number of classes and their labels

(Specify number of classes and their labels.)

5 Classes

Labels:

1. Kingsley Coman
2. Joshua Kimmich
3. Robert Lewandowski
4. Manuel Neuer
5. Leory Sane

#### c. Dataset Images Numbers and size

(The total number of images in dataset and the size of each.)

The data contains 230 images

Each image is 64 \* 64 Pixels

Data contains 5 Classes

#### d. Training, Validation and Testing

(The number of images used in training, validation and testing.)

**Number of Training data: 172 (75%)**

**Number of Testing data: 58 (25%)**

## 2. Implementation Details

### a. Extracted Features

(How many features were extracted, their names, the dimension of resulted features)

### b. Cross-validation

(Is cross-validation is used in any of implemented models? If yes, specify the number of fold and ratio of training/validation)

#### Cross Validation

```
In [27]: svm_model = SVC(kernel='linear', gamma='auto')
         scores = cross_val_score(svm_model,X,y, cv=10)
         np.average(scores)
```

```
Out[27]: 0.8434782608695652
```

### c. Artificial Neural Network (ANN)

#### ⌘ Hyper-parameters

(Specify all the hyper-parameters (initial learning rate, optimizer, regularization, batch size, no. of epochs...) with their specified value in implementation)

batch\_size = 10.

epochs = 8.

optimizer = Adam

learning\_rate = 0.0001.

loss = 'categorical\_crossentropy'.



## d. Support Vector Machine (SVM)

### ⌘ Hyper-parameters

(Specify all the hyper-parameters (optimizer, regularization, ...) with their specified value in implementation)

cv=10 , gamma ='auto' ,Kernel= 'linear'

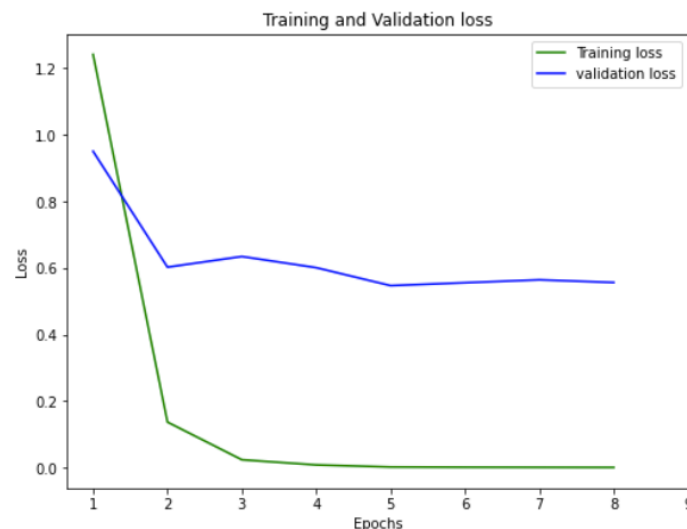
## 3. Models Results

**For each model you should show all these results for your model on testing data** (loss curve, accuracy, confusion matrix, ROC curve)

### a. ANN Results

Loss Curve

```
In [64]: plt.figure(figsize=(8,6))
loss_train = history.history['loss']
loss_val = history.history['val_loss']
epochs = range(1,9)
plt.plot(epochs, loss_train, 'g', label='Training loss')
plt.plot(epochs, loss_val, 'b', label='validation loss')
plt.title('Training and Validation loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.xticks(np.arange(1,10))
plt.legend()
plt.show()
```



### Accuracy of ANN

```
In [62]: nn_model.evaluate(X_test_nn, y_test_nn)
```

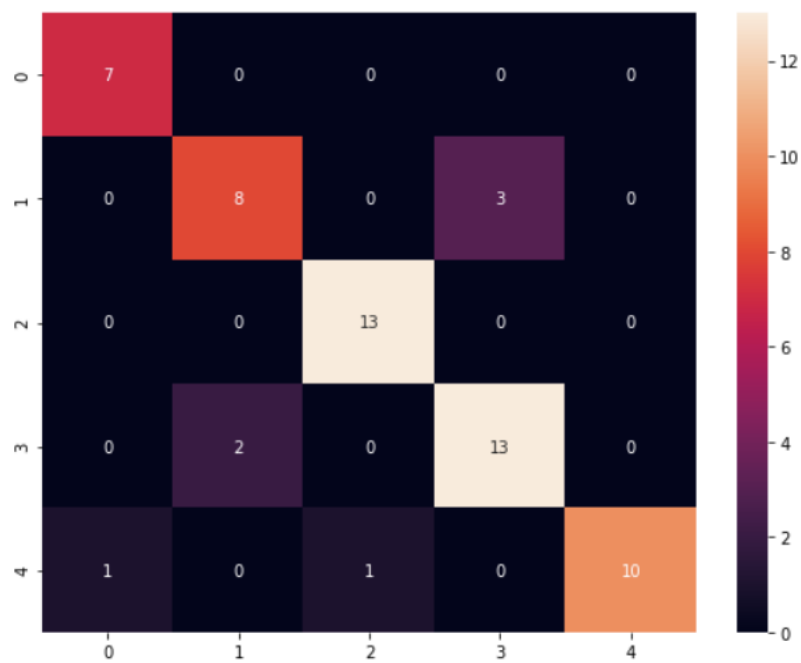
```
2/2 [=====] - 0s 55ms/step - loss: 0.5565 - acc: 0.8793
```

```
Out[62]: [0.5564684271812439, 0.8793103694915771]
```

### Confusion Matrix

```
In [73]: plt.figure(figsize=(9,7))  
sns.heatmap(cm, annot=True)
```

```
Out[73]: <AxesSubplot:>
```



## b.SVM Results

### Score of the model

```
In [29]: svm_model.score(X_test, y_test)
```

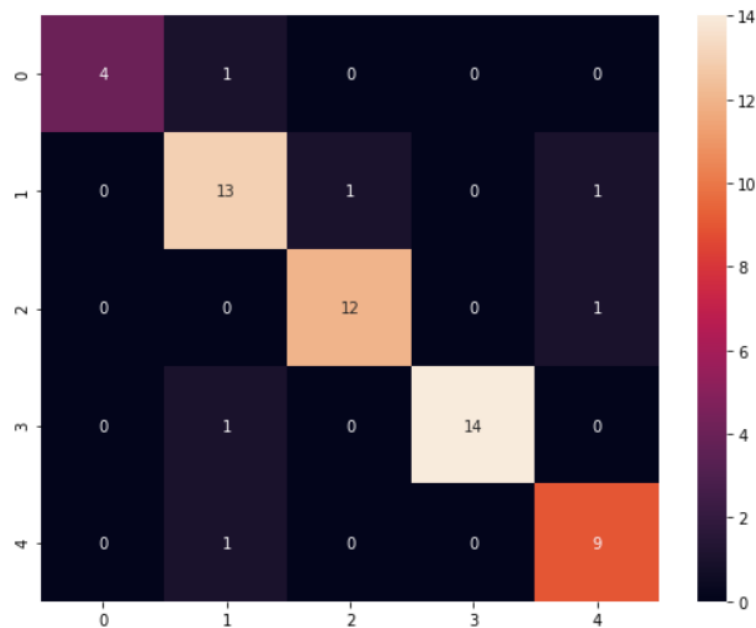
```
Out[29]: 0.896551724137931
```

### Confusion Matrix

```
FM In [30]: cm = confusion_matrix(y_test, svm_model.predict(X_test))  
cm
```

```
Out[30]: array([[ 4,  1,  0,  0,  0],  
               [ 0, 13,  1,  0,  1],  
               [ 0,  0, 12,  0,  1],  
               [ 0,  1,  0, 14,  0],  
               [ 0,  1,  0,  0,  9]], dtype=int64)
```

```
In [31]: plt.figure(figsize=(9,7))  
sns.heatmap(cm, annot=True);
```



## ROC Curve

```
In [46]: plt.figure(figsize=(10,7))
lw = 2
plt.plot(
    fpr[2],
    tpr[2],
    color="darkorange",
    lw=lw,
    label="ROC curve (area = %0.2f)" % roc_auc[2],
)
plt.plot([0, 1], [0, 1], color="navy", lw=lw, linestyle="--")
plt.xlabel("False Positive Rate")
plt.ylabel("True Positive Rate")
plt.title("Receiver operating characteristic")
plt.legend(loc="lower right")
plt.show()
```

