

# 10-26-2018 Keras API Functions

1  
`model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 1)))`

2  
`model.add(layers.MaxPooling2D((2, 2)))`

3  
`model.add(layers.Flatten())`

4  
`model.add(layers.Dense(64, activation='relu'))`

5  
`tf.keras.layers.Dropout(0.2)(fc_1)`

6  
`model.summary()`

# 10-26-2018 ConvNet Architecture Design

Deep Learning Oct. 26, 2018. 1/

Today's Topics:

1<sup>st</sup> Hands-On Programming ON Convnet.py.

Ref: [github/fualdi/opencv/IP120-AI-DL](https://github.com/fualdi/opencv/IP120-AI-DL)

Convnet.py. (Chollet's Style)

Compilation & Build this Program. Keras API Architect

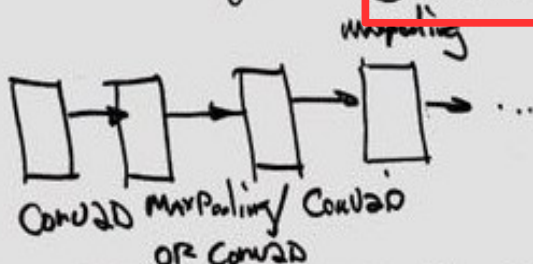
Network Architecture → The Project's Assignment.

Objectives: Design And Prototype A DL Convnet.

Approach: ① Network Diagram (from Examples) CAT I  
Ref: CAT Video & Image. 2018F-3-~ CAT II  
II & I CAT III

Basic Building Blocks

② 2D Conv. Layer	③ Flatten
④ Max Pooling	⑤ Dense (layer)



Observation I: To Design Convnet for Deep Learning  
we start with Convolution Layer followed very often by  
Max Pooling layer, then these layers can be repeated...

Step 1. Repeating  
conv2D( ) layer and  
MaxPooling( ) layer

Example

1  
model.add(layers.Conv2D(32, (3, 3),  
activation='relu', input\_shape=(28, 28, 1)))

2  
model.add(layers.MaxPooling2D((2, 2)))

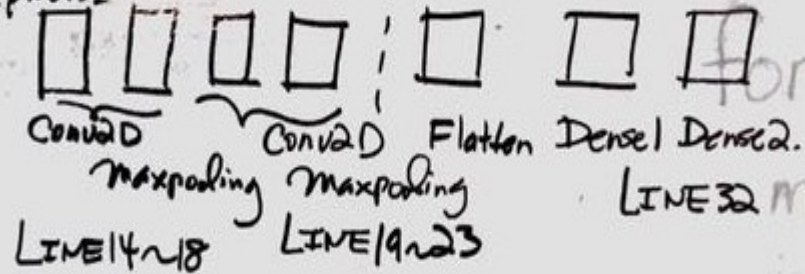
6  
model.summary()

↖  
To display the  
network architecture

# 10-26-2018 3 ConvNet Examples

Example ①: from Smodel.py. Architecture

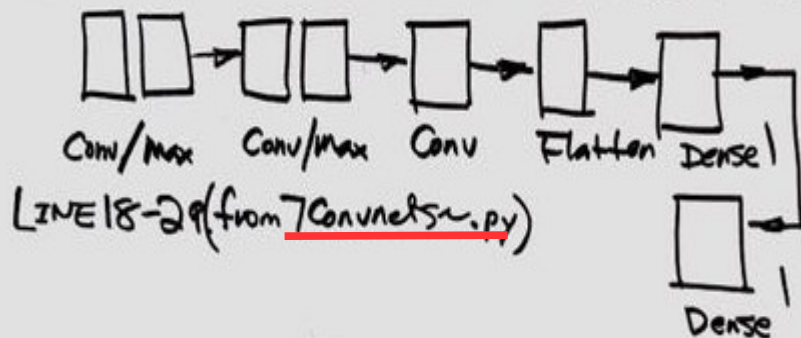
40K photos



Example: ② NIST For Digits (0-9) Recognition. 60K photos

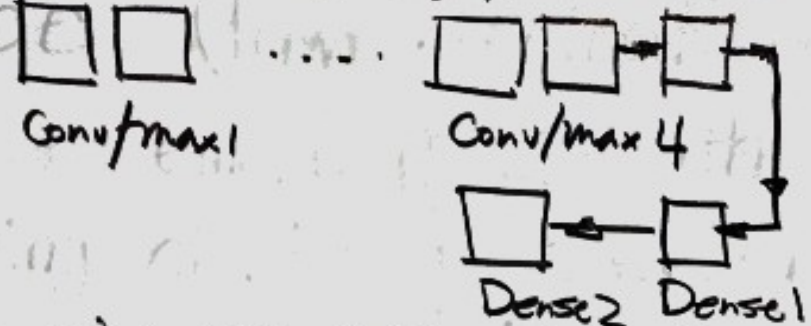
Note: Architecture

Conv/max → Conv/max → Dense → Dense → Flatten



Example ③: For 1000 Cats & Dogs Photos.

Ref: 8convnets.py. Flatten



Note: The tool to verify (Print) model is modelSummary()

Jerry On Python File Manipulations.  
Sample Code: 8convnet ~.py file manipulation.

Observation II: Generally, Deep Learning Architecture can be classified as convolutional layer(s) and NeuralNet layers.

Example: from the Summary of Example 1 to Example 3.

# 10-26-2018 Comparison of 3 ConvNets

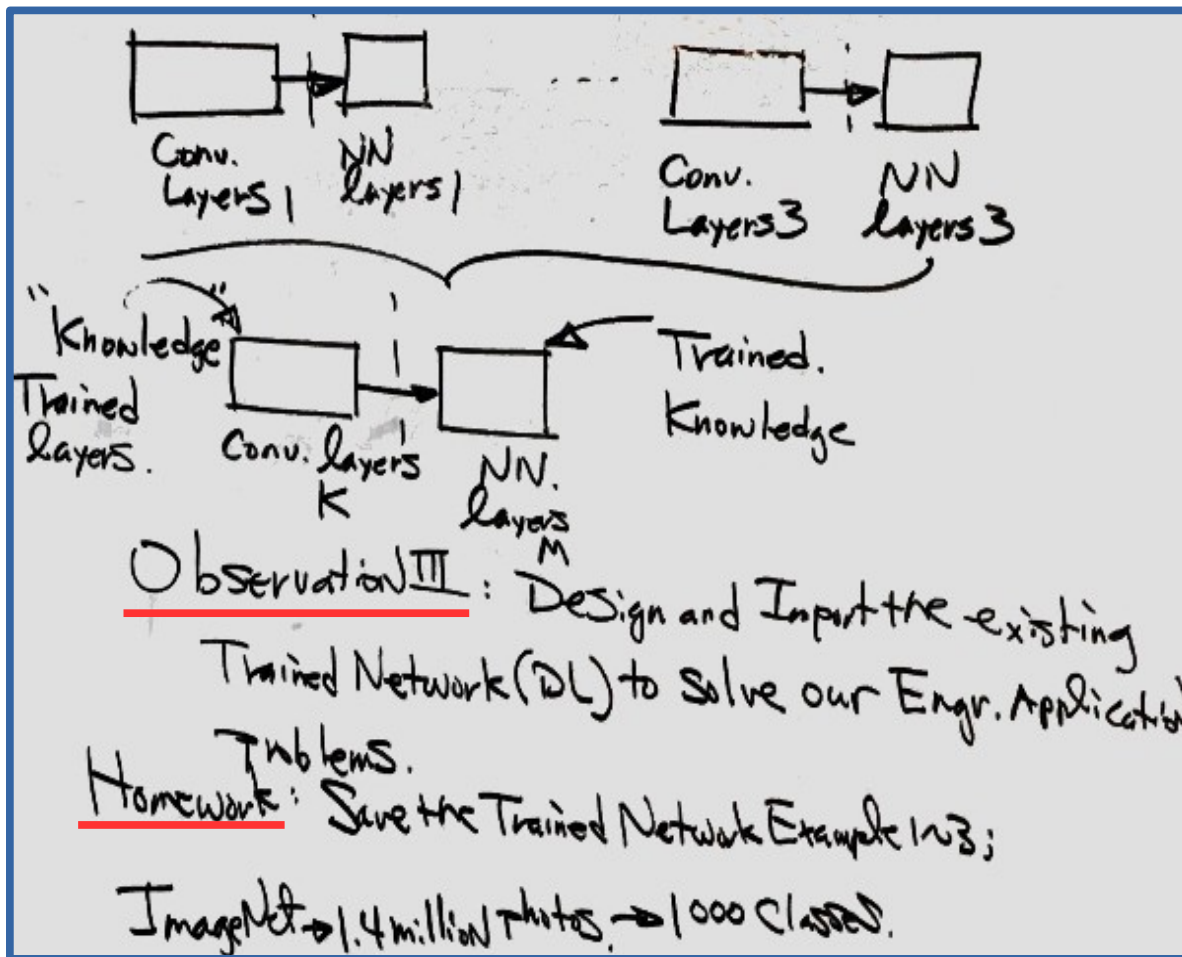
Table 1. Three well trained convnet examples

1 convnet1	Satellite imagery	40K images	ConvPool1+ConvPool2+Flatten+Den1+Den2
2 convnet2	NIST digits 0-9	60K images	ConvPool1+ConvPool2+Conv+Flatten+Den1+Den2
3 convnet3	Chollet cat-dog	1K images	ConvPool1+ConvPool2+ConvPool3 +ConvPool4 +Flatten+Den1+Den2

Table 2. Sample code for 3 well trained convnets

Network	Programs
convnet1	5model.py
convnet2	7convnets-NumeralDetection-ch05.py
convnet3	8convnets-SmallData-cats-dogs-ch05.py

# 10-26-2018 3 ConvNet Examples



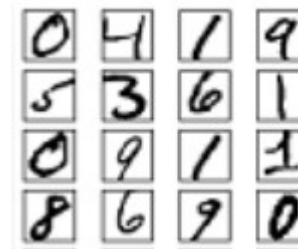


# 10-26-2018 Numerals Output Design

Example: NIST 10 digits (0-9) convnet Dense layer

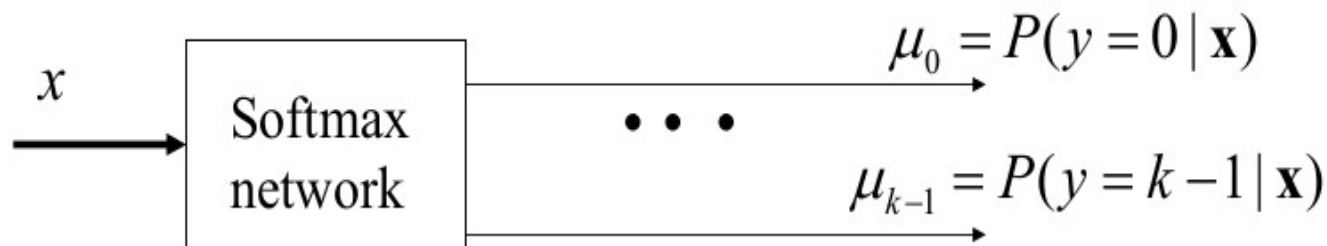
```
from keras import models
from keras import layers
```

```
network = models.Sequential()
network.add(layers.Dense(512, activation='relu', input_shape=(28 * 28,)))
network.add(layers.Dense(10, activation='softmax'))
```



Sample code :  
<https://github.com/fchollet/deep-learning-with-python-notebooks>

The network consists of 2 Dense layers, densely-connected ("fully-connected") neural layers. The output layer is a 10-way "softmax" layer, it returns an array of 10 probability scores (summing to 1). Each score will be the probability that the current digit image belongs to one of our 10 digit classes.



*Illustration of the softmax block diagram from Milos Hauskrecht,  
milos@cs.pitt.edu, 5329 Sennott Square*

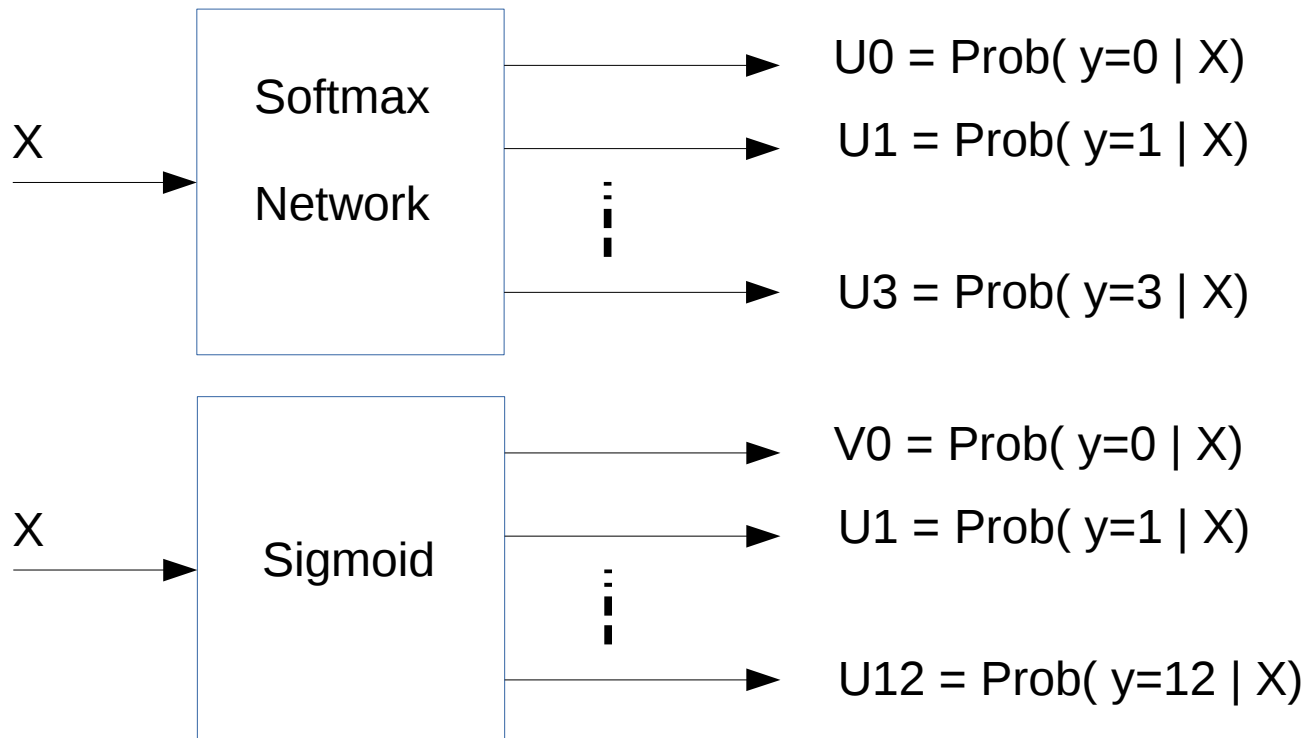
# 10-26-2018 Satellite Output Design

Example: Satellite Imagery Classification by convnet Dense layer

```
#separate outputs for the weather and the ground labels
weather_output = tf.keras.layers.Dense(4,
                                         activation='softmax',
                                         name='weather')(fc_2)
ground_output = tf.keras.layers.Dense(13,
                                       activation='sigmoid',
                                       name='ground')(fc_2)
```

Sample program: 5model.py

```
4
model.add(layers.Dense(64, activation='relu'))
```

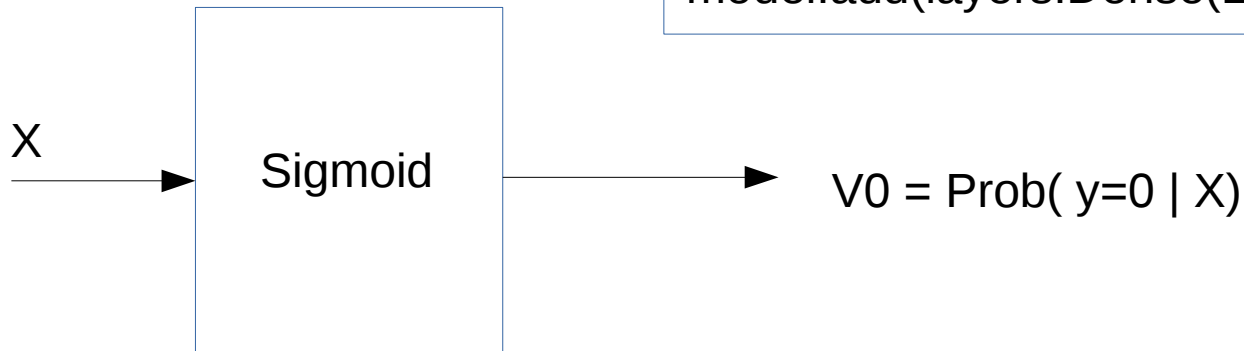


# 10-26-2018 Cats-Dogs Output Design

Example: Cats-dogs Imagery Classification by convnet Dense layer

Sample program: 8convnets-  
SmallData-cats-dogs-ch05

```
model.add(layers.Flatten())  
model.add(layers.Dense(512, activation='relu'))  
model.add(layers.Dense(1, activation='sigmoid'))
```





# 10-26-2018 Kaggle Data and NIST Data

Example: Satellite Imagery Classification by convnet Dense layer

```
class KagglePlanetSequence(tf.keras.utils.Sequence):
```

Sample program: 5model.py

```
# original_dataset_dir = '/Users/fchollet/Downloads/kaggle_original_data'  
original_dataset_dir = 'kaggle-cats-dogs'
```

```
# The directory to store smaller dataset  
base_dir = 'kaggle-cats-dogs-small'  
os.mkdir(base_dir)
```

Sample program: 8convnets-  
SmallData-cats-dogs-ch05

# 10-26-2018 Satellite Input Data Design

Example: Satellite Imagery Classification by convnet Dense layer

```
import tensorflow as tf  
IM_SIZE = 128
```

Sample program: 5model.py

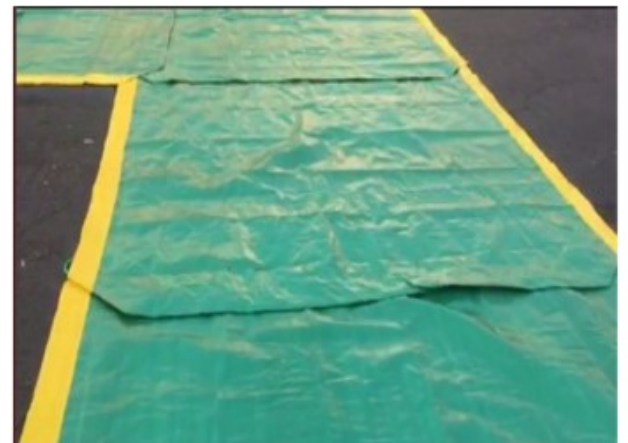
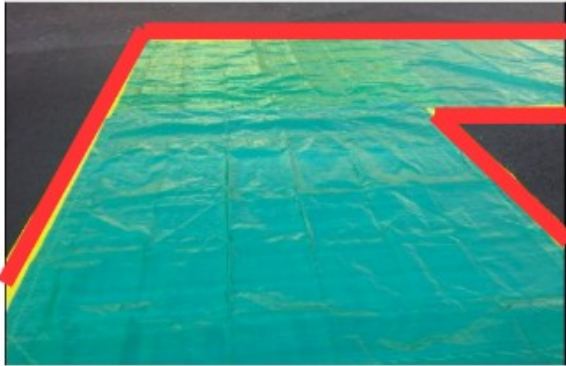
```
image_input = tf.keras.Input(shape=(IM_SIZE, IM_SIZE, 3), name='input_layer')
```

```
# Some convolutional layers
```

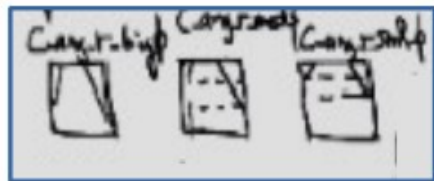
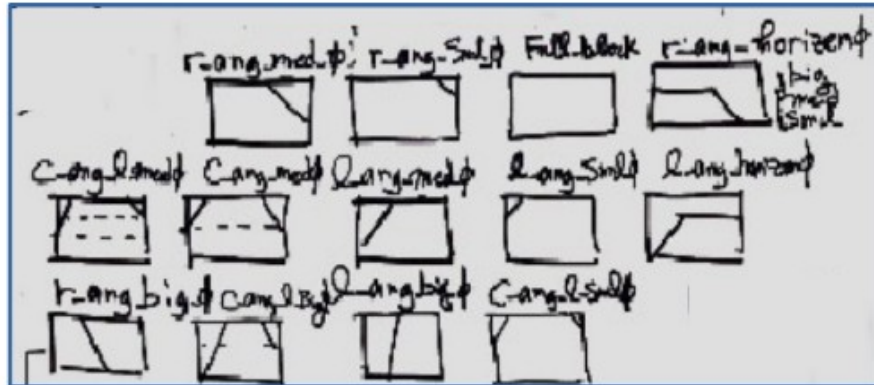
```
conv_1 = tf.keras.layers.Conv2D(32,  
                                kernel_size=(3, 3),  
                                padding='same',  
                                activation='relu')(image_input)
```

```
conv_1 = tf.keras.layers.MaxPooling2D(padding='same')(conv_1)
```

# CAT-II Path Primitives



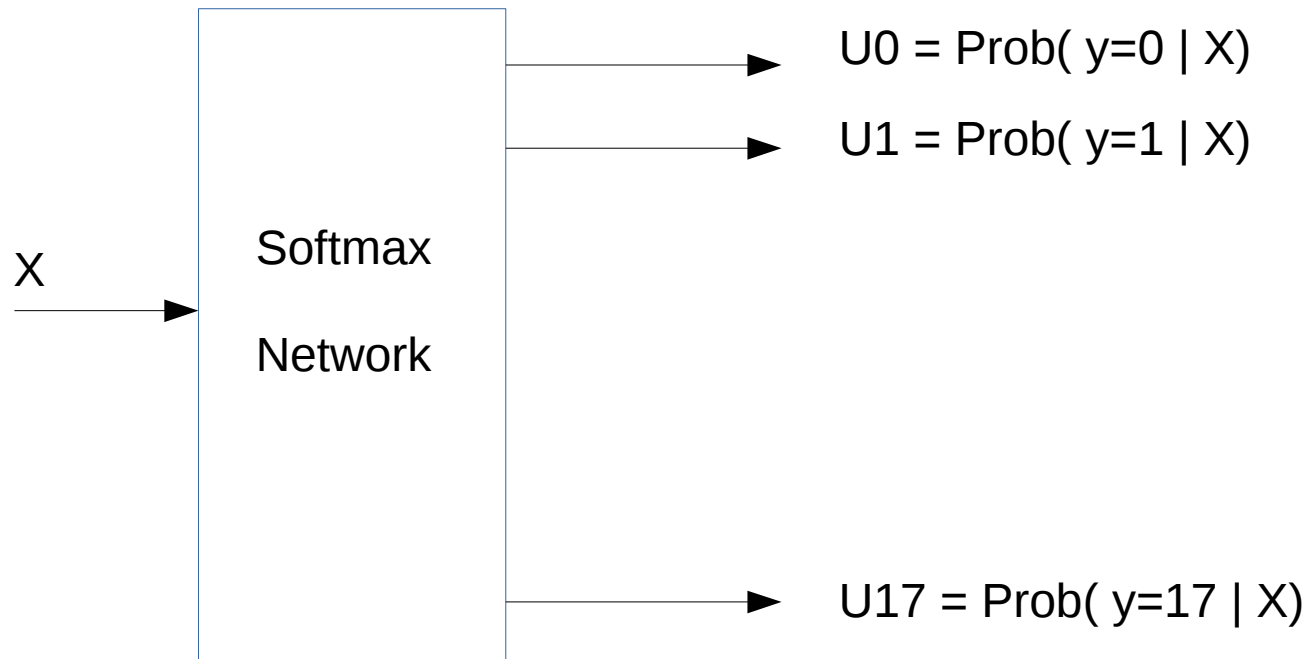
# 9-25-2018 Primitive Features for CAT-II Path Classification



1. r_ang_big	right angle big
2. r_ang_med	right angle medium
3. r_ang_sml	right angle small
4. r_ang_hor	right angle horizon
5. l_ang_big	left angle big
6. l_ang_med	left angle medium
7. l_ang_sml	left angle small
8. l_ang_hor	left angle horizon
9. cr_ang_big	central-right big
10. cr_ang_med	central angle medium
11. cr_ang_sml	central angle small
12. cl_ang_big	central-right angle
13. cl_ang_med	central angle medium
14. cl_ang_sml	central angle small
15. c_ang_big	central-angle big
16. c_ang_med	central-ang medium
17. c_ang_sml	central-ang small
18. full	full block

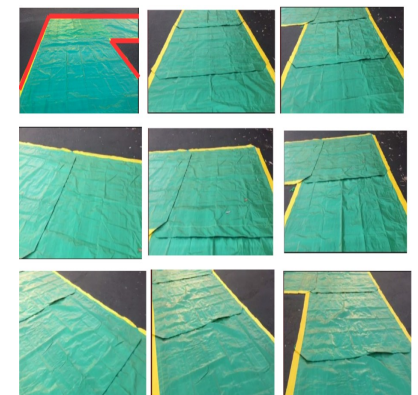
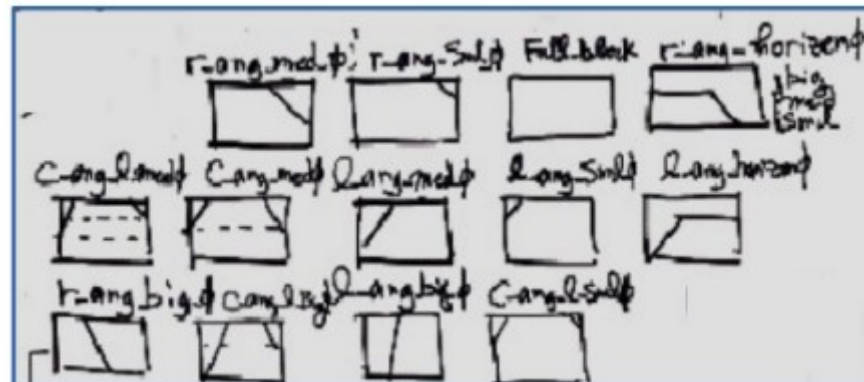
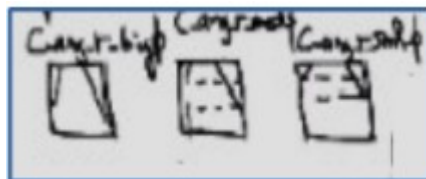


# 10-26-2018 Softmax Output Design for CATIII



1. r_ang_big	right angle big
2. r_ang_med	right angle medium
3. r_ang_sml	right angle small
4. r_ang_hor	right angle horizon
5. l_ang_big	left angle big
6. l_ang_med	left angle medium
7. l_ang_sml	left angle small
8. l_ang_hor	left angle horizon
9. cr_ang_big	central-right big
10. cr_ang_med	central angle medium
11. cr_ang_sml	central angle small
12. cl_ang_big	central-left big
13. cl_ang_med	central angle medium
14. cl_ang_sml	central angle small
15. c_ang_big	central-angle big
16. c_ang_med	central-angle medium
17. c_ang_sml	central-angle small
18. full	full block

Where  $\text{Prob}(y=0 | X) + \text{Prob}(y=1 | X) + \dots + \text{Prob}(y=17 | X) = 1$



# 10-29-2018 Keras API Functions

1

```
import input_data
```

2

```
Model.save('harry_convnet_mnist.h5')
```

3

```
from keras.models import load_model
```

4

```
model = load_model('my_model.h5')
```

5

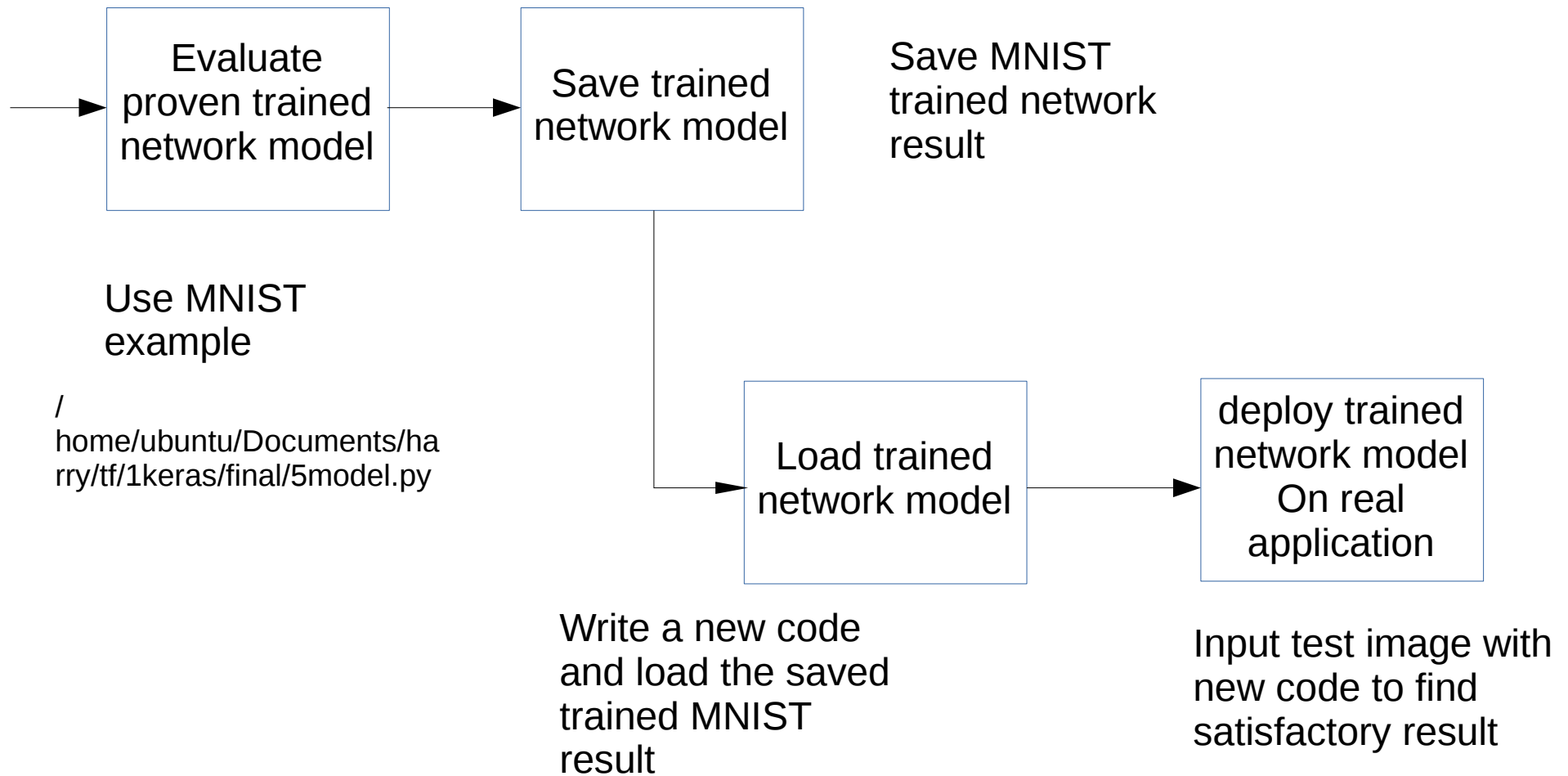
```
del model
```

6

```
import h5py  
h5py.run_tests()
```



# 10-29-2018 Road Map to Deploy CNN



# 10-29-2018 Save Keras Model

[https://www.google.com/search?](https://www.google.com/search?hl=en&ei=kFDZW__GM8SS0wLLhJSICg&q=save+model+json+keras&oq=save+Model+%28JSON%29&gs_l=psy-ab.1.0.0i22i30l7.4409.4409..7830...0.0..0.66.66.1.....0....1j2..gws-wiz.....0i71.fyoCDOFN8UU)

[hl=en&ei=kFDZW\\_\\_GM8SS0wLLhJSICg&q=save+model+json+keras&oq=save+Model+%28JSON%29&gs\\_l=psy-ab.1.0.0i22i30l7.4409.4409..7830...0.0..0.66.66.1.....0....1j2..gws-wiz.....0i71.fyoCDOFN8UU](https://www.google.com/search?hl=en&ei=kFDZW__GM8SS0wLLhJSICg&q=save+model+json+keras&oq=save+Model+%28JSON%29&gs_l=psy-ab.1.0.0i22i30l7.4409.4409..7830...0.0..0.66.66.1.....0....1j2..gws-wiz.....0i71.fyoCDOFN8UU)

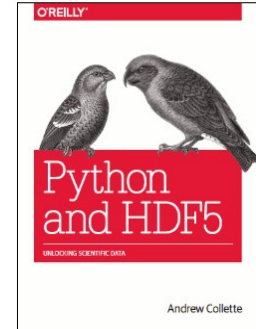
You can use `model.save(filepath)` to save a Keras model into a single HDF5 file which will contain: the architecture of the model, allowing to re-create the model. the weights of the model. the training configuration (loss, optimizer) the state of the optimizer, allowing to resume training exactly where you left off.

# 10-30-2018 Prepare h5py to Save Trained Network

<https://www.quora.com/How-do-I-save-a-convolution-neural-network-model-after-training-I-am-working-in-Python>

Keras for saving a model is just one line of code after training

Just make sure to have HDF5 for Python, use h5py package, which is a Pythonic interface to the HDF5 binary data format, which stores huge amounts of numerical data, and easily to be manipulated with NumPy. The files created can be exchanged including programs like IDL and MATLAB.



<http://www.h5py.org/>

First, install h5py, see the url link to the right and just follow the installation instructions, after the installation is done, be sure to use the following command to check:

```
import h5py
h5py.run_tests()
```

```
ubuntu@ubuntu-ThinkPad-Yoga-14:~/Documents$ python
Python 3.4.3 (default, Nov 28 2017, 16:41:13)
[GCC 4.8.4] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> import h5py
>>> h5py.run_tests()
.....X.....
.....X.....S...S...SS
.....SSSSSS
.....X...X.....X...X.....
.....
-----
Ran 457 tests in 0.632s

OK (skipped=14, expected failures=6)
<unittest.runner.TextTestResult run=457 errors=0 failures=0>
>>>
```

# 10-30-2018 Save Trained Network with h5py

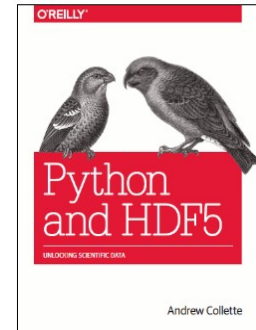
<https://www.quora.com/How-do-I-save-a-convolution-neural-network-model-after-training-I-am-working-in-Python>

Keras for saving a model is just one line of code after training

```
Model.save('harry_convnet_mnist.h5')
```

Example:

```
scores = model.evaluate(X_train,y_train,verbose=0)
print("Accuracy on train set: %.2f%%" % (scores[1]*100))
scores = model.evaluate(X_validation,y_validation,verbose=0)
print("Accuracy on validation set: %.2f%%" % (scores[1]*100))
scores = model.evaluate(X_test,y_test,verbose=0)
print("Accuracy on test set: %.2f%%" % (scores[1]*100))
model.save('ajits_model_d6_256.h5')
```



<http://www.h5py.org/>

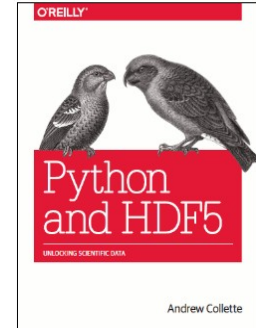
# 10-30-2018 Load Trained Network

<https://stackoverflow.com/questions/35074549/how-to-load-a-model-from-an-hdf5-file-in-keras>

Keras for loading a model is just one line of code

If you stored the complete model, not only the weights, in the HDF5 file, then load is as simple as

```
from keras.models import load_model  
model = load_model('model.h5')
```



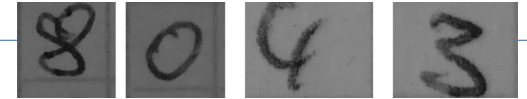
<http://www.h5py.org/>

No

```
from keras.models import load_model  
model.save('my_model.h5')  
# creates a HDF5 file 'my_model.h5'  
  
del model # deletes the existing model  
  
# returns a compiled model identical to the previous one  
model = load_model('my_model.h5')
```

# 10-29-2018 Deploy MNIST with Test Image (1)

<https://medium.com/@o.kroeger/tensorflow-mnist-and-your-own-handwritten-digits-4d1cd32bbab4>



8.png 0.png 4.png 3.png

```
# create an array where to store 4 pics with one numeral each
# each image consists of total 784 pixels
images = np.zeros((4,784))
# and the correct values
correct_vals = np.zeros((4,10))

# test images 8, 0, 4, 3
i = 0
for no in [8,0,4,3]:
    gray = cv2.imread("img/blog/own_"+str(no)+".png", cv2.CV_LOAD_IMAGE_GRAYSCALE)

    # resize the images and invert it (black background)
    gray = cv2.resize(255-gray, (28, 28))

    # save the processed images
    cv2.imwrite("pro-img/image_"+str(no)+".png", gray)
    """"

    all images in the training set ranges from 0-1
    not from 0-255 so divide the flatten images
    (a one dimensional vector with 784 pixels)
    """"

    flatten = gray.flatten() / 255.0
```



# 10-29-2018 Deploy MNIST with Test Image (2)

```
"""----- store flatten image and generate-----
the correct_vals array
correct_val for the first digit (9) would be
[0,0,0,0,0,0,0,0,0,1]
-----"""

images[i] = flatten
correct_val = np.zeros((10))
correct_val[no] = 1
correct_vals[i] = correct_val
i += 1

"""-----the prediction will be an array with four values-----
which show the predicted number
-----"""

prediction = tf.argmax(y,1)

"""-----run the prediction and the accuracy function-----
using our generated arrays (images and correct_vals)
-----"""

print sess.run(prediction, feed_dict={x: images, y_: correct_vals})
print sess.run(accuracy, feed_dict={x: images, y_: correct_vals})
```