

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),  
    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()
```

ASSIGNMENT 5:-

```
import numpy as np
import keras.backend as K
from keras.models import Sequential
from keras.layers import Dense, Embedding, Lambda
from keras.utils import np_utils
from keras.preprocessing import sequence
from keras.preprocessing.text import Tokenizer
import gensim

data = open("/content/corona.txt", "r")
covid_data= [text for text in data if text.count("")>=2]
vectorize=Tokenizer()
vectorize.fit_on_texts(covid_data)
covid_data=vectorize.texts_to_sequences(covid_data)
total_vocab=sum(len(s) for s in covid_data)
word_count=len(vectorize.word_index)+1
window_size=2

def cbow_model(data, windows_size, total_vocab):
    total_length=window_size*2
    for text in data:
        text_len=len(text)
        for idx, word in enumerate(text):
            context_word=[]
            target=[]
            begin=idx-window_size
            end=idx+window_size+1
            context_word.append([text[i] for i in range(begin, end) if 0<=
i< text_len and i!=idx])
            target.append(word)
            contextual = sequence.pad_sequences(context_word, total_length=to
tal_length)
            final_target=np_utils.to_categorical(target, total_vocab)
            yield(contextual, final_target)

model=Sequential()
model.add(Embedding(input_dim=total_vocab,output_dim=100,input_length=w
indow_size*2))
model.add(Lambda(lambda x:K.mean(x,axis=1), output_shape=(100,)))
model.add(Dense(total_vocab, activation="softmax"))
model.compile(loss="categorical_crossentropy", optimizer="adam")
for i in range(10):
    cost=0
    for x, y in cbow_model(data,window_size, total_vocab):
        cost+=model.train_on_batch(contextual, final_target)
```

```

    print(i, cost)

dimensions = 100
vect_file=open("/content/drive/MyDrive/vector.txt", "w")
vect_file.write('{} {} \n'.format(total_vocab, dimensions))

weight=model.get_weights()[0]
for text, i in vectorize.word_index.items():
    final_vec="".join(map(str, list(weight[i,:])))
    vect_file.write('{} {} \n'.format(text, final_vec))
vect_file.close()

cbow_output=gensim.models.KeyedVectors.load_word2vec_format("/content/drive/MyDrive/vector.txt", binary=False)
cbow_output.most_similar(positive=["virus"])

```

OUTPUT:-

```
0 0 1 0 2 0 3 0 4 0 5 0 6 0 7 0 8 0 9 0
```