\*Caffe: Convolutional Architecture for Fast Feature Embedding

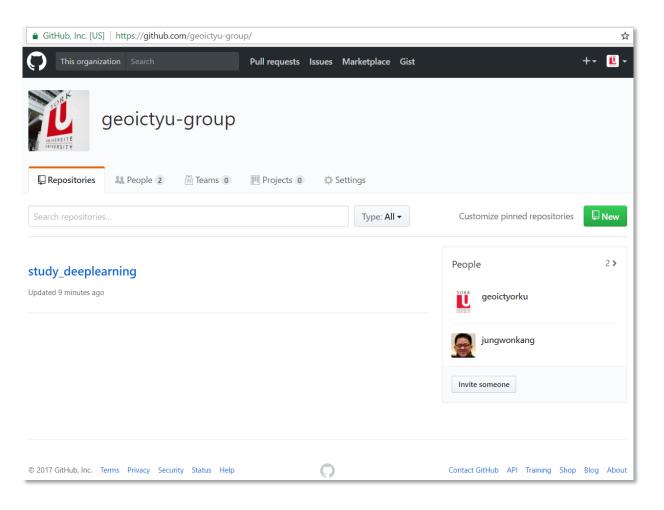
# **Caffe Programming Guide**

June 15 2017

Jungwon Kang

# Our Group Site Opens

https://github.com/geoictyu-group/



## **Presentation Schedule**

Date	Topic
6/1	Starting Caffe  • Caffe Installation  • Caffe Programming Overview
6/15	Caffe Programming Guide
TBD	Object Detection Using Caffe

## Contents

1. Review

2. Caffe Programming

■ 3. Conclusion

## Installation of Caffe

#### Recommendation

- NVIDIA Graphic Card (such as NVIDIA TITAN or GeForce Series)
- Linux: Ubuntu 14.04 or later

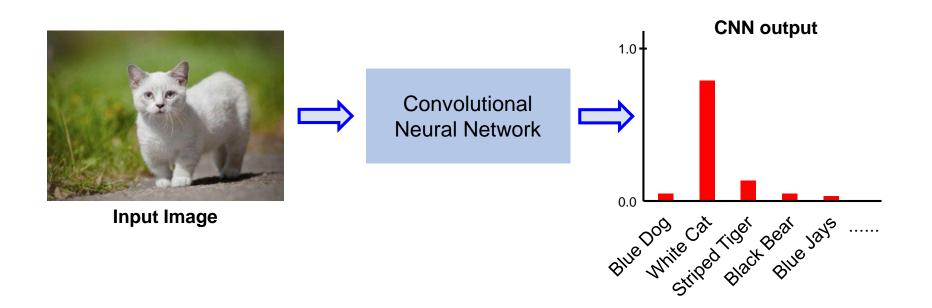
### • If not familiar with linux, you can install Caffe on Windows:

- [Code] Microsoft/caffe: Caffe on both Linux and Windows <a href="https://github.com/Microsoft/caffe">https://github.com/Microsoft/caffe</a>
- [Guide] How to install Caffe in windows in 5 min <u>https://youtu.be/nrzAF2sxHHM</u>
- [Guide] How to Build Caffe, Pycaffe, Matcaffe on Windows 10 <a href="https://youtu.be/AN2uXGRvw9E">https://youtu.be/AN2uXGRvw9E</a>

#### Other choices for Windows users:

Tensorflow, MatConvNet (Matlab), Keras (Wrapper for Tensorflow & Theano)

# Image Classification



# Common Implementation Using Caffe

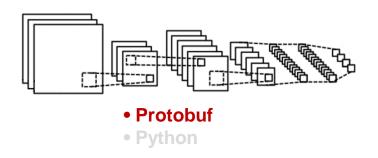
#### **Preparing Training Database**

# Images Blue Jays Black Bear Yellow Dog Striped Tiger Giraffe ... LMDB LevelDB Raw images & labels

(only if network definition is done using protobuf)

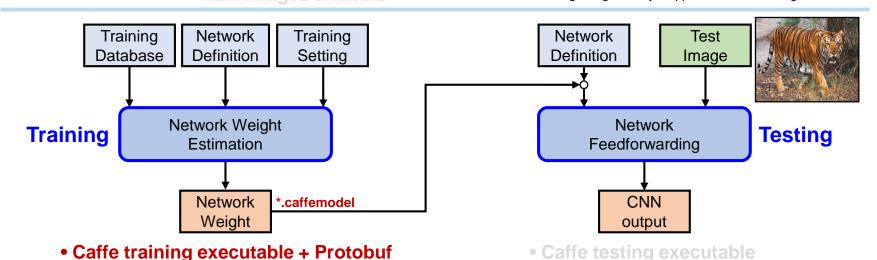
Python

#### **Defining a Network**



Protobuf: Protocol Buffers (Developed by Google)
LMDB: Lightning Memory-Mapped Database Manager

Python



## Contents

■ 1. Review

2. Caffe Programming

• 3. Conclusion

# Step 1: Preparing Training Database

#### **Preparing Training Database**

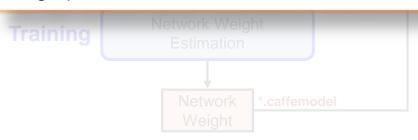
# Images Blue Jays Black Bear Yellow Dog Striped Tiger Giraffe ... LMDB LevelDB Raw images & labels

Defining a Network

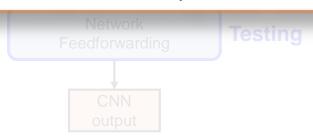


Protobu
 Puthon

What is LMDB? (https://symas.com/lightning-memory-mapped-database/)
Lightning Memory-Mapped Database (LMDB) is a software library that provides
a high-performance embedded transactional database in the form of a key-value store.



- Caffe training executable + Protobuf (only if network definition is done using protobuf)
- Pvthon



- Caffe testing executable (only if network definition is done using protobu
- Python

# **Preparing Training Database**

Preparing annotation file

Task	Annotation	(Common) Annotation file
Image classification	Class of each image	txt file
Object detection (Bounding box estimation)	Position of each bounding box in an image	JSON file
Image segmentation (Pixel-wise labeling)	Class of each pixel in an image	Labeled image file

- Better way for annotation?
   <a href="https://www.oreilly.com/ideas/data-preparation-in-the-age-of-deep-learning">https://www.oreilly.com/ideas/data-preparation-in-the-age-of-deep-learning</a>
- Converting annotated data(images & annotation file) into LMDB
  - Caffe tools: convert\_imageset.cpp (for image classification example)
  - LMDB: a directory containing data.mdb, lock.mdb

JSON: JavaScript Object Notation

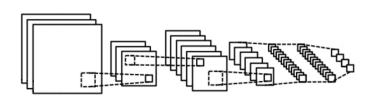
LMDB: Lightning Memory-Mapped Database Manager

# Step 2: Defining a Network

#### **Preparing Training Database**

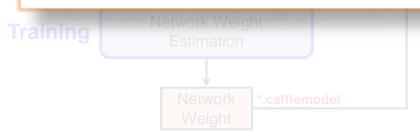


#### **Defining a Network**

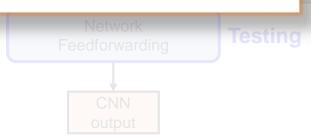


- Protobuf
- Python

What is Protocol buffers? (https://developers.google.com/protocol-buffers/) Protocol buffers are Google's language-neutral, platform-neutral, extensible mechanism for serializing structured data.



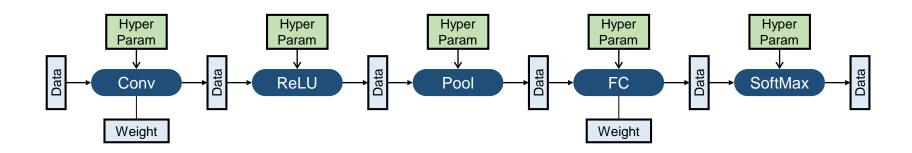
- Caffe training executable + Protobuf (only if network definition is done using protobuf)
- Pvthon



- Caffe testing executable (only if network definition is done using protobuf)
- Python

## **CNN**

CNN: a Set of 'Data Transformation'

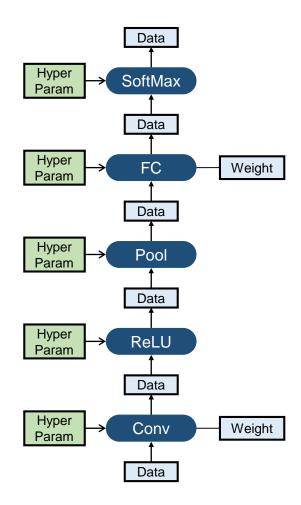


OOOO : Transformation

## CNN

CNN: a Set of 'Data Transformation'

Just re-drawn as *bottom-top* style description for Caffe notation →

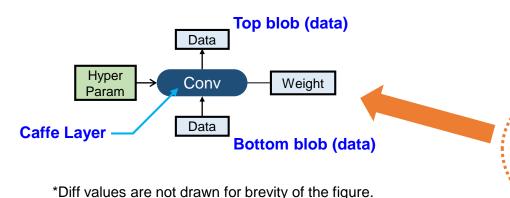


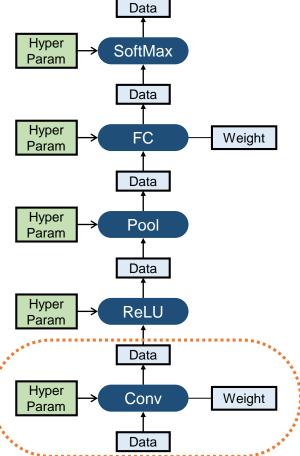
## CNN in Caffe (1/5)

- CNN in Caffe: a Set of Blob for Data & Caffe Layer for Transformation
  - Blob
    - ✓ N-D arrays for storing values
    - ✓ Includes data, weight and their diff values

#### Caffe Layer

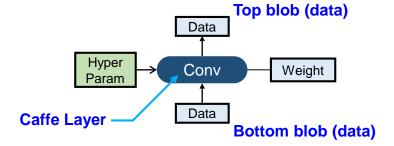
- ✓ Fundamental computation unit in Caffe
- ✓ Transforms bottom blobs to top blobs.
- ✓ Not equal to conventional neural network layer.
- \* Loading input data is also managed by Caffe layer.





## CNN in Caffe (2/5)

- Describing a Caffe Layer for Your CNN
  - Layer name (made by you)
  - Layer type
  - Bottom blob
  - Top blob
  - Hyper params



In your\_cnn.prototxt,

```
layer {
  name: "conv1"
  type: "Convolution"
  bottom: "data1"
  top: "conv1"
  convolution_param {
    num output: 20
    kernel size: 5
    weight_filler {
      type: "xavier"
```

## CNN in Caffe (3/5) [optional]

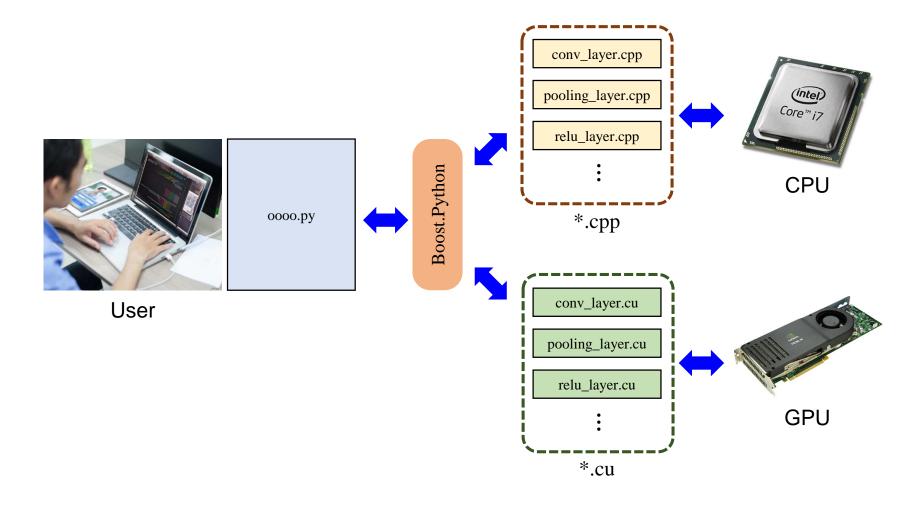
- Where is a low-level code for a Caffe Layer ?
  - Layer (Each implemented by Class)
    - √ header:
      - include/caffe/layers/oooo\_layer.hpp
    - ✓ source:
      - [For CPU] src/caffe/layers/oooo\_layer.cpp
      - [For GPU] src/caffe/layers/oooo\_layer.cu
    - ✓ major member functions
      - Reshape()
      - Forward(): Forward\_cpu(), Forward\_gpu()
      - Backward(): Backward\_cpu(), Backward\_gpu()
  - Hyper params
    - ✓ proto/caffe.proto

In your\_cnn.prototxt,

```
layer {
 name: "conv1"
  type: "Convolution"
 bottom: "data1"
  top: "conv1"
  convolution_param {
    num output: 20
    kernel size: 5
    weight_filler {
      type: "xavier"
```

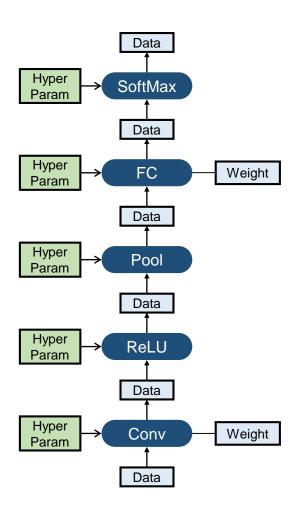
## CNN in Caffe (4/5) [optional]

How does python wrapper link with a Caffe Layer ?



## CNN in Caffe (5/5)

Defining a Complete CNN = Describing Whole Caffe Layers for the CNN



In your\_cnn.prototxt,

```
layer {
 name: "data"
 type: "Data"
 top: "data"
layer {
 name: "conv1"
 type: "Convolution"
 bottom: "data"
 top: "conv1"
layer {
 name: "relu1"
 type: "ReLU"
 bottom: "conv1"
 top: "conv1"
layer {
 name: "pool1"
 type: "Pooling"
 bottom: "conv1"
 top: "pool1"
layer {
 name: "fc1"
 type: "InnerProduct"
 bottom: "pool1"
 top: "fc1"
layer {
 name: "prob"
 type: "Softmax"
 bottom: "fc1"
  top: "prob"
```

# Step 3: Training

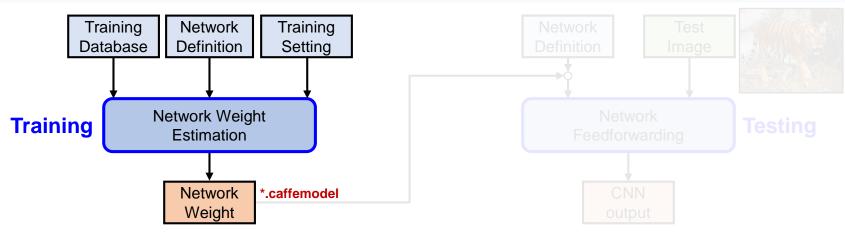
#### **Preparing Training Database**



#### Defining a Network



- Protobuf
- Pvthon



- Caffe training executable + Protobuf (only if network definition is done using protobuf)
- Python

- Carre testing executable (only if network definition is done using protobut
- Python

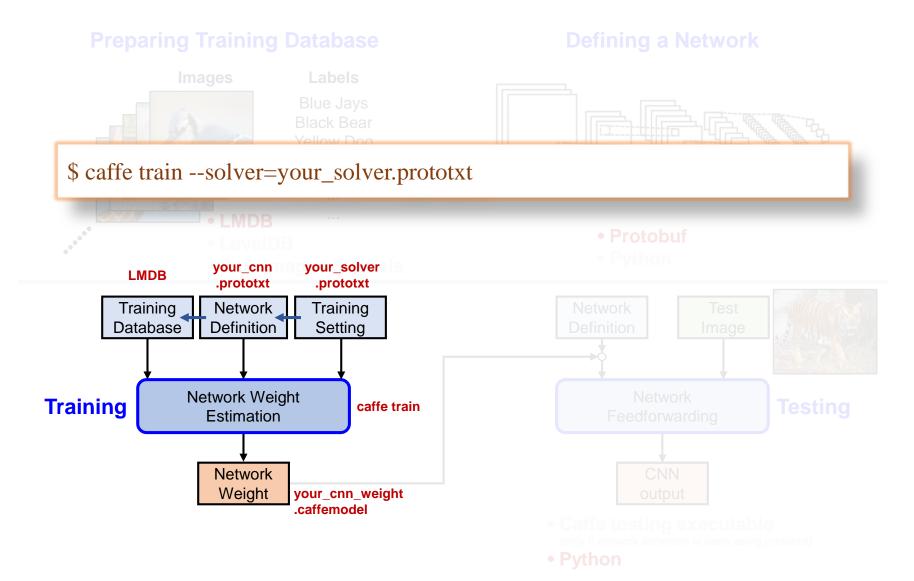
# Training Using caffe Tool

- One Line Command for Training
  - \$ caffe train --solver=your\_solver.prototxt

#### your\_solver.prototxt

```
# The train/test net protocol buffer definition
train net: "mnist/lenet auto train.prototxt"
test net: "mnist/lenet auto test.prototxt"
# test iter specifies how many forward passes the test should carry out.
test iter: 100
# Carry out testing every 500 training iterations.
test interval: 500
# The base learning rate, momentum and the weight decay of the network.
base 1r: 0.01
momentum: 0.9
weight decay: 0.0005
# The learning rate policy
lr policy: "inv"
gamma: 0.0001
power: 0.75
# Display every 100 iterations
display: 100
# The maximum number of iterations
max iter: 10000
# snapshot intermediate results
snapshot: 5000
snapshot prefix: "your dir/lenet"
# solver mode: CPU or GPU
solver mode: GPU
```

# Training Using caffe Tool



# Step 4: Testing

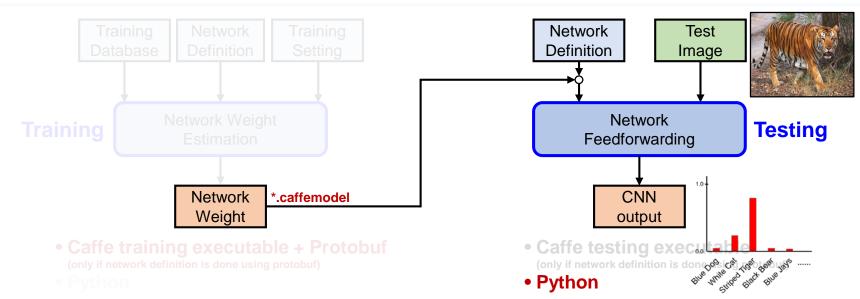
#### **Preparing Training Database**



#### Defining a Network



- Protobuf
- Pvthon



# Your Network Loaded in Python

- Blobs : Stored in Python 'ordered dictionary'
  - ex) net.blobs['conv1']
- Data in Blobs : Stored in Python NumPy's 'ndarray'
  - ex) net.blobs['conv1'].data

```
net = {Net} < caffe. caffe.Net object at 0x7f96e58fd788>
         ▶ ■ _blob_loss_weights = {DtypeVec} < caffe. _caffe. DtypeVec object at 0x7f96e673af30>
         blob_names = {StringVec} <caffe._caffe.StringVec object at 0x7f96e5b6ddd0>
         _blobs = {BlobVec} <caffe._caffe.BlobVec object at 0x7f96e5be2ad0>
         ■ _inputs = {IntVec} <caffe._caffe.IntVec object at 0x7f96e58d56d0>
         ■ _layer_names = {StringVec} <caffe._caffe.StringVec object at 0x7f96e58d5750>
         _outputs = {IntVec} <caffe._caffe.IntVec object at 0x7f96e58d57d0>
         ▶ ■ blob_loss_weights = {OrderedDict} OrderedDict([('data', 0.0), ('conv1', 0.0), ('pool1', 0.0), ('conv2', 0.0), ('pool2', 0.0), ('ip1', 0.0), ('ip2', 0.0), ('prob', 0.0)])
         ▼ ≣ blobs = {OrderedDict} OrderedDict([('data', <caffe._caffe.Blob object at 0x7F96e58deb90>), ('conv1', <caffe._caffe.Blob object at 0x7F96e58deb90>), ('
                               M __len__ = {int} 8
                    " = 'conv1' (140286073203616) = {Blob} < caffe._caffe.Blob object at 0x7f96e58deb18>
                                        In channels = {int} 20
                             ▶ ■ data = {ndarray}[[[[ 0. 0. 0. ..., 0. 0. 0.]\n [ 0. 0. 0. ...]\n 
                             ▶ ■ diff={ndarray}[[[[0. 0. 0. ..., 0. 0. 0.]\n [0. ...\n [0. ...\n [0. ...\n [0. ...\n [0. ...]\n [0. ..
                                         M height = {int} 24
                                         In num = {int} 1
                             ▶ = shape = {IntVec} < caffe. caffe.IntVec object at 0x7f96e5b6ddd0>

    width = {int} 24

                    ▶ ≡ 'ip1' (140286073501056) = {Blob} < caffe._caffe.Blob object at 0x7f96e58decf8>
                    ▶ ≡ 'ip2' (140286073501136) = {Blob} < caffe. caffe. Blob object at 0x7f96e58ded70>
                    ▶ | 'pool1' (140286073203664) = {Blob} < caffe. caffe. Blob object at 0x7f96e58deb90>
                    ▶ ■ 'pool2' (140286073203760) = {Blob} < caffe._caffe.Blob object at 0x7f96e58dec80>
                    "prob" (140286073203808) = {Blob} < caffe. caffe.Blob object at 0x7f96e58dede8>
                    ► ■ _OrderedDict_map = {dict} {'ip2': [[[[[[[[[...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...], [...
                    _OrderedDict_marker = {object} <object object at 0x7f972f3d50a0>
                    ▶ 월 _OrderedDict__root = {\text{list}} < type '\text{list}' > :[[[[[[[[...], [...], 'data'], [...], 'conv1'], [...], 'conv1'], [...], 'conv2'], [...], 'pool2'], '
```

# **Testing Using Python**

- Procedure for Testing
  - (1) Load a network definition & its weight
  - (2) Load a test image & Put the image into the network
  - (3) Perform feedforwarding
  - (4) Get result

your\_testing.py

```
import caffe
...
#(1)
model_def = caffe_root +'models/bvlc_alexnet.prototxt'
model_weights = caffe_root + 'models/bvlc_alexnet/bvlc_alexnet.caffemodel'
net = caffe.Net(model_def, model_weights, caffe.TEST)
...
#(2)
image = caffe.io.load_image(caffe_root + 'examples/images/cat.jpg')
net.blobs['data'].data[...] = image
...
#(3)
output = net.forward()
...
#(4)
output_prob = output['prob'][0]
print 'predicted class is:', output_prob.argmax()
...
```

• Partially extracted from: <a href="http://nbviewer.jupyter.org/github/BVLC/caffe/blob/master/examples/00-classification.ipynb">http://nbviewer.jupyter.org/github/BVLC/caffe/blob/master/examples/00-classification.ipynb</a>

## Contents

■ 1. Review

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# What We Need for Caffe Programming

- Data Preparation
  - One of the most important things in deep learning
- Protobuf
  - Description grammar used for defining a network & descripting a training setting
- Python
  - Major language for most deep learning frameworks including Caffe

## References (1/3)

#### General Introduction

- Stanford cs231n
   <a href="http://vision.stanford.edu/teaching/cs231n/slides/2015/caffe\_tutorial.pdf">http://vision.stanford.edu/teaching/cs231n/slides/2015/caffe\_tutorial.pdf</a>
- DIY Deep Learning for Vision-a Hands-On Tutorial with Caffe <a href="http://tutorial.caffe.berkeleyvision.org/">http://tutorial.caffe.berkeleyvision.org/</a>
- A Practical Introduction to Deep Learning with Caffe <a href="http://www.panderson.me/images/Caffe.pdf">http://www.panderson.me/images/Caffe.pdf</a>

### PyCaffe

- Deep learning tutorial on Caffe technology: basic commands, Python and C++ code
   <a href="http://christopher5106.github.io/deep/learning/2015/09/04/Deep-learning-tutorial-on-Caffe-Technology.html">http://christopher5106.github.io/deep/learning/2015/09/04/Deep-learning-tutorial-on-Caffe-Technology.html</a>
- Deep Learning With Caffe In Python <u>https://prateekvjoshi.com/2016/02/02/deep-learning-with-caffe-in-python-part-i-defining-a-layer/</u>

## References (2/3)

- PyCaffe (Conti.)
  - Tutorial for pycaffe, the Python API to the Neural Network framework, Caffe
    - https://github.com/nitnelave/pycaffe\_tutorial
  - A Practical Introduction to Deep Learning with Caffe and Python <a href="http://adilmoujahid.com/posts/2016/06/introduction-deep-learning-python-caffe/">http://adilmoujahid.com/posts/2016/06/introduction-deep-learning-python-caffe/</a>
  - pyCaffe Tools, Examples and Resources
     http://davidstutz.de/pycaffe-tools-examples-and-resources/
     https://github.com/davidstutz/caffe-tools
  - Training Multi-Layer Neural Network with Caffe
     <a href="http://nbviewer.jupyter.org/github/joyofdata/joyofdata-articles/blob/master/deeplearning-with-caffe/Neural-Networks-with-Caffe-on-the-GPU.ipynb">http://nbviewer.jupyter.org/github/joyofdata/joyofdata-articles/blob/master/deeplearning-with-caffe/Neural-Networks-with-Caffe-on-the-GPU.ipynb</a>

## References (3/3)

- Caffe Layer
  - Developing new layers <u>https://github.com/BVLC/caffe/wiki/Development</u>
  - Making a Caffe Layer <a href="https://chrischoy.github.io/research/making-caffe-layer/">https://chrischoy.github.io/research/making-caffe-layer/</a>
  - How to create your own layer in deep learning framework CAFFE <a href="https://yunmingzhang.wordpress.com/2015/01/19/how-to-create-your-own-layer-in-deep-learning-framework-caffe/">https://yunmingzhang.wordpress.com/2015/01/19/how-to-create-your-own-layer-in-deep-learning-framework-caffe/</a>
  - Simple Example: Sin Layer <a href="https://github.com/BVLC/caffe/wiki/Simple-Example:-Sin-Layer">https://github.com/BVLC/caffe/wiki/Simple-Example:-Sin-Layer</a>