# **Multinominal Classification**

# **Mnist dataset classification**

Classify 28x28 bit image as a digit from 0-9

In [0]:

```
import tensorflow as tf
tf.enable_eager_execution()
import matplotlib.pyplot as plt
import numpy as np
import random
from progressbar import progressbar
from keras.layers import Dense
from sklearn.model_selection import train_test_split
import time
```

# **Feed-Forward Neural Network Architecture**

```
In [0]:
```

```
def build_fc_model(activation):
    fc_model = tf.keras.Sequential([
        tf.keras.layers.Flatten(),
        # TODO: Define the rest of the model.
        tf.keras.layers.Dense(128, input_shape=(784,), activation=activation),
        tf.keras.layers.Dense(10, activation='softmax')

])
    return fc_model
```

# **Convolutional Neural Network Architecture**

```
In [0]:
```

```
def build_cnn_model(activation):
    cnn_model = tf.keras.Sequential([
        # TODO: Define the model.
        tf.keras.layers.Conv2D(24, 3, activation = activation, input_shape=(28,28,1,)),
        tf.keras.layers.MaxPooling2D(),
        tf.keras.layers.Conv2D(36, 3, activation = activation),
        tf.keras.layers.MaxPooling2D(),
        tf.keras.layers.Flatten(),
        tf.keras.layers.Dense(784, activation=activation),
        tf.keras.layers.Dense(128, activation=activation),
        tf.keras.layers.Dense(10, activation='softmax')
])
    return cnn_model
```

# Experiment code to be able to run different configurations

```
In [0]:
```

```
def run experiment(build model, activation, optimizer, loss, X, y, verbose = 1):
 model = build model(activation)
 BATCH SIZE = 64
 EPOCHS = 5
 train_X, val_X, train_y, val_y = train_test_split(X, y, train_size=0.8, random_state=0)
 # TODO compile and fit the model with the appropriate parameters.
 model.compile(loss=loss, metrics=['categorical_accuracy'], optimizer=optimizer)
 start = time.time()
 history = model.fit(train X, train y, verbose=verbose, batch size=BATCH SIZE, epochs=EPOCHS, validation da
ta = (val X, val y))
 end = time.time()
 model.summary()
 # Plot training & validation accuracy values
 plt.plot(history.history['categorical_accuracy'])
 plt.plot(history.history['val_categorical_accuracy'])
 plt.title('Model accuracy')
 plt.ylabel('Accuracy')
 plt.xlabel('Epoch')
 plt.legend(['Train', 'Test'], loc='upper left')
 plt.show()
 # Plot training & validation loss values
 plt.plot(history.history['loss'])
 plt.plot(history.history['val_loss'])
 plt.title('Model loss')
 plt.ylabel('Loss')
 plt.xlabel('Epoch')
 plt.legend(['Train', 'Test'], loc='upper left')
 plt.show()
 print('Last validation loss : ', history.history['val_loss'][-1], ' | last training loss : ', history.hist
ory['loss'][-1])
 print('Last validation accuracy : ', history.history['val_categorical_accuracy'][-1], ' | last training ac
curacy : ', history.history['categorical_accuracy'][-1])
 print('Time taken in training : ', end - start, ' sec')
  return model, history.history['val categorical accuracy'][-1]
```

# Get best configuration from model dictionary according to highest validation accuracy

```
In [0]:
```

```
def get_best_configuration(models):
    maxVal = 0
    maxkey = ''
    for key,(model, value) in models.items():
        if (value > maxVal):
            maxkey = key
            maxVal = value
    return maxkey
```

### Model evaluation code

```
In [0]:
```

```
def evaluate_model(model, test_X, test_y):
    BATCH_SIZE = 64
    start = time.time()
    test_loss, test_acc = model.evaluate(test_X, test_y, verbose = 0, batch_size = BATCH_SIZE)
    end = time.time()
    print('Test loss:', test_loss)
    print('Test accuracy:', test_acc)
    print('Time taken in training : ', end - start, ' sec')
```

# **Dataset setup**

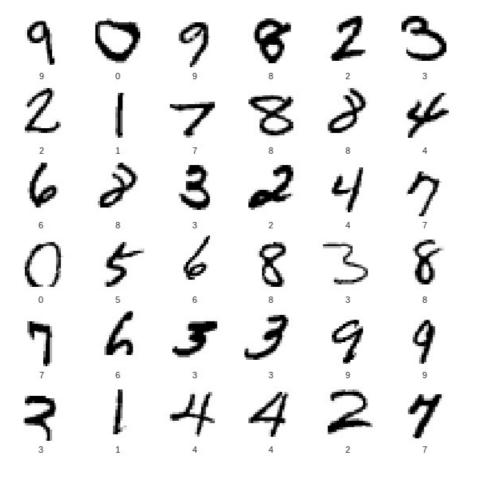
```
In [0]:
```

```
mnist = tf.keras.datasets.mnist
(train_images, train_labels), (test_images, test_labels) = mnist.load_data()
train_images = np.expand_dims(train_images, axis=-1)/255.
train_labels = np.int64(train_labels)
test_images = np.expand_dims(test_images, axis=-1)/255.
test_labels = np.int64(test_labels)
print(train_images.shape, train_labels.shape, test_images.shape, test_labels.shape)
train_labels = tf.keras.utils.to_categorical(train_labels, 10)
test_labels = tf.keras.utils.to_categorical(test_labels, 10)
print('After one hot encoding :', train_images.shape, train_labels.shape, test_images.shape, test_labels.shape)

(60000, 28, 28, 1) (60000,) (10000, 28, 28, 1) (10000,)
After one hot encoding : (60000, 28, 28, 1) (60000, 10) (10000, 28, 28, 1) (10000, 10)
```

In [0]:

```
plt.figure(figsize=(10,10))
random_inds = np.random.choice(60000,36)
for i in range(36):
    plt.subplot(6,6,i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    image_ind = random_inds[i]
    plt.imshow(np.squeeze(train_images[image_ind]), cmap=plt.cm.binary)
    plt.xlabel(np.where(1 == train_labels[image_ind])[0][0])
```



# Fully connected configuration experiments

All will use categorical crossentropy loss.

In [0]:

fc models = {}

#### **Default Adam with tanh**

#### In [0]:

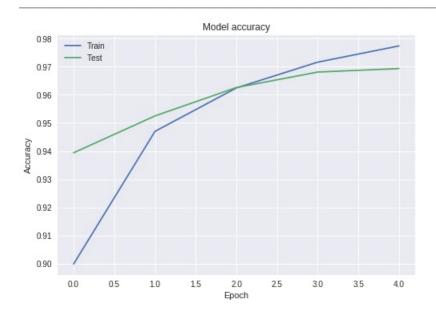
fc\_models['adam\_tanh'] = run\_experiment(build\_fc\_model, 'tanh', 'adam', 'categorical\_crossentropy', train\_im
ages, train\_labels)

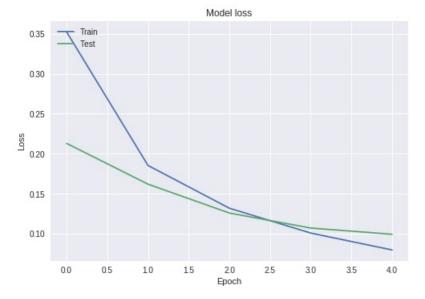
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F
rom version 0.21, test\_size will always complement train\_size unless both are specified.
 FutureWarning)

Train on 48000 samples, validate on 12000 samples Epoch 1/5 48000/48000 [================= ] - 7s 136us/sample - loss: 0.3522 - categorical\_acc uracy: 0.8999 - val\_loss: 0.2129 - val\_categorical\_accuracy: 0.9394 uracy: 0.9470 - val loss: 0.1620 - val categorical accuracy: 0.9525 Epoch 3/5 racy: 0.9624 - val loss: 0.1259 - val categorical accuracy: 0.9626 Epoch 4/5 48000/48000 [===== =====] - 5s 95us/sample - loss: 0.1009 - categorical accu racy: 0.9716 - val loss: 0.1072 - val categorical accuracy: 0.9681 Epoch 5/5 racy: 0.9774 - val\_loss: 0.0994 - val\_categorical\_accuracy: 0.9693

Layer (type)	Output Shape	Param #
flatten_51 (Flatten)	multiple	Θ
dense_122 (Dense)	multiple	100480
dense_123 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.09937453757723172 | last training loss : 0.07977328032627702

Last validation accuracy: 0.96933335 | last training accuracy: 0.977375

Time taken in training: 27.41925072669983 sec

#### Adam with learning rate 0.05 and tanh activation

#### In [0]:

```
fc_models['adam_0.005_tanh'] = run_experiment(build_fc_model, 'tanh', tf.keras.optimizers.Adam(0.005), 'cate
gorical crossentropy', train images, train labels)
```

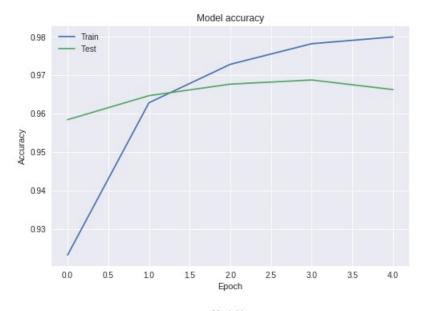
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

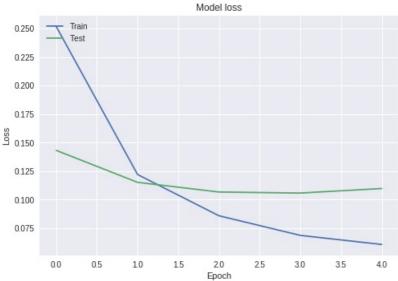
```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
uracy: 0.9232 - val loss: 0.1431 - val categorical accuracy: 0.9584
Epoch 2/5
uracy: 0.9628 - val_loss: 0.1151 - val_categorical_accuracy: 0.9647
Epoch 3/5
uracy: 0.9728 - val loss: 0.1066 - val categorical accuracy: 0.9677
Epoch 4/5
48000/48000 [============
                     =======] - 5s 97us/sample - loss: 0.0686 - categorical accu
racy: 0.9782 - val_loss: 0.1057 - val_categorical_accuracy: 0.9688
Epoch 5/5
                             5s 97us/sample - loss: 0.0606 - categorical accu
```

40000/40000 [=		====] - 58	9/us/sample -	1055: 0.0000	o - categoricat_ac	Cu
racv: 0.9800 -	val loss: 0.1097 - val c	categorical	accuracy: 0.96	663		
	141_1000. 01100. 141_0			,,,,		
				_		
Layer (type)	Output Sha	ape	Param #			

flatten_52 (Flatten)	multiple	Θ
dense_124 (Dense)	multiple	100480
dense_125 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.10969211432834466 | last training loss : 0.06063181223099431

Last validation accuracy: 0.96625 | last training accuracy: 0.97995836

Time taken in training : 28.098117113113403 sec

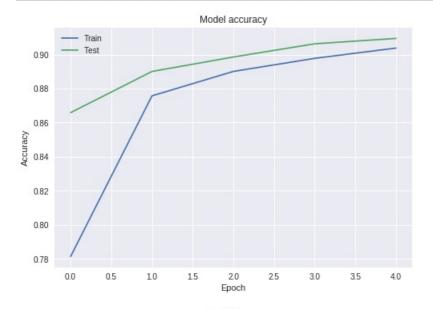
# SGD and tanh activation

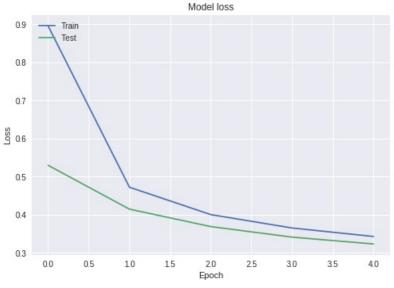
# In [0]:

fc\_models['sgd\_tanh'] = run\_experiment(build\_fc\_model, 'tanh', 'SGD', 'categorical\_crossentropy', train\_imag
es, train\_labels)

Layer (type)	Output Shape	Param #
flatten_53 (Flatten)	multiple	0
dense_126 (Dense)	multiple	100480
dense_127 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss: 0.32363501691818236 | last training loss: 0.34328800495465595 Last validation accuracy: 0.9095 | last training accuracy: 0.9038333 Time taken in training: 24.46976590156555 sec

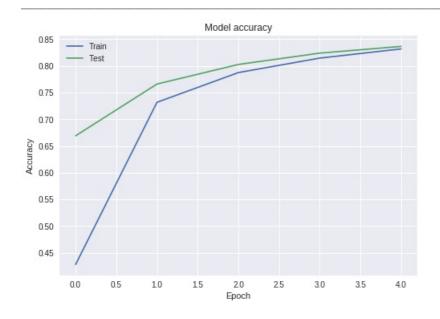
fc\_models['sgd\_0.001\_tanh'] = run\_experiment(build\_fc\_model, 'tanh', tf.keras.optimizers.SGD(0.001), 'catego
rical\_crossentropy', train\_images, train\_labels)

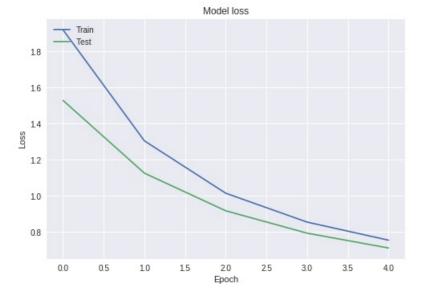
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F
rom version 0.21, test\_size will always complement train\_size unless both are specified.
 FutureWarning)

Train on 48000 samples, validate on 12000 samples Epoch 1/5 uracy: 0.4281 - val loss: 1.5291 - val categorical accuracy: 0.6692 Epoch 2/5 ======] - 4s 87us/sample - loss: 1.3050 - categorical accu 48000/48000 [===== racy: 0.7319 - val loss: 1.1259 - val categorical accuracy: 0.7661 Epoch 3/5 racy: 0.7875 - val\_loss: 0.9171 - val\_categorical\_accuracy: 0.8027 Epoch 4/5 racy: 0.8147 - val loss: 0.7930 - val categorical accuracy: 0.8240 Epoch 5/5 racy: 0.8319 - val\_loss: 0.7109 - val\_categorical\_accuracy: 0.8366

Layer (type)	Output Shape	Param #
flatten_54 (Flatten)	multiple	0
dense_128 (Dense)	multiple	100480
dense_129 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.7108835372924804 | last training loss : 0.7542476077079773

Last validation accuracy : 0.8365833 | last training accuracy : 0.8318958

Time taken in training: 24.809332847595215 sec

#### SGD with learning rate 0.00465 and tanh activation

#### In [0]:

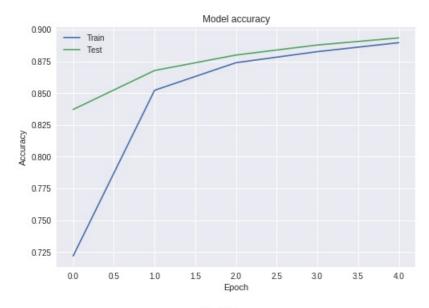
```
fc_models['sgd_0.00465_tanh'] = run_experiment(build_fc_model, 'tanh', tf.keras.optimizers.SGD(0.00465), 'ca
tegorical crossentropy', train images, train labels)
```

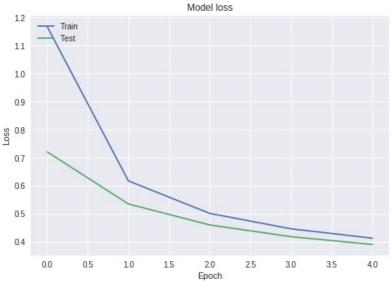
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
uracy: 0.7218 - val loss: 0.7213 - val categorical accuracy: 0.8372
Epoch 2/5
racy: 0.8523 - val loss: 0.5351 - val categorical accuracy: 0.8678
Epoch 3/5
racy: 0.8740 - val loss: 0.4603 - val categorical accuracy: 0.8800
Epoch 4/5
48000/48000 [=====
                    ======] - 4s 84us/sample - loss: 0.4465 - categorical accu
racy: 0.8827 - val_loss: 0.4184 - val_categorical_accuracy: 0.8879
Epoch 5/5
racy: 0.8898 - val loss: 0.3907 - val categorical accuracy: 0.8935
```

Layer (type)	Output Shape	Param #
flatten_55 (Flatten)	multiple	0
dense_130 (Dense)	multiple	100480
dense_131 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.39066985233624774 | last training loss : 0.41296696801980337

Last validation accuracy : 0.8935 | last training accuracy : 0.88975

Time taken in training: 24.42633819580078 sec

# Adam and relu activation

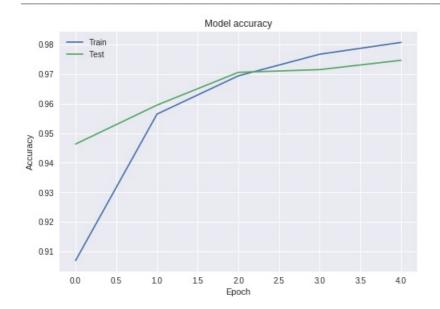
# In [0]:

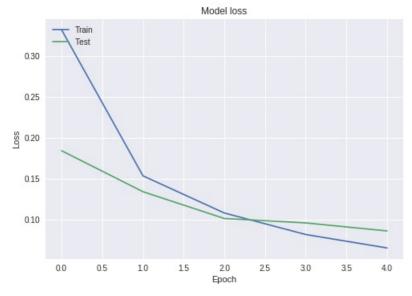
fc\_models['adam\_relu'] = run\_experiment(build\_fc\_model, 'relu', 'adam', 'categorical\_crossentropy', train\_im
ages, train\_labels)

Train on 48000 samples, validate on 12000 samples Epoch 1/5 uracy: 0.9069 - val\_loss: 0.1843 - val\_categorical\_accuracy: 0.9463 Epoch 2/5 48000/48000 [=============== ] - 5s 98us/sample - loss: 0.1536 - categorical\_accu racy: 0.9565 - val\_loss: 0.1342 - val\_categorical\_accuracy: 0.9595 Epoch 3/5 =======] - 5s 98us/sample - loss: 0.1081 - categorical accu 48000/48000 [======== racy: 0.9694 - val\_loss: 0.1014 - val\_categorical\_accuracy: 0.9706 Epoch 4/5 =======] - 5s 98us/sample - loss: 0.0818 - categorical accu 48000/48000 [======== racy: 0.9767 - val\_loss: 0.0960 - val\_categorical\_accuracy: 0.9715 Epoch 5/5

Layer (type)	Output Shape	Param #
flatten_56 (Flatten)	multiple	0
dense_132 (Dense)	multiple	100480
dense_133 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.08621248503774405 | last training loss : 0.0653957677303503 Last validation accuracy : 0.97466666 | last training accuracy : 0.98070836 Time taken in training : 27.997569799423218 sec

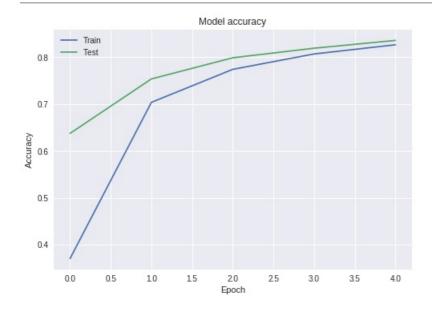
fc\_models['sgd\_0.001\_relu'] = run\_experiment(build\_fc\_model, 'relu', tf.keras.optimizers.SGD(0.001), 'catego
rical\_crossentropy', train\_images, train\_labels)

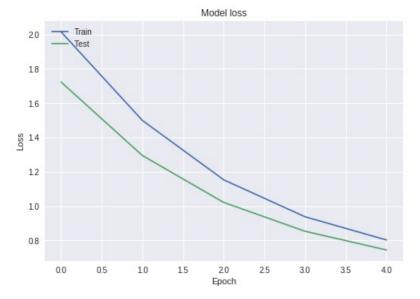
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

Train on 48000 samples, validate on 12000 samples Epoch 1/5 48000/48000 [=============== ] - 6s 130us/sample - loss: 2.0147 - categorical acc uracy: 0.3707 - val loss: 1.7225 - val categorical accuracy: 0.6377 Epoch 2/5 ======] - 4s 84us/sample - loss: 1.4991 - categorical accu 48000/48000 [===== racy: 0.7039 - val loss: 1.2947 - val categorical accuracy: 0.7538 Epoch 3/5 racy: 0.7742 - val\_loss: 1.0217 - val\_categorical\_accuracy: 0.7989 Epoch 4/5 racy: 0.8071 - val loss: 0.8538 - val categorical accuracy: 0.8193 racy: 0.8267 - val\_loss: 0.7450 - val\_categorical\_accuracy: 0.8360

Layer (type)	Output Shape	Param #
flatten_57 (Flatten)	multiple	0
dense_134 (Dense)	multiple	100480
dense_135 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss: 0.7449987587928772 | last training loss: 0.8031060738563538

Last validation accuracy: 0.836 | last training accuracy: 0.82666665

Time taken in training: 24.851763248443604 sec

#### SGD with learning rate 0.0009 and relu activation

#### In [0]:

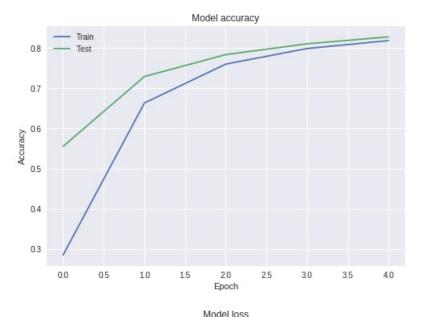
fc\_models['sgd\_0.0009\_relu'] = run\_experiment(build\_fc\_model, 'relu', tf.keras.optimizers.SGD(0.0009), 'cate
gorical crossentropy', train images, train labels)

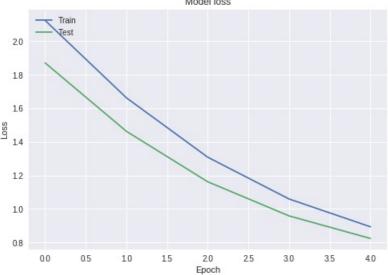
/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
48000/48000 [==============] - 6s 134us/sample - loss: 2.1246 - categorical acc
uracy: 0.2856 - val loss: 1.8711 - val categorical accuracy: 0.5554
Epoch 2/5
48000/48000 [================ ] - 4s 88us/sample - loss: 1.6628 - categorical accu
racy: 0.6632 - val loss: 1.4630 - val categorical accuracy: 0.7290
Epoch 3/5
racy: 0.7599 - val loss: 1.1620 - val categorical accuracy: 0.7834
Epoch 4/5
48000/48000 [=====
                          ======] - 4s 85us/sample - loss: 1.0593 - categorical accu
racy: 0.7985 - val_loss: 0.9586 - val_categorical_accuracy: 0.8102
Epoch 5/5
racy: 0.8183 - val loss: 0.8239 - val categorical accuracy: 0.8274
```

Layer (type)	Output Shape	Param #
flatten_58 (Flatten)	multiple	0
dense_136 (Dense)	multiple	100480
dense_137 (Dense)	multiple	1290

Total params: 101,770 Trainable params: 101,770 Non-trainable params: 0





Last validation loss : 0.8238914051055908 | last training loss : 0.8936570182641347

Last validation accuracy: 0.82741666 | last training accuracy: 0.81827086

Time taken in training : 25.428094625473022 sec

# Best fully connected configuration evaluation

In [0]:

key = get\_best\_configuration(fc\_models)
values = fc\_models[key]
print(key, ' has the best validation accuracy ', values[1])
evaluate\_model(values[0], test\_images, test\_labels)

adam\_relu has the best validation accuracy 0.97466666

Test loss: 0.0911650103919208

Test accuracy: 0.9705

Time taken in training : 0.512174129486084 sec

# **Convolutional configuration experiments**

All will use categorical crossentropy loss.

In [0]:

cnn\_models = {}

cnn models['adam tanh'] = run experiment(build cnn model, 'tanh', 'adam', 'categorical crossentropy', train images, train labels)

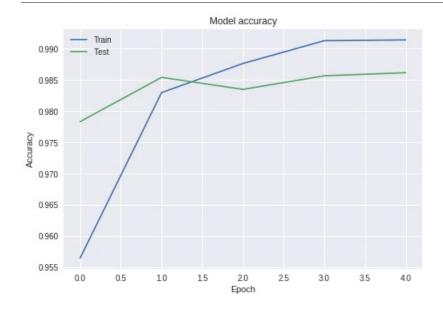
/usr/local/lib/python3.6/dist-packages/sklearn/model selection/ split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

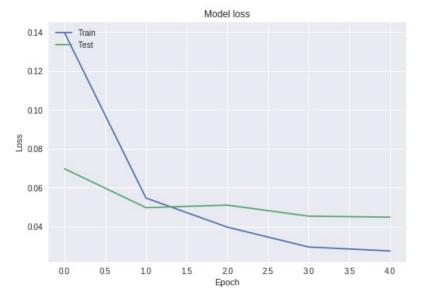
Train on 48000 samples, validate on 12000 samples curacy: 0.9565 - val\_loss: 0.0697 - val\_categorical\_accuracy: 0.9783 uracy: 0.9830 - val loss: 0.0497 - val categorical accuracy: 0.9854 uracy: 0.9876 - val loss: 0.0510 - val categorical accuracy: 0.9835 Epoch 4/5 uracy: 0.9913 - val loss: 0.0453 - val categorical accuracy: 0.9857 Epoch 5/5 

uracy: 0.9914 - val loss: 0.0448 - val categorical accuracy: 0.9862

Layer (type)	Output Shape	Param #
conv2d_40 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_40 (MaxPooling	(None, 13, 13, 24)	0
conv2d_41 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_41 (MaxPooling	(None, 5, 5, 36)	0
flatten_60 (Flatten)	(None, 900)	0
dense_140 (Dense)	(None, 784)	706384
dense_141 (Dense)	(None, 128)	100480
dense_142 (Dense)	(None, 10)	1290 ======

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss : 0.04478418886351088 | last training loss : 0.027422362679770836

Last validation accuracy: 0.98616666 | last training accuracy: 0.9914167

Time taken in training: 49.66046357154846 sec

#### Adam with learning rate 0.01 and tanh activation

#### In [0]:

cnn models['adam 0.01 tanh'] = run experiment(build cnn model, 'tanh', tf.keras.optimizers.Adam(0.01), 'categorical crossentropy', train images, train labels)

/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test size will always complement train size unless both are specified. FutureWarning)

```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
```

48000/48000 [=============] - 12s 247us/sample - loss: 0.2246 - categorical ac

curacy: 0.9355 - val loss: 0.1723 - val categorical accuracy: 0.9465

Epoch 2/5

uracy: 0.9375 - val loss: 0.2097 - val categorical accuracy: 0.9334

Epoch 3/5

uracy: 0.9232 - val loss: 0.2788 - val categorical accuracy: 0.9128

Epoch 4/5

48000/48000 [===== ======] - 9s 182us/sample - loss: 0.2916 - categorical acc

uracy: 0.9114 - val\_loss: 0.3056 - val\_categorical\_accuracy: 0.9094

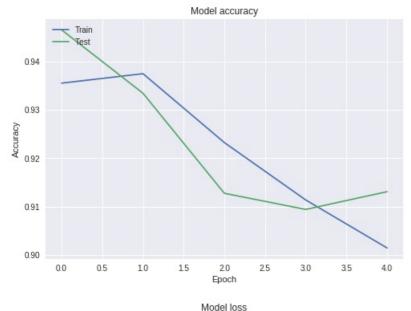
Epoch 5/5

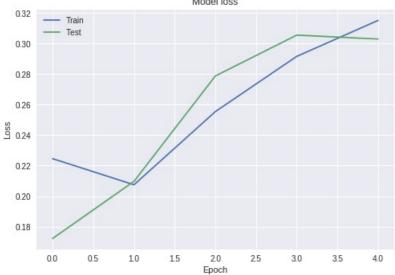
uracy: 0.9015 - val loss: 0.3030 - val categorical accuracy: 0.9131

Layer (type)	Output Shape	Param #
conv2d_42 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_42 (MaxPooling	(None, 13, 13, 24)	0
conv2d_43 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_43 (MaxPooling	(None, 5, 5, 36)	0
flatten_61 (Flatten)	(None, 900)	0
dense_143 (Dense)	(None, 784)	706384
dense_144 (Dense)	(None, 128)	100480
dense_145 (Dense) ====================================	(None, 10) ====================================	1290 =======

Total params: 816,206 Trainable params: 816,206

Non-trainable params: 0





Last validation loss : 0.30299906196196874 | last training loss : 0.3151985639979442

Last validation accuracy: 0.9130833 | last training accuracy: 0.9014583

Time taken in training: 50.07255673408508 sec

### Adam with learning rate 0.005 and tanh activation

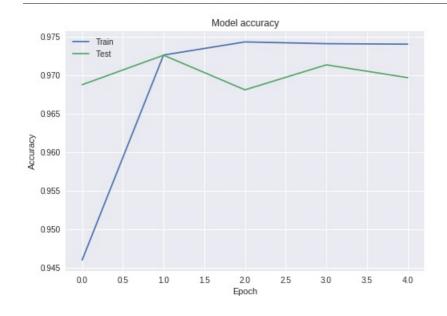
# In [0]:

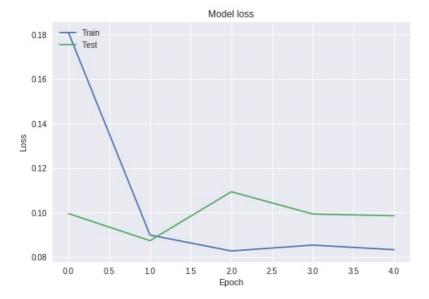
 $cnn_models['adam_0.005_tanh'] = run_experiment(build_cnn_model, 'tanh', tf.keras.optimizers.Adam(0.005), 'categorical_crossentropy', train_images, train_labels)$ 

Layer (type)	Output Shape	Param #
conv2d_44 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_44 (MaxPooling	(None, 13, 13, 24)	0
conv2d_45 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_45 (MaxPooling	(None, 5, 5, 36)	0
flatten_62 (Flatten)	(None, 900)	Θ
dense_146 (Dense)	(None, 784)	706384
dense_147 (Dense)	(None, 128)	100480
dense_148 (Dense)	(None, 10)	1290

uracy: 0.9740 - val loss: 0.0985 - val categorical accuracy: 0.9697

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss: 0.0985477549880743 | last training loss: 0.08330577919166535

Last validation accuracy : 0.96966666 | last training accuracy : 0.97402084

Time taken in training: 50.204779386520386 sec

#### SGD and tanh activation

#### In [0]:

cnn models['sgd tanh'] = run experiment(build cnn model, 'tanh', 'sgd', 'categorical crossentropy', train im ages, train labels)

/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test size will always complement train size unless both are specified. FutureWarning)

```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
```

48000/48000 [=============] - 10s 204us/sample - loss: 0.7161 - categorical ac

curacy: 0.8178 - val loss: 0.3181 - val categorical accuracy: 0.9096

Epoch 2/5

48000/48000 [=============== ] - 7s 150us/sample - loss: 0.2689 - categorical acc

uracy: 0.9241 - val loss: 0.2115 - val categorical accuracy: 0.9406

Epoch 3/5

uracy: 0.9439 - val loss: 0.1613 - val categorical accuracy: 0.9556

Epoch 4/5

48000/48000 [===== ======] - 7s 149us/sample - loss: 0.1548 - categorical acc

uracy: 0.9564 - val\_loss: 0.1290 - val\_categorical\_accuracy: 0.9643

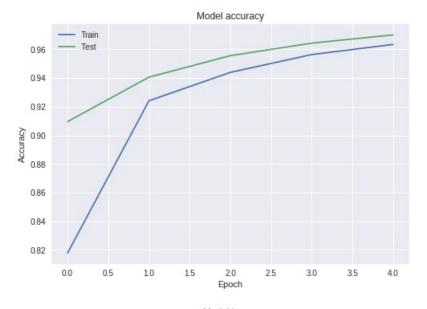
Epoch 5/5

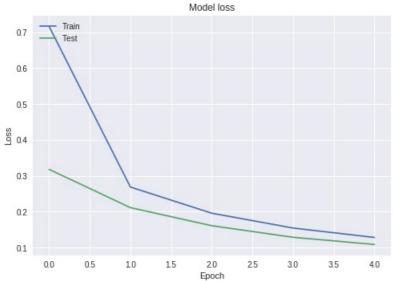
uracy: 0.9634 - val loss: 0.1090 - val categorical accuracy: 0.9701

Layer (type)	Output Shape	Param #
conv2d_46 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_46 (MaxPooling	(None, 13, 13, 24)	0
conv2d_47 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_47 (MaxPooling	(None, 5, 5, 36)	0
flatten_63 (Flatten)	(None, 900)	0
dense_149 (Dense)	(None, 784)	706384
dense_150 (Dense)	(None, 128)	100480
dense_151 (Dense)	(None, 10)	1290

Total params: 816,206 Trainable params: 816,206

Non-trainable params: 0





Last validation loss : 0.10904256647825242 | last training loss : 0.12853660616775353

Last validation accuracy: 0.97008336 | last training accuracy: 0.9634167

Time taken in training: 40.89606785774231 sec

# SGD with learning rate 0.001 and tanh activation

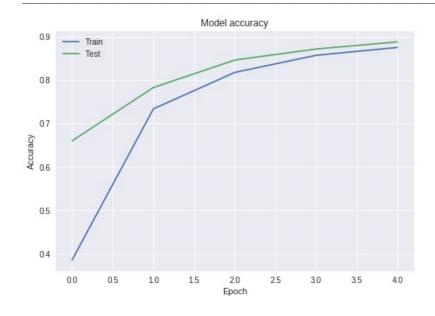
# In [0]:

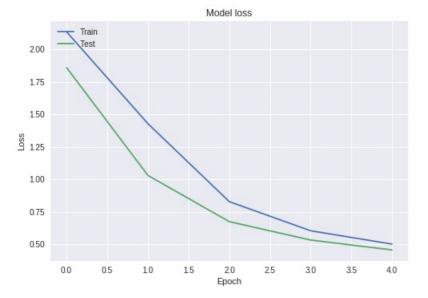
 $\label{local_constrain} $$\operatorname{cnn_models['sgd_0.001_tanh'] = run_experiment(build_cnn_model, 'tanh', tf.keras.optimizers.SGD(0.001), 'cate gorical_crossentropy', train_images, train_labels)}$ 

Layer (type)	Output Shape	Param #
conv2d_48 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_48 (MaxPooling	(None, 13, 13, 24)	0
conv2d_49 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_49 (MaxPooling	(None, 5, 5, 36)	0
flatten_64 (Flatten)	(None, 900)	0
dense_152 (Dense)	(None, 784)	706384
dense_153 (Dense)	(None, 128)	100480
dense_154 (Dense)	(None, 10)	1290

uracy: 0.8749 - val loss: 0.4555 - val categorical accuracy: 0.8877

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss: 0.45547159695625306 | last training loss: 0.5004099428653717

Last validation accuracy: 0.88766664 | last training accuracy: 0.8749167

Time taken in training: 42.349854707717896 sec

#### SGD with learning rate 0.005 and tanh activation

#### In [0]:

cnn models['sgd 0.005 tanh'] = run experiment(build cnn model, 'tanh', tf.keras.optimizers.SGD(0.005), 'cate gorical crossentropy', train images, train labels)

/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test size will always complement train size unless both are specified. FutureWarning)

Train on 48000 samples, validate on 12000 samples

Epoch 1/5

curacy: 0.7687 - val loss: 0.4473 - val categorical accuracy: 0.8854

Epoch 2/5

uracy: 0.8974 - val loss: 0.3124 - val categorical accuracy: 0.9144

Epoch 3/5

48000/48000 [============== ] - 8s 157us/sample - loss: 0.2915 - categorical acc

uracy: 0.9185 - val loss: 0.2542 - val categorical accuracy: 0.9296

Epoch 4/5

48000/48000 [===== ======] - 8s 156us/sample - loss: 0.2456 - categorical acc

uracy: 0.9305 - val\_loss: 0.2178 - val\_categorical\_accuracy: 0.9390

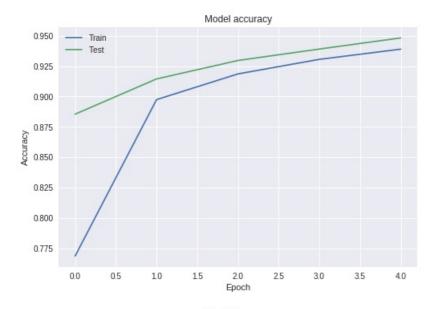
Epoch 5/5

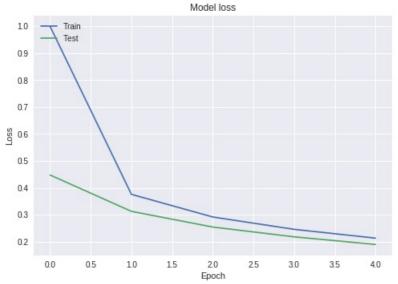
uracy: 0.9389 - val loss: 0.1895 - val categorical accuracy: 0.9482

Layer (type)	Output Shape	Param #
conv2d_50 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_50 (MaxPooling	(None, 13, 13, 24)	0
conv2d_51 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_51 (MaxPooling	(None, 5, 5, 36)	0
flatten_65 (Flatten)	(None, 900)	0
dense_155 (Dense)	(None, 784)	706384
dense_156 (Dense)	(None, 128)	100480
dense_157 (Dense)	(None, 10)	1290

Total params: 816,206 Trainable params: 816,206

Non-trainable params: 0





Last validation loss : 0.18947235067685445 | last training loss : 0.21299565037091572

Last validation accuracy: 0.94816667 | last training accuracy: 0.9389167

Time taken in training: 42.358307123184204 sec

#### SGD with learning rate 0.0055 and tanh activation

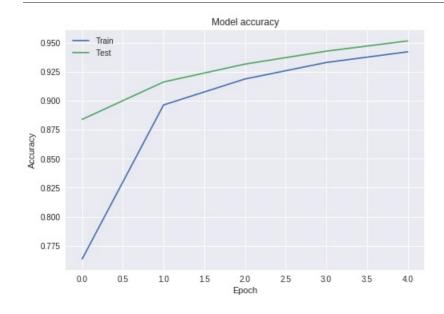
# In [0]:

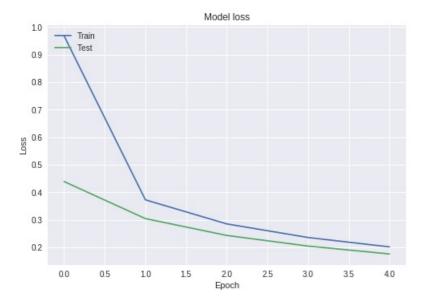
 $cnn\_models['sgd\_0.0055\_tanh'] = run\_experiment(build\_cnn\_model, 'tanh', tf.keras.optimizers.SGD(0.0055), 'categorical\_crossentropy', train\_images, train\_labels)$ 

Layer (type)	Output	Shape	Param #
conv2d_52 (Conv2D)	(None,	26, 26, 24)	240
max_pooling2d_52 (MaxPooling	(None,	13, 13, 24)	0
conv2d_53 (Conv2D)	(None,	11, 11, 36)	7812
max_pooling2d_53 (MaxPooling	(None,	5, 5, 36)	0
flatten_66 (Flatten)	(None,	900)	Θ
dense_158 (Dense)	(None,	784)	706384
dense_159 (Dense)	(None,	128)	100480
dense_160 (Dense)	(None,	10)	1290 =======

uracy: 0.9422 - val loss: 0.1758 - val categorical accuracy: 0.9516

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss : 0.17580808498462042 | last training loss : 0.2014609561562538

Last validation accuracy : 0.9515833 | last training accuracy : 0.9421875

Time taken in training: 42.91641426086426 sec

#### SGD and relu activation

#### In [0]:

cnn\_models['sgd\_relu'] = run\_experiment(build\_cnn\_model, 'relu', tf.keras.optimizers.SGD(), 'categorical\_cro
ssentropy', train\_images, train\_labels)

/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test\_size will always complement train\_size unless both are specified. FutureWarning)

```
Train on 48000 samples, validate on 12000 samples Epoch 1/5 48000/48000 [=======] - 10s
```

curacy: 0.7644 - val\_loss: 0.2536 - val\_categorical\_accuracy: 0.9273

Epoch 2/5

uracy: 0.9310 - val\_loss: 0.1614 - val\_categorical\_accuracy: 0.9510

Epoch 3/5

uracy: 0.9537 - val\_loss: 0.1200 - val\_categorical\_accuracy: 0.9638

Epoch 4/5

uracy: 0.9640 - val\_loss: 0.1016 - val\_categorical\_accuracy: 0.9691

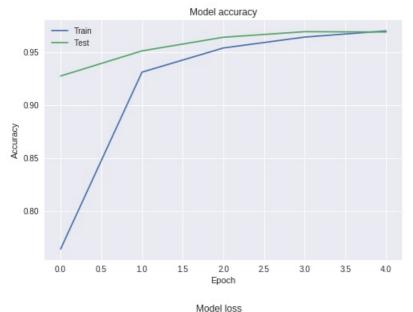
Epoch 5/5

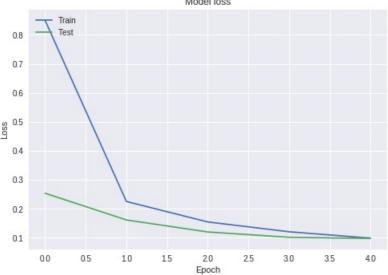
uracy: 0.9698 - val loss: 0.0975 - val categorical accuracy: 0.9688

Layer (type)	Output Shape	Param #
conv2d_54 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_54 (MaxPooling	(None, 13, 13, 24)	0
conv2d_55 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_55 (MaxPooling	(None, 5, 5, 36)	0
flatten_67 (Flatten)	(None, 900)	0
dense_161 (Dense)	(None, 784)	706384
dense_162 (Dense)	(None, 128)	100480
dense_163 (Dense)	(None, 10)	1290

Total params: 816,206 Trainable params: 816,206

Non-trainable params: 0





Last validation loss : 0.09753970853736003 | last training loss : 0.09872187430908283

Last validation accuracy: 0.96875 | last training accuracy: 0.9698125

Time taken in training: 42.44763159751892 sec

# SGD with learning rate 0.001 and relu activation

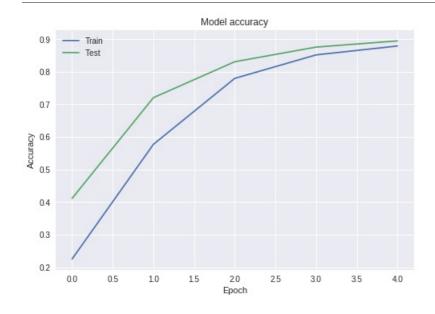
# In [0]:

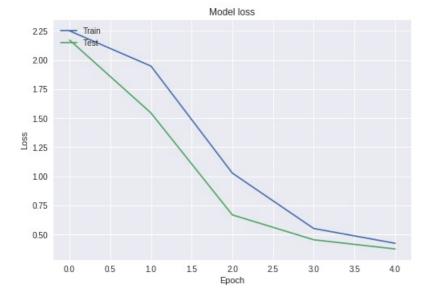
 $cnn\_models['sgd\_0.001\_relu'] = run\_experiment(build\_cnn\_model, 'relu', tf.keras.optimizers.SGD(0.001), 'cate gorical\_crossentropy', train\_images, train\_labels)$ 

Layer (type)	Output	Shape	Param #
conv2d_56 (Conv2D)	(None,	26, 26, 24)	240
max_pooling2d_56 (MaxPooling	(None,	13, 13, 24)	0
conv2d_57 (Conv2D)	(None,	11, 11, 36)	7812
max_pooling2d_57 (MaxPooling	(None,	5, 5, 36)	0
flatten_68 (Flatten)	(None,	900)	0
dense_164 (Dense)	(None,	784)	706384
dense_165 (Dense)	(None,	128)	100480
dense_166 (Dense)	(None,	10)	1290 ======

uracy: 0.8788 - val loss: 0.3766 - val categorical accuracy: 0.8944

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss : 0.37661323316891987 | last training loss : 0.4255420411427816

Last validation accuracy : 0.8944167 | last training accuracy : 0.8787917

Time taken in training: 43.33968257904053 sec

#### SGD with learning rate 0.00165 and relu activation

#### In [0]:

cnn models['sgd 0.00165 relu'] = run experiment(build cnn model, 'relu', tf.keras.optimizers.SGD(0.00165), ' categorical crossentropy', train images, train labels)

/usr/local/lib/python3.6/dist-packages/sklearn/model\_selection/\_split.py:2179: FutureWarning: F rom version 0.21, test size will always complement train size unless both are specified. FutureWarning)

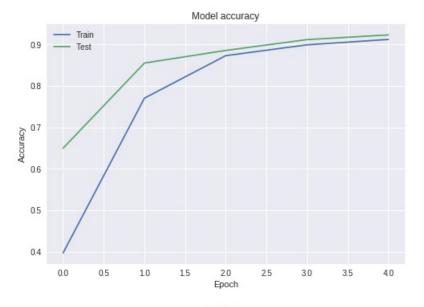
```
Train on 48000 samples, validate on 12000 samples
Epoch 1/5
curacy: 0.3973 - val loss: 1.7895 - val categorical accuracy: 0.6498
Epoch 2/5
48000/48000 [============== ] - 8s 159us/sample - loss: 1.0294 - categorical acc
uracy: 0.7704 - val loss: 0.5636 - val categorical accuracy: 0.8550
Epoch 3/5
48000/48000 [============== ] - 8s 159us/sample - loss: 0.4647 - categorical acc
uracy: 0.8728 - val loss: 0.3860 - val categorical accuracy: 0.8854
Epoch 4/5
48000/48000 [====
                             ======] - 8s 159us/sample - loss: 0.3521 - categorical acc
uracy: 0.8989 - val_loss: 0.3033 - val_categorical_accuracy: 0.9115
Epoch 5/5
```

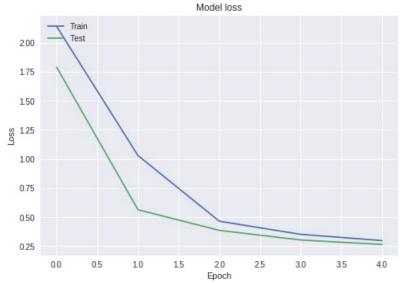
uracy: 0.9119 - val loss: 0.2650 - val categorical accuracy: 0.9228

Layer (type)	Output Shape	Param #
conv2d_58 (Conv2D)	(None, 26, 26, 24)	240
max_pooling2d_58 (MaxPooling	(None, 13, 13, 24)	0
conv2d_59 (Conv2D)	(None, 11, 11, 36)	7812
max_pooling2d_59 (MaxPooling	(None, 5, 5, 36)	0
flatten_69 (Flatten)	(None, 900)	0
dense_167 (Dense)	(None, 784)	706384
dense_168 (Dense)	(None, 128)	100480
dense_169 (Dense)	(None, 10)	1290

Total params: 816,206 Trainable params: 816,206

Non-trainable params: 0





Last validation loss : 0.26501761094729104 | last training loss : 0.29830213260650634

Last validation accuracy: 0.9228333 | last training accuracy: 0.9118958

Time taken in training : 43.8600389957428 sec

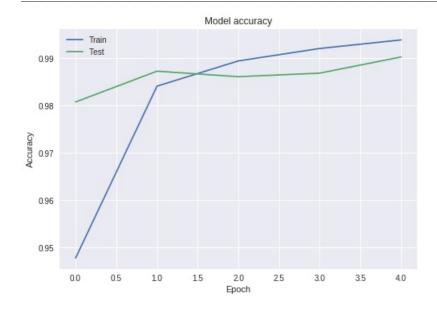
# In [0]:

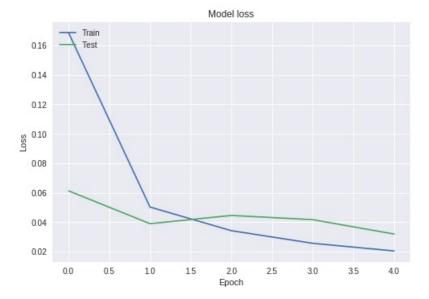
cnn\_models['adam\_0.0009\_relu'] = run\_experiment(build\_cnn\_model, 'relu', tf.keras.optimizers.Adam(0.0009), '
categorical\_crossentropy', train\_images, train\_labels)

Layer (type)	Output Sha	ape	Param #
conv2d_62 (Conv2D)	(None, 26	, 26, 24)	240
max_pooling2d_62 (MaxPooling	(None, 13	, 13, 24)	0
conv2d_63 (Conv2D)	(None, 11	, 11, 36)	7812
max_pooling2d_63 (MaxPooling	(None, 5,	5, 36)	0
flatten_71 (Flatten)	(None, 900	0)	0
dense_173 (Dense)	(None, 784	4)	706384
dense_174 (Dense)	(None, 128	8)	100480
dense_175 (Dense)	(None, 10)	)	1290

uracy: 0.9939 - val loss: 0.0319 - val categorical accuracy: 0.9902

Total params: 816,206 Trainable params: 816,206 Non-trainable params: 0





Last validation loss : 0.03189373222133145 | last training loss : 0.020388831514923367

Last validation accuracy : 0.99025 | last training accuracy : 0.99385417

Time taken in training: 50.58485269546509 sec

# **CNN** model evaluation

In [0]:

```
key = get_best_configuration(cnn_models)
values = cnn_models[key]
print(key, ' has the best validation accuracy ', values[1])
evaluate_model(values[0], test_images, test_labels)
```

Test loss: 0.03254169241283089

Test accuracy: 0.9891

Time taken in training: 0.7434859275817871 sec

# Last task

In [0]:

```
best_model = get_best_configuration(cnn_models)
predictions = cnn_models[best_model][0].predict(test_images)

print(predictions[0])
print('Predicted digit : ', np.where(predictions[0] == np.amax(predictions[0]))[0][0])

#TODO: identify the digit with the highest confidence prediction for the first image in the test dataset
print('Correct digit : ', np.where(1 == test_labels[0])[0][0])
```

```
[7.4170905e-09 6.3201037e-07 3.0762595e-07 3.4817189e-05 4.1947651e-08 2.1944260e-08 3.1757766e-10 9.9995959e-01 2.5718603e-07 4.3169143e-06] Predicted digit : 7 Correct digit : 7
```