```
Roll No: 08
Code:
import numpy as np
class HopfieldNetwork:
  def___init_(self, n_neurons):
     self.n\_neurons = n\_neurons
     self.weights = np.zeros((n_neurons, n_neurons))
  def train(self, patterns):
     for pattern in patterns:
       self.weights += np.outer(pattern, pattern)
     np.fill_diagonal(self.weights, 0)
  def predict(self, pattern):
     energy = -0.5 * np.dot(np.dot(pattern, self.weights), pattern)
     return np.sign(np.dot(pattern, self.weights) + energy)
if name == 'main ':
  patterns = np.array([
     [1, 1, -1, -1],
     [-1, -1, 1, 1],
     [1, -1, 1, -1],
     [-1, 1, -1, 1]
  1)
  n_neurons = patterns.shape[1]
  network = HopfieldNetwork(n_neurons)
  network.train(patterns)
  for pattern in patterns:
     prediction = network.predict(pattern)
     print('Input pattern:', pattern)
     print('Predicted pattern:', prediction)
```

Input pattern: [ 1 1 -1 -1]
Predicted pattern: [-1. -1. -1. -1.]
Input pattern: [-1 -1 1 1]
Predicted pattern: [-1. -1. -1. -1.]
Input pattern: [ 1 -1 1 -1]
Predicted pattern: [-1. -1. -1. -1.]
Input pattern: [-1 1 -1 1]
Predicted pattern: [-1. -1. -1. -1.]

Roll No: 08

# Code: import keras from keras.datasets import cifar10 from keras.models import Sequential from keras.layers import Dense, Dropout, Flatten from keras.layers import Conv2D, MaxPooling2D from keras.optimizers import SGD from keras.preprocessing.image import ImageDataGenerator # Load CIFAR-10 dataset (X\_train, y\_train), (X\_test, y\_test) = cifar10.load\_data() # Define the model model = Sequential() model.add(Conv2D(32, (3, 3), activation='relu', input\_shape=(32, 32, 3))) model.add(Conv2D(32, (3, 3), activation='relu')) model.add(MaxPooling2D(pool\_size=(2, 2))) model.add(Dropout(0.25))model.add(Conv2D(64, (3, 3), activation='relu')) model.add(Conv2D(64, (3, 3), activation='relu')) model.add(MaxPooling2D(pool\_size=(2, 2))) model.add(Dropout(0.25)) model.add(Flatten()) model.add(Dense(512, activation='relu')) model.add(Dropout(0.5))model.add(Dense(10, activation='softmax')) # Define data generators train\_datagen = ImageDataGenerator(rescale=1./255, shear\_range=0.2, zoom\_range=0.2, horizontal\_flip=True) test\_datagen = ImageDataGenerator(rescale=1./255) # Prepare the data train\_set = train\_datagen.flow(X\_train, y\_train, batch\_size=32) test\_set = test\_datagen.flow(X\_test, y\_test, batch\_size=32) # Compile the model

```
sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)
model.compile(loss='categorical_crossentropy', optimizer=sgd, metrics=['accuracy'])
# Train the model
model.fit_generator(train_set, steps_per_epoch=len(X_train)//32, epochs=100,
validation_data=test_set, validation_steps=len(X_test)//32)
# Evaluate the model
score = model.evaluate(test_set, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

```
Downloading data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
Epoch 1/100
/usr/local/lib/python3.10/dist-packages/keras/optimizers/legacy/gradient descent.py:114:
UserWarning: The `lr` argument is deprecated, use `learning_rate` instead.
super(). init (name, **kwargs)
<ipython-input-15-75bb0166727e>:40: UserWarning: `Model.fit generator` is deprecated and
will be removed in a future version. Please use 'Model.fit', which supports generators.
model.fit_generator(train_set, steps_per_epoch=len(X_train)//32, epochs=100,
validation_data=test_set, validation_steps=len(X_test)//32)
0.9977 - val_loss: nan - val_accuracy: 1.0000
Epoch 2/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 3/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 4/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 5/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 6/100
```

```
1.0000 - val loss: nan - val accuracy: 1.0000
Epoch 7/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 8/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 9/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 10/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 11/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 12/100
1.0000 - val loss: nan - val accuracy: 1.0000
Epoch 13/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
Epoch 14/100
1.0000 - val loss: nan - val accuracy: 1.0000
Epoch 15/100
1.0000 - val loss: nan - val accuracy: 1.0000
Epoch 16/100
1.0000 - val_loss: nan - val_accuracy: 1.0000
```

### Roll No: 08

### Code:

```
import tensorflow as tf
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.datasets import load_breast_cancer
df=load_breast_cancer()

X_train,X_test,y_train,y_test=train_test_split(df.data,df.target,test_size=0.20,random_state=42)
sc=StandardScaler()
X_train=sc.fit_transform(X_train)
X_test=sc.transform(X_test)

model=tf.keras.models.Sequential([tf.keras.layers.Dense(1,activation='sigmoid',input_shape=(X_train.shape[1],))])
model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
model.fit(X_train,y_train,epochs=5)
y_pred=model.predict(X_test)
```

test\_loss,test\_accuracy=model.evaluate(X\_test,y\_test)

print("accuracy is",test\_accuracy)

Epoch 1/5	
-	======] - 1s 2ms/step - loss: 0.5449 - accuracy: 0.7385
Epoch 2/5	
15/15 [====================================	======] - 0s 2ms/step - loss: 0.4896 - accuracy: 0.7802
Epoch 3/5	
15/15 [====================================	======] - 0s 2ms/step - loss: 0.4439 - accuracy: 0.8286
Epoch 4/5	
15/15 [====================================	======] - 0s 2ms/step - loss: 0.4074 - accuracy: 0.8462
Epoch 5/5	
15/15 [====================================	======] - 0s 3ms/step - loss: 0.3776 - accuracy: 0.8593
4/4 [===================================	=====] - 0s 5ms/step
4/4 [===================================	=====] - 0s 4ms/step - loss: 0.3090 - accuracy: 0.9298
accuracy is 0.9298245906829834	

### Roll No: 08

#### Code:

```
import tensorflow as tf
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
from tensorflow.keras.utils import to_categorical
(X_train, y_train), (X_test, y_test) = mnist.load_data()
X_{train} = X_{train.reshape}(-1, 28, 28, 1) / 255.0
X_{\text{test}} = X_{\text{test.reshape}}(-1, 28, 28, 1) / 255.0
y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
model = Sequential([
  Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)),
  MaxPooling2D((2, 2)),
  Conv2D(64, (3, 3), activation='relu'),
  MaxPooling2D((2, 2)),
  Conv2D(64, (3, 3), activation='relu'),
  Flatten(),
  Dense(64, activation='relu'),
  Dense(10, activation='softmax')
1)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.fit(X_train, y_train, batch_size=64, epochs=10, verbose=1)
loss, accuracy = model.evaluate(X_test, y_test)
print(f"Test Loss: {loss}")
print(f"Test Accuracy: {accuracy}")
```

```
Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
Epoch 1/10
0.9448
Epoch 2/10
0.9835
Epoch 3/10
0.9878
Epoch 4/10
0.9908
Epoch 5/10
0.9926
Epoch 6/10
0.9939
Epoch 7/10
0.9950
Epoch 8/10
0.9957
Epoch 9/10
0.9961
Epoch 10/10
0.9971
0.9921
Test Loss: 0.028454650193452835
Test Accuracy: 0.9921000003814697
```

```
Roll No: 08
Code:
import tensorflow as tf
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.optimizers import Adam
# Load and preprocess the MNIST dataset
(X_train, y_train), (X_test, y_test) = mnist.load_data()
X_{train} = X_{train} / 255.0
X_{\text{test}} = X_{\text{test}} / 255.0
# Define the model architecture
model = Sequential([
  Flatten(input_shape=(28, 28)),
  Dense(128, activation='relu'),
  Dense(10, activation='softmax')
])
# Compile the model
model.compile(optimizer=Adam(learning_rate=0.001),
        loss='sparse_categorical_crossentropy',
         metrics=['accuracy'])
# Train the model
model.fit(X_train, y_train, batch_size=64, epochs=10, verbose=1)
# Evaluate the model
loss, accuracy = model.evaluate(X_test, y_test)
print(f"Test Loss: {loss}")
print(f"Test Accuracy: {accuracy}")
```

```
Epoch 1/10
0.9153
Epoch 2/10
0.9612
Epoch 3/10
0.9723
Epoch 4/10
0.9783
Epoch 5/10
0.9833
Epoch 6/10
0.9864
Epoch 7/10
0.9892
Epoch 8/10
0.9913
Epoch 9/10
0.9927
Epoch 10/10
0.9944
0.9804
Test Loss: 0.06786014884710312
Test Accuracy: 0.98040002584457
```

