

# Assignment 5 - Convolutional Neural Network

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In this notebook, I implement a convolutional neural network for classifying images using Keras. We carry out two tasks:

1. Handwritten Digit Classification
2. Handwritten Letter Classification

```
In [20]: # keras import statements
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense
from keras.layers import Dropout
from keras.layers import Flatten
from keras.layers.convolutional import Conv2D
from keras.layers.convolutional import MaxPooling2D
from keras.utils import np_utils

# other imports
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt
from sklearn import preprocessing
from sklearn import model_selection
```

## 1. Handwritten digit recognition

### Get digits from MNIST Dataset

We also show some images for visualisation. The data is loaded a smaller version of the mnist dataset because the actual files are too large to download and reupload during submission.

```
In [21]: # Loading data from CSV file
X = pd.read_csv('mnist.csv')
y = X['label']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33,
, random_state=42)
```

### Preprocessing and Visualizing the Training Data

```
In [22]: # Storing the labels in a separate vector y_train
y_train = np.array(X_train['label'])

# Dropping the labels column
X_train = X_train.drop(['label'], axis=1)

# Normalizing the values of the pixels
X_train = X_train / 255

# Converting X_train dataframe to numpy array
X_train = np.array(X_train)

print("-----")
print("SAMPLE OF TRAINING DATA")
print("-----")

# Plotting the data
plt.figure(figsize=(8,8))
for i in range(16):
    plt.subplot(4,4,i+1)
    plt.axis('off')
    r = np.random.randint(X_train.shape[0])    # Get a random image to show
    plt.title('True Label: '+str(y_train[r])) # Show its label as title
    plt.imshow(X_train[r].reshape(28,28), cmap='magma') # Plotting the image
plt.show()

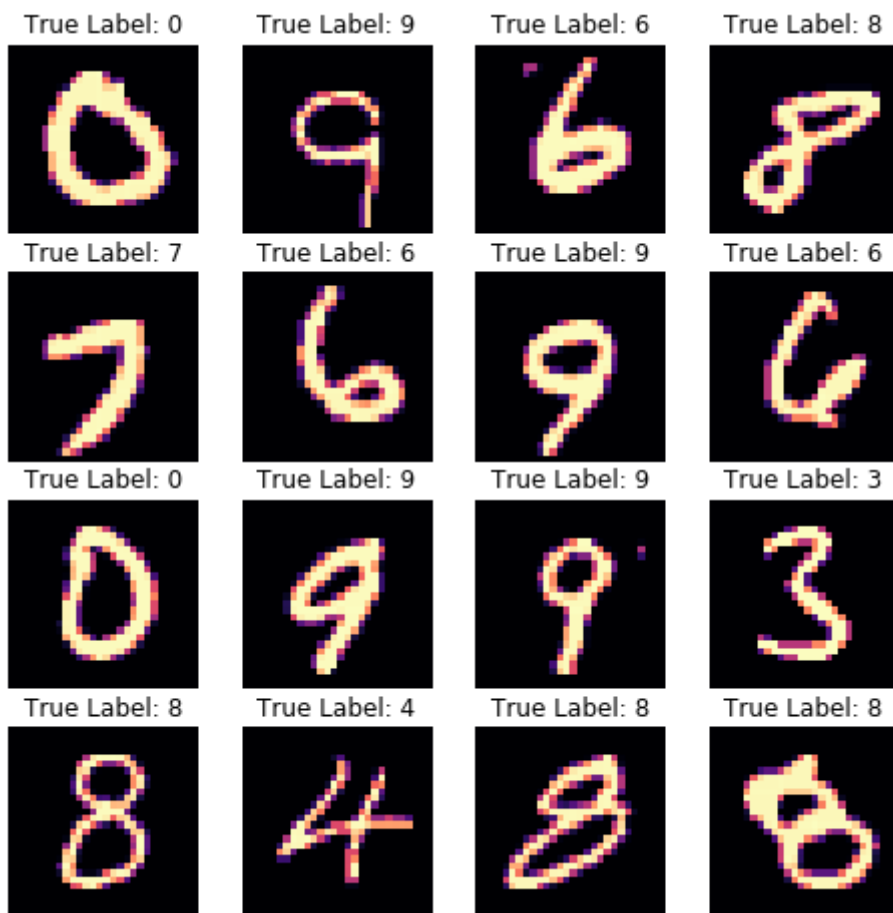
# Converting the labels vector into one hot encoded form
oh = np.zeros((y_train.size, y_train.max()+1))
oh[np.arange(y_train.size), y_train] = 1
y_train = np.array(oh)

# Reshaping X_train to the format of the network
X_train = X_train.reshape(X_train.shape[0], 28, 28, 1).astype('float32')
```

---

SAMPLE OF TRAINING DATA

---



### Preprocessing and Visualizing the Testing Data

```
In [23]: # Storing the labels in a separate vector y_train
y_test = np.array(X_test['label'])

# Dropping the labels column
X_test = X_test.drop(['label'], axis=1)

# Normalizing the values of the pixels
X_test = X_test/ 255

# Converting X_train dataframe to numpy array
X_test = np.array(X_test)

print("-----")
print("SAMPLE OF TESTING DATA")
print("-----")

# Plotting the data
plt.figure(figsize=(8,8))
for i in range(16):
    plt.subplot(4,4,i+1)
    plt.axis('off')
    r = np.random.randint(X_test.shape[0]) # Get a random image to show
    plt.title('True Label: '+str(y_test[r])) # Show its label as title
    plt.imshow(X_test[r].reshape(28,28), cmap='inferno') # Plotting the image
plt.show()

# Converting the labels vector into one hot encoded form
oh = np.zeros((y_test.size, y_test.max()+1))
oh[np.arange(y_test.size), y_test] = 1
y_test = np.array(oh)

# Reshaping X_train to the format of the network
X_test = X_test.reshape(X_test.shape[0], 28, 28, 1).astype('float32')

# number of classes
no_of_label_classes = y_test.shape[1]
```

---

SAMPLE OF TESTING DATA

---



### Designing the Convolutional Neural Network

For the architecture, I use a simple convolutional model due to hardware constraints, but it gives good performance anyway. The model has the following architecture:

- 2D convolutional layer of size 5x5 filters, stride 1 and 32 filters. This is followed by an activation function layer of ReLU.
- Max pooling layer of size 2x2
- Dropout for normalisation with factor = 0.2
- Affine (fully-connected) layer of 128 units, followed by a relu activation.
- Finally, another affine layer with the number of desired output units is used with a softmax activation, since it is a classification task.

I use the adam optimiser and categorical cross entropy loss in the model.

```
In [25]: def create_model():  
    '''  
        Function to create a sequential keras model  
    '''  
  
    model = Sequential()  
    model.add(Conv2D(32, (5, 5), input_shape=(28, 28, 1), activation='relu'))  
    model.add(MaxPooling2D(pool_size=(2, 2)))  
    model.add(Dropout(0.2))  
    model.add(Flatten())  
    model.add(Dense(128, activation='relu'))  
    model.add(Dense(no_of_label_classes, activation='softmax'))  
  
    # Compile model  
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])  
    return model
```

```
In [26]: # initialising hyperparameters  
iterations = 20  
b_size = 16
```

## Training the Model

Here I train the model, and for each epoch I also print out the time taken, loss, and accuracy. Likewise, after training, I print out the model summary and all of the weights of the neural network, as required for the assignment.

```
In [27]: # Creating the model
model = create_model()
# Fitting the model
history = model.fit(X_train, y_train, epochs=iterations, batch_size=batch_size, verbose=2)

model.summary()
for layer in model.layers: print(layer.get_config(), layer.get_weights())
```

```

Epoch 1/20
1759/1759 - 12s - loss: 0.1706 - accuracy: 0.9493
Epoch 2/20
1759/1759 - 11s - loss: 0.0607 - accuracy: 0.9808
Epoch 3/20
1759/1759 - 15s - loss: 0.0434 - accuracy: 0.9853
Epoch 4/20
1759/1759 - 15s - loss: 0.0287 - accuracy: 0.9904
Epoch 5/20
1759/1759 - 15s - loss: 0.0200 - accuracy: 0.9937
Epoch 6/20
1759/1759 - 15s - loss: 0.0183 - accuracy: 0.9939
Epoch 7/20
1759/1759 - 16s - loss: 0.0143 - accuracy: 0.9951
Epoch 8/20
1759/1759 - 11s - loss: 0.0119 - accuracy: 0.9954
Epoch 9/20
1759/1759 - 10s - loss: 0.0106 - accuracy: 0.9960
Epoch 10/20
1759/1759 - 11s - loss: 0.0101 - accuracy: 0.9966
Epoch 11/20
1759/1759 - 11s - loss: 0.0069 - accuracy: 0.9976
Epoch 12/20
1759/1759 - 11s - loss: 0.0067 - accuracy: 0.9978
Epoch 13/20
1759/1759 - 10s - loss: 0.0061 - accuracy: 0.9979
Epoch 14/20
1759/1759 - 11s - loss: 0.0075 - accuracy: 0.9974
Epoch 15/20
1759/1759 - 11s - loss: 0.0071 - accuracy: 0.9978
Epoch 16/20
1759/1759 - 10s - loss: 0.0047 - accuracy: 0.9985
Epoch 17/20
1759/1759 - 10s - loss: 0.0050 - accuracy: 0.9982
Epoch 18/20
1759/1759 - 10s - loss: 0.0066 - accuracy: 0.9982
Epoch 19/20
1759/1759 - 11s - loss: 0.0039 - accuracy: 0.9988
Epoch 20/20
1759/1759 - 10s - loss: 0.0026 - accuracy: 0.9990
Model: "sequential_1"

```

Layer (type)	Output Shape	Param #
conv2d_1 (Conv2D)	(None, 24, 24, 32)	832
max_pooling2d_1 (MaxPooling2D)	(None, 12, 12, 32)	0
dropout_1 (Dropout)	(None, 12, 12, 32)	0
flatten_1 (Flatten)	(None, 4608)	0
dense_2 (Dense)	(None, 128)	589952
dense_3 (Dense)	(None, 10)	1290
Total params: 592,074		



Trainable params: 592,074

Non-trainable params: 0

```

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```

## Testing the Model

```
In [28]: print("Evaluating on Test Data")
results = model.evaluate(X_test, y_test, batch_size=b_size, verbose=0)
print("Test Accuracy:", results[1])
```

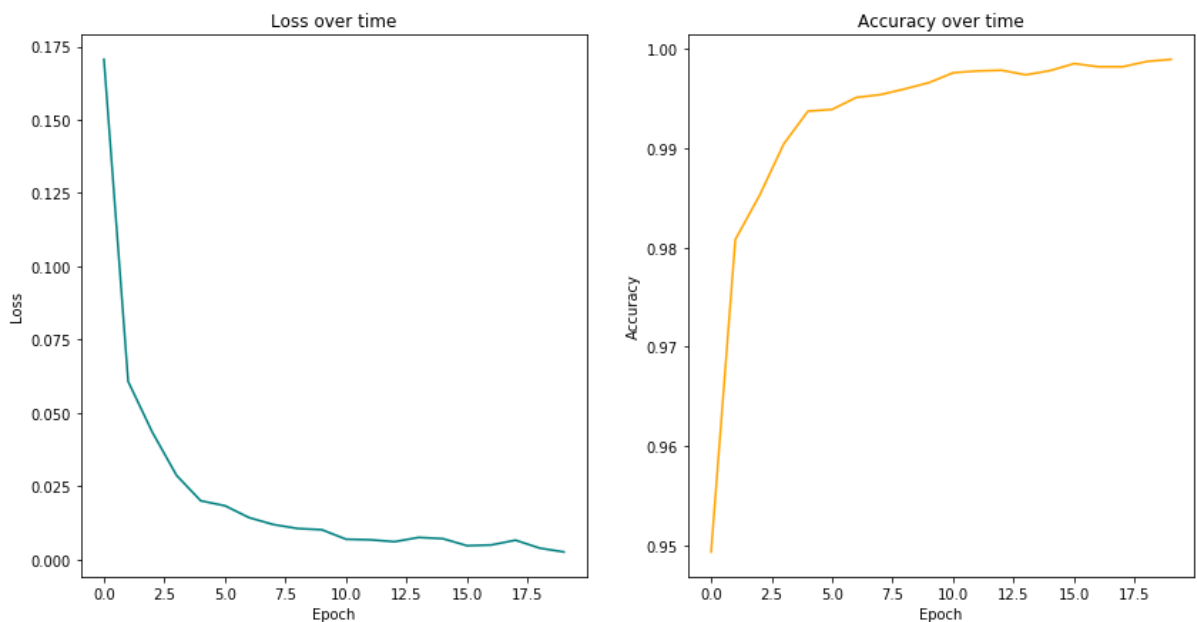
Evaluating on Test Data  
Test Accuracy: 0.9852092266082764

## Visualizing the Performance

```
In [31]: # accuracy vs epochs
plt.figure(figsize=(14,7))

plt.subplot(1,2,1)
plt.plot(history.history['loss'], c='teal')
plt.title('Loss over time')
plt.ylabel('Loss')
plt.xlabel('Epoch')

plt.subplot(1,2,2)
plt.plot(history.history['accuracy'], c='orange')
plt.title('Accuracy over time')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.show()
```



## 2. Handwritten character recognition



```
In [57]: # Loading data from CSV file
X = pd.read_csv('emnist-letters.csv')
y = X['label']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33
, random_state=42)
```

## Preprocessing and Visualizing the Training Data

```
In [58]: # Storing the labels in a separate vector y_train
y_train = np.array(X_train['label'])

# Dropping the labels column
X_train = X_train.drop(['label'], axis=1)

# Normalizing the values of the pixels
X_train = X_train / 255

# Converting X_train dataframe to numpy array
X_train = np.array(X_train)

print("-----")
print("SAMPLE OF TRAINING DATA")
print("-----")

# Plotting the data
plt.figure(figsize=(8,8))
for i in range(16):
    plt.subplot(4,4,i+1)
    plt.axis('off')
    r = np.random.randint(X_train.shape[0])    # Get a random image to show
    plt.title('True Label: '+str(chr(ord('@')+y_train[r]))) # Show its label as title
    plt.imshow(X_train[r].reshape(28,28), cmap='magma')    # Plotting the image
plt.show()

# Converting the labels vector into one hot encoded form
oh = np.zeros((y_train.size, y_train.max()+1))
oh[np.arange(y_train.size), y_train] = 1
y_train = np.array(oh)

# Reshaping X_train to the format of the network
X_train = X_train.reshape(X_train.shape[0], 28, 28, 1).astype('float32')
```

---

SAMPLE OF TRAINING DATA

---



### Preprocessing and Visualizing the Testing Data

```
In [59]: # Storing the labels in a separate vector y_train
y_test = np.array(X_test['label'])

# Dropping the labels column
X_test = X_test.drop(['label'], axis=1)

# Normalizing the values of the pixels
X_test = X_test/ 255

# Converting X_train dataframe to numpy array
X_test = np.array(X_test)

print("-----")
print("SAMPLE OF TESTING DATA")
print("-----")

# Plotting the data
plt.figure(figsize=(8,8))
for i in range(16):
    plt.subplot(4,4,i+1)
    plt.axis('off')
    r = np.random.randint(X_test.shape[0]) # Get a random image to show
    plt.title('True Label: '+str(chr(ord('@')+y_test[r]))) # Show its label as title
    plt.imshow(X_test[r].reshape(28,28), cmap='inferno') # Plotting the image
plt.show()

# Converting the labels vector into one hot encoded form
oh = np.zeros((y_test.size, y_test.max()+1))
oh[np.arange(y_test.size), y_test] = 1
y_test = np.array(oh)

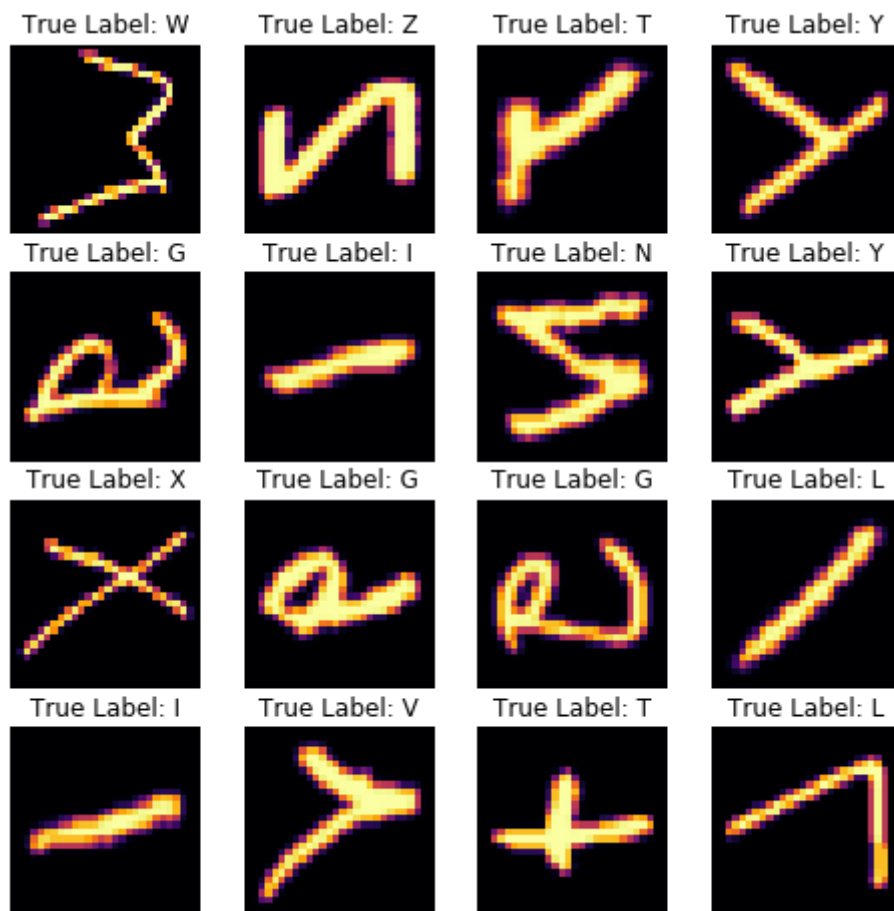
# Reshaping X_train to the format of the network
X_test = X_test.reshape(X_test.shape[0], 28, 28, 1).astype('float32')

# number of class labels
no_of_label_classes = y_test.shape[1]
```

---

SAMPLE OF TESTING DATA

---



### Initializing the Hyperparameters

```
In [60]: iterations = 10  
         b_size = 20
```

### Training the Model

Here I train the model, and for each epoch I also print out the time taken, loss, and accuracy. Likewise, after training, I print out the model summary and all of the weights of the neural network, as required for the assignment.

We use the same network as the first task.

```
In [66]: # Creating the model
model = create_model()
model.summary()

# Fitting the model
history = model.fit(X_train, y_train, epochs=iterations, batch_size=b_size
, verbose=2)

for layer in model.layers: print(layer.get_config(), layer.get_weights
())
```

Model: "sequential\_4"

Layer (type)	Output Shape	Param #
conv2d_4 (Conv2D)	(None, 24, 24, 32)	832
max_pooling2d_4 (MaxPooling2)	(None, 12, 12, 32)	0
dropout_4 (Dropout)	(None, 12, 12, 32)	0
flatten_4 (Flatten)	(None, 4608)	0
dense_8 (Dense)	(None, 128)	589952
dense_9 (Dense)	(None, 27)	3483
Total params: 594,267		
Trainable params: 594,267		
Non-trainable params: 0		

Epoch 1/10

235/235 - 2s - loss: 1.5737 - accuracy: 0.5335

Epoch 2/10

235/235 - 2s - loss: 0.7698 - accuracy: 0.7673

Epoch 3/10

235/235 - 2s - loss: 0.5056 - accuracy: 0.8443

Epoch 4/10

235/235 - 2s - loss: 0.3846 - accuracy: 0.8795

Epoch 5/10

235/235 - 2s - loss: 0.2851 - accuracy: 0.9104

Epoch 6/10

235/235 - 2s - loss: 0.2123 - accuracy: 0.9298

Epoch 7/10

235/235 - 2s - loss: 0.1798 - accuracy: 0.9386

Epoch 8/10

235/235 - 2s - loss: 0.1404 - accuracy: 0.9526

Epoch 9/10

235/235 - 2s - loss: 0.1153 - accuracy: 0.9616

Epoch 10/10

235/235 - 2s - loss: 0.0990 - accuracy: 0.9648

```
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```

## Testing the Model

```

In [67]: print("Evaluating on Test Data")
results = model.evaluate(X_test, y_test, batch_size=b_size, verbose=0)
print("Test Accuracy:", results[1])

```

```

Evaluating on Test Data
Test Accuracy: 0.8246753215789795

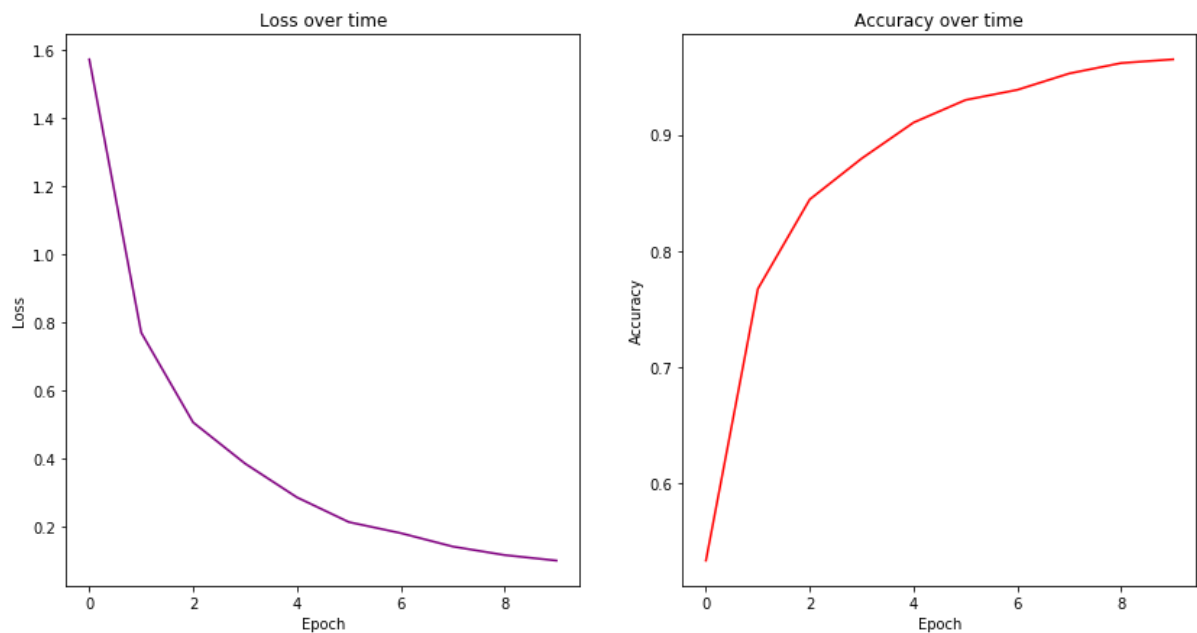
```

## Visualizing the Performance

```
In [69]: # accuracy vs epochs
plt.figure(figsize=(14,7))

plt.subplot(1,2,1)
plt.plot(history.history['loss'], c='purple')
plt.title('Loss over time')
plt.ylabel('Loss')
plt.xlabel('Epoch')

plt.subplot(1,2,2)
plt.plot(history.history['accuracy'], c='red')
plt.title('Accuracy over time')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.show()
```



**END OF ASSIGNMENT**

**Author: Rudraksh Kapil - 177154**

**Thank you for reading :)**