

In [1]:

```
# Credits: https://github.com/keras-team/keras/blob/master/examples/mnist\_cnn.py
```

```
from __future__ import print_function
import keras
from keras.datasets import mnist
from keras.models import Sequential
from keras.layers import Dense, Dropout, Flatten
from keras.layers import Conv2D, MaxPooling2D
from keras import backend as K

batch_size = 128
num_classes = 10
epochs = 12

# input image dimensions
img_rows, img_cols = 28, 28

# the data, split between train and test sets
(x_train, y_train), (x_test, y_test) = mnist.load_data()

if K.image_data_format() == 'channels_first':
    x_train = x_train.reshape(x_train.shape[0], 1, img_rows,
img_cols)
    x_test = x_test.reshape(x_test.shape[0], 1, img_rows, img
_cols)
    input_shape = (1, img_rows, img_cols)
else:
    x_train = x_train.reshape(x_train.shape[0], img_rows, img
_cols, 1)
    x_test = x_test.reshape(x_test.shape[0], img_rows, img_co
ls, 1)
    input_shape = (img_rows, img_cols, 1)
```

```
x_train = x_train.astype('float32')
x_test = x_test.astype('float32')
x_train /= 255
x_test /= 255
print('x_train shape:', x_train.shape)
print(x_train.shape[0], 'train samples')
print(x_test.shape[0], 'test samples')

# convert class vectors to binary class matrices
y_train = keras.utils.to_categorical(y_train, num_classes)
y_test = keras.utils.to_categorical(y_test, num_classes)
```

Using TensorFlow backend.

```
x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
```

# Architecture 1: Conlayers-2

## kernel size 3\*3

In [2]:

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
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(128, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

model.fit(x_train, y_train,
          batch_size=batch_size,
          epochs=epochs,
          verbose=1,
          validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

WARNING:tensorflow:From C:\Users\Public\Anaconda3\lib\site-packages\tensorflow\python\framework

ork\op\_def\_library.py:263: colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:

Colocations handled automatically by placer.

WARNING:tensorflow:From C:\Users\Public\Anaconda3\lib\site-packages\keras\backend\tensorflow\_backend.py:3445: calling dropout (from tensorflow.python.ops.nn\_ops) with keep\_prob is deprecated and will be removed in a future version.

Instructions for updating:

Please use `rate` instead of `keep\_prob`. Rate should be set to `rate = 1 - keep\_prob`.

WARNING:tensorflow:From C:\Users\Public\Anaconda3\lib\site-packages\tensorflow\python\ops\math\_ops.py:3066: to\_int32 (from tensorflow.python.ops.math\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.cast instead.

Train on 60000 samples, validate on 10000 samples

Epoch 1/12

60000/60000 [=====] -  
93s 2ms/step - loss: 0.2580 - acc: 0.9219 - val\_loss: 0.0524 - val\_acc: 0.9831

Epoch 2/12

60000/60000 [=====] -  
98s 2ms/step - loss: 0.0852 - acc: 0.9745 - val\_loss: 0.0391 - val\_acc: 0.9864

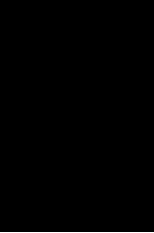
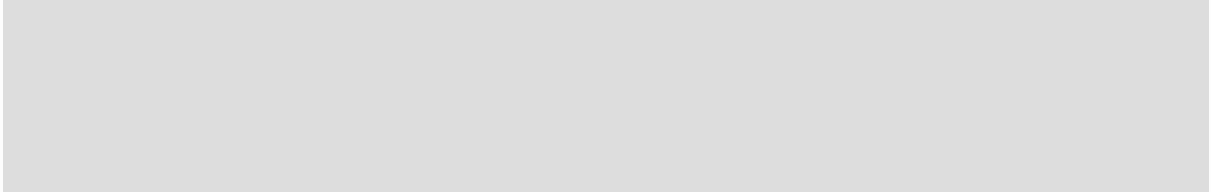
Epoch 3/12

60000/60000 [=====] -  
95s 2ms/step - loss: 0.0649 - acc: 0.9812 - val\_loss: 0.0346 - val\_acc: 0.9882

Epoch 4/12

60000/60000 [=====] -

```
92s 2ms/step - loss: 0.0548 - acc: 0.9838 - v
al_loss: 0.0304 - val_acc: 0.9905
Epoch 5/12
60000/60000 [=====] -
89s 1ms/step - loss: 0.0463 - acc: 0.9860 - v
al_loss: 0.0308 - val_acc: 0.9893
Epoch 6/12
60000/60000 [=====] -
88s 1ms/step - loss: 0.0421 - acc: 0.9871 - v
al_loss: 0.0275 - val_acc: 0.9896
Epoch 7/12
60000/60000 [=====] -
89s 1ms/step - loss: 0.0388 - acc: 0.9886 - v
al_loss: 0.0276 - val_acc: 0.9904
Epoch 8/12
60000/60000 [=====] -
88s 1ms/step - loss: 0.0342 - acc: 0.9899 - v
al_loss: 0.0297 - val_acc: 0.9907
Epoch 9/12
60000/60000 [=====] -
89s 1ms/step - loss: 0.0312 - acc: 0.9908 - v
al_loss: 0.0254 - val_acc: 0.9915
Epoch 10/12
60000/60000 [=====] -
91s 2ms/step - loss: 0.0295 - acc: 0.9907 - v
al_loss: 0.0320 - val_acc: 0.9896
Epoch 11/12
60000/60000 [=====] -
91s 2ms/step - loss: 0.0266 - acc: 0.9918 - v
al_loss: 0.0276 - val_acc: 0.9914
Epoch 12/12
60000/60000 [=====] -
92s 2ms/step - loss: 0.0258 - acc: 0.9918 - v
al_loss: 0.0293 - val_acc: 0.9917
Test loss: 0.029326565258706888
Test accuracy: 0.9917
```



## Architecture2: kernel 3\*3 Conlayers

### 3

In [3]:

```
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.4))
model.add(Conv2D(128, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.4))
model.add(Flatten())
model.add(Dense(256, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

model.fit(x_train, y_train,
          batch_size=batch_size,
          epochs=epochs,
          verbose=1,
          validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

Train on 60000 samples, validate on 10000 samples

Epoch 1/12

60000/60000 [=====] -  
125s 2ms/step - loss: 0.2741 - acc: 0.9131 -  
val\_loss: 0.0524 - val\_acc: 0.9829

Epoch 2/12

60000/60000 [=====] -  
120s 2ms/step - loss: 0.0807 - acc: 0.9754 -  
val\_loss: 0.0339 - val\_acc: 0.9883

Epoch 3/12

60000/60000 [=====] -  
120s 2ms/step - loss: 0.0608 - acc: 0.9812 -  
val\_loss: 0.0272 - val\_acc: 0.9911

Epoch 4/12

60000/60000 [=====] -  
118s 2ms/step - loss: 0.0501 - acc: 0.9846 -  
val\_loss: 0.0254 - val\_acc: 0.9915

Epoch 5/12

60000/60000 [=====] -  
118s 2ms/step - loss: 0.0430 - acc: 0.9869 -  
val\_loss: 0.0225 - val\_acc: 0.9919

Epoch 6/12

60000/60000 [=====] -  
120s 2ms/step - loss: 0.0403 - acc: 0.9877 -  
val\_loss: 0.0221 - val\_acc: 0.9921

Epoch 7/12

60000/60000 [=====] -  
119s 2ms/step - loss: 0.0357 - acc: 0.9892 -  
val\_loss: 0.0199 - val\_acc: 0.9932

Epoch 8/12

60000/60000 [=====] -  
121s 2ms/step - loss: 0.0339 - acc: 0.9894 -  
val\_loss: 0.0200 - val\_acc: 0.9934

Epoch 9/12

60000/60000 [=====] -  
119s 2ms/step - loss: 0.0307 - acc: 0.9903 -



```
val_loss: 0.0180 - val_acc: 0.9942
Epoch 10/12
60000/60000 [=====] -
  116s 2ms/step - loss: 0.0291 - acc: 0.9911 -
val_loss: 0.0186 - val_acc: 0.9938
Epoch 11/12
60000/60000 [=====] -
  117s 2ms/step - loss: 0.0265 - acc: 0.9916 -
val_loss: 0.0198 - val_acc: 0.9932
Epoch 12/12
60000/60000 [=====] -
  120s 2ms/step - loss: 0.0259 - acc: 0.9921 -
val_loss: 0.0171 - val_acc: 0.9941
Test loss: 0.017080856872715048
Test accuracy: 0.9941
```

## Architecture3: kernel 3\*3 Conlayers 3 droupout 0.25 and batch normalization

In [4]:

```
from keras.layers.normalization import BatchNormalization

model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
model.add(Conv2D(64, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(3, 3)))
model.add(Dropout(0.25))
model.add(BatchNormalization())

model.add(Conv2D(128, (3, 3), activation='relu'))
model.add(MaxPooling2D(pool_size=(3, 3)))
model.add(Dropout(0.25))
model.add(BatchNormalization())

model.add(Flatten())
model.add(Dense(256, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

model.fit(x_train, y_train,
```

```
        batch_size=batch_size,
        epochs=epochs,
        verbose=1,
        validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])
```

Train on 60000 samples, validate on 10000 samples

Epoch 1/12

```
60000/60000 [=====] -
 107s 2ms/step - loss: 0.2693 - acc: 0.9150 -
val_loss: 0.0537 - val_acc: 0.9830
```

Epoch 2/12

```
60000/60000 [=====] -
 104s 2ms/step - loss: 0.0773 - acc: 0.9764 -
val_loss: 0.0297 - val_acc: 0.9914
```

Epoch 3/12

```
60000/60000 [=====] -
 105s 2ms/step - loss: 0.0582 - acc: 0.9826 -
val_loss: 0.0282 - val_acc: 0.9910
```

Epoch 4/12

```
60000/60000 [=====] -
 104s 2ms/step - loss: 0.0445 - acc: 0.9860 -
val_loss: 0.0386 - val_acc: 0.9865
```

Epoch 5/12

```
60000/60000 [=====] -
 104s 2ms/step - loss: 0.0407 - acc: 0.9879 -
val_loss: 0.0264 - val_acc: 0.9923
```

Epoch 6/12

```
60000/60000 [=====] -
 104s 2ms/step - loss: 0.0369 - acc: 0.9885 -
val_loss: 0.0209 - val_acc: 0.9927
```

Epoch 7/12

```
60000/60000 [=====] -
 104s 2ms/step - loss: 0.0343 - acc: 0.9891 -
```

```
val_loss: 0.0211 - val_acc: 0.9942
Epoch 8/12
60000/60000 [=====] -
  104s 2ms/step - loss: 0.0307 - acc: 0.9909 -
val_loss: 0.0188 - val_acc: 0.9942
Epoch 9/12
60000/60000 [=====] -
  104s 2ms/step - loss: 0.0270 - acc: 0.9919 -
val_loss: 0.0200 - val_acc: 0.9941
Epoch 10/12
60000/60000 [=====] -
  105s 2ms/step - loss: 0.0280 - acc: 0.9915 -
val_loss: 0.0211 - val_acc: 0.9934
Epoch 11/12
60000/60000 [=====] -
  107s 2ms/step - loss: 0.0278 - acc: 0.9913 -
val_loss: 0.0216 - val_acc: 0.9934
Epoch 12/12
60000/60000 [=====] -
  104s 2ms/step - loss: 0.0240 - acc: 0.9927 -
val_loss: 0.0199 - val_acc: 0.9941
Test loss: 0.01993270480447845
Test accuracy: 0.9941
```

## Architecture 4: kernel 5\*5 Conlayers 5 batch normalization

In [14]:

```
from keras.layers.normalization import BatchNormalization

model = Sequential()
model.add(Conv2D(16, kernel_size=(5, 5),
                 activation='relu',
                 input_shape=input_shape))

model.add(Conv2D(32, (5, 5), padding='same', strides=1, activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2), strides=None, padding='same', data_format=None))
model.add(BatchNormalization())

model.add(Conv2D(64, (5, 5), padding='same', activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2), strides=1, padding='same', data_format=None))
model.add(Dropout(0.25))
model.add(BatchNormalization())

model.add(Conv2D(128, (5, 5), padding='same', strides=1, activation='relu'))
model.add(MaxPooling2D(pool_size=(2, 2), strides=1, padding='same', data_format=None))
model.add(Dropout(0.25))
model.add(BatchNormalization())
```

```

model.add(Conv2D(256, (5, 5), activation='relu'))
model.add(Dropout(0.4))
model.add(BatchNormalization())

model.add(Flatten())
model.add(Dense(512, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(num_classes, activation='softmax'))

model.compile(loss=keras.losses.categorical_crossentropy,
              optimizer=keras.optimizers.Adadelta(),
              metrics=['accuracy'])

model.fit(x_train, y_train,
        batch_size=batch_size,
        epochs=epochs,
        verbose=1,
        validation_data=(x_test, y_test))
score = model.evaluate(x_test, y_test, verbose=0)
print('Test loss:', score[0])
print('Test accuracy:', score[1])

```

WARNING:tensorflow:From C:\Users\Public\Anaconda3\lib\site-packages\tensorflow\python\ops\math\_ops.py:3066: to\_int32 (from tensorflow.python.ops.math\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

Use tf.cast instead.

Train on 60000 samples, validate on 10000 samples

Epoch 1/12

```

60000/60000 [=====] -
 735s 12ms/step - loss: 0.4790 - acc: 0.9320 -
 val_loss: 0.0606 - val_acc: 0.9864

```

Epoch 2/12

```

60000/60000 [=====] -

```

```
733s 12ms/step - loss: 0.0751 - acc: 0.9815 -  
val_loss: 0.1089 - val_acc: 0.9766  
Epoch 3/12  
60000/60000 [=====] -  
703s 12ms/step - loss: 0.0545 - acc: 0.9867 -  
val_loss: 0.0681 - val_acc: 0.9850  
Epoch 4/12  
60000/60000 [=====] -  
706s 12ms/step - loss: 0.0432 - acc: 0.9892 -  
val_loss: 0.0295 - val_acc: 0.9933  
Epoch 5/12  
60000/60000 [=====] -  
725s 12ms/step - loss: 0.0348 - acc: 0.9908 -  
val_loss: 0.0304 - val_acc: 0.9938  
Epoch 6/12  
60000/60000 [=====] -  
720s 12ms/step - loss: 0.0299 - acc: 0.9924 -  
val_loss: 0.0428 - val_acc: 0.9900  
Epoch 7/12  
60000/60000 [=====] -  
724s 12ms/step - loss: 0.0265 - acc: 0.9928 -  
val_loss: 0.0239 - val_acc: 0.9949  
Epoch 8/12  
60000/60000 [=====] -  
765s 13ms/step - loss: 0.0251 - acc: 0.9934 -  
val_loss: 0.0256 - val_acc: 0.9948  
Epoch 9/12  
60000/60000 [=====] -  
763s 13ms/step - loss: 0.0201 - acc: 0.9948 -  
val_loss: 0.0315 - val_acc: 0.9939  
Epoch 10/12  
60000/60000 [=====] -  
761s 13ms/step - loss: 0.0193 - acc: 0.9951 -  
val_loss: 0.0244 - val_acc: 0.9948  
Epoch 11/12  
60000/60000 [=====] -  
768s 13ms/step - loss: 0.0173 - acc: 0.9961 -
```

```
val_loss: 0.0265 - val_acc: 0.9950
Epoch 12/12
60000/60000 [=====] -
  769s 13ms/step - loss: 0.0162 - acc: 0.9961 -
  val_loss: 0.0284 - val_acc: 0.9954
Test loss: 0.02835287011177452
Test accuracy: 0.9954
```

In [4]:

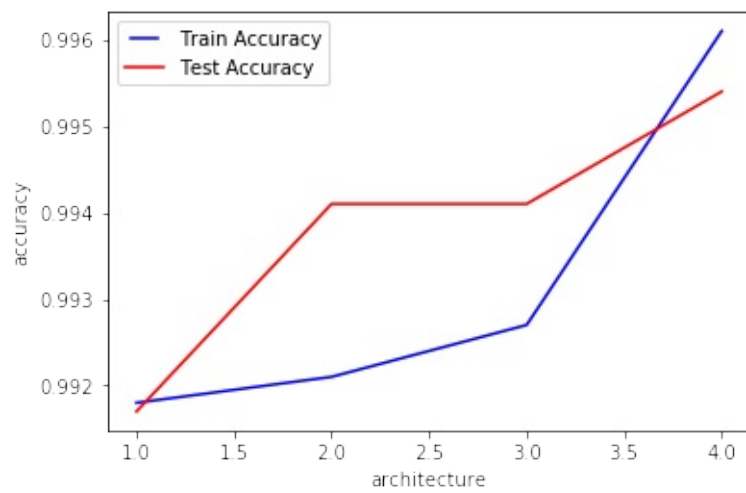
```
#plotting error plots for all the models

values =[1,2,3,4]
train_results=[0.9918,0.9921,0.9927,0.9961]
test_results=[0.9917,0.9941,0.9941,0.9954]

import matplotlib.pyplot as plt
from matplotlib.legend_handler import HandlerLine2D
line1, = plt.plot(values, train_results, 'b', label="Train Ac
curacy")
line2, = plt.plot(values, test_results, 'r', label="Test Accu
racy")

plt.legend(handler_map={line1: HandlerLine2D(numpoints=2)})
plt.ylabel("accuracy")
plt.xlabel("architecture")
plt.show()
```





**1.we can observe that for architecture 3 we get best results even model 2 give good results but among them model 3 is best .**

2.For model 4 and model 1, we could observe that test accuracy is more train accuracy which indicates the accuracy for unseen data is more than the train data which is not common.

# conclusions

In [1]:

```
from prettytable import PrettyTable

#If you get a ModuleNotFoundError error , install prettytable
using: pip3 install prettytable

x = PrettyTable()
x.field_names = ["CNN- architecture ", "convolution layers",
"kernel size", "Regularization", "Train accuracy", "Test Acc
uracy"]

x.add_row(["Architecture 1", 2, "3*3", "Droupout", 0.9918, 0.
9917])
x.add_row(["Architecture 2", 3, "3*3", "Droupout", 0.9921, 0.
9941])
x.add_row(["Architecture 3", 3, "3*3", "Batch", 0.9927, 0.994
1])
x.add_row(["Architecture 4", 5, "5*5", "Droupout+batch",0.996
1 ,0.9954 ])

print(x)
```

```
+-----+-----+-----+
-----+-----+-----+
-----+
| CNN- architecture | convolution layers | ke
rnel size | Regularization | Train accuracy |
Test Accuracy |
+-----+-----+-----+
-----+-----+-----+
```

```

-----+
| Architecture 1 |          2          |
| 3*3      | Droupout      | 0.9918      |
| 0.9917      |                          |
| Architecture 2 |          3          |
| 3*3      | Droupout      | 0.9921      |
| 0.9941      |                          |
| Architecture 3 |          3          |
| 3*3      | Batch        | 0.9927      |
| 0.9941      |                          |
| Architecture 4 |          5          |
| 5*5      | Droupout+batch | 0.9961      |
| 0.9954      |                          |
+-----+-----+-----+
-----+-----+-----+
-----+

```

In [ ]:

```

#We could observe with increase in kernel size as performance
is increasing but it is taking longer time to execute

```