

```
import tensorflow.compat.v1 as tf
tf.disable_v2_behavior()
```

⏏ WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow\_core/python Instructions for updating:  
non-resource variables are not supported in the long term

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from scipy.io import loadmat
from skimage import color
from skimage import io
from sklearn.model_selection import train_test_split
```

```
import tensorflow as tf
%matplotlib inline
```

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

⏏ Go to this URL in a browser: [https://accounts.google.com/o/oauth2/auth?client\\_id=9473](https://accounts.google.com/o/oauth2/auth?client_id=9473)

Enter your authorization code:  
.....  
Mounted at /content/drive

```
import h5py
```

```
# reading the h1py file
data = h5py.File('/content/drive/My Drive/22dec/SVHN_single_grey1.h5', 'r')
```

```
list(data.keys())
```

⏏ ['X\_test', 'X\_train', 'X\_val', 'y\_test', 'y\_train', 'y\_val']

```
# respective values for each key in the h1py file
list(data.values())
```

⏏ [<HDF5 dataset "X\_test": shape (18000, 32, 32), type "<f4">,  
<HDF5 dataset "X\_train": shape (42000, 32, 32), type "<f4">,  
<HDF5 dataset "X\_val": shape (60000, 32, 32), type "<f4">,  
<HDF5 dataset "y\_test": shape (18000,), type "|u1">,  
<HDF5 dataset "y\_train": shape (42000,), type "|u1">,  
<HDF5 dataset "y\_val": shape (60000,), type "|u1">]

```
derive = list(data.keys())
```

```
# Data fetching and understand the train/val/test splits.
```

```
# Get the data
```

```
X_test = list(data[derive[0]])
```

```
X_train= list(data[derive[1]])
```

```
X_val=list(data[derive[2]])
```

```
y_test=list(data[derive[3]])
```

```
y_train=list(data[derive[4]])
```

```
y_val=list(data[derive[5]])
```

```
from numpy import array
```

```
x_train= array(X_train)
```

```
x_test=array(X_test)
```

```
y_train=array(y_train)
```

```
y_test=array(y_test)
```

```
x_train[0]
```

```
↳ array([[ 33.0704,  30.2601,  26.852 , ...,  71.4471,  58.2204,  42.9939],
        [ 25.2283,  25.5533,  29.9765, ..., 113.0209, 103.3639,  84.2949],
        [ 26.2775,  22.6137,  40.4763, ..., 113.3028, 121.775 , 115.4228],
        ...,
        [ 28.5502,  36.212 ,  45.0801, ...,  24.1359,  25.0927,  26.0603],
        [ 38.4352,  26.4733,  23.2717, ...,  28.1094,  29.4683,  30.0661],
        [ 50.2984,  26.0773,  24.0389, ...,  49.6682,  50.853 ,  53.0377]],
      dtype=float32)
```

```
y_train
```

```
↳ array([2, 6, 7, ..., 7, 0, 4], dtype=uint8)
```

```
#Converting train and test labels to one hot vectors
```

```
trainY = tf.keras.utils.to_categorical(y_train, num_classes=10)
```

```
testY = tf.keras.utils.to_categorical(y_test, num_classes=10)
```

```
x_train/= 255
```

```
x_test/= 255
```

```
x_train = x_train.reshape(x_train.shape[0], 1024)
```

```
x_test = x_test.reshape(x_test.shape[0], 1024)
```

```
#● Implement and apply an optimal k-Nearest Neighbor (kNN) classifier (7.5 points)
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
from sklearn import metrics
```

```
import numpy as np
```

```
import pandas as pd
```

```

P2=np.arange(1,30,2)

accuracy_scores=[]
for i in P2:
    # Instantiate the model .
    knn = KNeighborsClassifier(n_neighbors=i,algorithm='brute')
    # Fit the model on the training data.
    knn=knn.fit(x_train, y_train)
    # Predict the response for test dataset
    y_pred = knn.predict(x_test)
    # Storing the results
    accuracy_scores.append(metrics.accuracy_score(y_test,y_pred))

knn_model = KNeighborsClassifier(n_neighbors=27,algorithm='brute')
knn_model = knn_model.fit(x_train, y_train)
y_pred_final = knn_model.predict(x_test)

# Getting the accuracy score for test data
knn_model.score(x_test, y_test)*100

```

↳ 53.227777777777774

```

print(metrics.classification_report(y_test, y_pred_final))
#Precision - Accuracy of positive predictions
#Recall - Fraction of positives That were correctly identified.
#F1 Score - harmonic mean of precision and recall

```

↳

	precision	recall	f1-score	support
0	0.46	0.71	0.56	1814
1	0.46	0.73	0.56	1828
2	0.64	0.54	0.58	1803
3	0.47	0.43	0.45	1719
4	0.64	0.65	0.65	1812
5	0.53	0.40	0.45	1768
6	0.51	0.41	0.46	1832
7	0.71	0.62	0.66	1808
8	0.47	0.37	0.41	1812
9	0.55	0.44	0.49	1804
accuracy			0.53	18000
macro avg	0.54	0.53	0.53	18000
weighted avg	0.54	0.53	0.53	18000

```

# Implement and apply a deep neural network classifier including (feedforward neural netwo

#Clear out tensorflow memory
tf.keras.backend.clear_session()

#Initialize Sequential model
model1 = tf.keras.models.Sequential()

```

```
model1 = tf.keras.models.Sequential()
```

```
#Normalize the data
```

```
model1.add(tf.keras.layers.BatchNormalization())
```

```
#Add 1st hidden layer
```

```
model1.add(tf.keras.layers.Dense(200, activation='relu'))
```

```
#Add 2nd hidden layer
```

```
model1.add(tf.keras.layers.Dense(100, activation='relu'))
```

```
#Add OUTPUT layer
```

```
model1.add(tf.keras.layers.Dense(10, activation='softmax'))
```

```
#Create optimizer with non-default learning rate
```

```
sgd_optimizer = tf.keras.optimizers.SGD(lr=0.03)
```

```
#Compile the model
```

```
model1.compile(optimizer=sgd_optimizer, loss='categorical_crossentropy', metrics=['accuracy'])
```

```
model1.fit(x_train,trainY,  
          validation_data=(x_test,testY),  
          epochs=50,  
          batch_size=24)
```



```
42000/42000 [=====] - 7s 175us/sample - loss: 0.4887 - acc:
Epoch 23/50
42000/42000 [=====] - 7s 176us/sample - loss: 0.4828 - acc:
Epoch 24/50
42000/42000 [=====] - 7s 176us/sample - loss: 0.4760 - acc:
Epoch 25/50
42000/42000 [=====] - 7s 174us/sample - loss: 0.4779 - acc:
Epoch 26/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4623 - acc:
Epoch 27/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4573 - acc:
Epoch 28/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4564 - acc:
Epoch 29/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4424 - acc:
Epoch 30/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.4514 - acc:
Epoch 31/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4435 - acc:
Epoch 32/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4352 - acc:
Epoch 33/50
42000/42000 [=====] - 7s 170us/sample - loss: 0.4288 - acc:
Epoch 34/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.4299 - acc:
Epoch 35/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4216 - acc:
Epoch 36/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.4166 - acc:
Epoch 37/50
42000/42000 [=====] - 8s 180us/sample - loss: 0.4095 - acc:
Epoch 38/50
42000/42000 [=====] - 8s 180us/sample - loss: 0.4144 - acc:
Epoch 39/50
42000/42000 [=====] - 7s 173us/sample - loss: 0.4084 - acc:
Epoch 40/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.4010 - acc:
Epoch 41/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.3978 - acc:
Epoch 42/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.4015 - acc:
Epoch 43/50
42000/42000 [=====] - 7s 174us/sample - loss: 0.3977 - acc:
Epoch 44/50
42000/42000 [=====] - 7s 171us/sample - loss: 0.3884 - acc:
Epoch 45/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.3903 - acc:
Epoch 46/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.3873 - acc:
Epoch 47/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.3820 - acc:
Epoch 48/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.3830 - acc:
Epoch 49/50
42000/42000 [=====] - 7s 172us/sample - loss: 0.3777 - acc:
Epoch 50/50
42000/42000 [=====] - 7s 170us/sample - loss: 0.3759 - acc:
<tensorflow.python.keras.callbacks.History at 0x7f72765f1668>
```

```
predictionsNN = model1.predict(x_test)

ClassificationReportNN = metrics.classification_report(testY.argmax(axis=1), predictionsNN)

print(ClassificationReportNN)
```