

In [9]:

```
# 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_cols, 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)

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'])

history=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])
```

```
x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
```

```

Epoch 1/12
60000/60000 [=====] - 228s 4ms/step - loss: 0.2652 - acc: 0.9174 - val_loss: 0.0601 - val_acc: 0.9810
Epoch 2/12
60000/60000 [=====] - 218s 4ms/step - loss: 0.0887 - acc: 0.9736 - val_loss: 0.0391 - val_acc: 0.9874
Epoch 3/12
60000/60000 [=====] - 242s 4ms/step - loss: 0.0663 - acc: 0.9803 - val_loss: 0.0397 - val_acc: 0.9872
Epoch 4/12
60000/60000 [=====] - 230s 4ms/step - loss: 0.0540 - acc: 0.9838 - val_loss: 0.0344 - val_acc: 0.9883
Epoch 5/12
60000/60000 [=====] - 232s 4ms/step - loss: 0.0474 - acc: 0.9862 - val_loss: 0.0312 - val_acc: 0.9897
Epoch 6/12
60000/60000 [=====] - 269s 4ms/step - loss: 0.0421 - acc: 0.9871 - val_loss: 0.0288 - val_acc: 0.9907
Epoch 7/12
60000/60000 [=====] - 265s 4ms/step - loss: 0.0370 - acc: 0.9890 - val_loss: 0.0294 - val_acc: 0.9916
Epoch 8/12
60000/60000 [=====] - 272s 5ms/step - loss: 0.0341 - acc: 0.9895 - val_loss: 0.0257 - val_acc: 0.9919
Epoch 9/12
60000/60000 [=====] - 268s 4ms/step - loss: 0.0308 - acc: 0.9905 - val_loss: 0.0257 - val_acc: 0.9927
Epoch 10/12
60000/60000 [=====] - 260s 4ms/step - loss: 0.0289 - acc: 0.9906 - val_loss: 0.0290 - val_acc: 0.9912
Epoch 11/12
60000/60000 [=====] - 261s 4ms/step - loss: 0.0284 - acc: 0.9914 - val_loss: 0.0265 - val_acc: 0.9927
Epoch 12/12
60000/60000 [=====] - 270s 5ms/step - loss: 0.0265 - acc: 0.9915 - val_loss: 0.0257 - val_acc: 0.9924
Test loss: 0.025672027044459447
Test accuracy: 0.9924

```

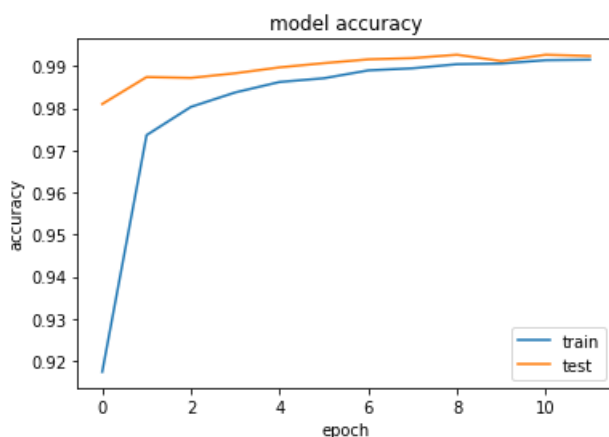
In [10]:

```

# list all data in history
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-not-subscriptable
import matplotlib.pyplot as plt

# summarize history for accuracy
plt.plot(model.history.history['acc'])
plt.plot(model.history.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'])
plt.show()

```



In [11]:

```

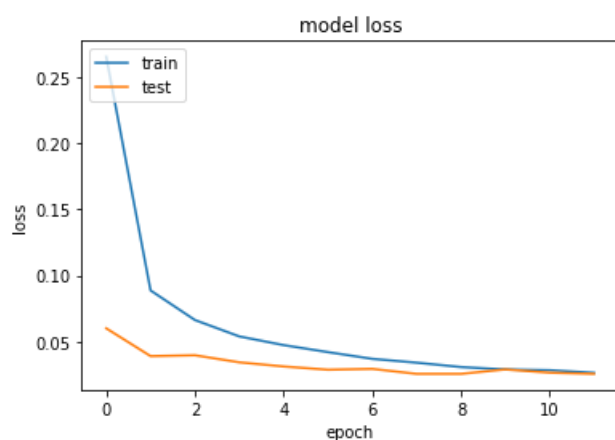
# summarize history for loss
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-not-subscriptable

```

```

https://stackoverflow.com/questions/5175120/python-neural-network-typeerror-history-object-is-no-t-subscriptable
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()

```



2 x 2 kernel with maxpooling,dropout

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(Dropout(0.5))
model.add(Conv2D(128, (3,3), activation='relu'))
model.add(Dropout(0.5))
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'])

history=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 [=====] - 778s 13ms/step - loss: 0.2742 - acc: 0.9153 - val_loss: 0.0731 - val_acc: 0.9820

Epoch 2/12

60000/60000 [=====] - 740s 12ms/step - loss: 0.0944 - acc: 0.9722 - val_loss: 0.0472 - val_acc: 0.9884

Epoch 3/12

60000/60000 [=====] - 698s 12ms/step - loss: 0.0703 - acc: 0.9790 - val_loss: 0.0420 - val_acc: 0.9873

Epoch 4/12

60000/60000 [=====] - 698s 12ms/step - loss: 0.0576 - acc: 0.9827 - val_loss: 0.0318 - val_acc: 0.9919

Epoch 5/12

60000/60000 [=====] - 710s 12ms/step - loss: 0.0526 - acc: 0.9845 - val_loss: 0.0318 - val_acc: 0.9919

```

oss: 0.0333 - val_acc: 0.9917
Epoch 6/12
60000/60000 [=====] - 851s 14ms/step - loss: 0.0447 - acc: 0.9865 - val_l
oss: 0.0344 - val_acc: 0.9906
Epoch 7/12
60000/60000 [=====] - 754s 13ms/step - loss: 0.0428 - acc: 0.9874 - val_l
oss: 0.0304 - val_acc: 0.9916
Epoch 8/12
60000/60000 [=====] - 786s 13ms/step - loss: 0.0398 - acc: 0.9876 - val_l
oss: 0.0301 - val_acc: 0.9900
Epoch 9/12
60000/60000 [=====] - 777s 13ms/step - loss: 0.0375 - acc: 0.9890 - val_l
oss: 0.0268 - val_acc: 0.9910
Epoch 10/12
60000/60000 [=====] - 742s 12ms/step - loss: 0.0343 - acc: 0.9893 - val_l
oss: 0.0264 - val_acc: 0.9932
Epoch 11/12
60000/60000 [=====] - 734s 12ms/step - loss: 0.0335 - acc: 0.9900 - val_l
oss: 0.0253 - val_acc: 0.9916
Epoch 12/12
60000/60000 [=====] - 714s 12ms/step - loss: 0.0321 - acc: 0.9908 - val_l
oss: 0.0238 - val_acc: 0.9921
Test loss: 0.023821924547245726
Test accuracy: 0.9921

```

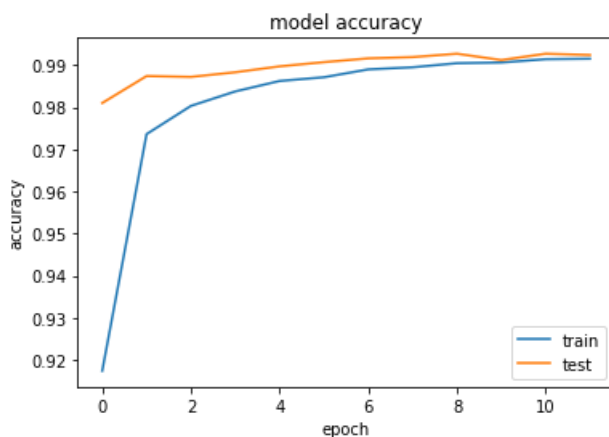
In [13]:

```

# list all data in history
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-no
t-subscriptable
import matplotlib.pyplot as plt
print(history.history.keys())
# summarize history for accuracy
plt.plot(history.history['acc'])
plt.plot(history.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'])
plt.show()

```

```
dict_keys(['val_loss', 'val_acc', 'loss', 'acc'])
```



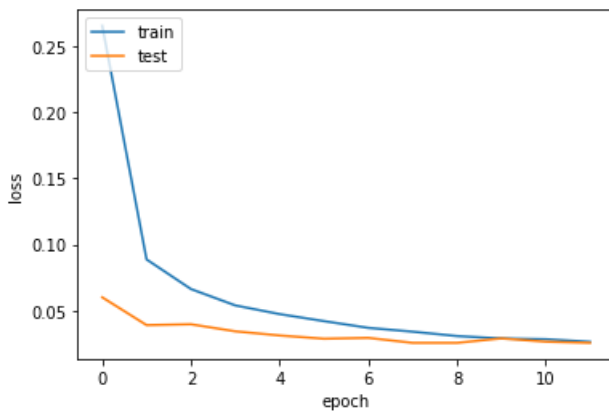
In [14]:

```

# summarize history for loss
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-no
t-subscriptable
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()

```

model loss



With batch normalization,max pooling and dropout

with three layers

In [6]:

```
from keras.layers import BatchNormalization

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_cols, 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)
model = Sequential()
model.add(Conv2D(32, kernel_size=(3, 3),
                 activation='relu',
                 input_shape=input_shape))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.5))

model.add(Conv2D(64, (3,3), activation='relu'))
model.add(BatchNormalization())
```

```

model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(2, 2)))
model.add(Dropout(0.5))

model.add(Conv2D(128, (3,3), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(pool_size=(3, 3)))
model.add(Dropout(0.5))

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'])

history=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])

```

```

x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
Train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [=====] - 404s 7ms/step - loss: 1.4731 - acc: 0.5175 - val_loss: 0.2578 - val_acc: 0.9272
Epoch 2/12
60000/60000 [=====] - 394s 7ms/step - loss: 0.4277 - acc: 0.8670 - val_loss: 0.1228 - val_acc: 0.9628
Epoch 3/12
60000/60000 [=====] - 417s 7ms/step - loss: 0.2774 - acc: 0.9179 - val_loss: 0.0733 - val_acc: 0.9779
Epoch 4/12
60000/60000 [=====] - 401s 7ms/step - loss: 0.2183 - acc: 0.9352 - val_loss: 0.0562 - val_acc: 0.9837
Epoch 5/12
60000/60000 [=====] - 385s 6ms/step - loss: 0.1828 - acc: 0.9456 - val_loss: 0.0481 - val_acc: 0.9861
Epoch 6/12
60000/60000 [=====] - 381s 6ms/step - loss: 0.1597 - acc: 0.9523 - val_loss: 0.0493 - val_acc: 0.9866
Epoch 7/12
60000/60000 [=====] - 375s 6ms/step - loss: 0.1490 - acc: 0.9554 - val_loss: 0.0407 - val_acc: 0.9887
Epoch 8/12
60000/60000 [=====] - 396s 7ms/step - loss: 0.1370 - acc: 0.9606 - val_loss: 0.0412 - val_acc: 0.9886
Epoch 9/12
60000/60000 [=====] - 392s 7ms/step - loss: 0.1276 - acc: 0.9637 - val_loss: 0.0366 - val_acc: 0.9892
Epoch 10/12
60000/60000 [=====] - 408s 7ms/step - loss: 0.1220 - acc: 0.9643 - val_loss: 0.0407 - val_acc: 0.9882
Epoch 11/12
60000/60000 [=====] - 413s 7ms/step - loss: 0.1138 - acc: 0.9678 - val_loss: 0.0389 - val_acc: 0.9890
Epoch 12/12
60000/60000 [=====] - 384s 6ms/step - loss: 0.1111 - acc: 0.9680 - val_loss: 0.0333 - val_acc: 0.9908
Test loss: 0.03330518811953516
Test accuracy: 0.9908

```

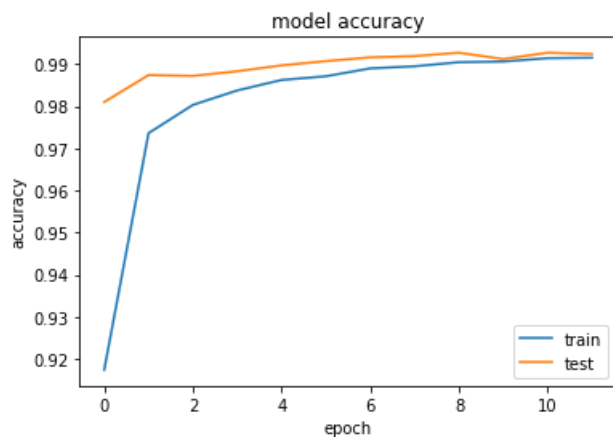
In [15]:

```

# summarize history for accuracy
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-not-subscriptable
plt.plot(history.history['acc'])

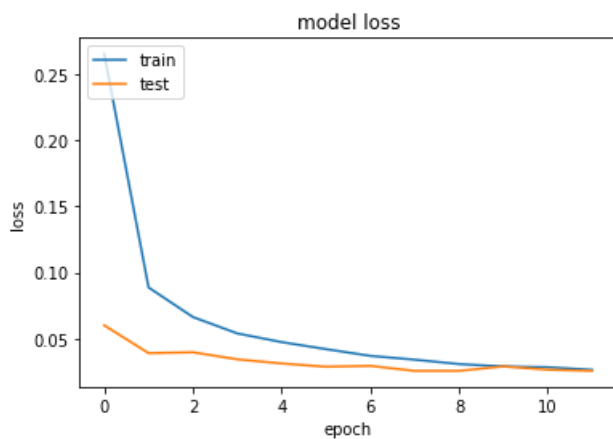
```

```
plt.plot(history.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'])
plt.show()
```



In [16]:

```
# summarize history for loss
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-not-subscriptable
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()
```



With five convolutional layers

In [1]:

```
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_cols, 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)
model = Sequential()
model.add(Conv2D(524, kernel_size=(3, 3),
                activation='relu',
                input_shape=input_shape))
model.add(Conv2D(360, (3,3), activation='relu'))
model.add(Conv2D(128, (3,3), activation='relu'))
model.add(Conv2D(64, (3,3), activation='relu'))
model.add(Conv2D(32, (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'])

history1=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])

```

C:\Users\Shashank\Anaconda3\lib\site-packages\h5py__init__.py:36: FutureWarning: Conversion of the second argument of issubdtype from `float` to `np.floating` is deprecated. In future, it will be treated as `np.float64 == np.dtype(float).type`.

from ._conv import register_converters as _register_converters

Using TensorFlow backend.

```

x_train shape: (60000, 28, 28, 1)
60000 train samples
10000 test samples
WARNING:tensorflow:From C:\Users\Shashank\Anaconda3\lib\site-
packages\tensorflow\python\framework\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\Shashank\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\Shashank\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

```



```

train on 60000 samples, validate on 10000 samples
Epoch 1/12
60000/60000 [=====] - 13754s 229ms/step - loss: 0.3211 - acc: 0.9012 - va
l_loss: 0.0561 - val_acc: 0.9815
Epoch 2/12
60000/60000 [=====] - 11489s 191ms/step - loss: 0.0886 - acc: 0.9738 - va
l_loss: 0.0364 - val_acc: 0.9879
Epoch 3/12
60000/60000 [=====] - 11691s 195ms/step - loss: 0.0642 - acc: 0.9808 - va
l_loss: 0.0424 - val_acc: 0.9869
Epoch 4/12
60000/60000 [=====] - 10287s 171ms/step - loss: 0.0533 - acc: 0.9846 - va
l_loss: 0.0265 - val_acc: 0.9912
Epoch 5/12
60000/60000 [=====] - 34341s 572ms/step - loss: 0.0462 - acc: 0.9861 - va
l_loss: 0.0321 - val_acc: 0.9896
Epoch 6/12
60000/60000 [=====] - 10178s 170ms/step - loss: 0.0410 - acc: 0.9880 - va
l_loss: 0.0226 - val_acc: 0.9935
Epoch 7/12
60000/60000 [=====] - 11371s 190ms/step - loss: 0.0369 - acc: 0.9890 - va
l_loss: 0.0227 - val_acc: 0.9931
Epoch 8/12
60000/60000 [=====] - 12045s 201ms/step - loss: 0.0342 - acc: 0.9896 - va
l_loss: 0.0253 - val_acc: 0.9918
Epoch 9/12
60000/60000 [=====] - 11697s 195ms/step - loss: 0.0316 - acc: 0.9906 - va
l_loss: 0.0190 - val_acc: 0.9933
Epoch 10/12
60000/60000 [=====] - 22333s 372ms/step - loss: 0.0288 - acc: 0.9917 - va
l_loss: 0.0177 - val_acc: 0.9935
Epoch 11/12
60000/60000 [=====] - 14696s 245ms/step - loss: 0.0263 - acc: 0.9919 - va
l_loss: 0.0210 - val_acc: 0.9930
Epoch 12/12
60000/60000 [=====] - 11577s 193ms/step - loss: 0.0248 - acc: 0.9921 - va
l_loss: 0.0206 - val_acc: 0.9937
Test loss: 0.02058376550391504
Test accuracy: 0.9937

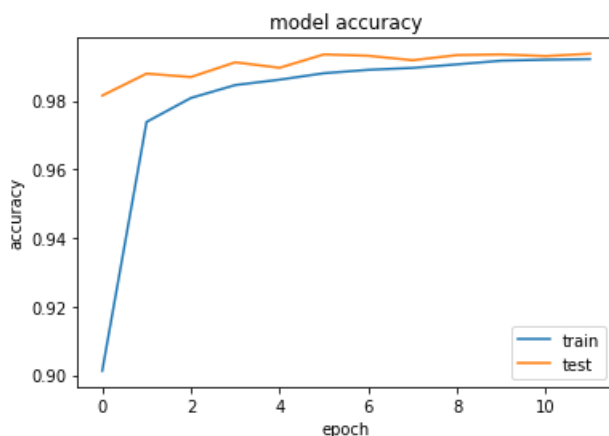
```

In [5]:

```

# summarize history for accuracy
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-no
t-subscriptable
import matplotlib.pyplot as plt
plt.plot(history1.history['acc'])
plt.plot(history1.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'])
plt.show()

```



In [4]:

```

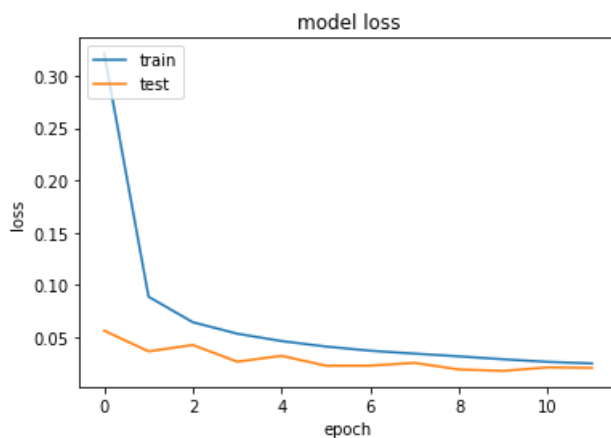
# summarize history for loss
#https://stackoverflow.com/questions/51731207/python-neural-network-typeerror-history-object-is-no
t-subscriptable

```

```

c=Subscriptable
plt.plot(history1.history['loss'])
plt.plot(history1.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()

```



In [6]:

```

# http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

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

x = PrettyTable()
x.field_names = ["Model", "Test Loss", "Test Accuracy"]
x.add_row([ "2 convolutional layers accuracy", 0.0256, 0.992])
x.add_row([ "2 convolutional layers + dropout+ max pooling accuracy", 0.0238, 0.992])
x.add_row([ "2 convolutional layers + dropout+ max pooling+batch normalization accuracy", 0.0333, 0.99])
x.add_row([ "3 convolutional layers accuracy", 0.033, 99.23])
x.add_row([ "5 convolutional layers accuracy+max pooling", 0.0205, 0.993])

print(x)

```

Model	Test Loss	Test Accuracy
2 convolutional layers accuracy	0.0256	0.992
2 convolutional layers + dropout+ max pooling accuracy	0.0238	0.992
2 convolutional layers + dropout+ max pooling+batch normalization accuracy	0.0333	0.99
3 convolutional layers accuracy	0.033	99.23
5 convolutional layers accuracy+max pooling	0.0205	0.993

Conclusion

1.In the first case we have two convolutional layers with dropoutout and max pooling.Here the value of dropout taken is 0.5 .

2.As we can see in the graph that train accuracy increases sharply upto .97 and hereafter train test loss remains constant .the gap between both train and tes loss gets narrowed.

3. From the graph it can be inferred that train loss decreases sharply upto 0.10 and from thereon the gap between train and test loss reduces and at some point it merges.

4. The same case happens with the three convolutional layers with slight changes which can be inferred from the graph

5. Coming to the five layers one can see that accuracy increases a bit and no such significant changes as the dataset for cnn is very small.