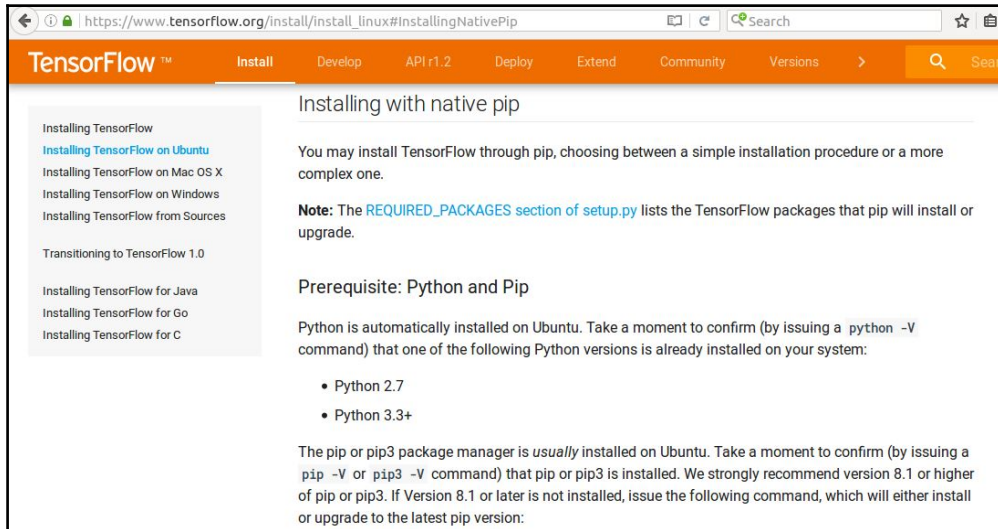


Chapter 1: Getting Started



The screenshot shows the TensorFlow website's 'Installing with native pip' page. The left sidebar contains links for installing TensorFlow on various operating systems (Ubuntu, Mac OS X, Windows, Sources) and for Java, Go, and C. The main content area is titled 'Installing with native pip' and explains that TensorFlow can be installed via pip. It includes a note about the `REQUIRED_PACKAGES` section in `setup.py` and a prerequisite section for Python and Pip, listing Python 2.7 and Python 3.3+.

TensorFlow™ **Install** Develop API r1.2 Deploy Extend Community Versions > Search

Installing TensorFlow
[Installing TensorFlow on Ubuntu](#)
Installing TensorFlow on Mac OS X
Installing TensorFlow on Windows
Installing TensorFlow from Sources
Transitioning to TensorFlow 1.0
Installing TensorFlow for Java
Installing TensorFlow for Go
Installing TensorFlow for C

Installing with native pip

You may install TensorFlow through pip, choosing between a simple installation procedure or a more complex one.

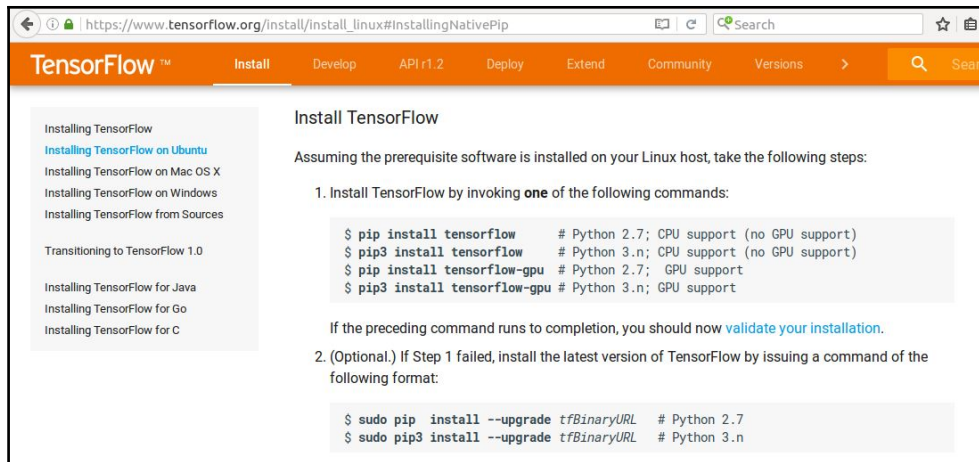
Note: The [REQUIRED_PACKAGES](#) section of `setup.py` lists the TensorFlow packages that pip will install or upgrade.

Prerequisite: Python and Pip

Python is automatically installed on Ubuntu. Take a moment to confirm (by issuing a `python -V` command) that one of the following Python versions is already installed on your system:

- Python 2.7
- Python 3.3+

The pip or pip3 package manager is *usually* installed on Ubuntu. Take a moment to confirm (by issuing a `pip -V` or `pip3 -V` command) that pip or pip3 is installed. We strongly recommend version 8.1 or higher of pip or pip3. If Version 8.1 or later is not installed, issue the following command, which will either install or upgrade to the latest pip version:



The screenshot shows the 'Install TensorFlow' section of the TensorFlow website. It provides a list of commands to install TensorFlow for different Python versions and GPU support. It also includes instructions on how to validate the installation and how to upgrade to the latest version.

TensorFlow™ **Install** Develop API r1.2 Deploy Extend Community Versions > Search

Installing TensorFlow
[Installing TensorFlow on Ubuntu](#)
Installing TensorFlow on Mac OS X
Installing TensorFlow on Windows
Installing TensorFlow from Sources
Transitioning to TensorFlow 1.0
Installing TensorFlow for Java
Installing TensorFlow for Go
Installing TensorFlow for C

Install TensorFlow

Assuming the prerequisite software is installed on your Linux host, take the following steps:

1. Install TensorFlow by invoking **one** of the following commands:

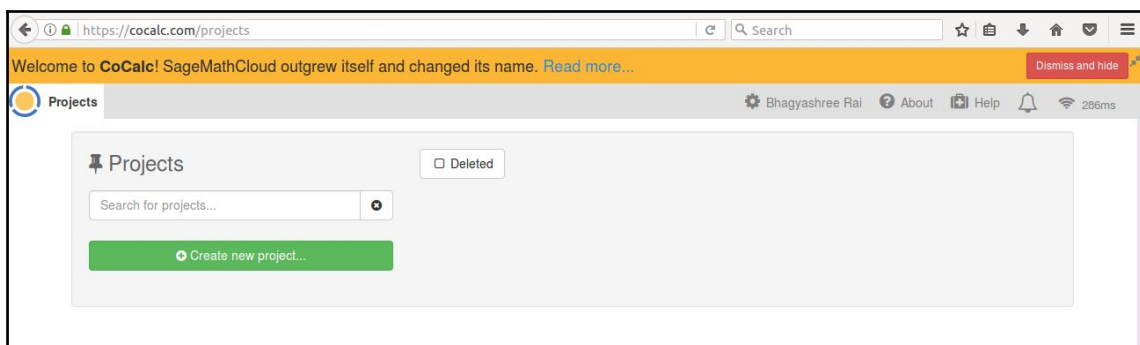
```
$ pip install tensorflow # Python 2.7; CPU support (no GPU support)
$ pip3 install tensorflow # Python 3.n; CPU support (no GPU support)
$ pip install tensorflow-gpu # Python 2.7; GPU support
$ pip3 install tensorflow-gpu # Python 3.n; GPU support
```

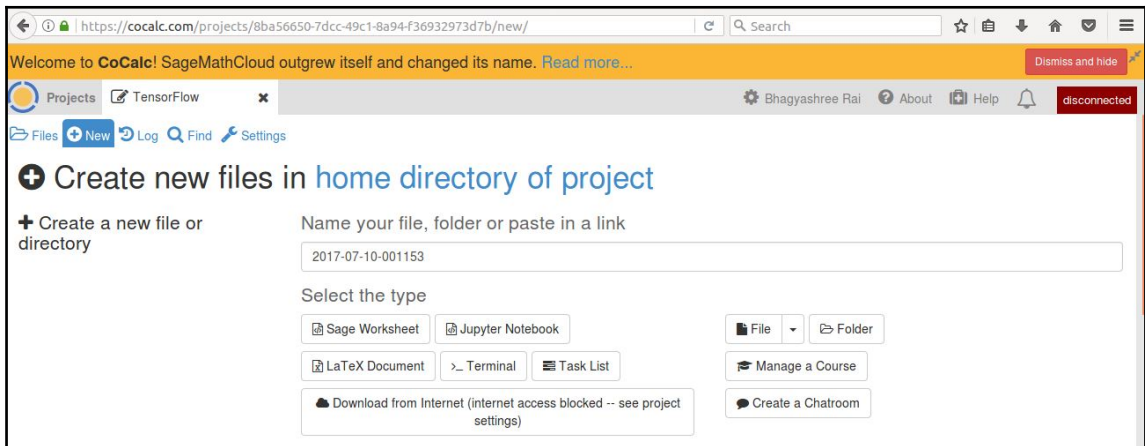
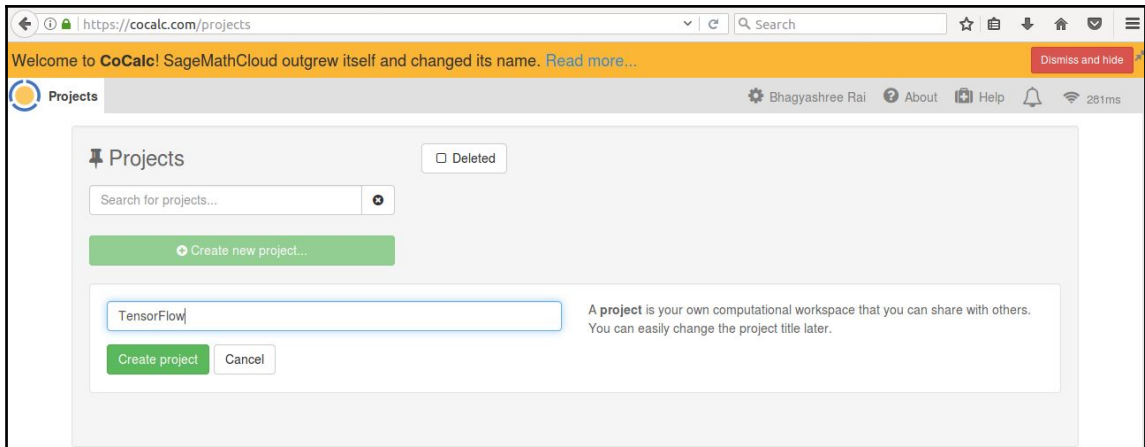
If the preceding command runs to completion, you should now [validate your installation](#).

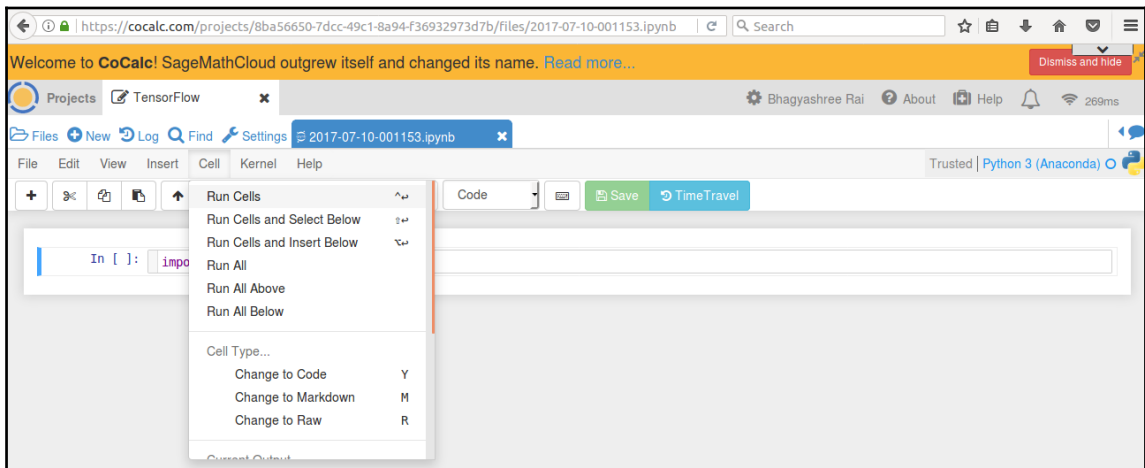
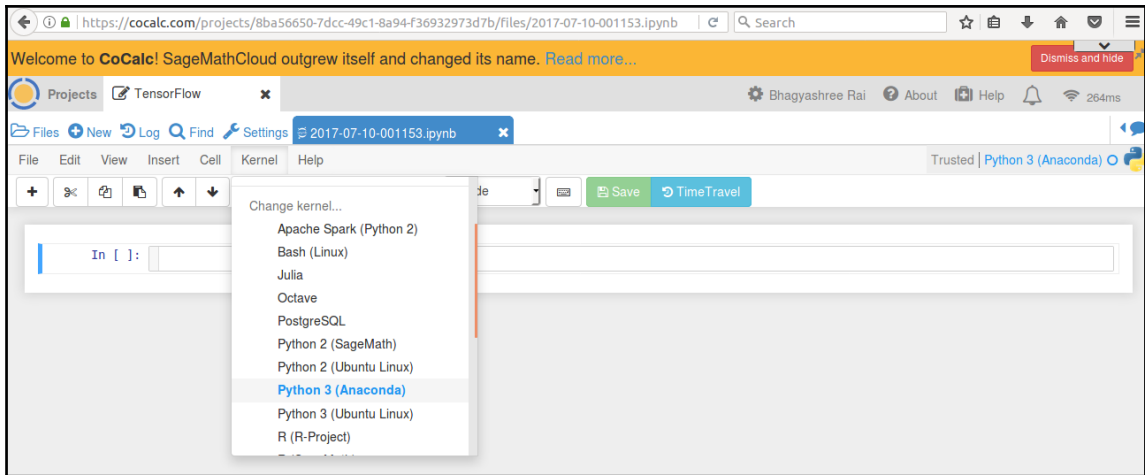
2. (Optional.) If Step 1 failed, install the latest version of TensorFlow by issuing a command of the following format:

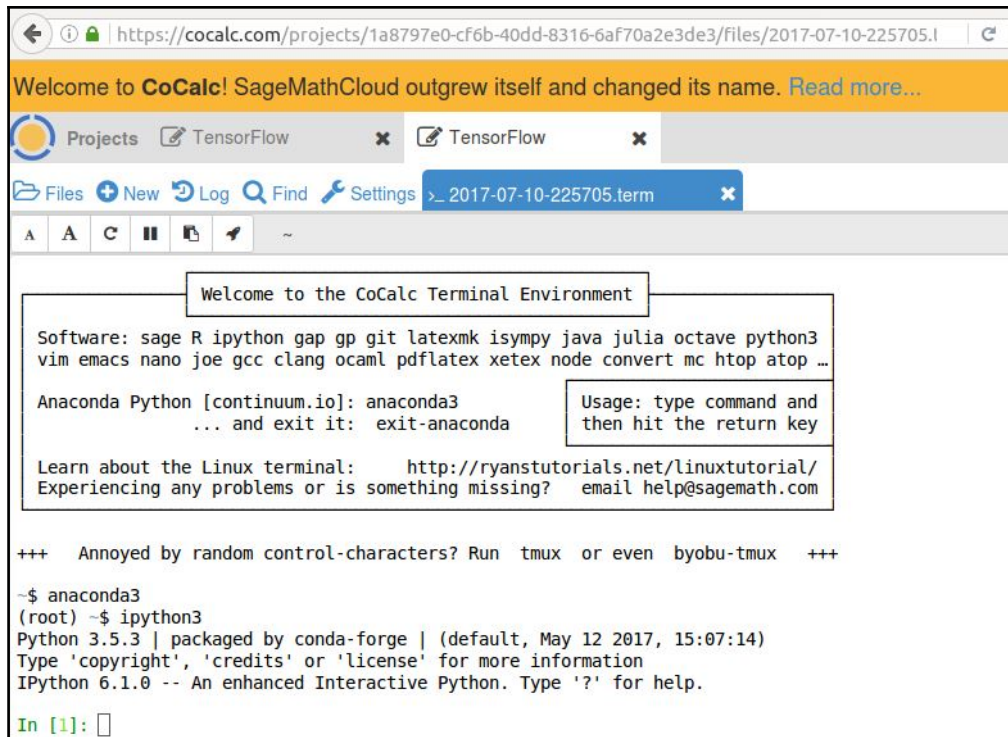
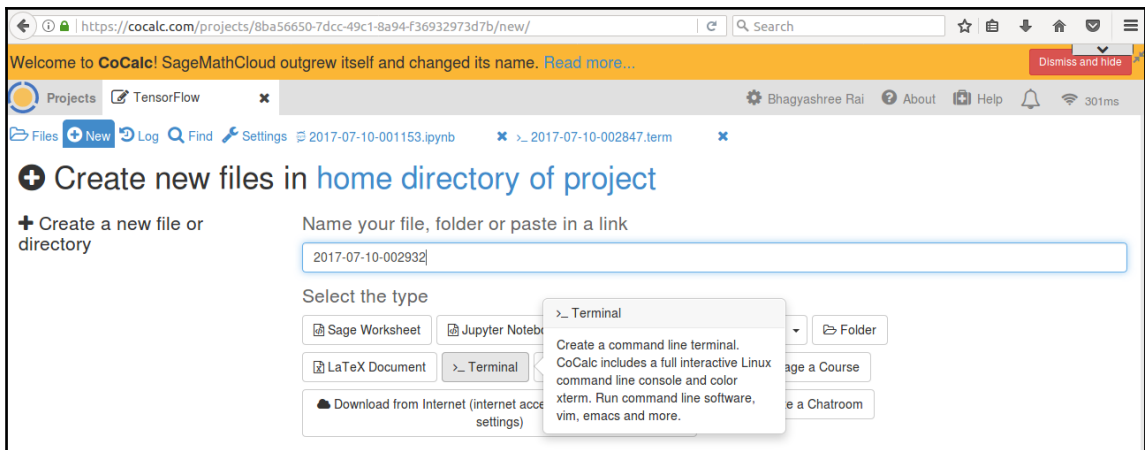
```
$ sudo pip install --upgrade tfBinaryURL # Python 2.7
$ sudo pip3 install --upgrade tfBinaryURL # Python 3.n
```

```
bhagya@bhagya-VirtualBox: ~/Downloads
bhagya@bhagya-VirtualBox:~/Downloads$ sudo pip3 install ./tensorflow-1.2.1-cp35-
cp35m-linux_x86_64.whl
The directory '/home/bhagya/.cache/pip/http' or its parent directory is not owne
d by the current user and the cache has been disabled. Please check the permissi
ons and owner of that directory. If executing pip with sudo, you may want sudo's
-H flag.
The directory '/home/bhagya/.cache/pip' or its parent directory is not owned by
the current user and caching wheels has been disabled. check the permissions and
owner of that directory. If executing pip with sudo, you may want sudo's -H fla
g.
Processing ./tensorflow-1.2.1-cp35-cp35m-linux_x86_64.whl
Collecting html5lib==0.9999999 (from tensorflow==1.2.1)
  Downloading html5lib-0.9999999.tar.gz (889kB)
    100% |██████████████████████████████| 890kB 129kB/s
Collecting bleach==1.5.0 (from tensorflow==1.2.1)
  Downloading bleach-1.5.0-py2.py3-none-any.whl
Requirement already satisfied (use --upgrade to upgrade): wheel>=0.26 in /usr/li
b/python3/dist-packages (from tensorflow==1.2.1)
Requirement already satisfied (use --upgrade to upgrade): six>=1.10.0 in /usr/li
b/python3/dist-packages (from tensorflow==1.2.1)
Collecting numpy>=1.11.0 (from tensorflow==1.2.1)
  Downloading numpy-1.13.1-cp35-cp35m-manylinux1_x86_64.whl (16.9MB)
    100% |██████████████████████████████| 16.9MB 51kB/s
Collecting protobuf>=3.2.0 (from tensorflow==1.2.1)
```









```
In [16]: print("NN is:")  
NN is:
```

```
In [17]: print(output)  
[[ 7.  10.]  
 [ 15. 22.]]
```

```
In [18]:
```

```
In [26]: print("W is:")  
W is:
```

```
In [27]: print(W.eval())  
0
```

```
In [28]: W += a

In [29]: print("W after adding a:")
W after adding a:

In [30]: print(W.eval())
1
```

```
In [31]: W += a

In [32]: print("W after adding a again:")
W after adding a again:

In [33]: print(W.eval())
2

In [34]:
```

```
In [37]: print("E and d:")
E and d:

In [38]: print(sess.run([E,d]))
[4, 2]
```

```
In [39]: print("E with custom d=4:")
E with custom d=4:

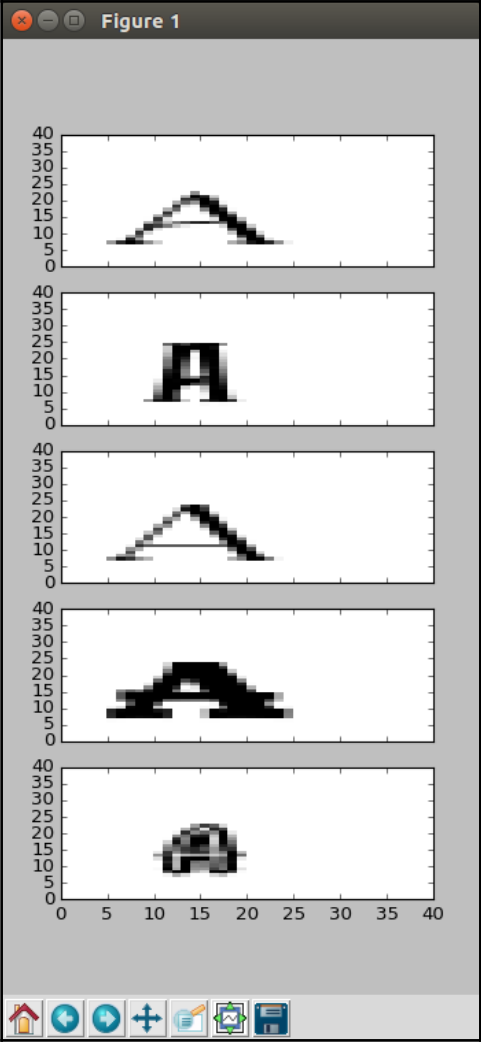
In [40]: print(sess.run(E, feed_dict = {d:4.}))
6

In [41]:
```

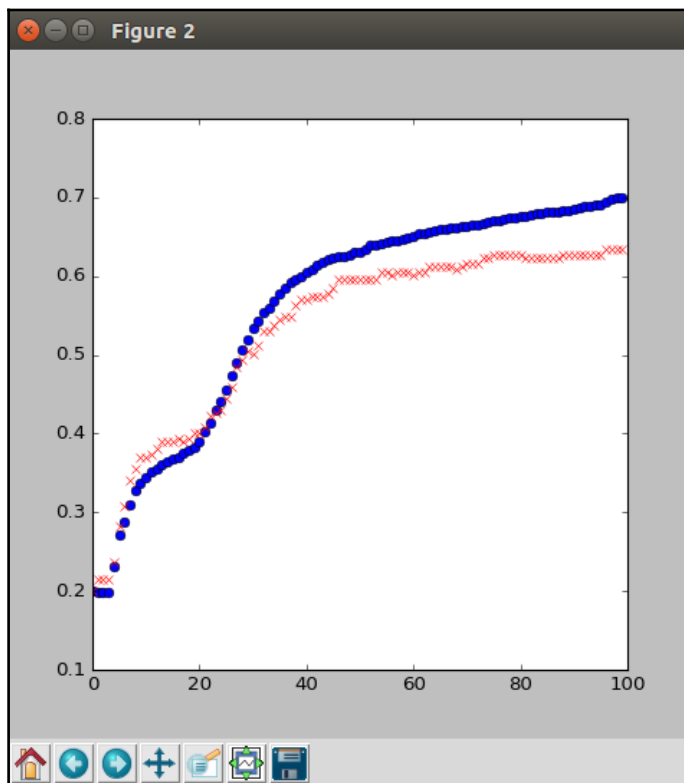
```
In [8]: print(train[0])
[[ 0.  0.  0. ...,  0.  0.  0.]
 [ 0.  0.  0. ...,  0.  0.  0.]
 [ 0.  0.  0. ...,  0.  0.  0.]
 ...,
 [ 0.  0.  0. ...,  0.  0.  0.]
 [ 0.  0.  0. ...,  0.  0.  0.]
 [ 0.  0.  0. ...,  0.  0.  0.]]

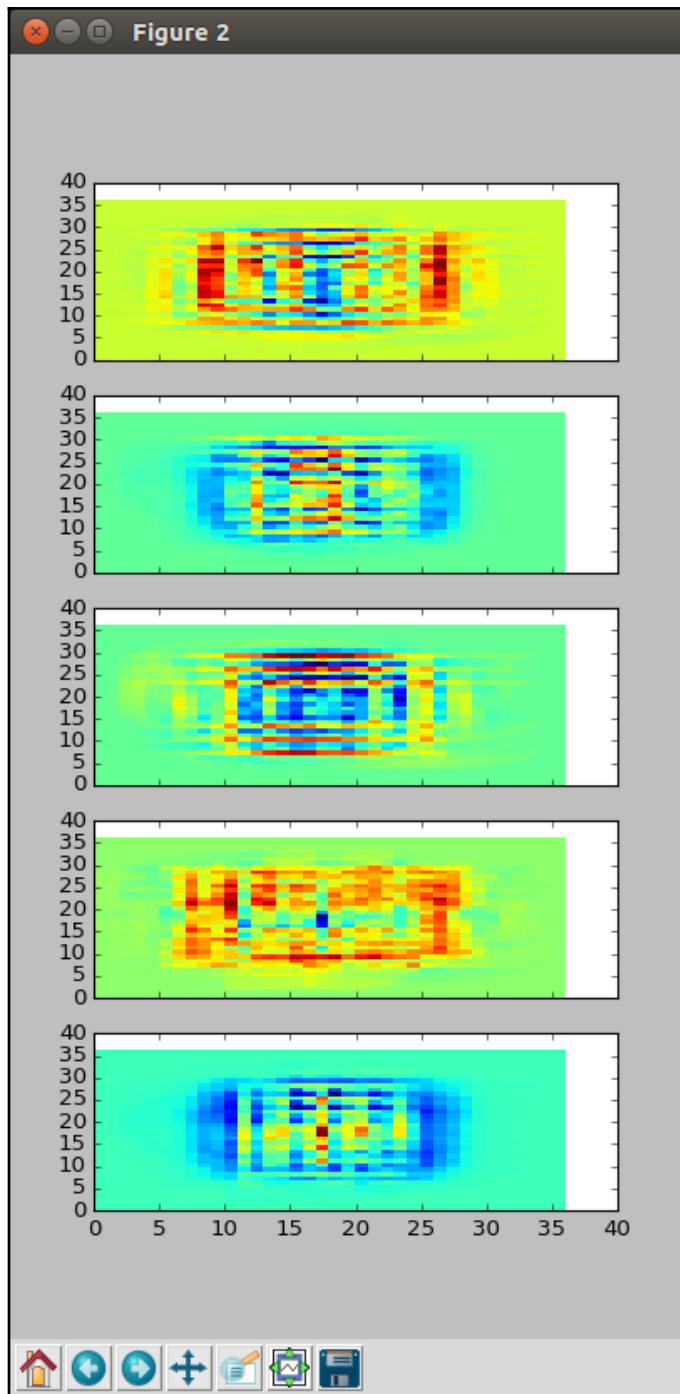
In [9]: print(labels[0])
0

In [10]:
```



$$Pr(Y_i=c)=\frac{\beta_c X_i}{\sum_h \beta_h X_i}$$





Chapter 2: Deep Neural Networks

```
In [11]: print(x2.eval())  
[-1.59633303 -1.39370716 -1.11756158 -0.93147004 -1.30868506]  
  
In [12]: print(sqx2.eval())  
[ 2.54827905  1.94241965  1.24894392  0.86763644  1.71265662]  
  
In [13]: █
```

```
In [14]: print(x1.eval())  
[ 3.2928977  3.11865115  2.75602937  2.55065155  2.60228252]  
  
In [15]: print(logx1.eval())  
[ 1.19176793  1.13740063  1.01379097  0.9363488  0.95638895]  
  
In [16]: █
```

```
In [17]: print(x3.eval())  
[-0.24215472 -0.26575294 -0.30768225  0.0072251  -0.1542311 ]  
  
In [18]: print(sigx3.eval())  
[ 0.43975541  0.43395001  0.4236806  0.50180626  0.4615185 ]  
  
In [19]: █
```

```

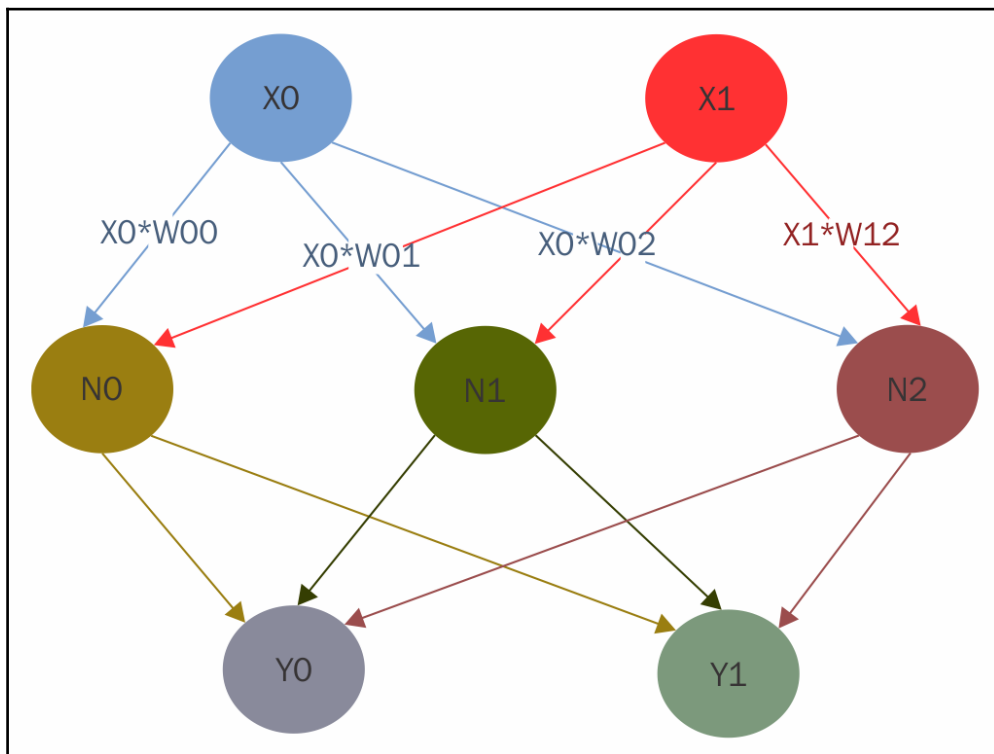
In [24]: print((w1*x1).eval())
[ 0.25280735  0.23198548  0.32570502  0.32763374  0.29772624]

In [25]: print((w2*x2).eval())
[-0.16239884 -0.26094452 -0.24706605 -0.228054  -0.14377597]

In [26]: print(n1.eval())
[ 0.52258676  0.49276075  0.51964962  0.52487439  0.53841174]

In [27]: 

```



```

In [23]: num_hidden = 128

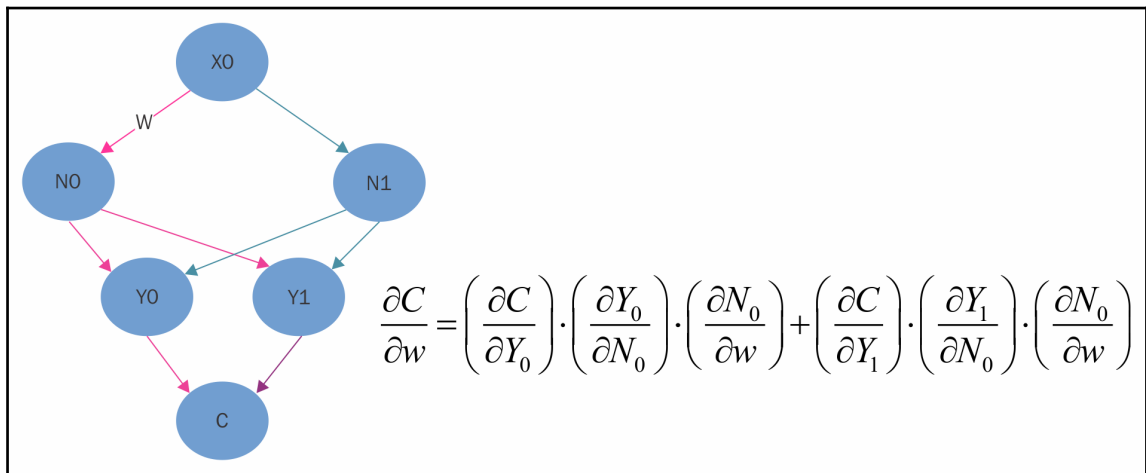
In [24]: W1 = tf.Variable(tf.truncated_normal([1296, num_hidden],
      ....:                                     stddev=1./math.sqrt(1296)))

In [25]: b1 = tf.Variable(tf.constant(0.1,shape=[num_hidden]))

In [26]: h1 = tf.sigmoid(tf.matmul(x,W1) + b1)

In [27]: 

```

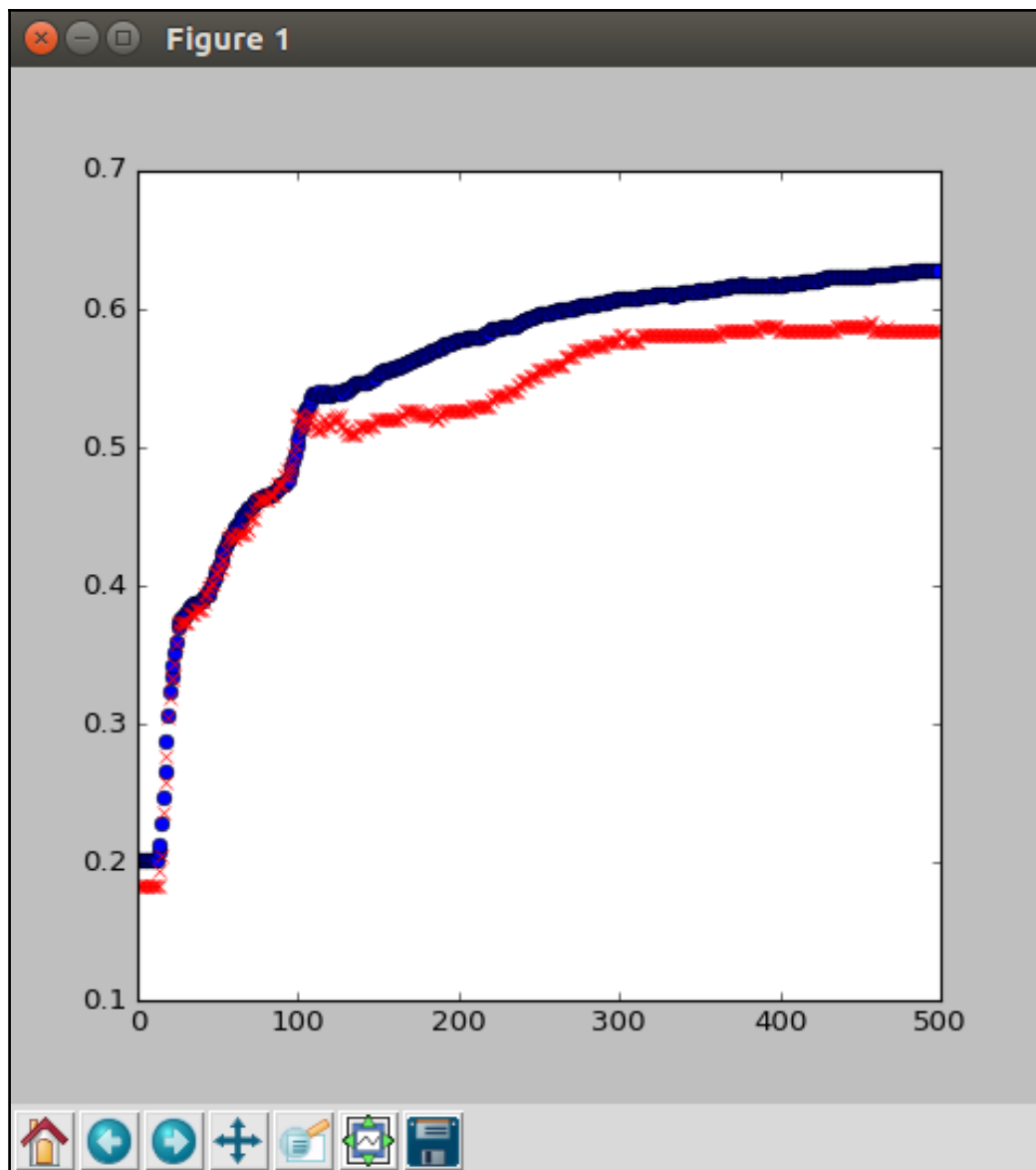


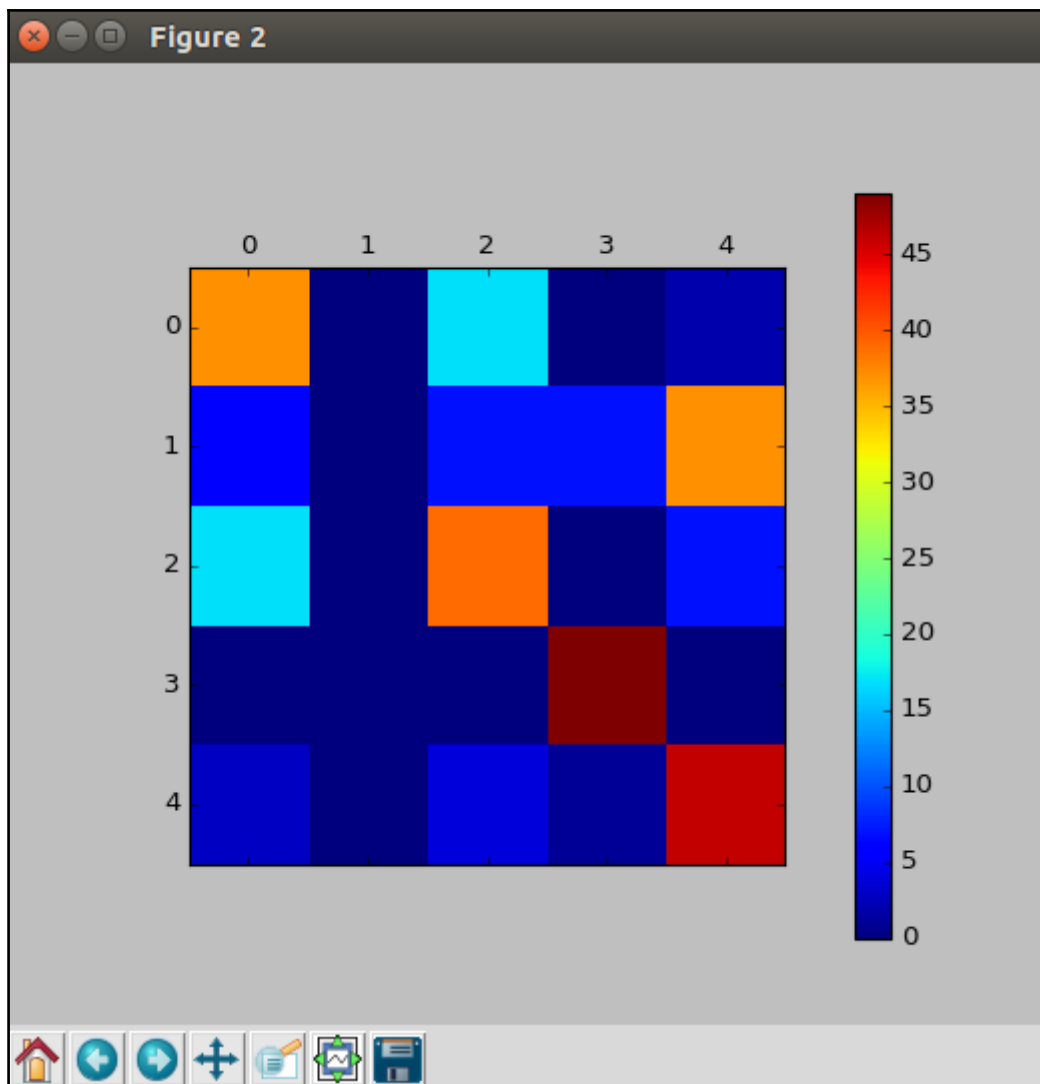
```
In [42]: epochs = 5000

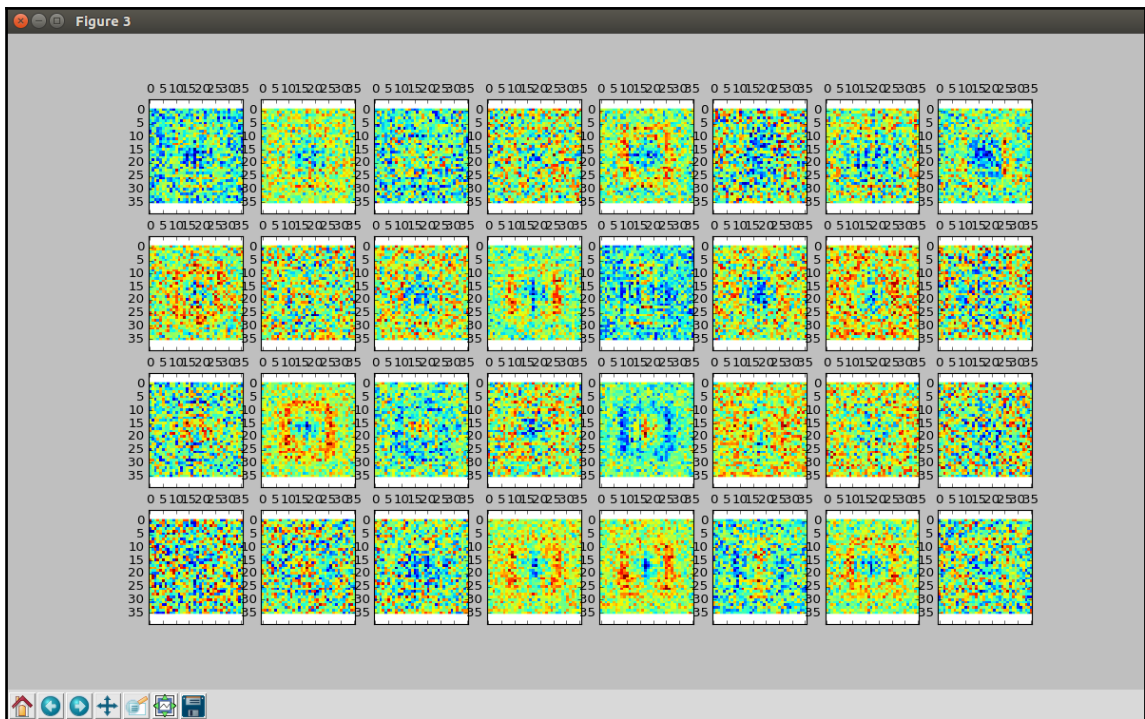
In [43]: train_acc = np.zeros(epochs//10)

In [44]: test_acc = np.zeros(epochs//10)

In [45]: for i in tqdm(range(epochs), ascii = True):
.....:     # Record summary data, and the accuracy
.....:     if i % 10 == 0:
.....:         # Check accuracy on train set
.....:         A = accuracy.eval(feed_dict={
.....:             x: train.reshape([-1,1296]),
.....:             y_: onehot_train})
.....:         train_acc[i//10] = A
.....:         # And now the validation set
.....:         A = accuracy.eval(feed_dict={
.....:             x: test.reshape([-1,1296]),
.....:             y_: onehot_test})
.....:         test_acc[i//10] = A
.....:     train_step.run(feed_dict={
.....:         x: train.reshape([-1,1296]),
.....:         y_: onehot_train})
.....:
```







In [2]:

```
import tensorflow as tf
import numpy as np
import math
%autoindent
try:
    from tqdm import tqdm
except ImportError:
    def tqdm(x, *args, **kwargs):
        return x

# Load data
data = np.load('data_with_labels.npz')
train = data['arr_0']/255.
labels = data['arr_1']

# Look at some data
print(train[0])
print(labels[0])

# If you have matplotlib installed
import matplotlib.pyplot as plt
plt.ion()

def to_onehot(labels, nclasses = 5):
    '''
    Convert labels to "one-hot" format.

    >>> a = [0,1,2,3]
    >>> to_onehot(a,5)
    array([[ 1.,  0.,  0.,  0.,  0.],
           [ 0.,  1.,  0.,  0.,  0.],
           [ 0.,  0.,  1.,  0.,  0.],
           [ 0.,  0.,  0.,  1.,  0.]])
    '''
```

```
In [30]: W3 = tf.Variable(tf.truncated_normal([num_hidden2, 5],
....:                                     stddev=1./math.sqrt(5)))

In [31]: b3 = tf.Variable(tf.constant(0.1,shape=[5]))

In [32]: sess.run(tf.global_variables_initializer())

In [33]: y = tf.nn.softmax(tf.matmul(h2,W3) + b3)

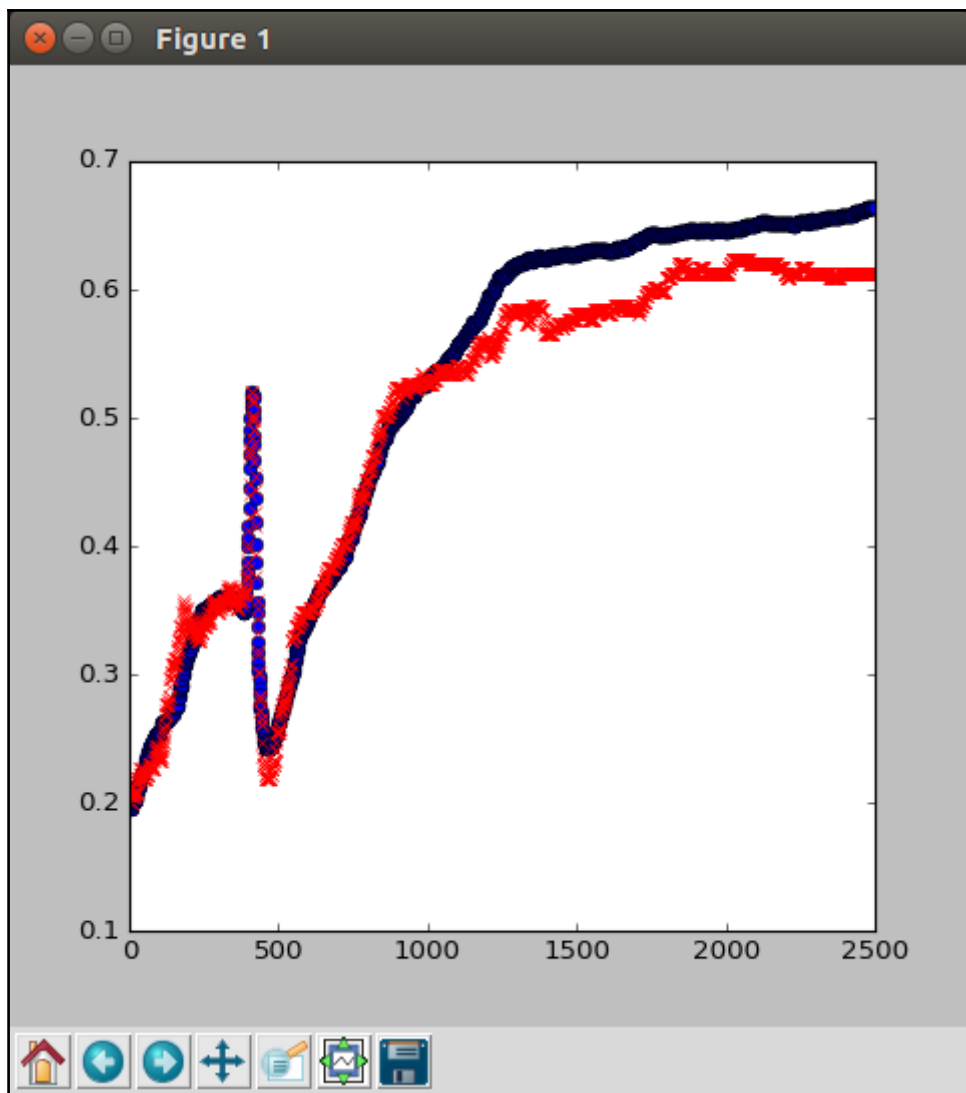
In [34]: █
```

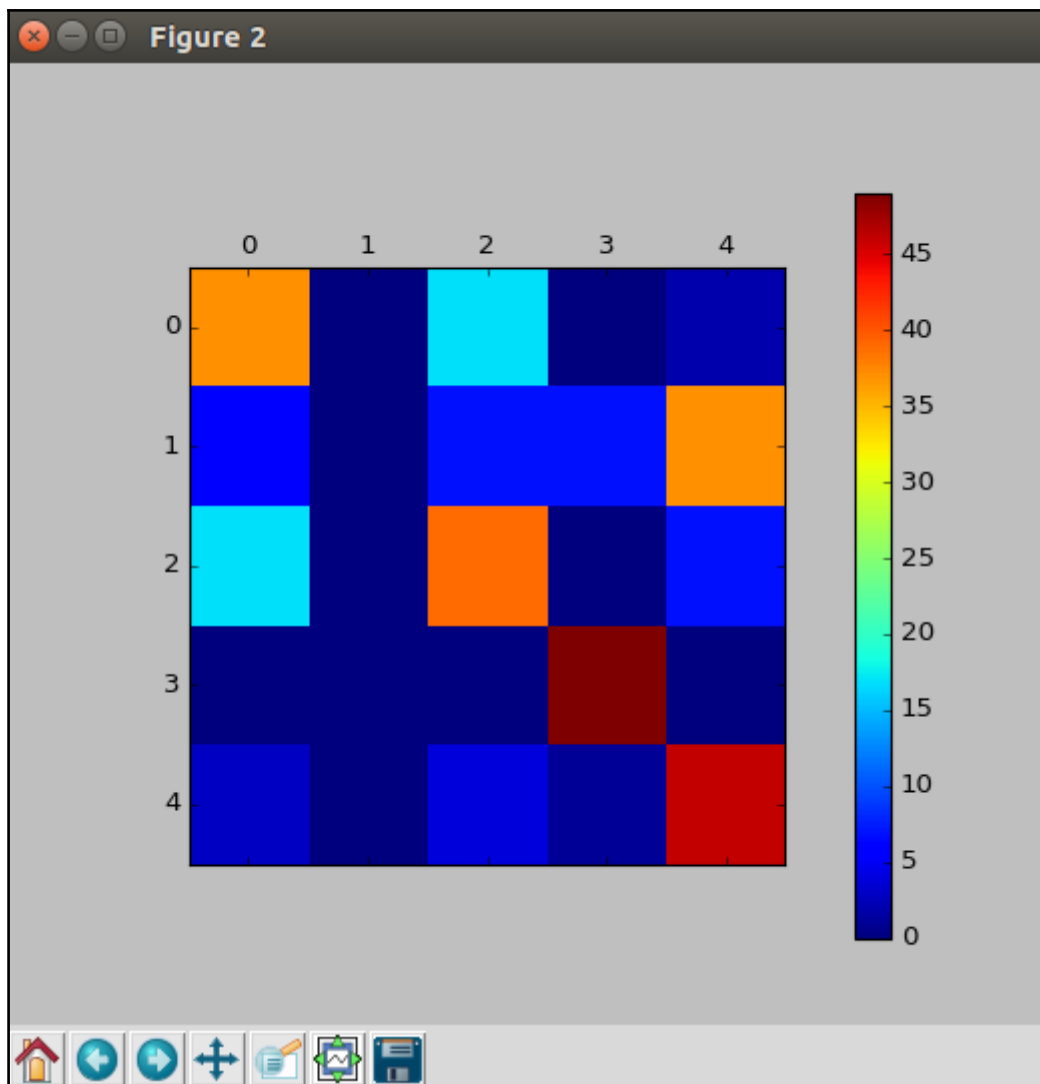
```
In [38]: epochs = 25000

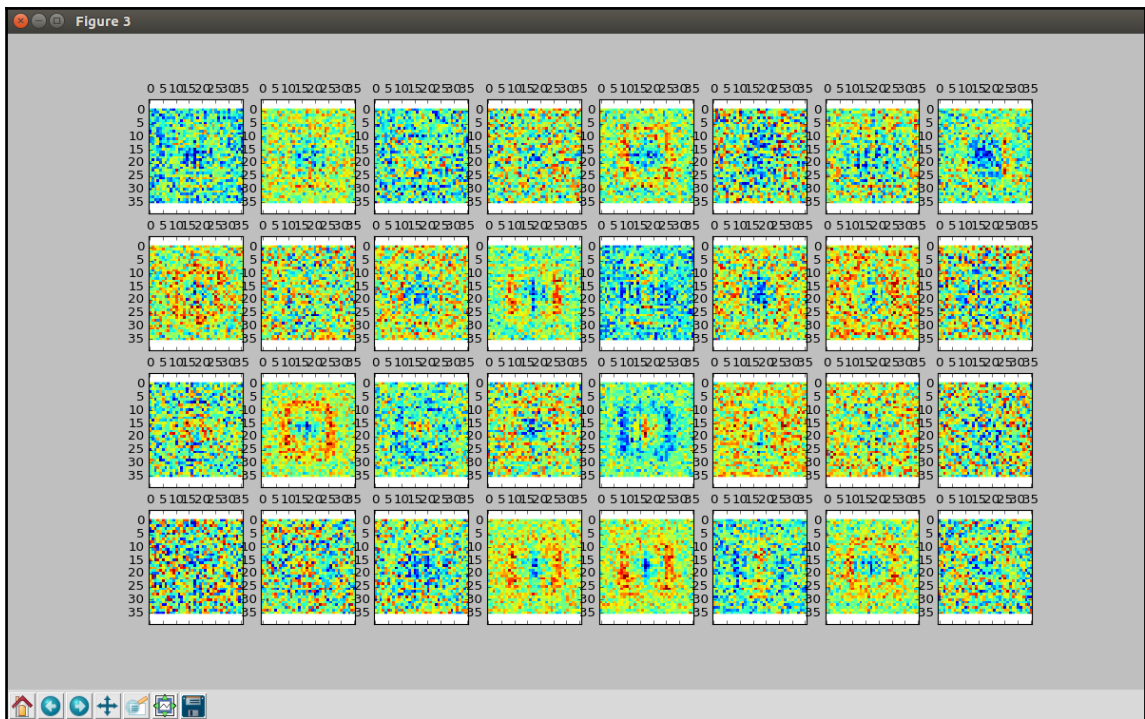
In [39]: train_acc = np.zeros(epochs//10)

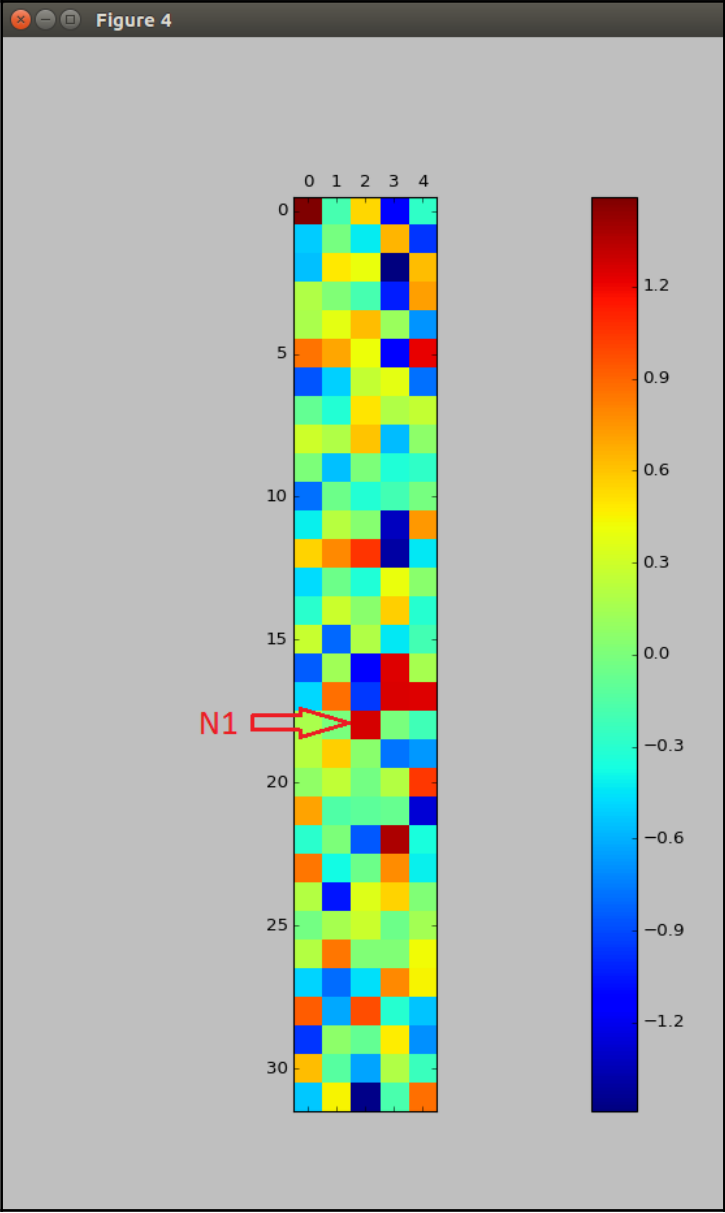
In [40]: test_acc = np.zeros(epochs//10)

In [41]: for i in tqdm(range(epochs), ascii = True):
....:     # Record summary data, and the accuracy
....:     if i % 10 == 0:
....:         # Check accuracy on train set
....:         A = accuracy.eval(feed_dict={
....:             x: train.reshape([-1,1296]),
....:             y_: onehot_train})
....:         train_acc[i//10] = A
....:         # And now the validation set
....:         A = accuracy.eval(feed_dict={
....:             x: test.reshape([-1,1296]),
....:             y_: onehot_test})
....:         test_acc[i//10] = A
....:     train_step.run(feed_dict={
....:         x: train.reshape([-1,1296]),
....:         y_: onehot_train})
....:
```

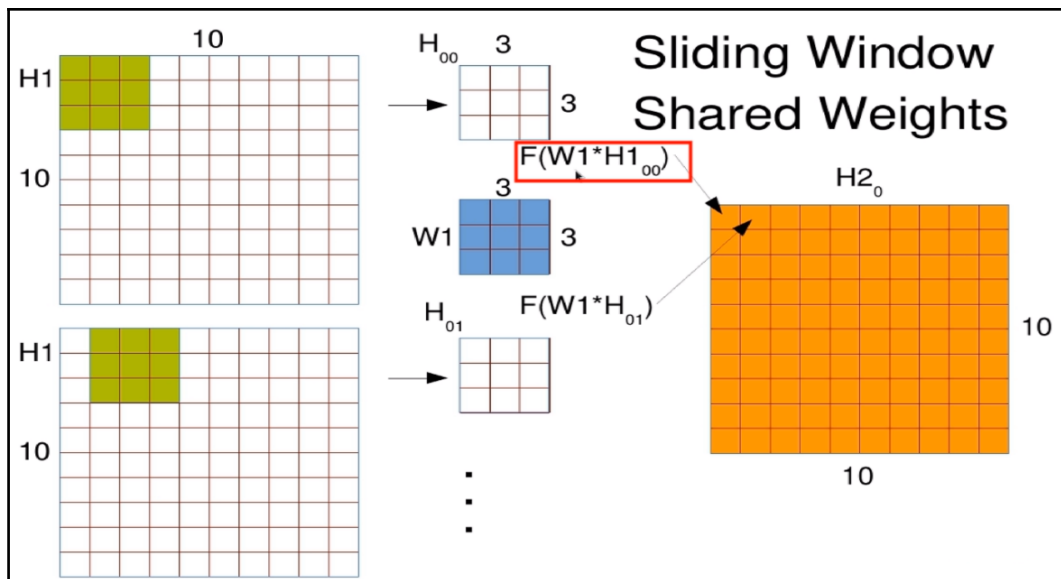
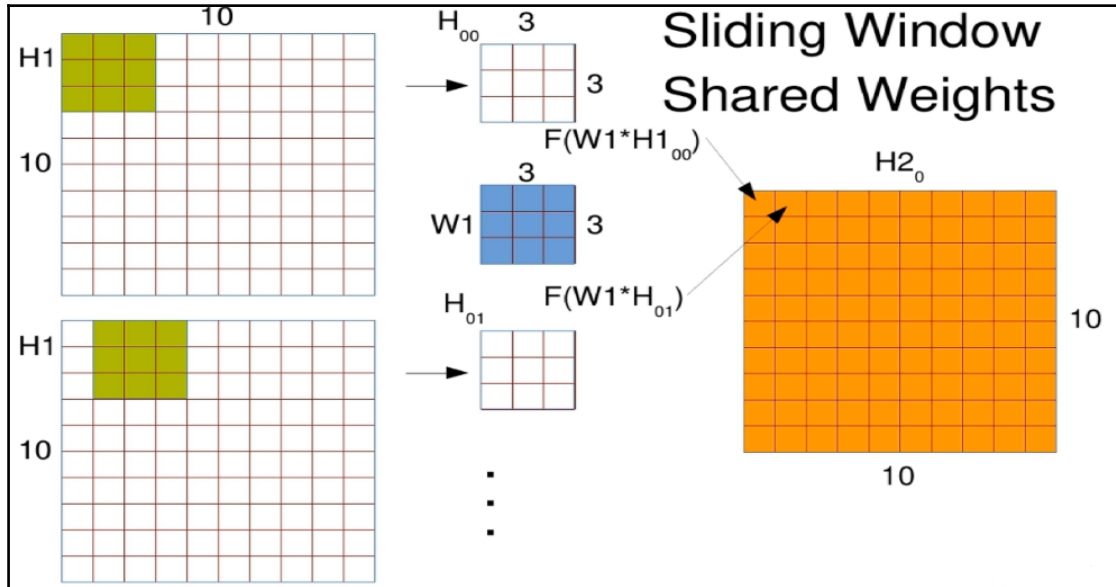


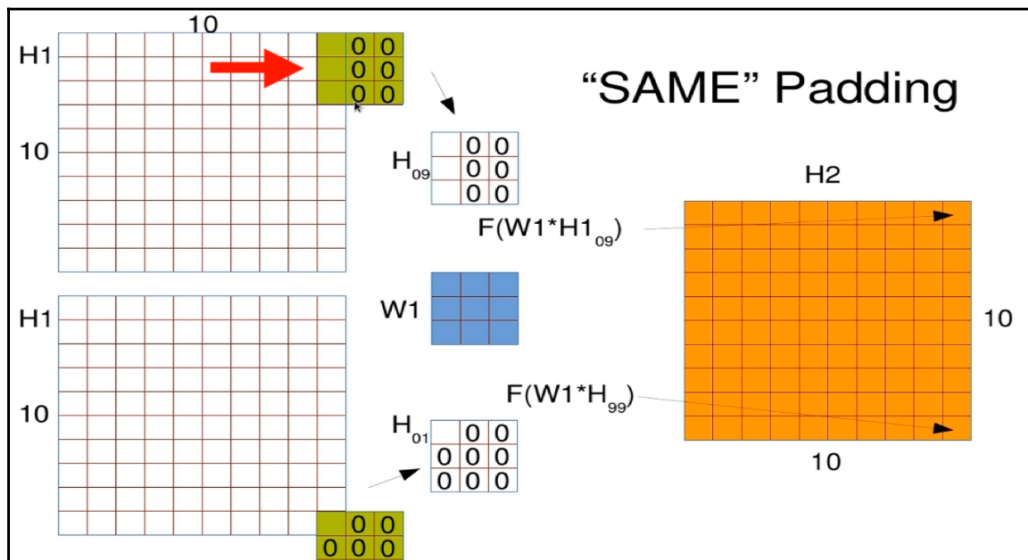
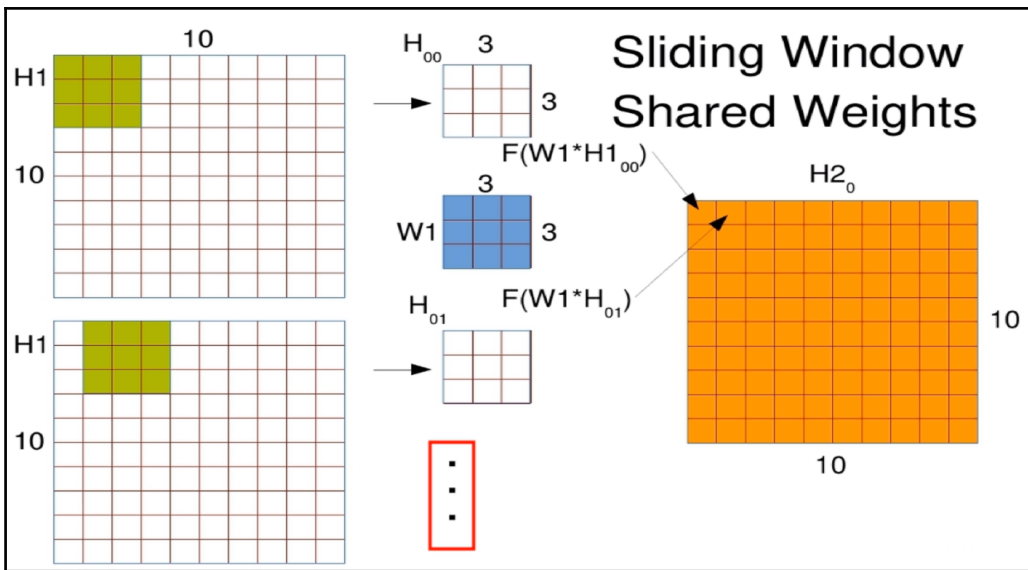


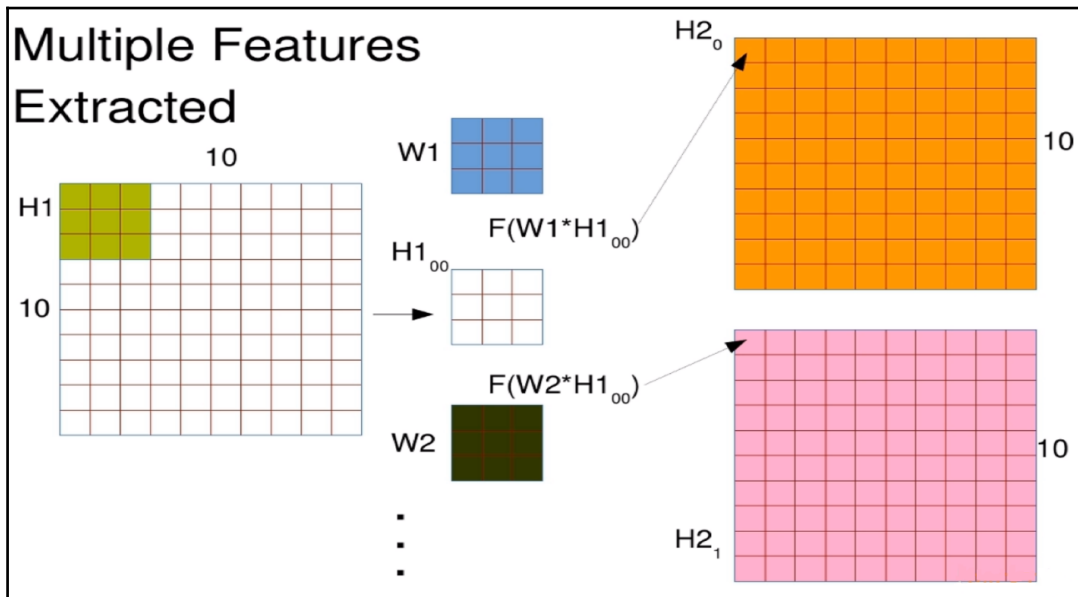
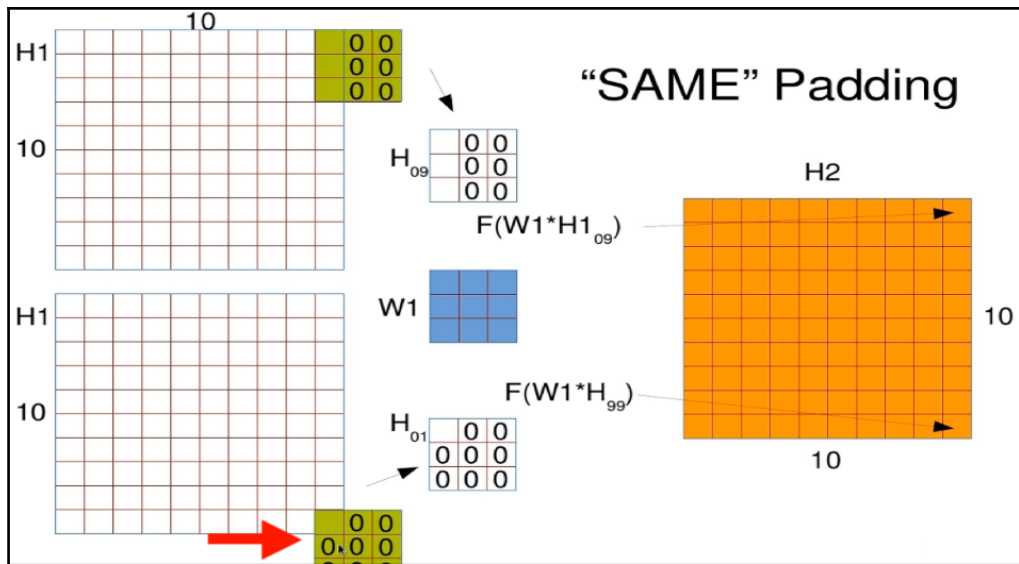


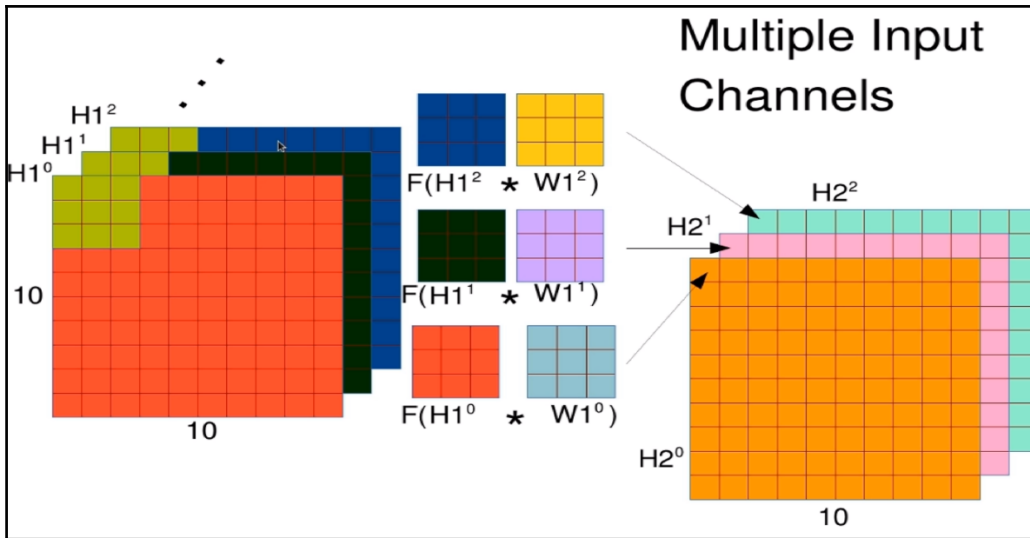


Chapter 3: Convolutional Neural Networks









```
In [1]: import tensorflow as tf
import math
import numpy as np

sess = tf.InteractiveSession()

# Make some fake data, 1 data points
image = np.random.randint(10,size=[1,10,10]) + np.eye(10)*10

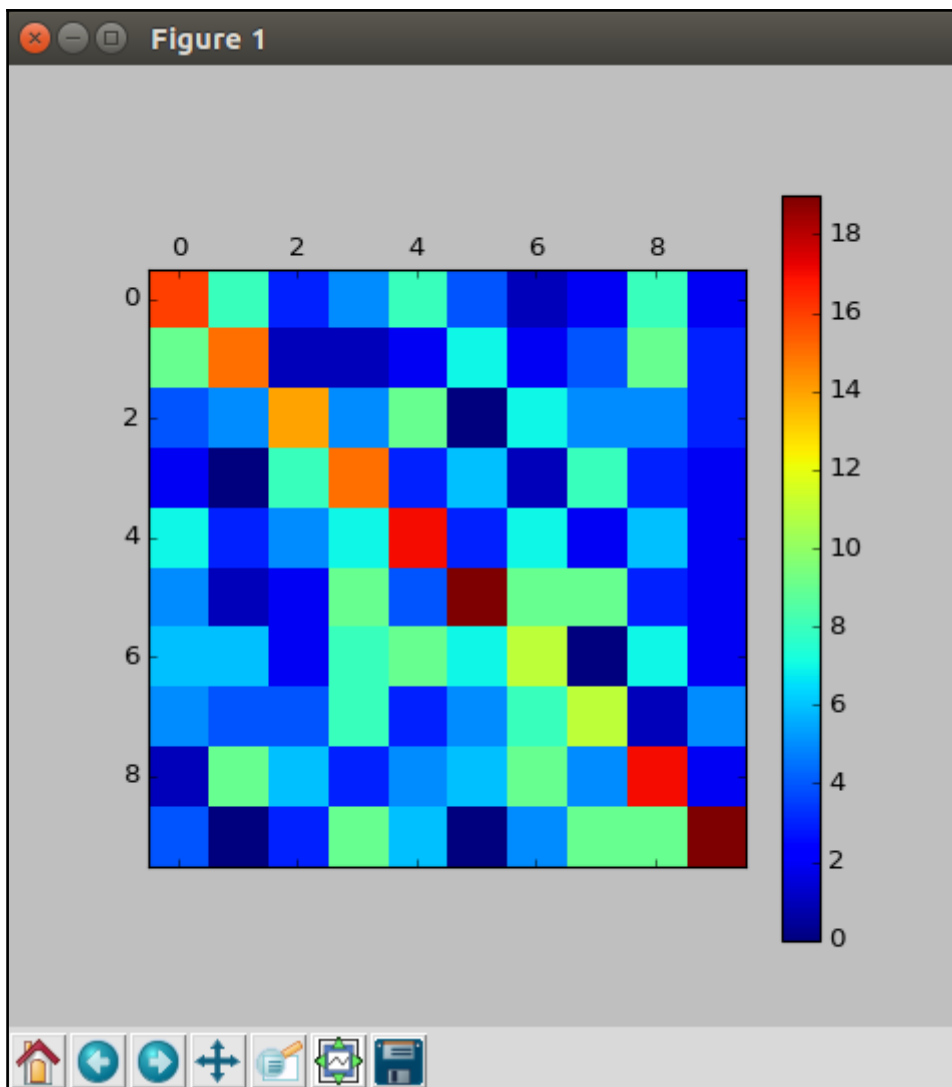
# TensorFlow placeholder
# None is for batch processing
# (-1 keeps same size)
# 10x10 is the shape
# 1 is the number of "channels"
# (like RGB colors or gray)
x = tf.placeholder("float", [None, 10, 10])
x_tm = tf.reshape(x, [-1,10,10,1])

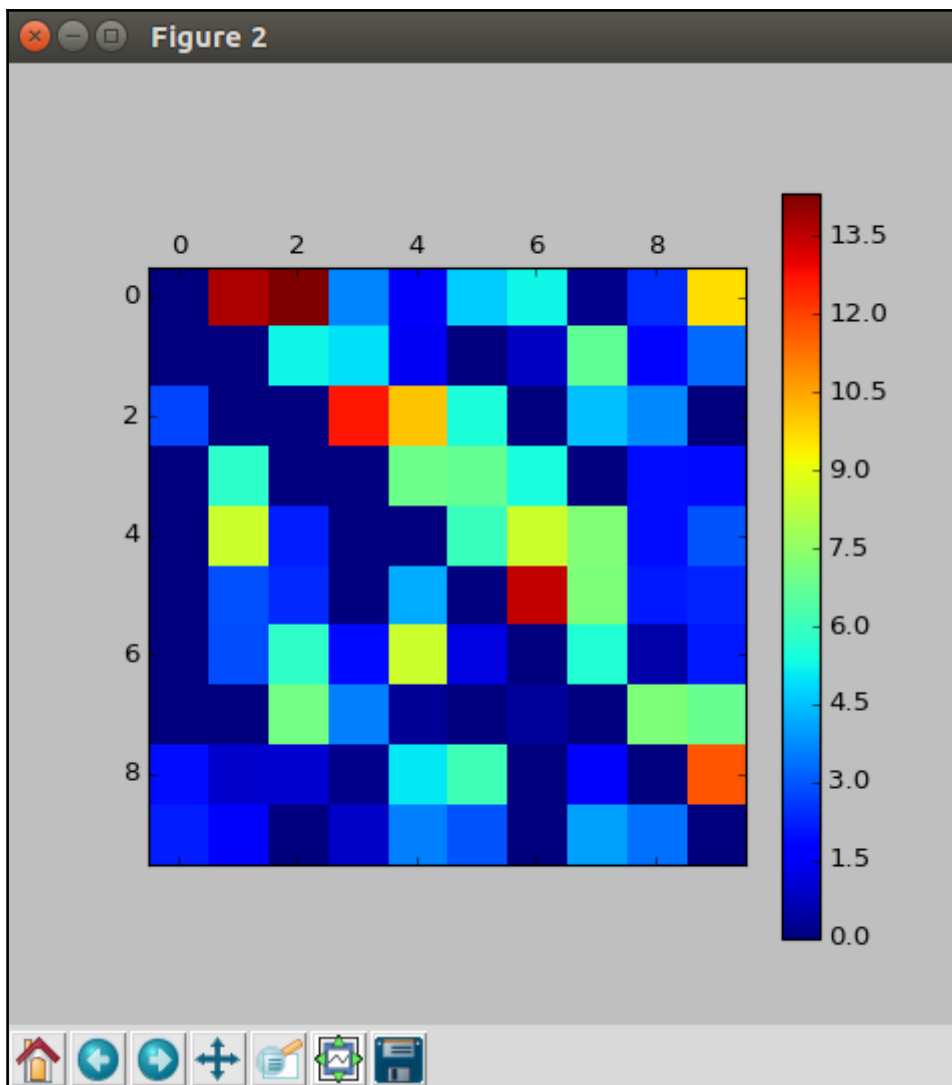
### Convolutional Layer

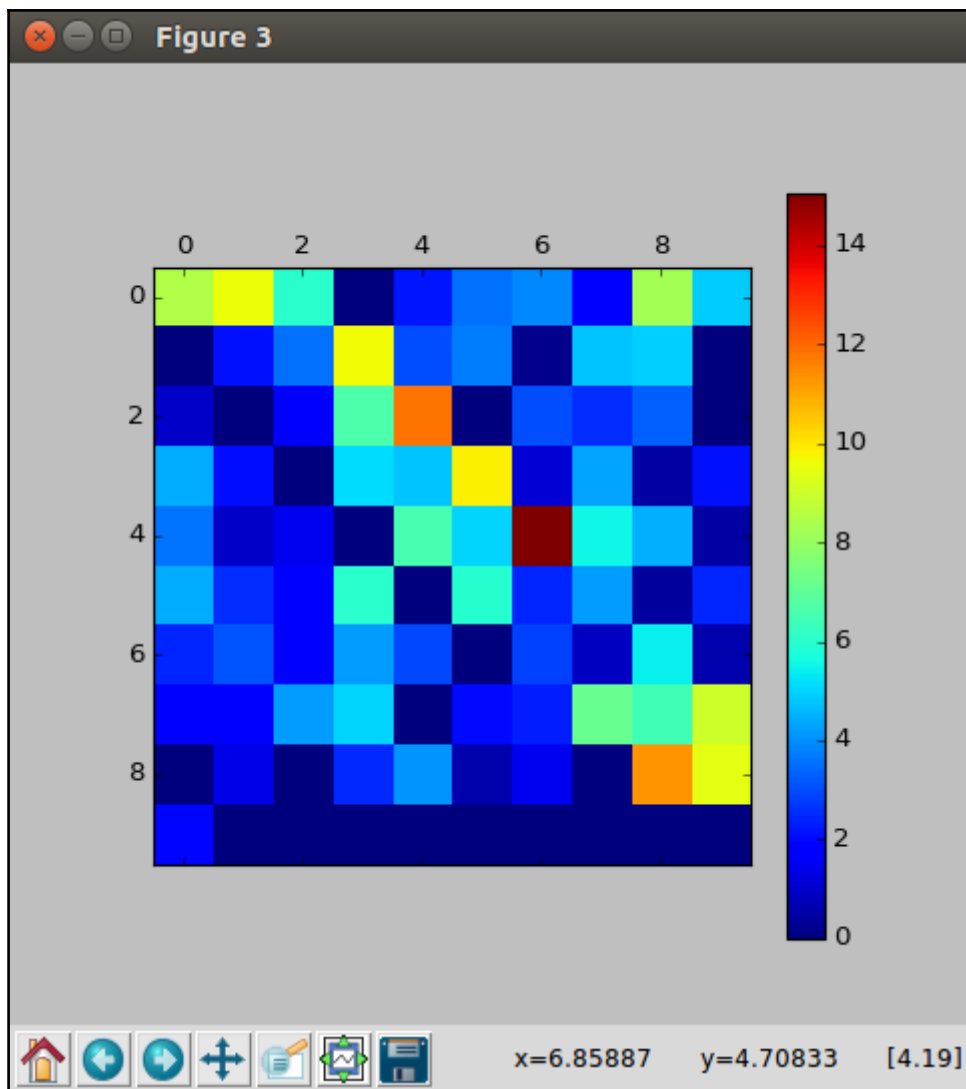
# Window size to use, 3x3 here
winx = 3
winy = 3

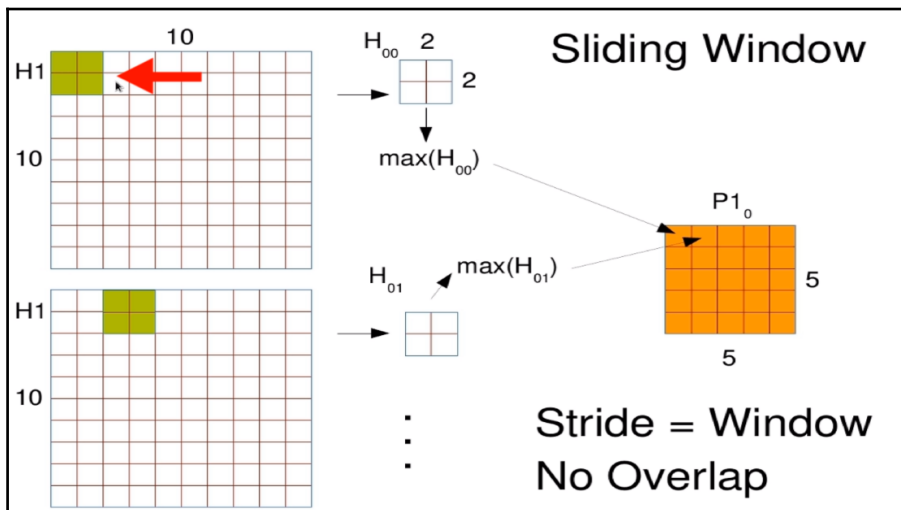
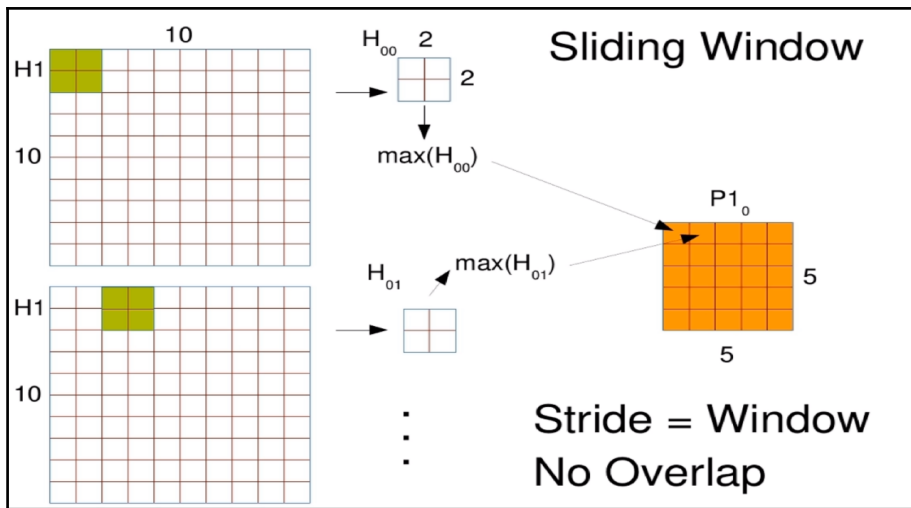
# How many features to compute on the window
num_filters = 2

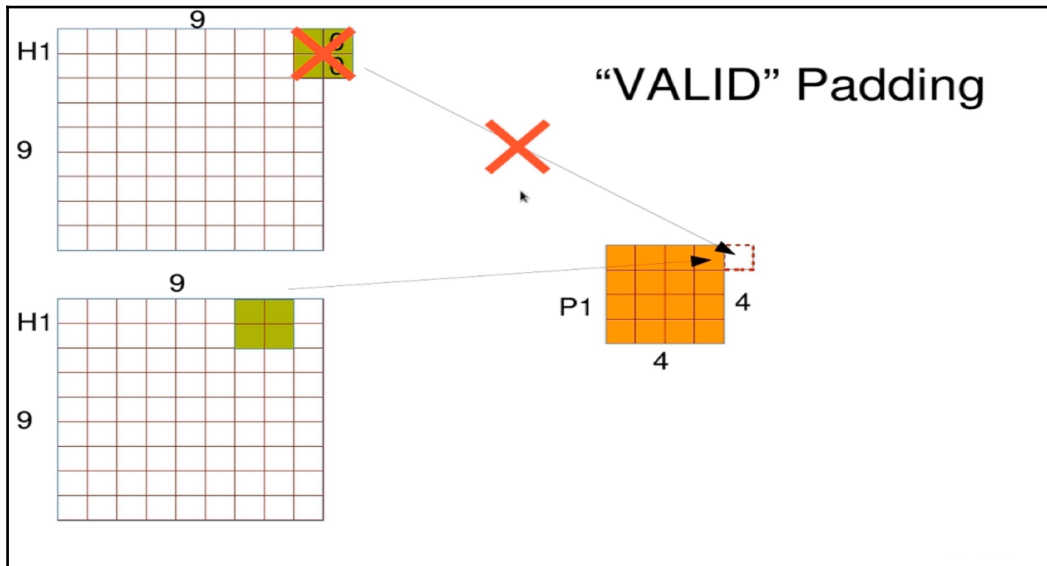
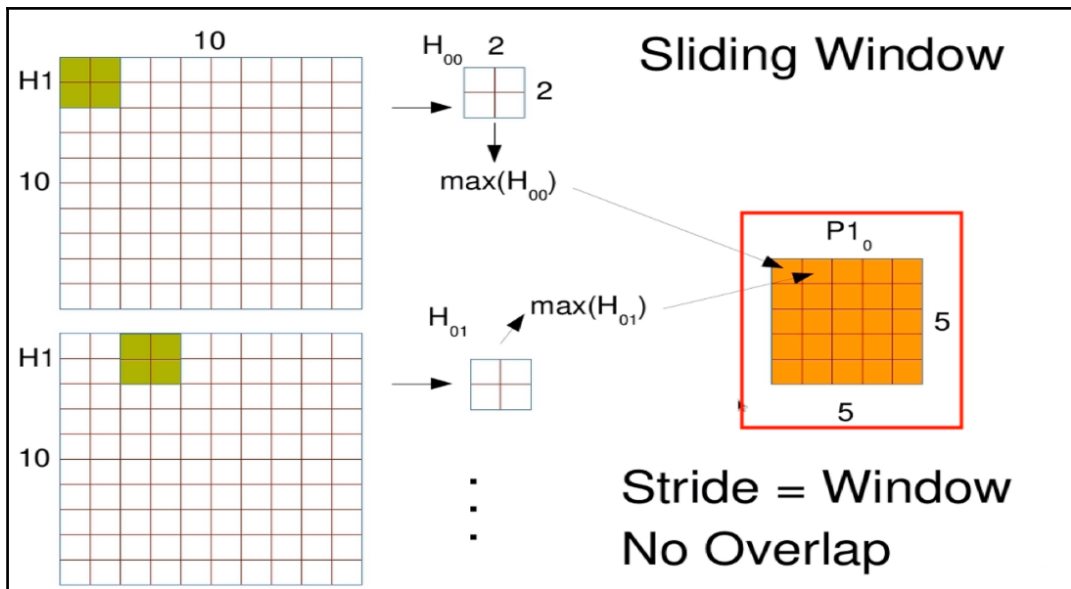
# Weight shape should match window size
# The '1' represents the number of
# input "channels" (colors)
W1 = tf.Variable(tf.truncated_normal(
    [winx, winy,1, num_filters],
    stddev=1./math.sqrt(winx*winy)))
b1 = tf.Variable(tf.constant(
    0.1,shape=[num_filters]))
```

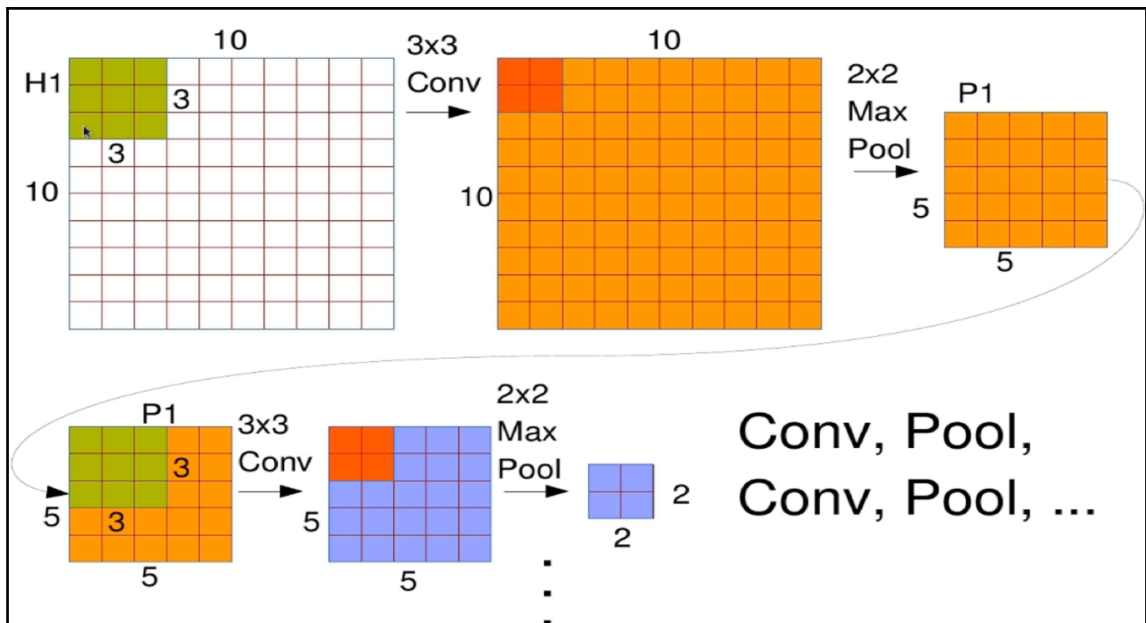


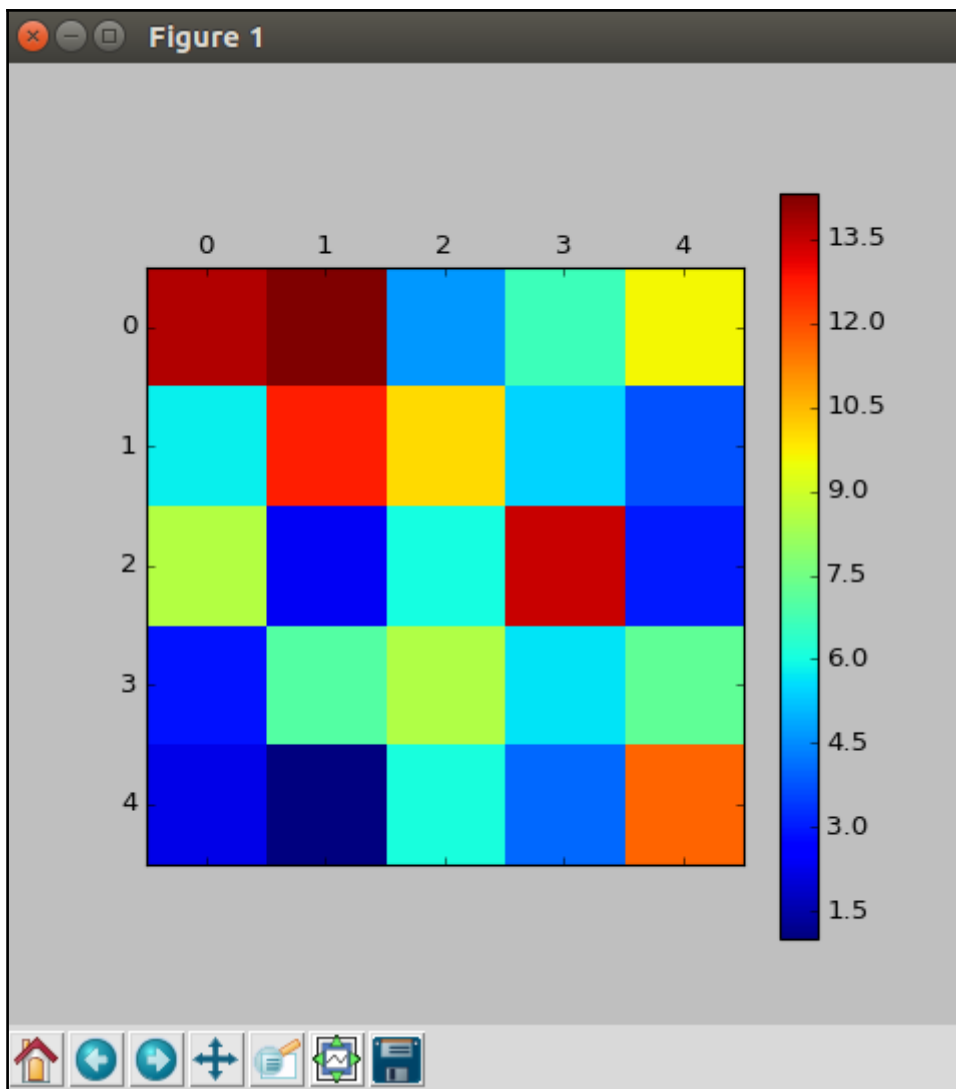


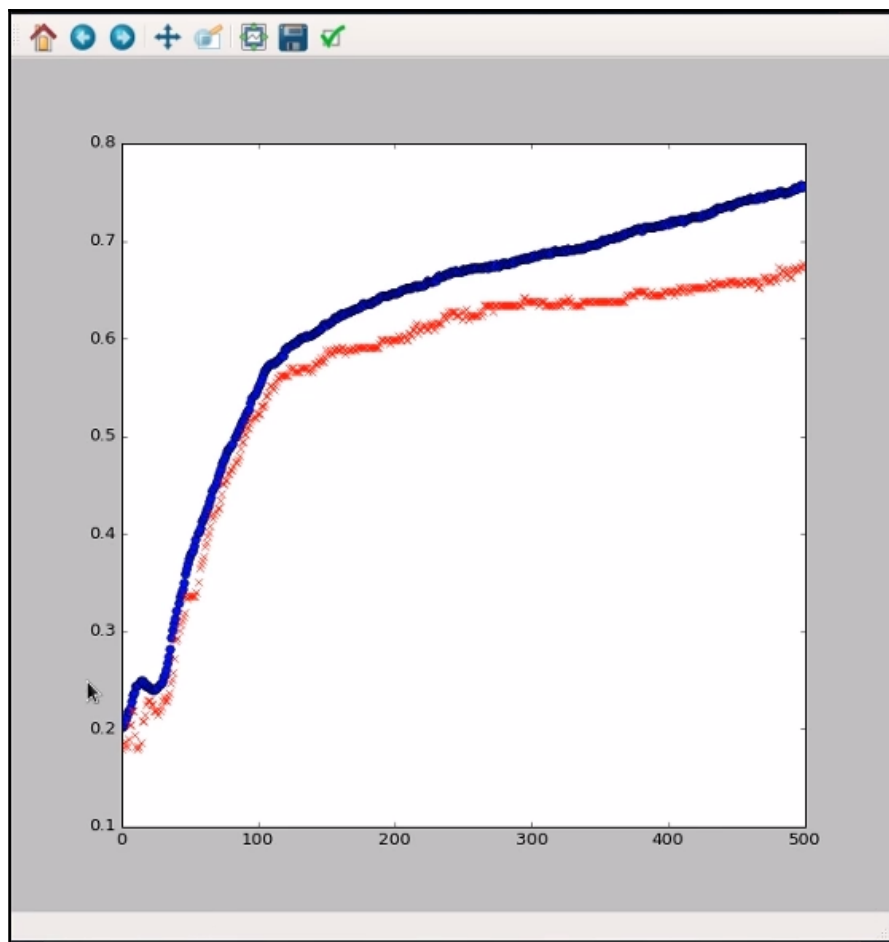


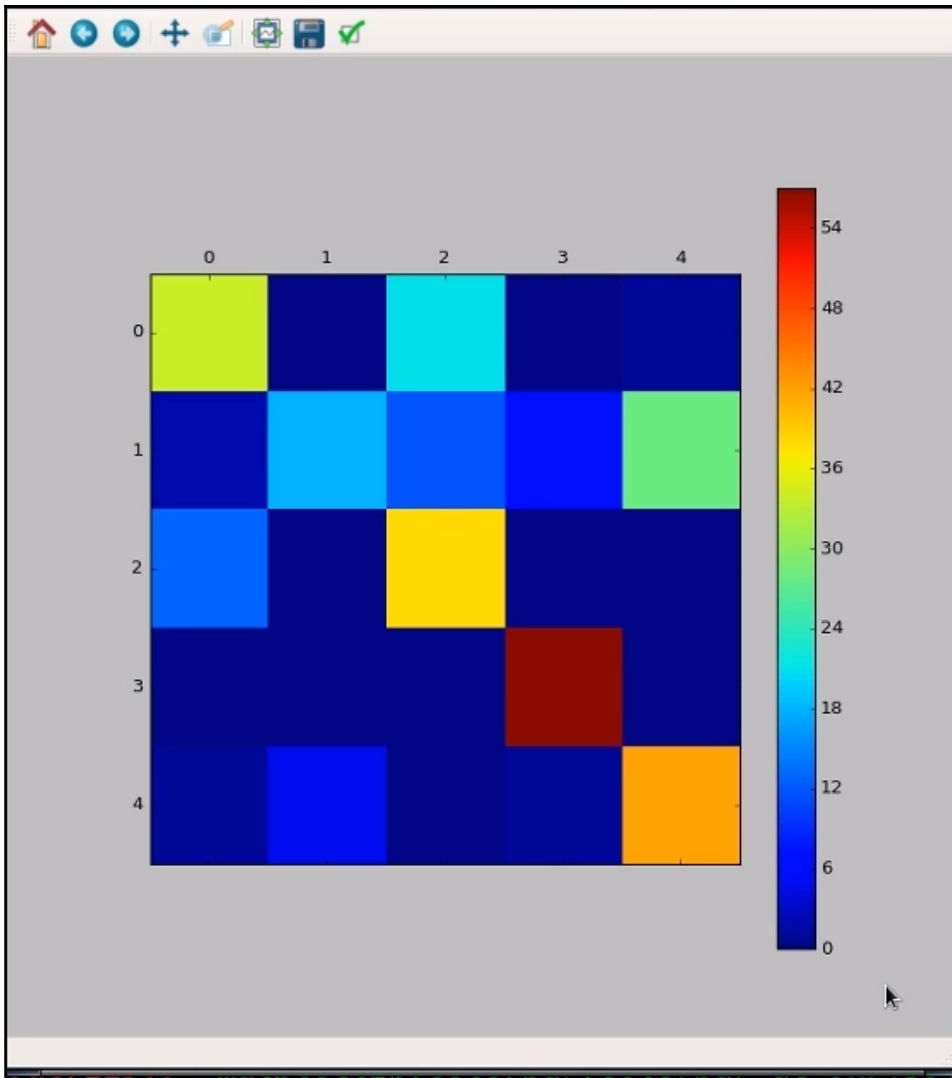


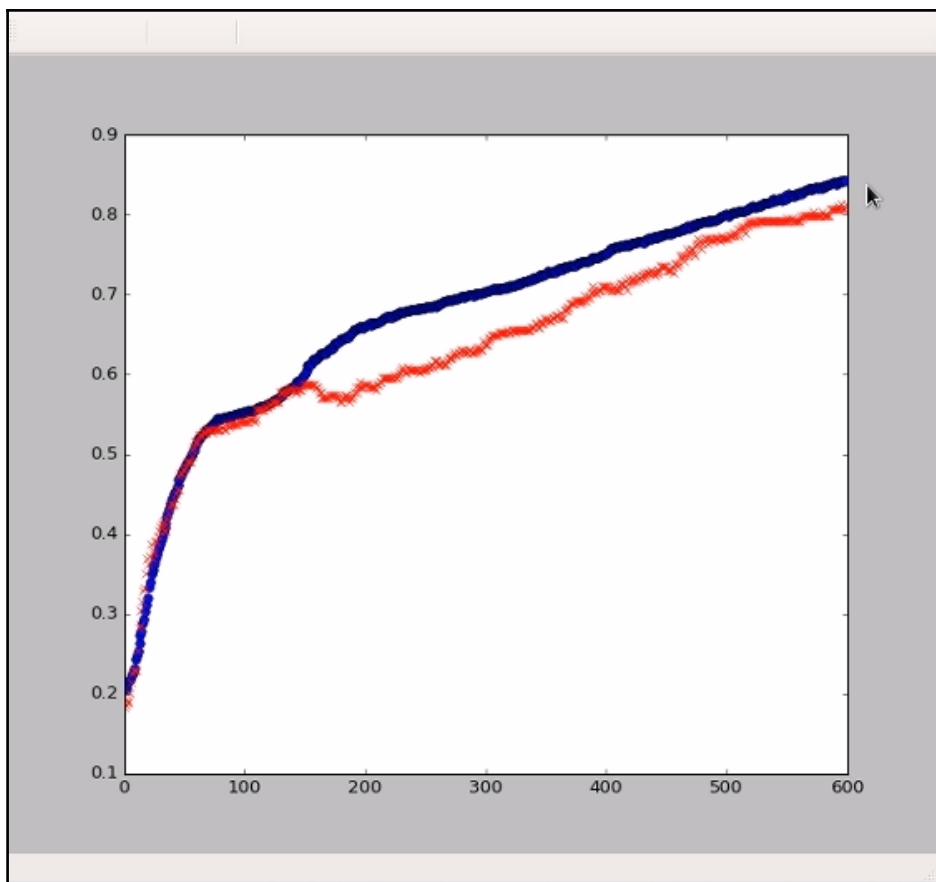


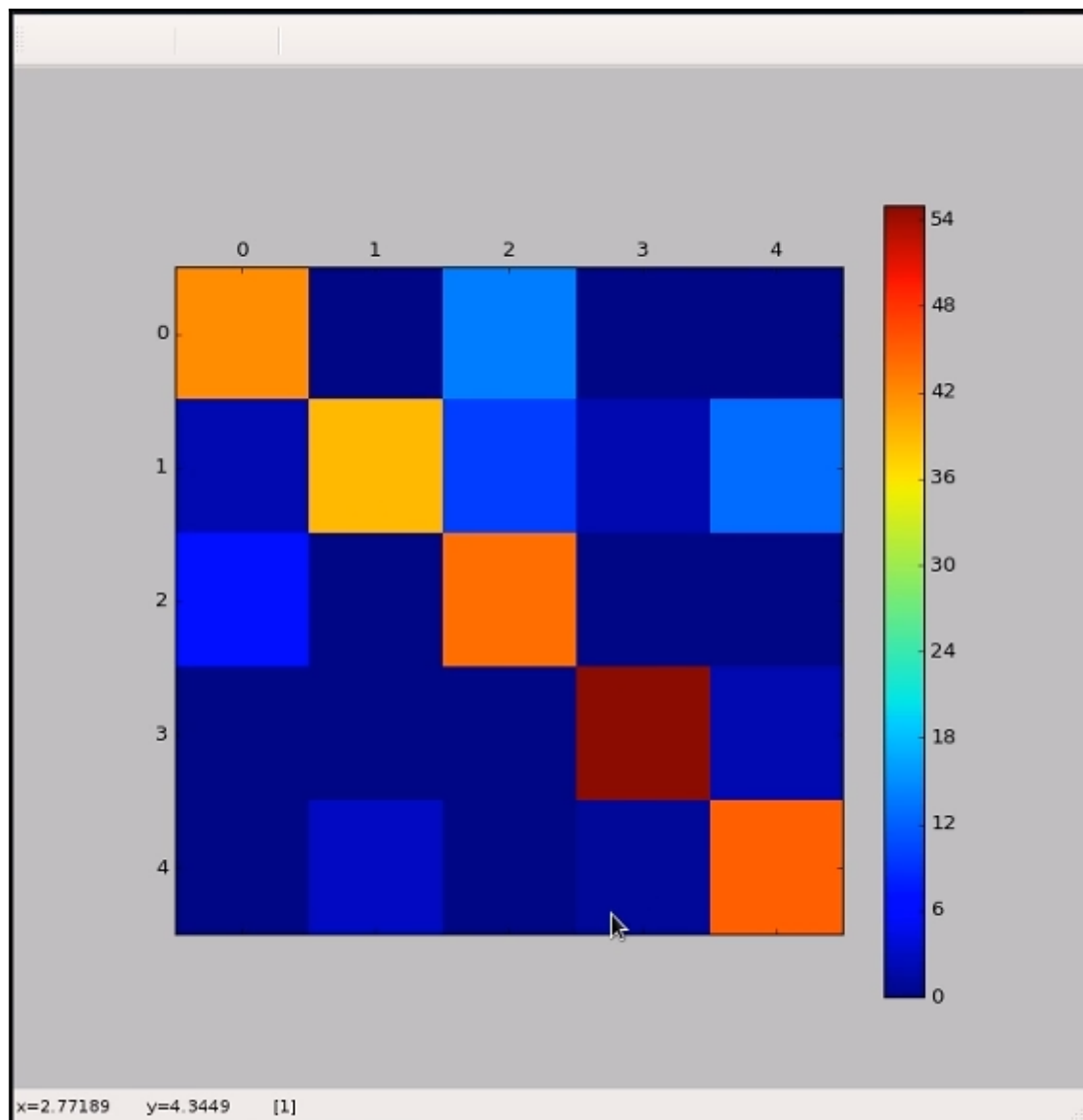


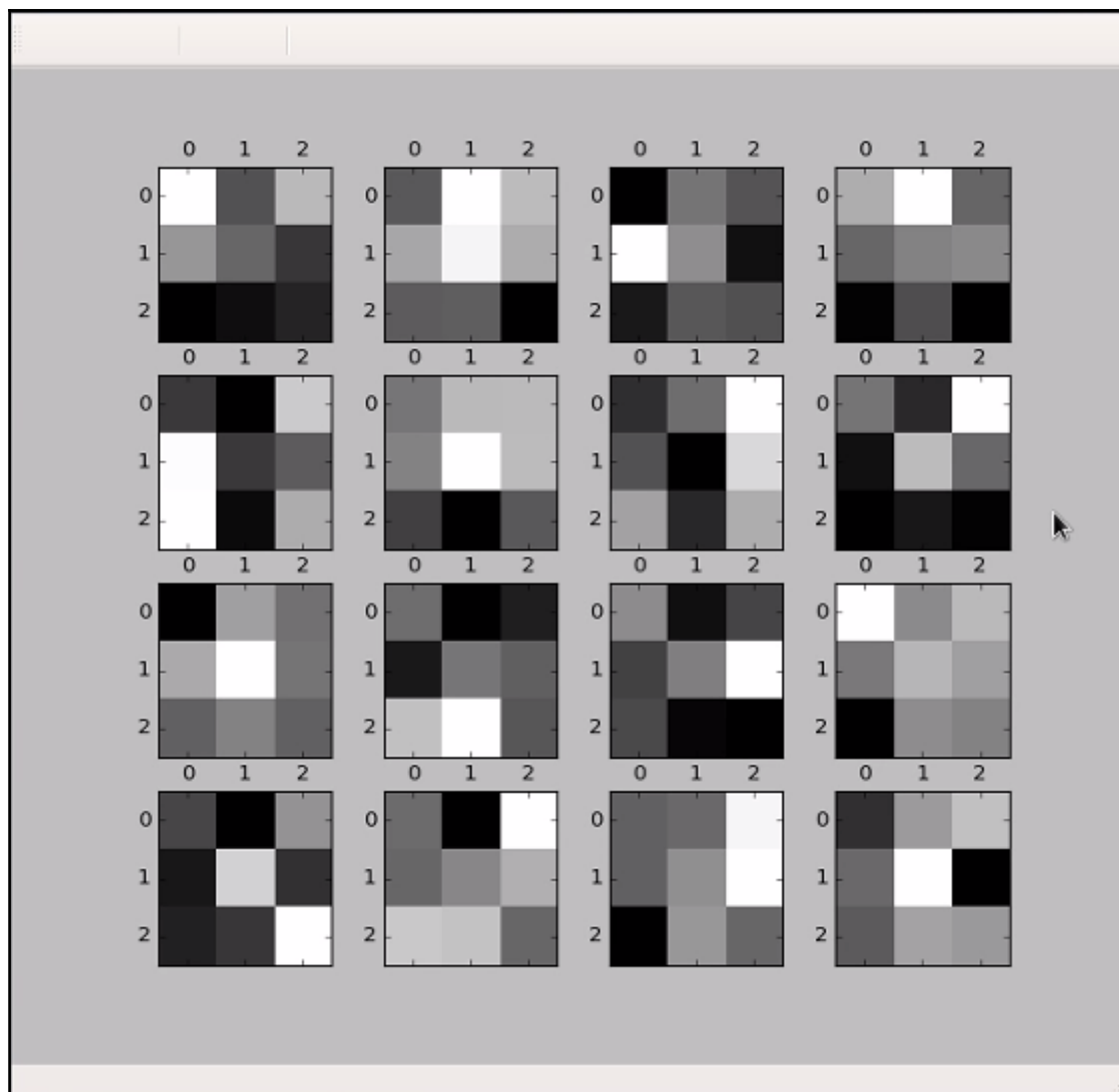


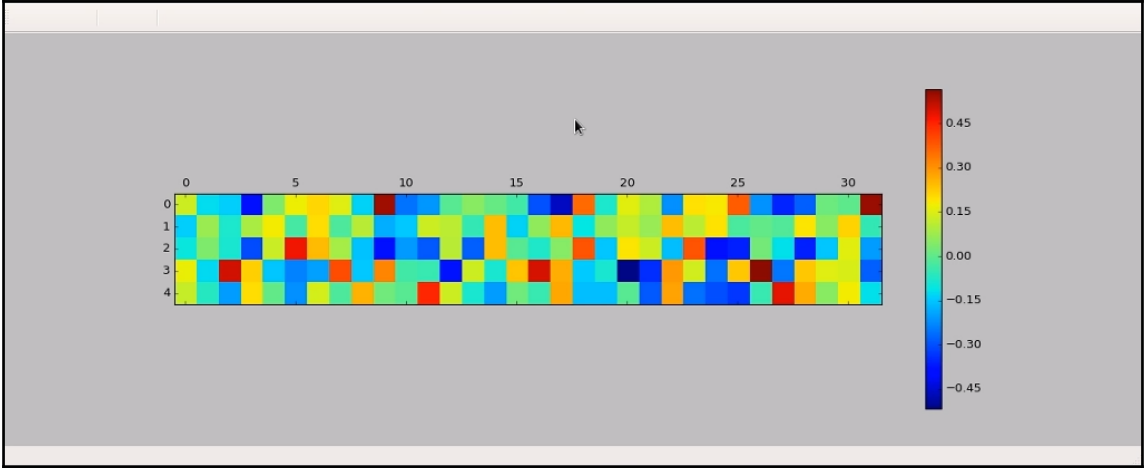




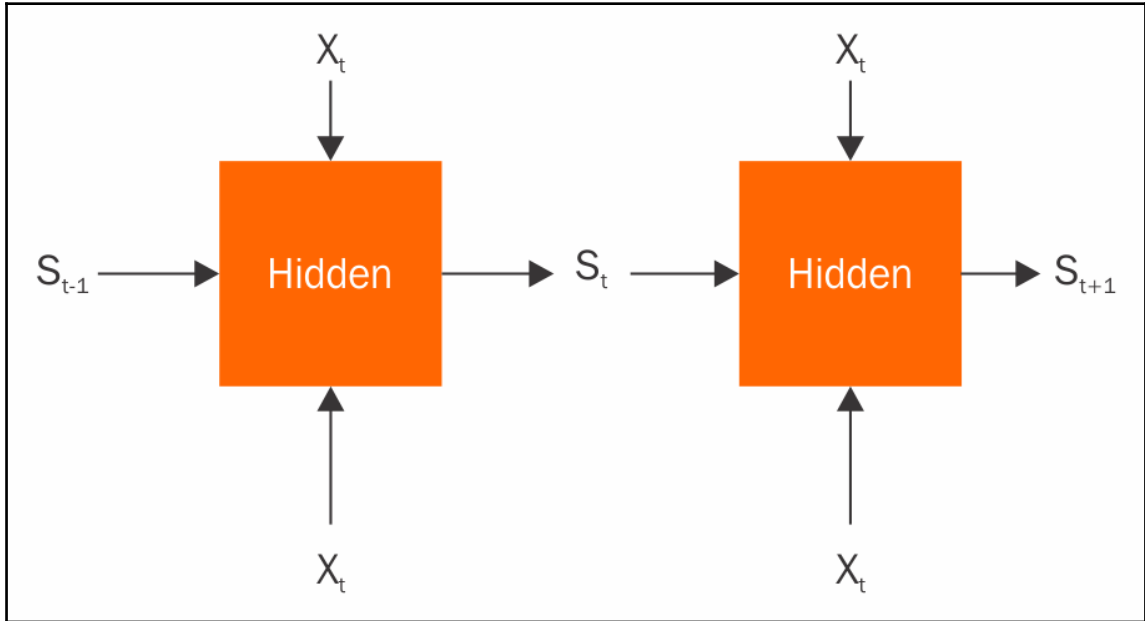


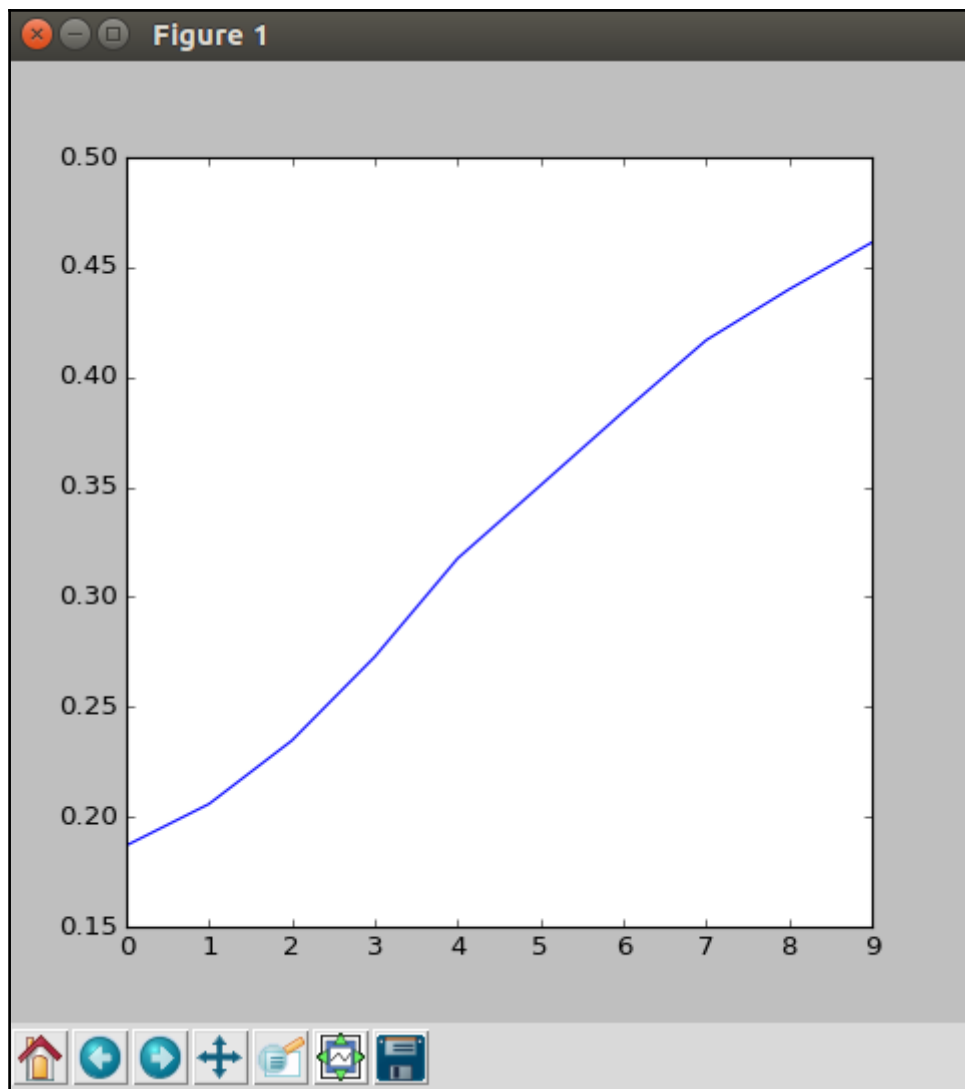




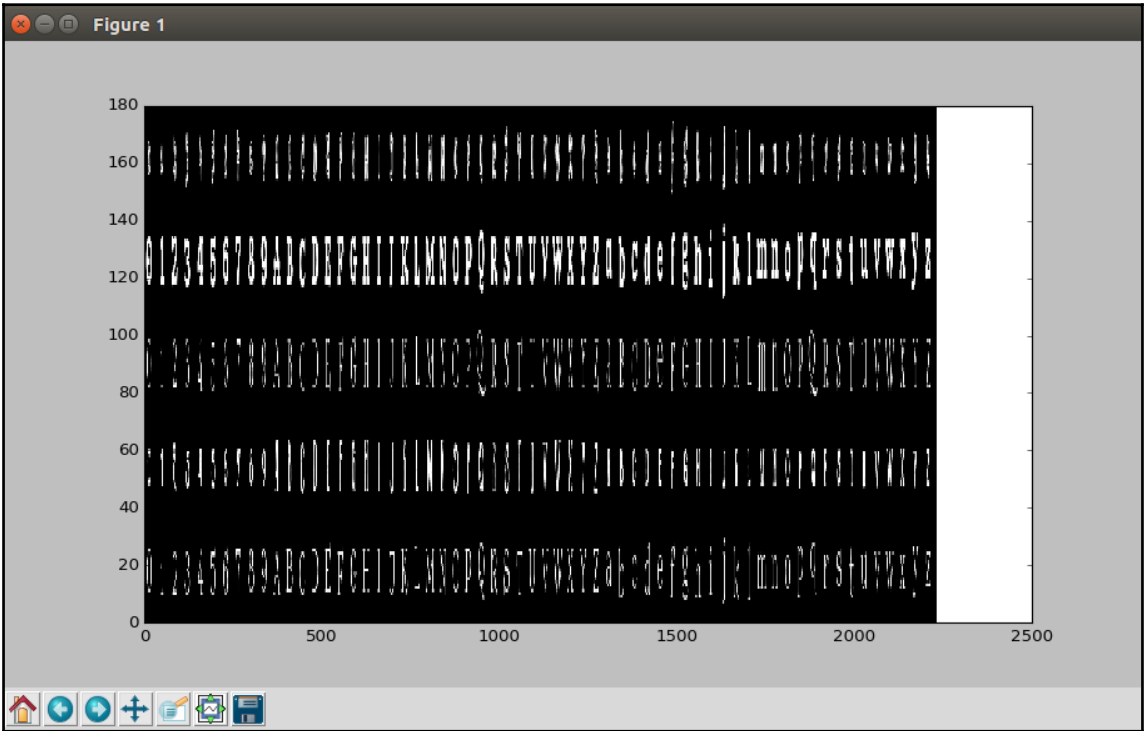


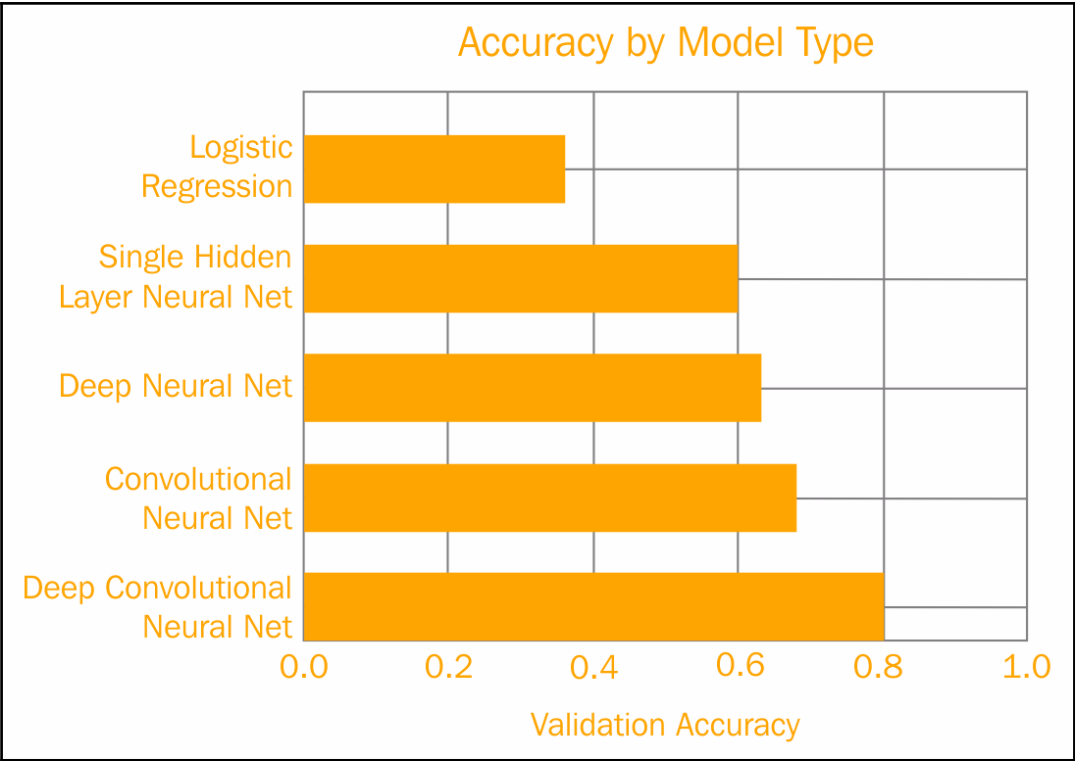
Chapter 4: Introducing Recurrent Neural Networks

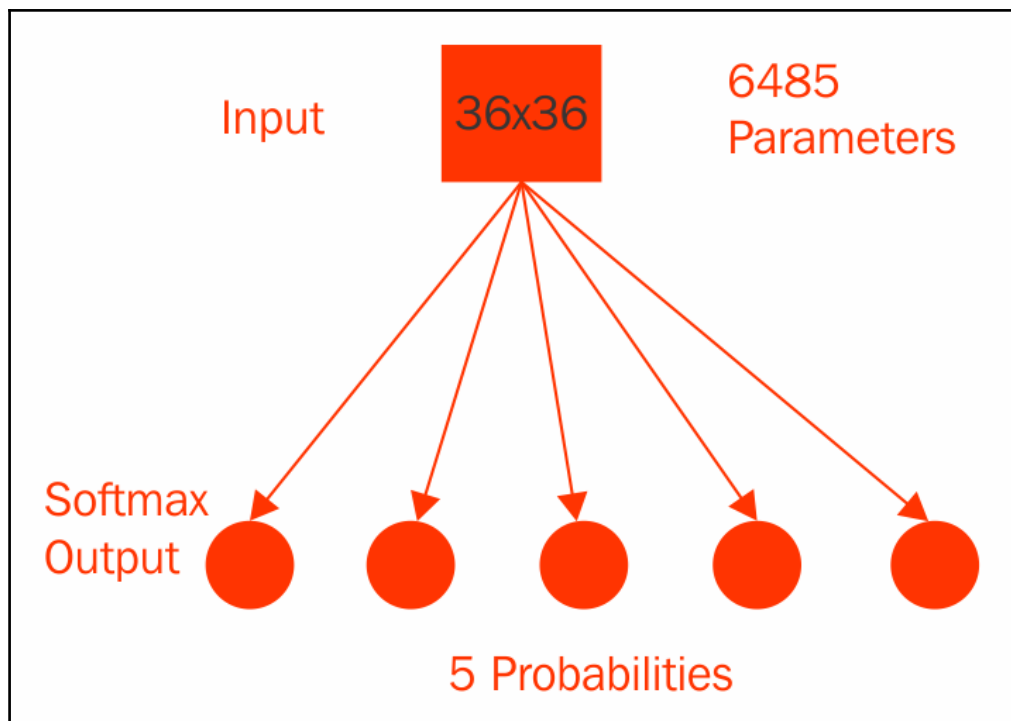


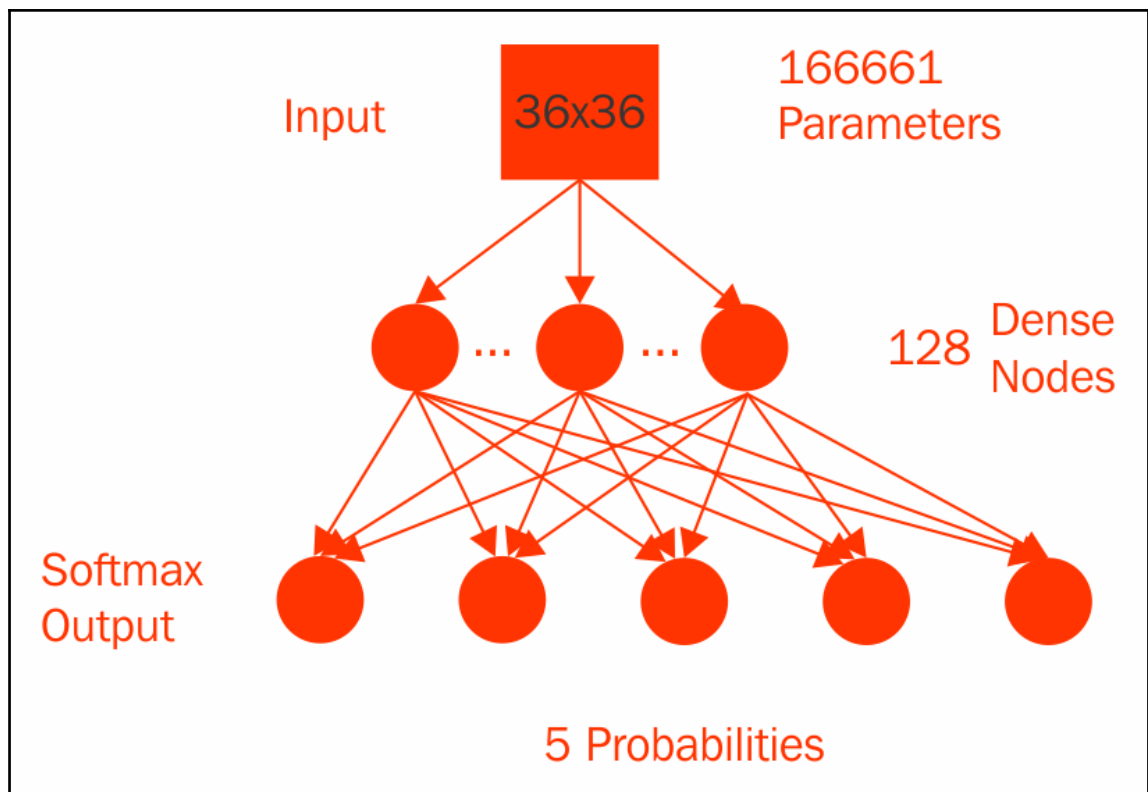


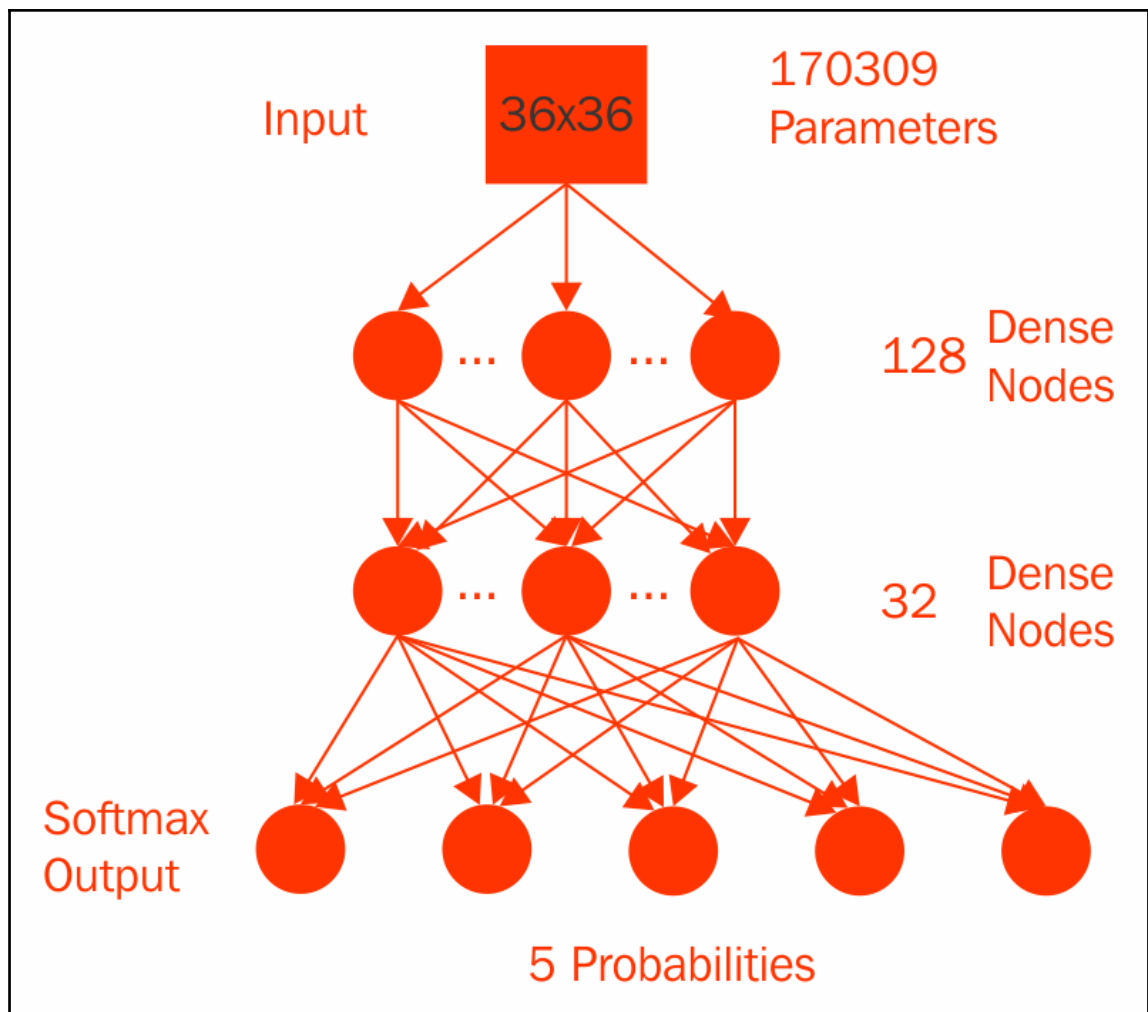
Chapter 5: Wrapping Up

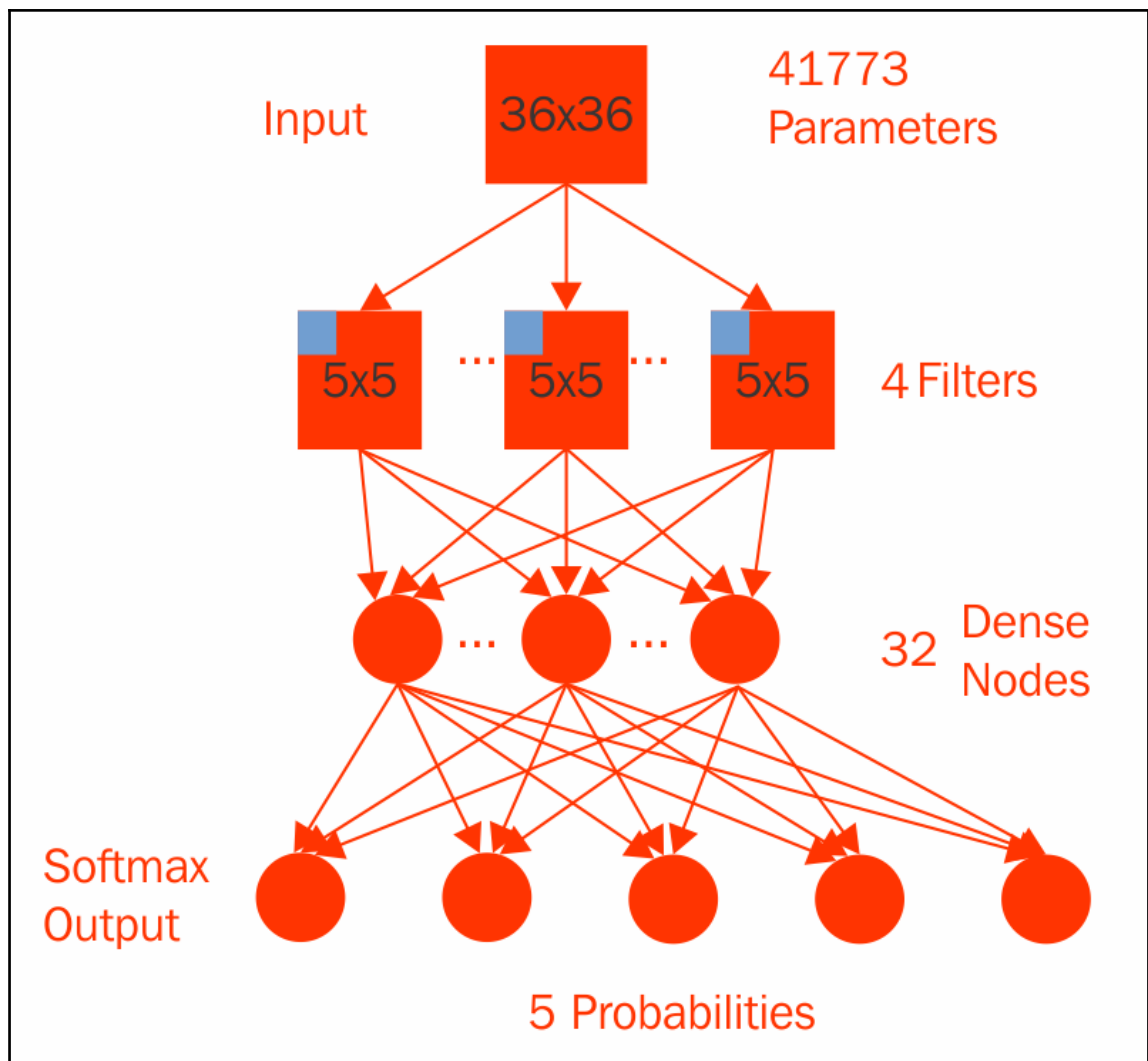


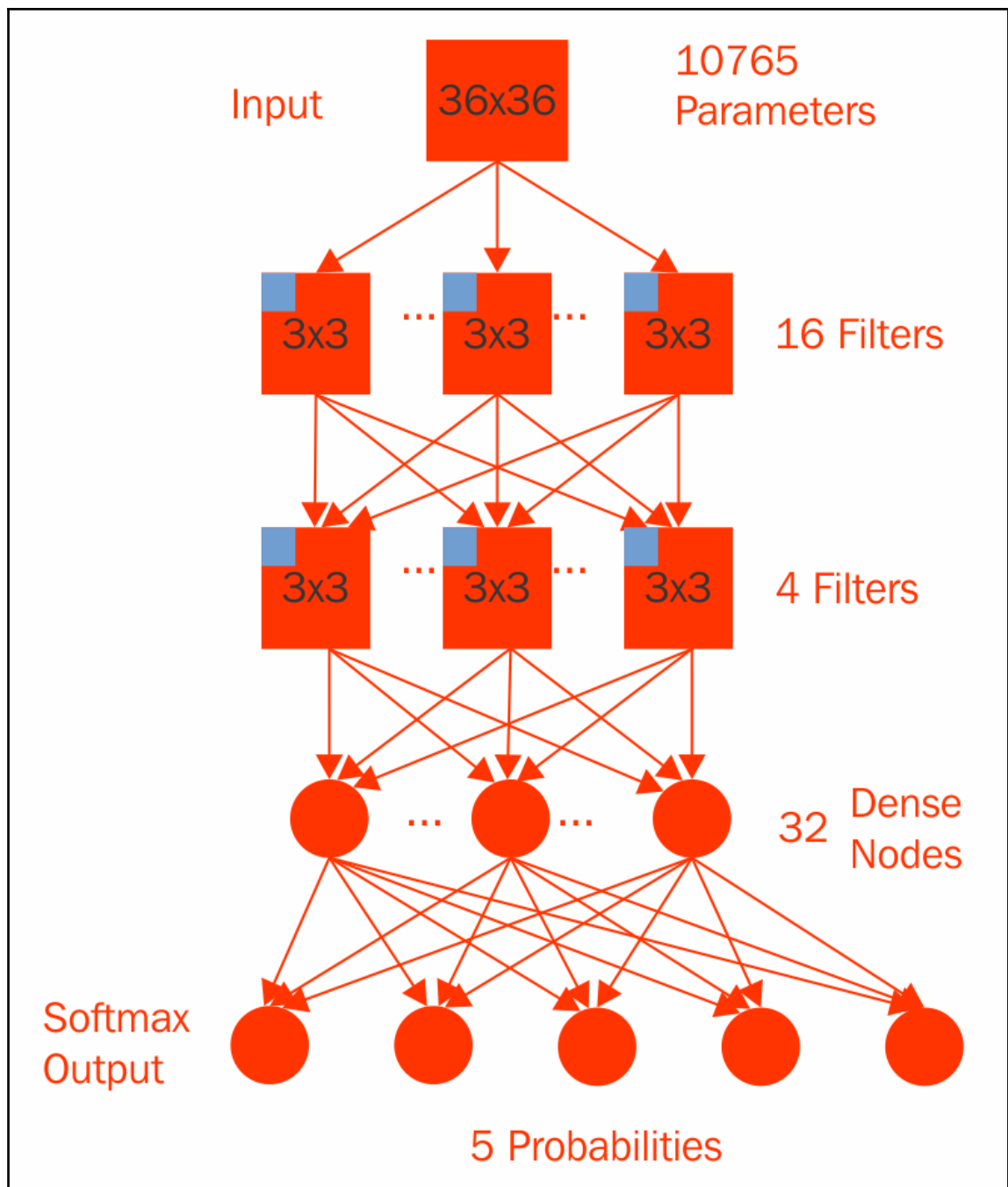













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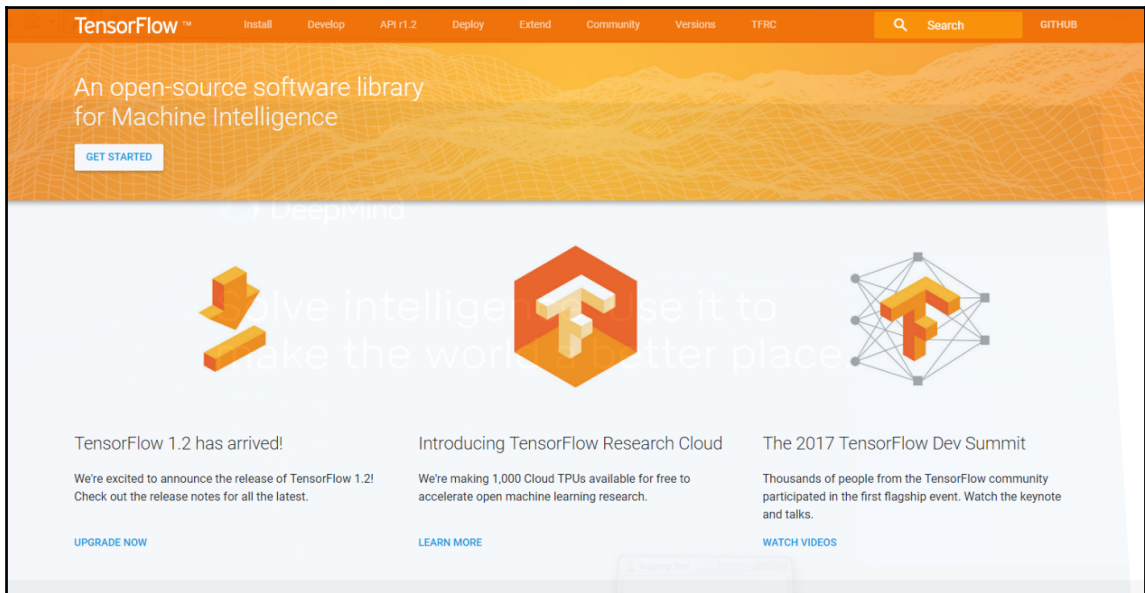
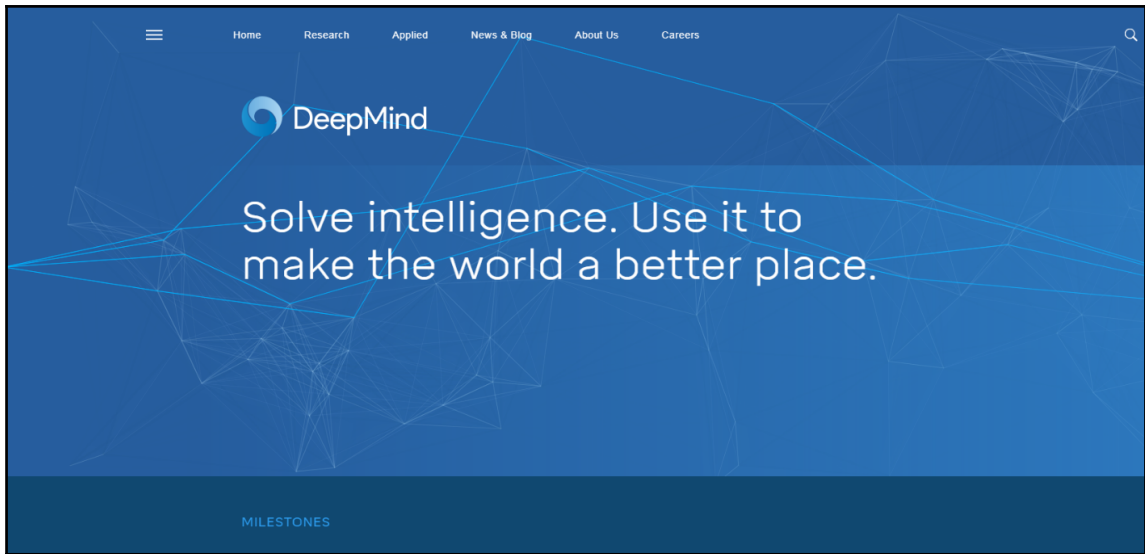
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 guscmue committed with gnan fix gpu build (#11647) Latest commit d3edb8c 3 hours ago

tensorflow	fix gpu build (#11647)	3 hours ago
third_party	Update toolchain configuration artifacts to work with latest version ...	10 days ago
tools	add a new config option sycl_nodouble for SYCL build (#11234)	6 days ago
util/python	Remove deleted files.	3 months ago
.gitignore	Merge changes from github.	2 months ago
ACKNOWLEDGMENTS	TensorFlow: Improve performance of Alexnet	2 years ago
ADOPTERS.md	Internal file cleanup.	9 months ago
AUTHORS	Merge changes from github.	a year ago
BUILD	Depend on protobuf's header only library when building custom ops	5 months ago
CODEOWNERS	Update CODEOWNERS, add owners to Windows, TPU components (#11485)	8 days ago
CODE_OF_CONDUCT.md	Add a code of conduct. (#11458)	a day ago
CONTRIBUTING.md	Merge changes from github.	23 days ago
ISSUE_TEMPLATE.md	Merge changes from github.	10 days ago




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