3. Applying the Convolution Neural Network on computer vision problems.

Step 1: Import Required Libraries

Start by importing TensorFlow and necessary modules.

Program:

import tensorflow as tf

from tensorflow.keras import layers, models

from tensorflow.keras.datasets import mnist

Step 2: Load and Preprocess the Data

Load the MNIST dataset and normalize the pixel values to be between 0 and 1.

Program:

Load the dataset

Reshape the data to add a single channel (since images are grayscale)

```
x train = x train.reshape((x train.shape[0], 28, 28, 1))
```

$$x_{test} = x_{test.reshape}((x_{test.shape}[0], 28, 28, 1))$$

Normalize the pixel values to be between 0 and 1

$$x_{train}, x_{test} = x_{train} / 255.0, x_{test} / 255.0$$

Step 3: Define a Simple CNN Model

Create a simple CNN with one convolutional layer followed by max pooling, and a dense layer for classification.

Program:

Define a basic CNN model

model = models.Sequential([

```
layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)), # 32 filters, 3x3 kernel
layers.MaxPooling2D((2, 2)), # Pooling layer with 2x2 pool size
layers.Flatten(), # Flatten the output from the convolutional layer
layers.Dense(64, activation='relu'), # Fully connected layer with 64 neurons
layers.Dense(10, activation='softmax') # Output layer for 10 classes
])
```

Step 4: Compile the Model

Compile the model using the Adam optimizer, a suitable loss function, and accuracy as the metric.

Program:

```
# Compile the model

model.compile(optimizer='adam',

loss='sparse_categorical_crossentropy',

metrics=['accuracy'])
```

Step 5: Train the Model

Train the CNN model on the training dataset for 5 epochs.

Program:

```
# Train the model
model.fit(x_train, y_train, epochs=5, batch_size=64)
```

Step 6: Evaluate the Model

After training, evaluate the model's performance on the test dataset.

Program:

```
# Evaluate the model
test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')
```

4. Image classification on MNIST dataset (CNN model with a fully connected layer).

Step 1: Import Required Libraries

Start by importing TensorFlow and necessary modules.

Program:

import tensorflow as tf

from tensorflow.keras import layers, models

from tensorflow.keras.datasets import mnist

Step 2: Load and Preprocess the Data

Load the MNIST dataset and normalize the pixel values to be between 0 and 1.

Program:

```
# Load the dataset
```

Reshape the data to add a single channel (since images are grayscale)

$$x_{test} = x_{test.reshape}((x_{test.shape}[0], 28, 28, 1))$$

Normalize the pixel values to be between 0 and 1

Step 3: Define a Simple CNN Model

Create a simple CNN with one convolutional layer followed by max pooling, and a dense layer for classification.

Program:

```
# Define a basic CNN model
```

```
model = models.Sequential([
```

First convolutional layer layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)),

layers.MaxPooling2D((2, 2)), # Second convolutional layer

layers.Conv2D(64, (3, 3), activation='relu'),

```
layers.MaxPooling2D((2, 2)), # Third convolutional layer layers.Conv2D(64, (3, 3), activation='relu'), layers.Flatten(), layers.Dense(64, activation='relu'), layers.Dense(10, activation='softmax')
])
```

Step 4: Compile the Model

Compile the model using the Adam optimizer, a suitable loss function, and accuracy as the metric.

Program:

```
model.compile(optimizer='adam',

loss='sparse_categorical_crossentropy',

metrics=['accuracy'])
```

Step 5: Train the Model

Train the CNN model on the training dataset for 5 epochs.

Program:

```
model.fit(x_train, y_train, epochs=5, batch_size=64)
```

Step 6: Evaluate the Model

After training, evaluate the model's performance on the test dataset.

Program:

```
test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test_acc}')
```

Step 7: Make Predictions

You can also use the model to predict the digit class of a test image.

Program:

```
predictions = model.predict(x_test)
predicted_label = tf.argmax(predictions[0])
print(f'Predicted label: {predicted_label.numpy()}')
```

4

5. Applying the Deep Learning Models in the field of Natural Language Processing.

Step 1: Import Required Libraries

We will use TensorFlow/Keras for building the model, and the IMDb dataset provided by Keras for the text data.

Program:

import tensorflow as tf

from tensorflow.keras import layers, models

from tensorflow.keras.datasets import imdb

from tensorflow.keras.preprocessing import sequence

Step 2: Load and Preprocess the IMDb Dataset

Program:

```
# Load the IMDb dataset (only top 10,000 most frequent words)
```

num words = 10000

(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=num_words)

Set the maximum length of each review to 500 words (truncating or padding shorter/longer reviews)

max len = 500

x_train = sequence.pad_sequences(x_train, maxlen=max_len)

x_test = sequence.pad_sequences(x_test, maxlen=max_len)

Step 3: Build a Deep Learning Model for Text Classification

Program:

Define the model

model = models.Sequential([

Embedding layer: Converts words into dense vectors

layers.Embedding(input_dim=num_words, output_dim=64, input length=max len),

LSTM layer: Captures sequential patterns

```
layers.LSTM(64),
  # Dense layer with sigmoid activation for binary classification
  layers.Dense(1, activation='sigmoid')
])
# Compile the model
model.compile(optimizer='adam',
        loss='binary crossentropy',
        metrics=['accuracy'])
# Display the model architecture
model.summary()
Step 4: Train the Model
Program:
# Train the model
model.fit(x train, y train, epochs=5, batch size=64, validation split=0.2)
Step 5: Evaluate the Model
Program:
# Evaluate the model
test_loss, test_acc = model.evaluate(x_test, y_test)
print(f'Test accuracy: {test acc}')
Step 6: Make Predictions on New Text
Program:
# Example review (preprocessed as integer sequences)
new review = [1, 14, 22, 16, 43, 530, 973, 1622, 1385, 65, 4581, 66, 394, 2,
530, 973]
new_review = sequence.pad_sequences([new_review], maxlen=max_len)
# Predict sentiment (1: Positive, 0: Negative)
prediction = model.predict(new review)
print(f'Predicted sentiment: {"Positive" if prediction >= 0.5 else "Negative"}')
```

6. Train a sentiment analysis model on IMDB dataset, use RNN layers with LSTM/GRU notes.

Step 1: Import Required Libraries

Start by importing TensorFlow, Keras, and necessary modules.

Program:

import tensorflow as tf

from tensorflow.keras import layers, models

from tensorflow.keras.datasets import imdb

from tensorflow.keras.preprocessing import sequence

Step 2: Load and Preprocess the Data

Load the IMDb dataset and preprocess the reviews by limiting the vocabulary and padding the sequences.

Program:

```
# Load the IMDb dataset with a limit of 10,000 words

num_words = 10000

(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=num_words)

# Set the maximum length for each review

max_len = 500

x_train = sequence.pad_sequences(x_train, maxlen=max_len)

x_test = sequence.pad_sequences(x_test, maxlen=max_len)

Step 3: Define the Model Architecture Using LSTM

Program:# Define the model using LSTM

model = models.Sequential([
    layers.Embedding(input_dim=num_words, output_dim=64, input_length=max_len),
    layers.LSTM(64), # LSTM layer with 64 units
    layers.Dense(1, activation='sigmoid') # Output layer for binary classification
])
```

```
# Compile the model
model.compile(optimizer='adam',
        loss='binary crossentropy',
        metrics=['accuracy'])
# Display the model architecture
model.summary()
Step 4: Train the Model
Program:
# Train the model
model.fit(x train, y train, epochs=5, batch size=64, validation split=0.2)
Step 5: Evaluate the Model on the Test Set
Program:
# Evaluate the model
test loss, test acc = model.evaluate(x test, y test)
print(f'Test accuracy: {test acc}')
Step 6: Example Prediction
Program:
# Example review (preprocessed as integer sequences)
new review = [1, 14, 22, 16, 43, 530, 973, 1622, 1385, 65, 4581, 66, 394, 2,
530, 973]
new review = sequence.pad sequences([new review], maxlen=max len)
# Predict sentiment (1: Positive, 0: Negative)
prediction = model.predict(new review)
print(f'Predicted sentiment: {"Positive" if prediction >= 0.5 else "Negative"}')
generate images(generator, 10)
```