

Exp No: 7

## BUILD AUTOENCODERS WITH KERAS/TENSORFLOW

Aim:

To build autoencoders with Keras/TensorFlow.

Procedure:

1. Download and load the dataset.
2. Perform analysis and preprocessing of the dataset.
3. Build a simple neural network model using Keras/TensorFlow.
4. Compile and fit the model.
5. Perform prediction with the test dataset.
6. Calculate performance metrics.

Program:

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
from keras import layers from
```

```
keras.datasets import mnist from
```

```
keras.models import Model
```

```
def preprocess(array):
```

```
    """Normalizes the supplied array and reshapes it."""
```

```
    array = array.astype("float32") / 255.0
```

```
    array = np.reshape(array, (len(array), 28, 28, 1))
```

```
    return array
```

```
def noise(array):
```

```
    """Adds random noise to each image in the supplied array."""
```

```

noise_factor = 0.4
noisy_array = array + noise_factor * np.random.normal(
    loc=0.0, scale=1.0, size=array.shape
)
return np.clip(noisy_array, 0.0, 1.0)

```

```

def display(array1, array2):
    """Displays ten random images from each array."""
    n = 10
    indices = np.random.randint(len(array1),
        size=n)
    images1 = array1[indices, :]
    images2 = array2[indices, :]

    plt.figure(figsize=(20, 4))
    for i, (image1, image2) in enumerate(zip(images1, images2)):
        ax = plt.subplot(2, n, i + 1)
        plt.imshow(image1.reshape(28, 28))
        plt.gray()
        ax.get_xaxis().set_visible(False)
        ax.get_yaxis().set_visible(False)

        ax = plt.subplot(2, n, i + 1 + n)
        plt.imshow(image2.reshape(28, 28))
        plt.gray()
        ax.get_xaxis().set_visible(False)
        ax.get_yaxis().set_visible(False)

    plt.show()

# Since we only need images from the dataset to encode and decode, we

```

```
# won't use the labels.
(train_data, _), (test_data, _) = mnist.load_data()

# Normalize and reshape the data
train_data = preprocess(train_data)
test_data = preprocess(test_data)

# Create a copy of the data with added noise
noisy_train_data = noise(train_data)
noisy_test_data = noise(test_data)

# Display the train data and a version of it with added noise
display(train_data, noisy_train_data)
input = layers.Input(shape=(28, 28, 1))

# Encoder
x = layers.Conv2D(32, (3, 3), activation="relu", padding="same")(input)
x = layers.MaxPooling2D((2, 2), padding="same")(x)
x = layers.Conv2D(32, (3, 3), activation="relu", padding="same")(x)
x = layers.MaxPooling2D((2, 2), padding="same")(x)

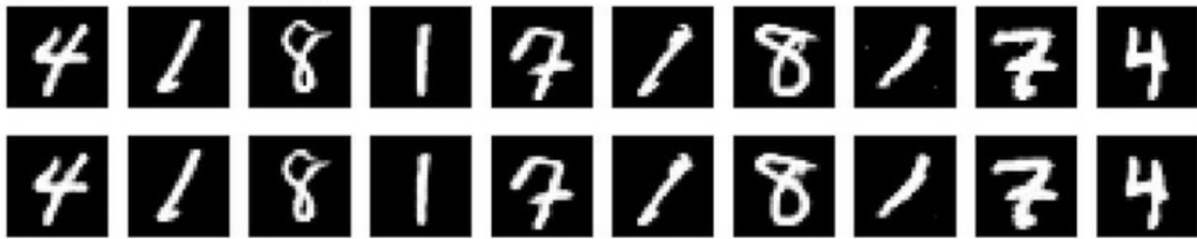
# Decoder
x = layers.Conv2DTranspose(32, (3, 3), strides=2, activation="relu",
padding="same")(x)
x = layers.Conv2DTranspose(32, (3, 3), strides=2, activation="relu",
padding="same")(x)
x = layers.Conv2D(1, (3, 3), activation="sigmoid", padding="same")(x)

# Autoencoder
autoencoder = Model(input, x)
```

```
autoencoder.compile(optimizer="adam", loss="binary_crossentropy")
autoencoder.summary() autoencoder.fit( x=train_data, y=train_data,
epochs=10, batch_size=128, shuffle=True, validation_data=
(test_data, test_data),
)
predictions = autoencoder.predict(test_data)
display(test_data, predictions)
autoencoder.fit(
```

```
x=noisy_train_data,
y=train_data,
epochs=10,
batch_size=128,
shuffle=True,
validation_data=(noisy_test_data, test_data),
)
predictions = autoencoder.predict(noisy_test_data)
display(noisy_test_data, predictions)
```

Output:



Result:

Autocoder has been successfully built using tensorflow/keras.