

```
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

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

import h5py

# reading the h5py 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 h5py 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.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 [=====] - 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)
```