

Lab Report 4

Convolution Neural Networks

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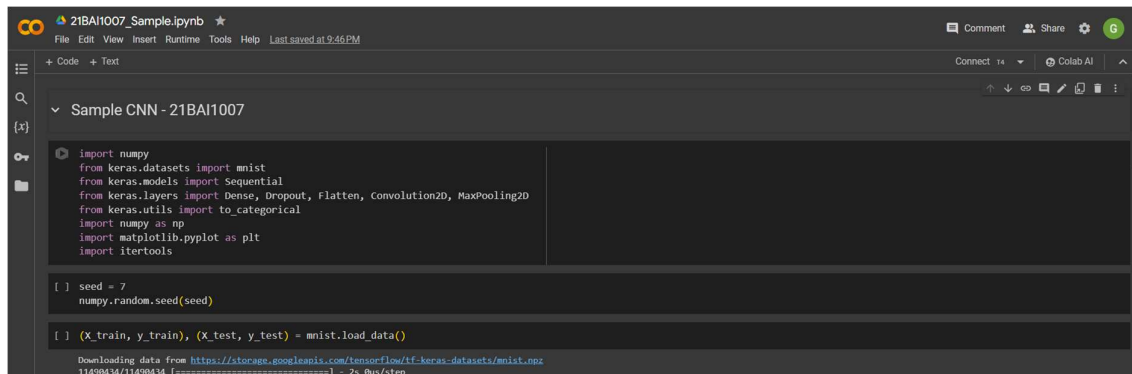
21BA11007

Aim

1. To create a simple CNN model for MNIST Digit classification dataset.
2. To compare 5 different CNN models for the MNIST Dataset.
3. Visualize filter and feature for the CNN model.
4. Create a CNN for another customized dataset (CIFAR-10).
5. Compare CNN with a Fully Connected Dense model.

Observations

For aim 1: To create a simple CNN model for MNIST Digit classification dataset.



```
import numpy
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Convolution2D, MaxPooling2D
from keras.utils import to_categorical
import numpy as np
import matplotlib.pyplot as plt
import itertools

[ ] seed = 7
numpy.random.seed(seed)

[ ] (X_train, y_train), (X_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490434/11490434 [=====] - 2s 0us/step
```



```
[ ] X_train = X_train.reshape(X_train.shape[0], 28, 28, 1).astype('float32')
X_test = X_test.reshape(X_test.shape[0], 28, 28, 1).astype('float32')

[ ] X_train = X_train/255
X_test = X_test/255

[ ] y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
num_classes = y_test.shape[1]
(X_train, y_train1), (X_test, y_test1) = mnist.load_data()

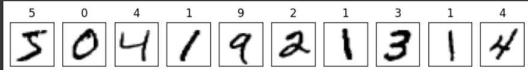
[ ] X_train[1].shape

(28, 28, 1)
```

```
21BAI1007_Sample.ipynb
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fig, ax = plt.subplots(1, 10, figsize=(10, 10))
for i in range(0, 10):
    ax[i].axis.set_visible(False)
    ax[i].yaxis.set_visible(False)
    ax[i].set_title(y_train[i])
    ax[i].imshow(x_train[i], cmap=plt.cm.binary)

def baseline_model():
    model = Sequential()
    model.add(Convolution2D(32, (5, 5), padding='valid', strides=(1,1), 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(1152, activation='relu'))
    model.add(Dense(128, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model
```



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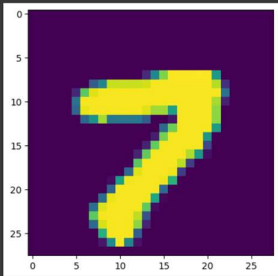
model = baseline_model()
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=200, verbose=2)
scores = model.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
```

```
Epoch 1/10
300/300 - loss: 0.1750 - accuracy: 0.9475 - val_loss: 0.0575 - val_accuracy: 0.9796 - 8s/epoch - 26ms/step
Epoch 2/10
300/300 - loss: 0.0488 - accuracy: 0.9848 - val_loss: 0.0442 - val_accuracy: 0.9861 - 2s/epoch - 8ms/step
Epoch 3/10
300/300 - loss: 0.0322 - accuracy: 0.9897 - val_loss: 0.0388 - val_accuracy: 0.9878 - 3s/epoch - 10ms/step
Epoch 4/10
300/300 - loss: 0.0215 - accuracy: 0.9929 - val_loss: 0.0403 - val_accuracy: 0.9871 - 2s/epoch - 8ms/step
Epoch 5/10
300/300 - loss: 0.0183 - accuracy: 0.9940 - val_loss: 0.0367 - val_accuracy: 0.9884 - 2s/epoch - 8ms/step
Epoch 6/10
300/300 - loss: 0.0127 - accuracy: 0.9959 - val_loss: 0.0330 - val_accuracy: 0.9897 - 3s/epoch - 9ms/step
Epoch 7/10
300/300 - loss: 0.0118 - accuracy: 0.9962 - val_loss: 0.0338 - val_accuracy: 0.9906 - 3s/epoch - 11ms/step
Epoch 8/10
300/300 - loss: 0.0095 - accuracy: 0.9969 - val_loss: 0.0434 - val_accuracy: 0.9886 - 2s/epoch - 8ms/step
Epoch 9/10
300/300 - loss: 0.0098 - accuracy: 0.9967 - val_loss: 0.0397 - val_accuracy: 0.9894 - 2s/epoch - 6ms/step
Epoch 10/10
300/300 - loss: 0.0072 - accuracy: 0.9974 - val_loss: 0.0439 - val_accuracy: 0.9889 - 2s/epoch - 6ms/step
CNN Error: 1.11%
```

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num = 79;
plt.imshow(x_test[num])
testSample = x_test[num].reshape(1, x_test[num].shape[0], 28, 1).astype('float32')
pred = model.predict(testSample)
for i in range(0, 10):
    if pred[i][1]==1:
        print(i);
```

```
1/1 [=====] - 0s 106ms/step
7
```



For aim 2: To compare 5 different CNN models for the MNIST Dataset.

```
21BA1007_Q1_Compare.ipynb
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Comparing 5 different Models on the MNIST Dataset - 21BA1007

[ ] import numpy
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Convolution2D, MaxPooling2D
from keras.utils import to_categorical
import numpy as np
import matplotlib.pyplot as plt
import itertools

[ ] seed = 7
numpy.random.seed(seed)

[ ] (X_train, y_train), (X_test, y_test) = mnist.load_data()

Downloading data from https://storage.googleapis.com/tensorflow/tf-keras-datasets/mnist.npz
11490414/11490414 [=====] - 2s 0us/step
```

```
21BA1007_Q1_Compare.ipynb
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[ ] X_train = X_train.reshape(X_train.shape[0], 28, 28, 1).astype('float32')
X_test = X_test.reshape(X_test.shape[0], 28, 28, 1).astype('float32')

[ ] X_train = X_train/255
X_test = X_test/255

[ ] y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
num_classes = y_test.shape[1]
(x_train, y_train1), (x_test, y_test1) = mnist.load_data()

[ ] X_train[1].shape

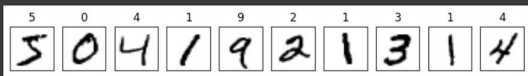
(28, 28, 1)
```

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21BA1007_Q1_Compare.ipynb
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Model 1 - 21BA1007

[ ] fig, ax = plt.subplots(1, 10, figsize=(10, 10))
for i in range(0, 10):
    ax[i].axis.set_visible(False)
    ax[i].yaxis.set_visible(False)
    ax[i].set_title(y_train1[i])
    ax[i].imshow(X_train[i], cmap=plt.cm.binary)

def baseline_model():
    model = Sequential()
    model.add(Convolution2D(32, (5, 5), padding='valid', strides=(1,1), 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(1152, activation='relu'))
    model.add(Dense(128, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model
```



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[ ] model = baseline_model()
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=200, verbose=2)
scores = model.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))

Epoch 1/10
300/300 - 7s - loss: 0.1704 - accuracy: 0.9481 - val_loss: 0.0519 - val_accuracy: 0.9833 - 7s/epoch - 25ms/step
Epoch 2/10
300/300 - 2s - loss: 0.0508 - accuracy: 0.9842 - val_loss: 0.0445 - val_accuracy: 0.9854 - 2s/epoch - 6ms/step
Epoch 3/10
300/300 - 2s - loss: 0.0327 - accuracy: 0.9894 - val_loss: 0.0318 - val_accuracy: 0.9894 - 2s/epoch - 6ms/step
Epoch 4/10
300/300 - 2s - loss: 0.0228 - accuracy: 0.9929 - val_loss: 0.0439 - val_accuracy: 0.9857 - 2s/epoch - 6ms/step
Epoch 5/10
300/300 - 2s - loss: 0.0190 - accuracy: 0.9938 - val_loss: 0.0319 - val_accuracy: 0.9878 - 2s/epoch - 6ms/step
Epoch 6/10
300/300 - 2s - loss: 0.0123 - accuracy: 0.9961 - val_loss: 0.0392 - val_accuracy: 0.9870 - 2s/epoch - 7ms/step
Epoch 7/10
300/300 - 2s - loss: 0.0122 - accuracy: 0.9958 - val_loss: 0.0337 - val_accuracy: 0.9897 - 2s/epoch - 6ms/step
Epoch 8/10
300/300 - 2s - loss: 0.0089 - accuracy: 0.9969 - val_loss: 0.0335 - val_accuracy: 0.9895 - 2s/epoch - 7ms/step
Epoch 9/10
300/300 - 2s - loss: 0.0092 - accuracy: 0.9969 - val_loss: 0.0445 - val_accuracy: 0.9879 - 2s/epoch - 7ms/step
Epoch 10/10
300/300 - 2s - loss: 0.0094 - accuracy: 0.9968 - val_loss: 0.0287 - val_accuracy: 0.9916 - 2s/epoch - 6ms/step
CNN Error: 0.84%
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Second Model - 21BAI1007

def model2():
    model = Sequential()
    model.add(Convolution2D(64, (5, 5), padding='valid', strides=(1,2), input_shape=(28,28,1), activation='relu'))
    model.add(MaxPooling2D(pool_size=(3, 3)))
    model.add(Dropout(0.2))
    model.add(Flatten())
    model.add(Dense(1024, activation='relu'))
    model.add(Dense(512, activation='relu'))
    model.add(Dense(256, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Dense(128, activation='relu'))
    model.add(Dense(64, activation='relu'))
    model.add(Dense(32, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

model2 = model2()
model2.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=200, verbose=2)
scores = model2.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))

Epoch 1/10
300/300 - 5s - loss: 0.2776 - accuracy: 0.9120 - val_loss: 0.0009 - val_accuracy: 0.9759 - 5s/epoch - 10ms/step
Epoch 2/10
300/300 - 2s - loss: 0.0716 - accuracy: 0.9788 - val_loss: 0.0477 - val_accuracy: 0.9853 - 2s/epoch - 5ms/step
Epoch 3/10
300/300 - 2s - loss: 0.0508 - accuracy: 0.9852 - val_loss: 0.0413 - val_accuracy: 0.9864 - 2s/epoch - 5ms/step
Epoch 4/10
300/300 - 2s - loss: 0.0366 - accuracy: 0.9895 - val_loss: 0.0410 - val_accuracy: 0.9878 - 2s/epoch - 5ms/step
Epoch 5/10
300/300 - 2s - loss: 0.0339 - accuracy: 0.9898 - val_loss: 0.0413 - val_accuracy: 0.9880 - 2s/epoch - 5ms/step
Epoch 6/10
300/300 - 2s - loss: 0.0301 - accuracy: 0.9908 - val_loss: 0.0416 - val_accuracy: 0.9888 - 2s/epoch - 5ms/step
Epoch 7/10
300/300 - 2s - loss: 0.0276 - accuracy: 0.9918 - val_loss: 0.0416 - val_accuracy: 0.9888 - 2s/epoch - 5ms/step
Epoch 8/10
300/300 - 2s - loss: 0.0251 - accuracy: 0.9928 - val_loss: 0.0416 - val_accuracy: 0.9888 - 2s/epoch - 5ms/step
Epoch 9/10
300/300 - 2s - loss: 0.0226 - accuracy: 0.9938 - val_loss: 0.0416 - val_accuracy: 0.9888 - 2s/epoch - 5ms/step
Epoch 10/10
300/300 - 2s - loss: 0.0201 - accuracy: 0.9948 - val_loss: 0.0416 - val_accuracy: 0.9888 - 2s/epoch - 5ms/step
```

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Third model - 21BAI1007

def model3():
    model = Sequential()
    model.add(Convolution2D(32, (3, 3), padding='valid', strides=(1,1), input_shape=(28, 28, 1), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(Flatten())
    model.add(Dense(64, activation='relu'))
    #model.add(Dropout(0.2))
    #model.add(Dense(32, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

model3 = model3()
model3.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=20, batch_size=64, verbose=1)
scores = model3.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))

Epoch 1/20
938/938 [=====] - 8s 6ms/step - loss: 0.1879 - accuracy: 0.9421 - val_loss: 0.0496 - val_accuracy: 0.9832
Epoch 2/20
938/938 [=====] - 5s 5ms/step - loss: 0.0500 - accuracy: 0.9844 - val_loss: 0.0538 - val_accuracy: 0.9845
Epoch 3/20
938/938 [=====] - 5s 6ms/step - loss: 0.0365 - accuracy: 0.9884 - val_loss: 0.0365 - val_accuracy: 0.9884
Epoch 4/20
938/938 [=====] - 5s 5ms/step - loss: 0.0292 - accuracy: 0.9905 - val_loss: 0.0282 - val_accuracy: 0.9919
Epoch 5/20
938/938 [=====] - 5s 5ms/step - loss: 0.0229 - accuracy: 0.9928 - val_loss: 0.0354 - val_accuracy: 0.9883
Epoch 6/20
938/938 [=====] - 5s 6ms/step - loss: 0.0185 - accuracy: 0.9942 - val_loss: 0.0302 - val_accuracy: 0.9912
Epoch 7/20
938/938 [=====] - 5s 5ms/step - loss: 0.0152 - accuracy: 0.9952 - val_loss: 0.0291 - val_accuracy: 0.9912
Epoch 8/20
938/938 [=====] - 5s 5ms/step - loss: 0.0127 - accuracy: 0.9962 - val_loss: 0.0281 - val_accuracy: 0.9912
Epoch 9/20
938/938 [=====] - 5s 5ms/step - loss: 0.0102 - accuracy: 0.9972 - val_loss: 0.0271 - val_accuracy: 0.9912
Epoch 10/20
938/938 [=====] - 5s 5ms/step - loss: 0.0077 - accuracy: 0.9982 - val_loss: 0.0261 - val_accuracy: 0.9912
Epoch 11/20
938/938 [=====] - 5s 5ms/step - loss: 0.0052 - accuracy: 0.9992 - val_loss: 0.0251 - val_accuracy: 0.9912
Epoch 12/20
938/938 [=====] - 5s 5ms/step - loss: 0.0027 - accuracy: 1.0000 - val_loss: 0.0241 - val_accuracy: 0.9912
Epoch 13/20
938/938 [=====] - 5s 5ms/step - loss: 0.0002 - accuracy: 1.0000 - val_loss: 0.0231 - val_accuracy: 0.9912
Epoch 14/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0221 - val_accuracy: 0.9912
Epoch 15/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0211 - val_accuracy: 0.9912
Epoch 16/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0201 - val_accuracy: 0.9912
Epoch 17/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0191 - val_accuracy: 0.9912
Epoch 18/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0181 - val_accuracy: 0.9912
Epoch 19/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0171 - val_accuracy: 0.9912
Epoch 20/20
938/938 [=====] - 5s 5ms/step - loss: 0.0000 - accuracy: 1.0000 - val_loss: 0.0161 - val_accuracy: 0.9912
```

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Fourth Model - ALEXNET - 21BAI1007

def model4():
    model = Sequential()
    model.add(Convolution2D(96, (3,3), strides=1, input_shape=(28,28,1), activation='relu'))
    model.add(MaxPooling2D((3,3), strides=2))
    model.add(Convolution2D(256, (5,5), padding='same', activation='relu'))
    model.add(MaxPooling2D((3,3), strides=2))
    model.add(Convolution2D(384, (3,3), padding='same', activation='relu'))
    model.add(Convolution2D(384, (3,3), padding='same', activation='relu'))
    model.add(Convolution2D(256, (3,3), padding='same', activation='relu'))
    model.add(MaxPooling2D((3,3), strides=2))
    model.add(Flatten())
    model.add(Dense(4096, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Dense(4096, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

model4 = model4()
model4.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=30, batch_size=64, verbose=1)
scores = model4.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))

Epoch 3/30
938/938 [=====] - 22s 23ms/step - loss: 0.0480 - accuracy: 0.9878 - val_loss: 0.0424 - val_accuracy: 0.9889
Epoch 4/30
938/938 [=====] - 22s 23ms/step - loss: 0.0403 - accuracy: 0.9895 - val_loss: 0.0411 - val_accuracy: 0.9893
Epoch 5/30
938/938 [=====] - 22s 24ms/step - loss: 0.0351 - accuracy: 0.9914 - val_loss: 0.0590 - val_accuracy: 0.9866
Epoch 6/30
938/938 [=====] - 22s 24ms/step - loss: 0.0301 - accuracy: 0.9924 - val_loss: 0.0603 - val_accuracy: 0.9843
Epoch 7/30
938/938 [=====] - 22s 23ms/step - loss: 0.0287 - accuracy: 0.9925 - val_loss: 0.0333 - val_accuracy: 0.9907
Epoch 8/30
938/938 [=====] - 22s 23ms/step - loss: 0.0233 - accuracy: 0.9942 - val_loss: 0.0256 - val_accuracy: 0.9939
Epoch 9/30
938/938 [=====] - 22s 23ms/step - loss: 0.0251 - accuracy: 0.9940 - val_loss: 0.0309 - val_accuracy: 0.9928
```

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Fifth Model - VGG16 - 21BAI1007

[] from keras.layers import ZeroPadding2D

def model5():
    model = Sequential()
    model.add(Input((28, 28, 1)))
    model.add(ZeroPadding2D((2,2)))
    model.add(Convolution2D(64, (3,3), strides=1, padding='same', activation='relu'))
    model.add(Convolution2D(64, (3,3), strides=1, padding='same', activation='relu'))
    model.add(MaxPooling2D((2,2), strides=2))
    model.add(Convolution2D(128, (3,3), strides=1, padding='same', activation='relu'))
    model.add(Convolution2D(128, (3,3), strides=1, padding='same', activation='relu'))
    model.add(MaxPooling2D((2,2), strides=2))
    model.add(Convolution2D(256, (3,3), strides=1, padding='same', activation='relu'))
    model.add(Convolution2D(256, (3,3), strides=1, padding='same', activation='relu'))
    model.add(Convolution2D(256, (3,3), strides=1, padding='same', activation='relu'))
    model.add(MaxPooling2D((2,2), strides=2))
    model.add(Flatten())
    model.add(Dense(4096, activation='relu'))
    model.add(Dense(128, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

model5 = model5()
model5.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=15, batch_size=64, verbose=1)
scores = model5.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))
```

```
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[]

Epoch 1/15
938/938 [=====] - 24s 22ms/step - loss: 0.1972 - accuracy: 0.9367 - val_loss: 0.0410 - val_accuracy: 0.9872
Epoch 2/15
938/938 [=====] - 20s 21ms/step - loss: 0.0550 - accuracy: 0.9843 - val_loss: 0.0301 - val_accuracy: 0.9891
Epoch 3/15
938/938 [=====] - 21s 22ms/step - loss: 0.0406 - accuracy: 0.9882 - val_loss: 0.0446 - val_accuracy: 0.9864
Epoch 4/15
938/938 [=====] - 21s 22ms/step - loss: 0.0343 - accuracy: 0.9899 - val_loss: 0.0362 - val_accuracy: 0.9894
Epoch 5/15
938/938 [=====] - 20s 22ms/step - loss: 0.0308 - accuracy: 0.9918 - val_loss: 0.0390 - val_accuracy: 0.9896
Epoch 6/15
938/938 [=====] - 20s 21ms/step - loss: 0.0260 - accuracy: 0.9928 - val_loss: 0.0342 - val_accuracy: 0.9903
Epoch 7/15
938/938 [=====] - 20s 21ms/step - loss: 0.0226 - accuracy: 0.9936 - val_loss: 0.0272 - val_accuracy: 0.9916
Epoch 8/15
938/938 [=====] - 20s 22ms/step - loss: 0.0206 - accuracy: 0.9943 - val_loss: 0.0312 - val_accuracy: 0.9897
Epoch 9/15
938/938 [=====] - 20s 22ms/step - loss: 0.0203 - accuracy: 0.9945 - val_loss: 0.0319 - val_accuracy: 0.9926
Epoch 10/15
938/938 [=====] - 20s 22ms/step - loss: 0.0177 - accuracy: 0.9950 - val_loss: 0.0294 - val_accuracy: 0.9933
Epoch 11/15
938/938 [=====] - 20s 22ms/step - loss: 0.0173 - accuracy: 0.9952 - val_loss: 0.0357 - val_accuracy: 0.9908
Epoch 12/15
938/938 [=====] - 20s 21ms/step - loss: 0.0176 - accuracy: 0.9954 - val_loss: 0.0317 - val_accuracy: 0.9917
Epoch 13/15
938/938 [=====] - 20s 21ms/step - loss: 0.0137 - accuracy: 0.9962 - val_loss: 0.0409 - val_accuracy: 0.9987
Epoch 14/15
938/938 [=====] - 20s 22ms/step - loss: 0.0145 - accuracy: 0.9959 - val_loss: 0.0369 - val_accuracy: 0.9928
Epoch 15/15
938/938 [=====] - 20s 21ms/step - loss: 0.0133 - accuracy: 0.9963 - val_loss: 0.0501 - val_accuracy: 0.9914
CNN Error: 0.86%
Accuracy: 99.14%
```

```
21BAI1007_Q1_Compare.ipynb
File Edit View Insert Runtime Tools Help All changes saved
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Reconnect 14 Colab AI

Results - 21BAI1007

First Model:
Accuracy: 99.16%
Error %: 0.84%

Second Model
Accuracy: 98.79%
Error %: 1.21%

Third Model
Accuracy: 99.27%
Error %: 0.73%

Fourth Model
Accuracy: 99.17%
Error %: 0.83%

Fifth Model
Accuracy: 99.14%
Error %: 0.86%
```

Model	Accuracy	Error
First Model	99.16%	0.84%
Second Model	98.79%	1.21%
Third Model	99.27%	10.73%
Fourth Model	99.17%	0.63%
Fifth Model	99.14%	0.86%

For aim 3: Visualize filter and feature for the CNN model.

```
21BA1007-visualize.ipynb
File Edit View Insert Runtime Tools Help Last saved at February 28

+ Code + Text

Visualizing filters and features - 21BA1007

[] import numpy
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Convolution2D, MaxPooling2D
from keras.utils import to_categorical
import numpy as np
import matplotlib.pyplot as plt
import itertools

[] seed = 7
numpy.random.seed(seed)

[] (X_train, y_train), (X_test, y_test) = mnist.load_data()

[] X_train = X_train.reshape(X_train.shape[0], 28, 28, 1).astype('float32')
X_test = X_test.reshape(X_test.shape[0], 28, 28, 1).astype('float32')

[] X_train = X_train/255
X_test = X_test/255

[] y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
num_classes = y_test.shape[1]
(x_train, y_train), (x_test, y_test) = mnist.load_data()
```

```
21BA1007-visualize.ipynb
File Edit View Insert Runtime Tools Help Last saved at February 28

+ Code + Text

fig, ax = plt.subplots(1, 10, figsize=(10, 10))
for i in range(0, 10):
    ax[i].axis.set_visible(False)
    ax[i].axis.set_visible(False)
    ax[i].set_title(y_train[i])
    ax[i].imshow(X_train[i], cmap=plt.cm.binary)

def baseline_model():
    model = Sequential()
    model.add(Convolution2D(32, (5, 5), padding='valid', strides=(1,1), 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(10, activation='softmax'))
    model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

[] model = baseline_model()
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=200, verbose=2)
scores = model.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))

Epoch 1/10
300/300 - loss: 0.1722 - accuracy: 0.9491 - val_loss: 0.0490 - val_accuracy: 0.9848 - 185s/epoch - 350ms/step
Epoch 2/10
300/300 - loss: 0.0491 - accuracy: 0.9844 - val_loss: 0.0439 - val_accuracy: 0.9859 - 96s/epoch - 321ms/step
Epoch 3/10
300/300 - loss: 0.0323 - accuracy: 0.9807 - val_loss: 0.0288 - val_accuracy: 0.9908 - 95s/epoch - 317ms/step
Epoch 4/10
300/300 - loss: 0.0244 - accuracy: 0.9928 - val_loss: 0.0365 - val_accuracy: 0.9887 - 95s/epoch - 317ms/step
Epoch 5/10
300/300 - loss: 0.0179 - accuracy: 0.9939 - val_loss: 0.0316 - val_accuracy: 0.9907 - 96s/epoch - 320ms/step
Epoch 6/10
300/300 - loss: 0.0144 - accuracy: 0.9955 - val_loss: 0.0377 - val_accuracy: 0.9879 - 94s/epoch - 313ms/step
```

```
21BA1007-visualize.ipynb
File Edit View Insert Runtime Tools Help Last saved at February 28

+ Code + Text

import matplotlib.pyplot as plt
import numpy as np
from tensorflow.keras import backend as K
from tensorflow.keras.models import Model

def visualize_filters(model, layer_name, n_filters=8, filter_size=(3, 3), figsize=(10, 10)):
    layer = model.get_layer(layer_name)
    filters = layer.get_weights()[0]

    if len(filters.shape) == 3:
        filters = np.expand_dims(filters, axis=-1)

    plt.figure(figsize=figsize)

    for i in range(n_filters):
        plt.subplot(n_filters // 4, 4, i + 1)
        plt.imshow(filters[:, :, 0, i], cmap='viridis')
        plt.axis('off')

    plt.show()

def visualize_feature_maps(model, layer_name, input_image):
    layer = model.get_layer(layer_name)
    intermediate_model = Model(inputs=model.input, outputs=layer.output)

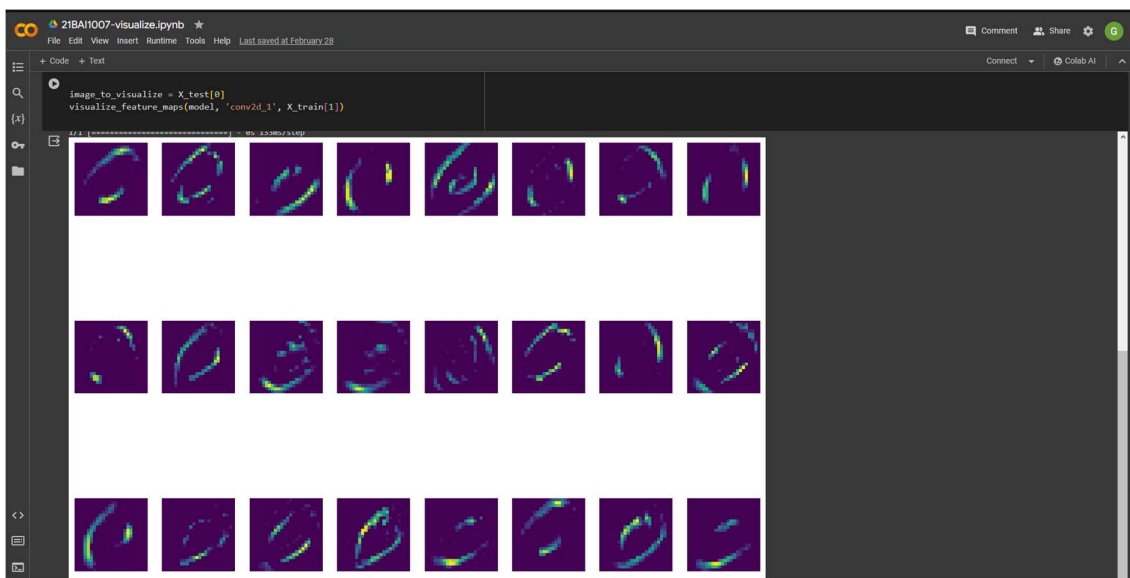
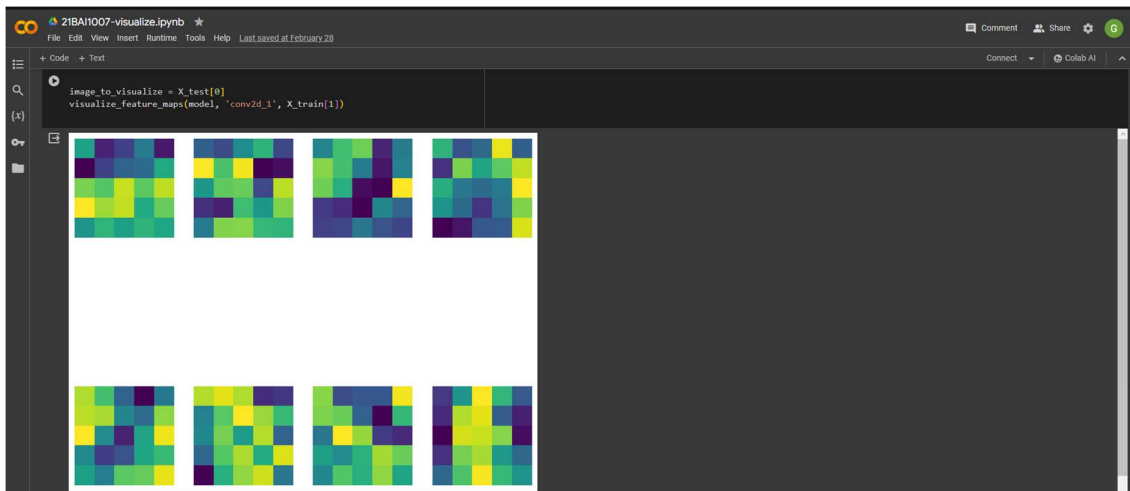
    input_image = np.expand_dims(input_image, axis=0)
    feature_maps = intermediate_model.predict(input_image)

    n_features = feature_maps.shape[-1]
    plt.figure(figsize=(15, 15))

    for i in range(n_features):
        plt.subplot(n_features // 8, 8, i + 1)
        plt.imshow(feature_maps[0, :, :, i], cmap='viridis')
        plt.axis('off')

    plt.show()

visualize_filters(model, 'conv2d_1')
```



For aim 4: Create a CNN for another customized dataset (CIFAR-10).

```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text
CIFAR-10 dataset classification using CNN
Goutham Krishnan - 21BA1007

[ ] import cv2 as cv
import numpy as np
import matplotlib.pyplot as plt
from tensorflow.keras import datasets
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Convolution2D, MaxPooling2D

Loading the preprocessing the dataset - 21BA1007

[ ] (training_images, training_labels), (testing_images, testing_labels) = datasets.cifar10.load_data()

Download data from https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz
170498071/170498071 [-----] - 13s 0us/step

[ ] training_images.shape
(50000, 32, 32, 3)

[ ] training_labels.shape
(50000, 1)
```




```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text
plt.imshow(training_images[1])
matplotlib.image.AxesImage at 0x7bc8184af30
0
5
10
15
20
25
30
0 5 10 15 20 25 30

[ ] training_images, testing_images = training_images / 255, testing_images / 255

[ ] class_names = ['Plane', 'Car', 'Bird', 'Cat', 'Deer', 'Dog', 'Frog', 'Horse', 'Ship', 'Truck']

[ ] for i in range(10):
    plt.subplot(4, 4, i+1)
    plt.xticks([])
    plt.yticks([])
    plt.imshow(training_images[i], cmap=plt.cm.binary)
    plt.xlabel(class_names[training_labels[i][0]])

plt.show()
```

```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text

[ ] X_train = training_images
y_train = training_labels
X_test = testing_images
y_test = testing_labels

[ ] X_train.shape
(10000, 32, 32, 3)

[ ] X_test.shape
(1000, 32, 32, 3)
```

```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text
Creating the model - 21BA1007

def model():
    model = Sequential()
    model.add(Convolution2D(32, (3, 3), padding='valid', strides=(1,1), input_shape=(32, 32, 3), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(Flatten())
    model.add(Dense(64, activation='relu'))
    model.add(Dropout(0.2))
    model.add(Dense(32, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model

model = model()
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=20, batch_size=64, verbose=1)
scores = model.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))

Epoch 1/20
782/782 [=====] - 7s 7ms/step - loss: 1.6148 - accuracy: 0.4096 - val_loss: 1.3142 - val_accuracy: 0.5311
Epoch 2/20
782/782 [=====] - 4s 5ms/step - loss: 1.2400 - accuracy: 0.5575 - val_loss: 1.1809 - val_accuracy: 0.5787
Epoch 3/20
782/782 [=====] - 4s 5ms/step - loss: 1.0741 - accuracy: 0.6225 - val_loss: 1.0384 - val_accuracy: 0.6336
Epoch 4/20
782/782 [=====] - 5s 6ms/step - loss: 0.9690 - accuracy: 0.6612 - val_loss: 1.0095 - val_accuracy: 0.6582
Epoch 5/20
782/782 [=====] - 4s 5ms/step - loss: 0.8900 - accuracy: 0.6919 - val_loss: 0.9654 - val_accuracy: 0.6681
Epoch 6/20
782/782 [=====] - 4s 5ms/step - loss: 0.8327 - accuracy: 0.7082 - val_loss: 0.8900 - val_accuracy: 0.6923
Epoch 7/20
782/782 [=====] - 5s 6ms/step - loss: 0.7813 - accuracy: 0.7277 - val_loss: 0.8892 - val_accuracy: 0.6919
Epoch 8/20
782/782 [=====] - 4s 6ms/step - loss: 0.7349 - accuracy: 0.7444 - val_loss: 0.8537 - val_accuracy: 0.7016
Epoch 9/20
782/782 [=====] - 4s 5ms/step - loss: 0.6976 - accuracy: 0.7565 - val_loss: 0.8533 - val_accuracy: 0.7186
Epoch 10/20
```



```
21BA1007-Q2-CIFAR10.ipynb
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+ Code + Text
782/782 [=====] - 5s 7ms/step - loss: 0.4424 - accuracy: 0.8426 - val_loss: 0.9724 - val_accuracy: 0.7156
Epoch 10/20
782/782 [=====] - 5s 6ms/step - loss: 0.4182 - accuracy: 0.8523 - val_loss: 0.9788 - val_accuracy: 0.7091
Epoch 20/20
782/782 [=====] - 4s 6ms/step - loss: 0.3980 - accuracy: 0.8590 - val_loss: 0.9911 - val_accuracy: 0.7179
CNN Error: 28.21%
Accuracy: 71.79%

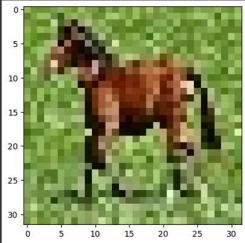
loss, accuracy = model.evaluate(X_test, y_test)
print(loss, accuracy*100, "%")

313/313 [=====] - 1s 3ms/step - loss: 0.9911 - accuracy: 0.7179
0.50180055186676 71.7899978108582 %

Predictions - 21BA1007
The image of a horse is given as input and the prediction is retrieved

+ Code + Text
### Testing with a random image of a horse
img = cv.imread('resized_horse.jpg')
img = cv.cvtColor(img, cv.COLOR_BGR2RGB)

plt.imshow(img, cmap=plt.cm.binary)
```

```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text


prediction = model.predict(np.array([img])/255)

1/1 [=====] - 0s 424ms/step

prediction

array([[4.6116588e-10, 9.2871959e-09, 1.3860985e-07, 5.9190910e-07,
        1.8527507e-03, 7.8039847e-05, 7.4427282e-03, 9.8862457e-03,
        1.4382377e-07, 1.0928353e-07]], dtype=float32)
```

```
21BA1007-Q2-CIFAR10.ipynb
File Edit View Insert Runtime Tools Help Last edited on February 27
+ Code + Text

prediction

array([[4.6116588e-10, 9.2871959e-09, 1.3860985e-07, 5.9190910e-07,
        1.8527507e-03, 7.8039847e-05, 7.4427282e-03, 9.8862457e-03,
        1.4382377e-07, 1.0928353e-07]], dtype=float32)

index = np.argmax(prediction)

index

7

print("Prediction : {class_names[index]}")

Prediction : Horse

+ Code + Text
Start coding on notebook with AI
```

For aim 5: Compare CNN with a fully connected dense model.

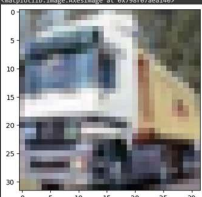
```
21BA1007_CNNvsDense.ipynb
File Edit View Insert Runtime Tools Help Last saved at 6:38 AM
+ Code + Text
Comparing CNN and Dense Network on CIFAR-10 dataset
Goutham Krishnan 21BA1007

+ Code + Text

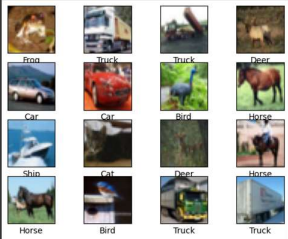
import cv2 as cv
import numpy as np
import matplotlib.pyplot as plt
from keras import datasets
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten, Convolution2D, MaxPooling2D

(training_images, training_labels), (testing_images, testing_labels) = datasets.cifar10.load_data()

plt.imshow(training_images[1])

matplotlib.figure.dfigure at 0x70f07f0e3480

```

```
21BA1007_CNNvsDense.ipynb
File Edit View Insert Runtime Tools Help Last saved at 6:30 AM
+ Code + Text
[ ] training_images, testing_images = training_images / 255, testing_images / 255
[ ] class_names = ['Plane', 'Car', 'Bird', 'Cat', 'Deer', 'Dog', 'Frog', 'Horse', 'Ship', 'Truck']
for i in range(16):
    plt.subplot(4, 4, i+1)
    plt.axis('off')
    plt.imshow(training_images[i], cmap=plt.cm.binary)
    plt.xlabel(class_names[training_labels[i][0]])
plt.show()
```



```
21BA1007_CNNvsDense.ipynb
File Edit View Insert Runtime Tools Help Last saved at 6:30 AM
+ Code + Text
[ ] X_train = training_images
    y_train = training_labels
    X_test = testing_images
    y_test = testing_labels
Creating the model using CNN - 21BA1007
[ ] def model():
    model = Sequential()
    model.add(Convolution2D(32, (3, 3), padding='valid', strides=(1,1), input_shape=(32, 32, 3), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(MaxPooling2D(pool_size=(2, 2)))
    model.add(Convolution2D(64, (3, 3), activation='relu'))
    model.add(Flatten())
    model.add(Dense(64, activation='relu'))
    #model.add(Dropout(0.2))
    #model.add(Dense(32, activation='relu'))
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model
model = model()
model.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=64, verbose=1)
scores = model.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))
```

```
21BA1007_CNNvsDense.ipynb
File Edit View Insert Runtime Tools Help Last saved at 6:30 AM
+ Code + Text
Epoch 3/10
[ ] 782/782 [=====] - 68s 77ms/step - loss: 1.0756 - accuracy: 0.6218 - val_loss: 1.0429 - val_accuracy: 0.6388
Epoch 4/10
782/782 [=====] - 68s 77ms/step - loss: 0.9750 - accuracy: 0.6575 - val_loss: 1.1180 - val_accuracy: 0.6185
Epoch 5/10
782/782 [=====] - 41s 78ms/step - loss: 0.9825 - accuracy: 0.6837 - val_loss: 0.9618 - val_accuracy: 0.6635
Epoch 6/10
782/782 [=====] - 58s 75ms/step - loss: 0.8429 - accuracy: 0.7828 - val_loss: 0.8888 - val_accuracy: 0.6945
Epoch 7/10
782/782 [=====] - 62s 79ms/step - loss: 0.7902 - accuracy: 0.7221 - val_loss: 0.8778 - val_accuracy: 0.6956
Epoch 8/10
782/782 [=====] - 59s 76ms/step - loss: 0.7534 - accuracy: 0.7355 - val_loss: 0.8571 - val_accuracy: 0.7869
Epoch 9/10
782/782 [=====] - 62s 79ms/step - loss: 0.7115 - accuracy: 0.7512 - val_loss: 0.9293 - val_accuracy: 0.6854
Epoch 10/10
782/782 [=====] - 62s 79ms/step - loss: 0.6762 - accuracy: 0.7612 - val_loss: 0.8596 - val_accuracy: 0.7878
CNN Error: 29.22%
Accuracy: 70.78%
[ ] loss, accuracy = model.evaluate(X_test, y_test)
print(loss, accuracy*100, "%")
313/313 [=====] - 4s 11ms/step - loss: 0.8596 - accuracy: 0.7878
0.859600084084753 70.77999711836682 %
Creating a dense connected ANN - 21BA1007
[ ] def model2():
    model.add(Dense(64, activation='relu', input_shape=(32, 32, 3)))
    model.add(Dropout(0.2))
    model.add(Dense(32, activation='relu'))
    model.add(Flatten())
    model.add(Dense(10, activation='softmax'))
    model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
    return model
model2 = model2()
model2.fit(X_train, y_train, validation_data=(X_test, y_test), epochs=10, batch_size=64, verbose=1)
scores = model2.evaluate(X_test, y_test, verbose=0)
print("CNN Error: %.2f%%" % (100-scores[1]*100))
print("Accuracy: %.2f%%" % (scores[1]*100))
```

21BA1007_CNNvsDense.ipynb

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+ Code + Text

```
[ ] 782/782 [=====] - 50s 79ms/step - loss: 0.7188 - accuracy: 0.7889 - val_loss: 0.9482 - val_accuracy: 0.7385
Epoch 7/10
782/782 [=====] - 68s 76ms/step - loss: 0.6721 - accuracy: 0.7913 - val_loss: 0.9638 - val_accuracy: 0.7129
Epoch 8/10
782/782 [=====] - 56s 72ms/step - loss: 0.6434 - accuracy: 0.8000 - val_loss: 0.9777 - val_accuracy: 0.7131
Epoch 9/10
782/782 [=====] - 59s 76ms/step - loss: 0.6191 - accuracy: 0.8070 - val_loss: 0.9453 - val_accuracy: 0.7129
Epoch 10/10
782/782 [=====] - 57s 73ms/step - loss: 0.5968 - accuracy: 0.8143 - val_loss: 0.9574 - val_accuracy: 0.7196
Omni Server: 28.84%
Accuracy: 71.96%
```

[] loss, accuracy = model.evaluate(X_test, y_test)
print(loss, accuracy*100,"%")

```
313/313 [=====] - 4s 12ms/step - loss: 0.9574 - accuracy: 0.7196
0.9574005007741835 71.96000218391418 %
```

Results

For CNN Network:
- Accuracy: 70.77%

For Dense Network
- Accuracy: 71.96%