

ASSIGNMENT: - 2

```
import tensorflow as tf
from tensorflow import keras
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import random
get_ipython().run_line_magic("matplotlib", "inline")
```

```
mnist = tf.keras.datasets.mnist
(x_train, y_train), (x_test, y_test) = mnist.load_data()
```

```
len(x_train)
len(x_test)
x_train.shape
x_test.shape
x_train[0]
```

```
plt.matshow(x_train[11]) #we can change it by changing the argument
```

```
x_train = x_train/255
x_test = x_test/255
```

```
x_train[11]
```

```
model = keras.Sequential([
    keras.layers.Flatten(input_shape=(28, 28)),
    keras.layers.Dense(128, activation='relu'),
    keras.layers.Dense(10, activation='softmax')
])
```

```
model.summary()
```

```
model.compile(optimizer='sgd',
              loss='sparse_categorical_crossentropy',
              metrics=['accuracy'])
```

```
history=model.fit(x_train, y_train,validation_data=(x_test,y_test),epochs=10)
```

```
test_loss, test_acc=model.evaluate(x_test,y_test)
print("Loss=%.3f" %test_loss)
print("Accuracy=%.3f" %test_acc)
```

```
n=random.randint(0,9999)
plt.imshow(x_test[n])
plt.show()
```

```
predicted_value=model.predict(x_test)
print("Handwritten nuber in the image is= %d" %np.argmax(predicted_value))
```

```
get_ipython().run_line_magic('pinfo2','history.history')
history.history.keys()
```

```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
```

```
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
```

```
plt.title('Training Loss and accuracy')
plt.ylabel('accuracy/Loss')
plt.xlabel('epoch')
plt.legend(['accuracy', 'val_accuracy', 'loss', 'val_loss'])
plt.show()
```

```
keras_model_path="/content/sample_data"
model.save(keras_model_path)
```

```
restored_keras_model = tf.keras.models.load_model(keras_model_path)
```

ASSIGNMENT: - 3

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

```
import numpy as np
import pandas as pd
import os
import random
```

```
import matplotlib.image as mpimg
import matplotlib.pyplot as plt
import seaborn as sns
import cv2
```

```
import tensorflow
```

```
from keras.preprocessing.image import ImageDataGenerator
```

```
%matplotlib inline
```

```
TrainingImagePath="/content/drive/MyDrive/Image /train"
TestImagePath="/content/drive/MyDrive/Image /test"
```

ValidationImagePath="/content/drive/MyDrive/Image /valid"

```
train_datagen = ImageDataGenerator(
    rescale = 1./255,
    shear_range=0.1,
    zoom_range=0.1,
    horizontal_flip=True
)
test_datagen = ImageDataGenerator(rescale=1./255)

training_set = train_datagen.flow_from_directory(
    TrainingImagePath,
    target_size=(128,128),
    batch_size=32,
    class_mode="categorical"
)

test_set = test_datagen.flow_from_directory(
    TestImagePath,
    target_size = (128,128),
    batch_size=32,
    class_mode="categorical"
)

valid_set = test_datagen.flow_from_directory(
    ValidationImagePath,
    target_size=(128,128),
    batch_size=32,
    class_mode="categorical"
)
```

```
def showImages(class_name):
    random_index = random.choice(list(range(1,49)))
    folder_path = os.path.join(TrainingImagePath, class_name)
    try:
        image_path = os.path.join(folder_path, str(random_index).zfill(3)+".jpg")
        plt.imshow(mping.imread(image_path))
    except:
        image_path = os.path.join(folder_path, str(random_index).zfill(2)+".jpg")
        plt.imshow(mping.imread(image_path))
    plt.title(class_name)
    plt.axis(False)
```

```
plt.figure(figsize = (20,20))
for labels,number in training_set.class_indices.items():
    plt.subplot(6,6,number+1)
    showImages(labels)

test_set.class_indices
```

```
'''##### Creating lookup table for all balls #####
#####'''
# class_indices have the numeric tag for each balls
TrainClasses=training_set.class_indices

# Storing the face and the numeric tag for future reference
ResultMap={}
for ballValue,ballName in zip(TrainClasses.values(),TrainClasses.keys()
):
    ResultMap[ballValue]=ballName

# Saving the face map for future reference
import pickle
with open(R"E:\Data Sets\Balls Classification\ResultsMap.pkl", 'wb') as
f:
    pickle.dump(ResultMap, f, pickle.HIGHEST_PROTOCOL)

print("Mapping of Face and its ID",ResultMap)

# The number of neurons for the output layer is equal to the number of
faces
OutputNeurons=len(ResultMap)
print('\n The Number of output neurons: ', OutputNeurons)
```

```
from keras.models import Sequential
from keras.layers import Convolution2D
from keras.layers import MaxPool2D
from keras.layers import Flatten
from keras.layers import Dense
```

```
classifier= Sequential()
```

```
classifier.add(Convolution2D(32, kernel_size=(3, 3), strides=(1, 1), input_shape=(128,128,3), activation='relu'))
```

```
classifier.add(MaxPool2D(pool_size=(2,2)))
```

```
''
```

```
classifier.add(Convolution2D(64, kernel_size=(3, 3), strides=(1, 1), activation='relu'))
```

```
classifier.add(MaxPool2D(pool_size=(2,2)))
```

```
classifier.add(Flatten())
```

```
classifier.add(Dense(256, activation='relu'))
```

```
classifier.add(Dense(OutputNeurons, activation='softmax'))
```

```
classifier.compile(loss='categorical_crossentropy', optimizer = 'rmsprop', metrics=["accuracy"])
```

```
classifier.summary()
```

```
import time
```

```
# Measuring the time taken by the model to train
```

```
StartTime=time.time()
```

```
# Starting the model training
```

```
model_history=classifier.fit_generator(
```

```
    training_set,
```

```
    steps_per_epoch=len(training_set),
```

```
t),
```

```
    epochs=20,
```

```
    validation_data=valid_set,
```

```
    validation_steps=len(valid_set)
```

```
,
```

```
    verbose=1)
```

```
EndTime=time.time()
```

```
print("##### Total Time Taken: ", round((EndTime-StartTime)/60), 'Minutes #####')
```

```
accuracy = model_history.history['accuracy']
val_accuracy = model_history.history['val_accuracy']

loss = model_history.history['loss']
val_loss = model_history.history['val_loss']
```

```
plt.figure(figsize=(15,10))

plt.subplot(2, 2, 1)
plt.plot(accuracy, label = "Training accuracy")
plt.plot(val_accuracy, label="Validation accuracy")
plt.legend()
plt.title("Training vs validation accuracy")

plt.subplot(2,2,2)
plt.plot(loss, label = "Training loss")
plt.plot(val_loss, label="Validation loss")
plt.legend()
plt.title("Training vs validation loss")

plt.show()
```

ASSIGNMENT: -4

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import tensorflow as tf

from sklearn.metrics import accuracy_score, precision_score, recall_score
from sklearn.model_selection import train_test_split
from tensorflow.keras import layers, losses
from tensorflow.keras.datasets import fashion_mnist
from tensorflow.keras.models import Model
```

```
(x_train, _), (x_test, _) = fashion_mnist.load_data()
```

```
x_train = x_train/255.  
x_test = x_test/255.
```

```
print(x_train.shape)  
print(x_test.shape)
```

```
latent_dim = 64  
class Autoencoder(Model):  
    def __init__(self, latent_dim):  
        super(Autoencoder, self).__init__()  
        self.latent_dim = latent_dim  
        self.encoder = tf.keras.Sequential([  
            layers.Flatten(),  
            layers.Dense(latent_dim, activation='relu'),  
        ])  
        self.decoder = tf.keras.Sequential([  
            layers.Dense(784, activation='sigmoid'),  
            layers.Reshape((28, 28))  
        ])  
  
    def call(self, x):  
        encoded = self.encoder(x)  
        decoded = self.decoder(encoded)  
        return decoded  
  
autoencoder = Autoencoder(latent_dim)
```

```
autoencoder.compile(optimizer='adam', loss=losses.MeanSquaredError())  
  
autoencoder.fit(x_train, x_train,  
                epochs=10,  
                shuffle=True,  
                validation_data=(x_test, x_test))
```

```
encoded_imgs = autoencoder.encoder(x_test).numpy()  
decoded_imgs = autoencoder.decoder(encoded_imgs).numpy()  
  
n = 10  
plt.figure(figsize=(20, 4))  
for i in range(n):  
    # display original  
    ax = plt.subplot(2, n, i + 1)  
    plt.imshow(x_test[i])  
    plt.title("original")
```



```
plt.gray()
ax.get_xaxis().set_visible(False)
ax.get_yaxis().set_visible(False)

# display reconstruction
ax = plt.subplot(2, n, i + 1 + n)
plt.imshow(decoded_imgs[i])
plt.title("reconstructed")
plt.gray()
ax.get_xaxis().set_visible(False)
ax.get_yaxis().set_visible(False)
plt.show()
```