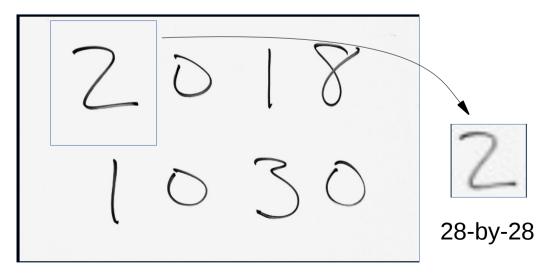
#### Prepare Image For Testing Trained MNIST convnet

Find out the input image size to MNIST convnet, from 2 sources: (1) from 7convnets-NumeralDetection-ch05.py (code below) (2) from the code, the dimension of the conv2d\_1, kernel size is 3x3, from model.summary, the output shape is 26x26.

# model = models.Sequential() model.add(layers.Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1)))

So, (1) use screen capture program to locate ROI as shown below, and then (2) we wrote opency program to take any size input image and convert it to 28-by-28.



>>> model.summary()			
Layer (type)	Output	Shape	Param #
conv2d_1 (Conv2D)	(None,	26, 26, 32)	320
max_pooling2d_1 (MaxPooling2	(None,	13, 13, 32)	0
conv2d_2 (Conv2D)	(None,	11, 11, 64)	18496
max_pooling2d_2 (MaxPooling2	(None,	5, 5, 64)	0
conv2d_3 (Conv2D)	(None,	3, 3, 64)	36928
flatten_1 (Flatten)	(None,	576)	0
dense_1 (Dense)	(None,	64)	36928
dense_2 (Dense)	(None,	10)	650
Total params: 93,322 Trainable params: 93,322 Non-trainable params: 0			

Harry Li, Ph.D. 2018

# 10-26-2018 Keras API Functions

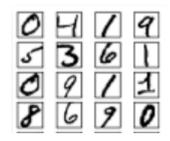
```
model.add(layers.Conv2D(32, (3, 3), activation='relu', input shape=(28, 28, 1)))
2
model.add(layers.MaxPooling2D((2, 2)))
3
model.add(layers.Flatten())
4
model.add(layers.Dense(64, activation='relu'))
 5
 tf.keras.layers.Dropout(0.2)(fc_1)
 6
 model.summary()
```

# 10-26-2018 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.

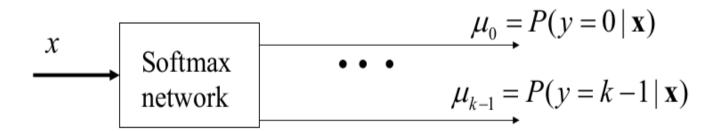


Illustration of the softmax block diagram from Milos Hauskrecht, milos@cs.pitt.edu,5329 Sennott Square

## Convnet MNIST Architecture

Using TensorFlow backend.

Layer (type)	Output Shape Pa	ıram #		
conv2d_1 (Conv2D)	(None, 26, 26, 32)	320		
max_pooling2d_1 (Ma	exPooling2 (None, 13, 13,	32) 0		
conv2d_2 (Conv2D)	(None, 11, 11, 64)	18496		
max_pooling2d_2 (MaxPooling2 (None, 5, 5, 64) 0				
conv2d_3 (Conv2D)	(None, 3, 3, 64)	36928	=======================================	===

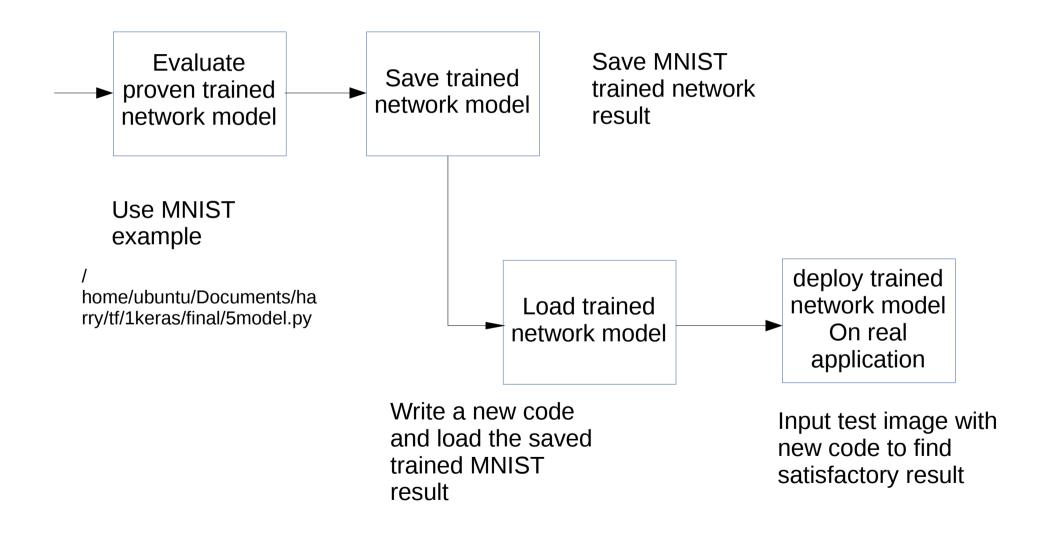
Total params: 55,744

Trainable params: 55,744 Non-trainable params: 0

# Oct-Nov-2018 Keras API Functions

```
import input_data
                                             result_arr = model.predict(test_image)
model.save('harryTest.h5')
3
from keras.models import load_model
4
model = load_model('harryTest.h5')
5
del model
import h5py
h5py.run_tests()
```

# 10-29-2018 Road Map to Deploy CNN



### 10-29-2018 Save Keras Model

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

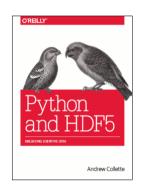
You can use model.save(filepath) to save a Keras model into a single HDF5 file which will contain: the architecture of the model, allowing to re-create the model, the weights of the model, the training configuration (loss, optimizer) the state of the optimizer, allowing to resume training exactly where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

# 10-30-2018 Save Trained Network with h5py

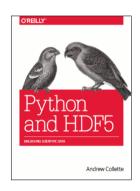
https://www.quora.com/How-do-I-save-a-convolution-neural-network-model-after-training-I-am-working-in-Python

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/

Example: 7-1convnets-NumeralDet-saveTrained.py

```
#-----*
import h5py
model.save('harryTest.h5')
#-end

Just add 2
lines of code
```

### 10-30-2018 Load Trained Network

https://stackoverflow.com/guestions/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

#### Example:

#-----load trained model----from keras.models import load\_model model = load\_model('harryTest.h5') model.summary() #check the model

>>> model.summary()			
Layer (type)	0utput	Shape	Param #
conv2d_1 (Conv2D)	(None,	26, 26, 32)	320
max_pooling2d_1 (MaxPooling2	(None,	13, 13, 32)	0
conv2d_2 (Conv2D)	(None,	11, 11, 64)	18496
max_pooling2d_2 (MaxPooling2	(None,	5, 5, 64)	0
conv2d_3 (Conv2D)	(None,	3, 3, 64)	36928
flatten_1 (Flatten)	(None,	576)	0
dense_1 (Dense)	(None,	64)	36928
dense_2 (Dense)	(None,	10)	650
Total params: 93,322 Trainable params: 93,322			

Non-trainable params: 0

#### OpenCV Program To Resize 28x28 Images

7-3ResizeImage.py

```
import cv2
import numpy as np

gray = cv2.imread('test.png',cv2.IMREAD_GRAYSCALE)
gray = cv2.resize(gray, (28,28))
cv2.imshow('image',gray)
cv2.imwrite('resized_harryTest.jpg',gray, [int(cv2.IMWRITE_JPEG_QUALITY),90])

k = cv2.waitKey(0)
if k == 27:  # wait for ESC key to exit
    cv2.destroyAllWindows()
```

# 11-2-2018 Deploy MNIST with Test Image (1)

#### Example:

```
model = load_model('Numeral_detector.h5')
test_image= cv2.imread('resized_LCTest.jpg', cv2.IMREAD_GRAYSCALE)
test_image = np.reshape(test_image, [1, 28, 28, 1])
result_arr = model.predict(test_image)
result = model.predict_classes(test_image)
print(result_arr,' ', result)
```

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

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

```
# create an array where to store 4 pics with one numeral each
# each image consists of total 784 pixels
                                                                    8.png 0.png 4.png 3.png
images = np.zeros((4,784))
# and the correct values
                                           Image size: 28x28 = 784
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})
```

# 11-2-2018 Architecture Aspects on Deployment

Deep Learning its. 2nd, 2018. 11. Harry LI. Today's Topics: 1. Architectural Design. 20 Homework (Build Convnet, Prototype + Ne Convnet - tinish the Design, Deployment) Ref: 1) git hub/Rudili, 7 convnets. - & MNIST A Deploy 7-2 Load Trained

Simpley 7-3 - Open CU

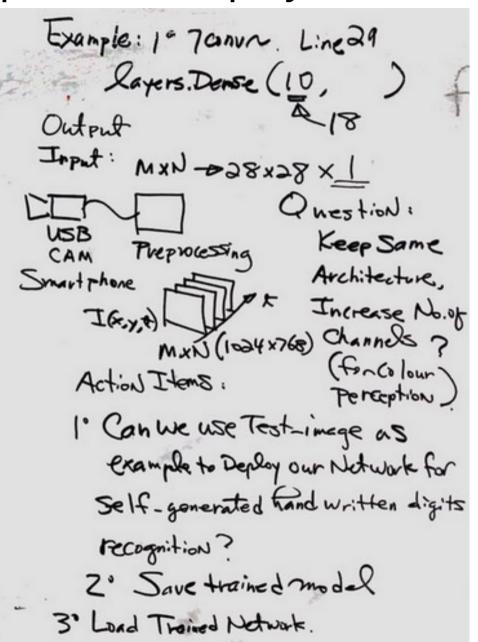
Conuncts. 7-4 ~ Deployment

Design - Trained - Prepar Systems. 
Model S. Latter S. (Sm bodded Systems) (C1+C2+101+D2 ...() \* C1+ (2+C2)+D1+D2...(a) Kagely Deployment of C1+ ... + C4+D2+D2...(3) ImageNet

[AT |Oby(3) | Note: Use Architecture(8)

[apt Output layer | 10 Wegaries of output layer;

[apt Output layer | 10 Wegaries of output layer; as a starting pt. due to its



# 11-2-2018 Architecture Aspects on Deployment

```
Homework:

I USE Architecture of the Choice,
and the Self written digits to Test

The Pretrained Network.

Z. Digitize 18 Categories of Images

from our Sample video (Show of tell
inclass Nextweek).
```

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

2
import matplotlib.pyplot as plt

plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()
plt.figure()
plt.show()
```

# 11-2-2018 Architecture Aspects on Deployment

```
from keras import models
from keras import layers

2
model = models.Sequential()
model.add(conv_base)
model.add(layers.Flatten())
model.add(layers.Dense(256, activation='relu'))
```

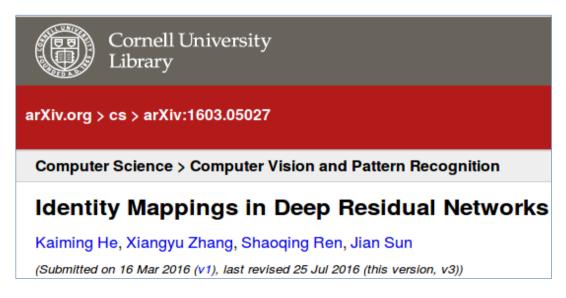
#### 11-6-2018 Keras Supported CNN Architectures

https://www.pyimagesearch.com/2017/03/20/imagenet-vggnet-resnet-inception-xception-keras/

Xception
InceptionV3
ResNet50
VGG16
VGG19
MobileNet

ResNet architecture which can be successfully trained at depths of 50-200 for ImageNet and over 1,000 for CIFAR-10

Two major drawbacks with VGGNet: (1) painfully slow to train. (2) Network architecture weights themselves are quite large (in terms of disk/bandwidth). Due to its depth and number of fully-connected nodes, VGG16 is over 533MB and VGG19 is 574MB. This makes deploying VGG a tiresome task.



https://github.com/KaimingHe/resnet-1k-layers

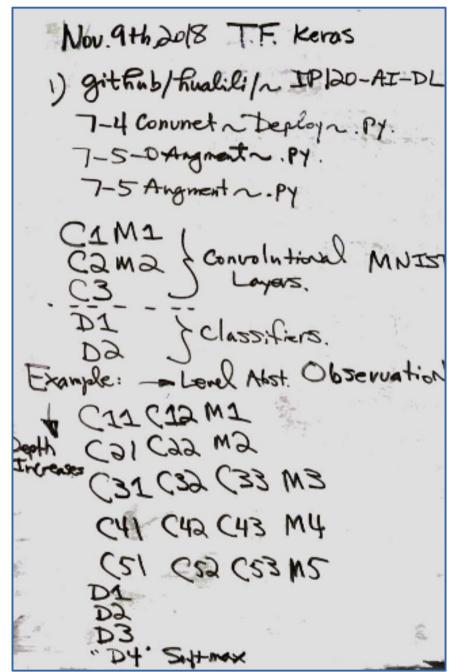
### 11-6-2018 Use Keras Supported CNN

https://www.pyimagesearch.com/2017/03/20/imagenet-vggnet-resnet-inception-xception-keras/

# import the necessary packages
from keras.applications import ResNet50
from keras.applications import InceptionV3
from keras.applications import VGG16
from keras.applications import VGG19
from keras.applications import imagenet\_utils
from keras.applications.inception\_v3 import preprocess\_input
from keras.preprocessing.image import img\_to\_array
from keras.preprocessing.image import load\_img
import numpy as np
import argparse
import cv2

Note: Weights for VGG16 and VGG19 are > 500MB. ResNet weights are ~100MB, while Inception and Xception weights are between 90-100MB. If this is the first time you are running this script for a given network, these weights will be (automatically) downloaded and cached to your local disk. Depending on your internet speed, this may take awhile. However, once the weights are downloaded, they will not need to be downloaded again, allowing subsequent runs of classify\_image.py to be much faster.

### 11-6-2018 Keras Supported CNN Architectures



Example: Augmentation st Images. Observation: Generate Augmented Image via (1) Gaussian Blur (K-4-K, 8) (2) Randomized Rotations. Honework (Test Digits from pag; (3) CAT-II. Video W/Ref to the P.P.T (18 CATS) -> TO GENERATE TEST Imago. Example: 8 convnet ~ py Small Artiset Applications OSave trained model; (2) (reste A New Tragram to load (3) Deploy the model Pre-trained model; 4) CATS Dogs Other Aninols. 2018-11-9 ~.45

### 11-6-2018 Deploy Cats Dogs Detection CNN

```
import time
import keras
keras. version
import numpy as np
from keras import models
import cv2
from keras.models import load model
model = load model('cats and dogs small 1.h5')
#model.summary() #check the model
image =cv2.imread('1.jpg',cv2.IMWRITE_JPEG_QUALITY)
image = cv2.resize(image,(150,150))
image=np.reshape(image,[1,150,150,3])
                                          One image with
                                          resolution 150x150
time curr=int(round(time.time()*1000))
                                          with 3 channels (r, g,
output=model.predict classes(image)
                                          b) colors
time end=int(round(time.time()*1000))
time=time end-time curr
print("")
print(output)
print("")
print(time)
print("")
```

# 11-9-2018 Python Imports for Augmentation To Create Train Data Set (1)

https://leemendelowitz.github.io/blog/how-does-python-find-packages.html

Python finds packages: Python imports by searching the directories listed in sys.path

```
>>> import sys
```

>>> print '\n'.join(sys.path)

/usr/lib/python2.7

/usr/lib/python2.7/plat-x86\_64-linux-gnu

/usr/lib/python2.7/lib-tk

/usr/lib/python2.7/lib-old

/usr/lib/python2.7/lib-dynload

/usr/local/lib/python2.7/dist-packages

/usr/lib/python2.7/dist-packages

/usr/lib/python2.7/dist-packages/PILcompat

/usr/lib/python2.7/dist-packages/gtk-2.0

/usr/lib/python2.7/dist-packages/ubuntu-sso-client

>>>

sys.path.append() adds the path for the code to access.

#### 11-9-2018 Python Common Path Name Manipulations

https://docs.python.org/3/library/os.path.html

posixpath for UNIX-style paths ntpath for Windows paths macpath for old-style MacOS paths os2emxpath for OS/2 EMX paths Example: check if in the right directory

```
>>> print os.path.isdir(data_dir)
True
>>>
```

Example: to join the directory path

```
>>> path = os.path.join('CTIOne_harry_Training_Data', 'level1','level2')
>>> print path
CTIOne_harry_Training_Data/level1/level2
>>>
```

Example: List what is in the directory. (Just like Is command)

```
>>> print os.listdir(data_dir)
['test.txt']
```