

Caffe: Convolution Architecture For Feature Extraction

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Convolution Architecture For Feature Extraction

- C++/CUDA framework for deep learning and vision
 - library of layers that compose into models
 - fast stochastic gradient descent (SGD) solver
 - tools, demos, and recipes
- Seamless switch between CPU and GPU
 - Caffe::set_mode(Caffe::CPU);

Train Models



Experiment/Prototype



Inference at Scale



All with essentially the same code!

Convolution Architecture For Feature Extraction

- Model schemas
 - Define the model and solving strategy and let Caffe take care of the rest
 - Adapt already learned models to new problems in one step

State-of-the-art solving in 14 lines.

1	<pre>train_net: "imagenet_train.prototxt"</pre>
2	<pre>test_net: "imagenet_val.prototxt"</pre>
3	test_iter: 1000
4	test_interval: 1000
5	base_lr: 0.01
6	lr_policy: "step"
7	gamma: 0.1
8	stepsize: 100000
9	display: 20
10	max_iter: 450000
11	momentum: 0.9
12	weight_decay: 0.0005
13	snapshot: 10000
14	<pre>snapshot_prefix: "caffe_imagenet_train"</pre>

Models are schema, not code.

```
1 name: "CaffeNet"
 2 input: "data"
 3 input_dim: 10
   input_dim: 3
   input_dim: 227
 6 input_dim: 227
   # convolution 1: 96 filters
  layers {
     layer {
10
       name: "conv1"
11
       type: "conv"
12
       num_output: 96
13
       kernelsize: 11
14
       stride: 4
15
       weight_filler {
         type: "gaussian"
17
         std: 0.01
18
19
       bias_filler {
         type: "constant"
20
21
         value: 0.
22
       }
23
       blobs_lr: 1.
24
       blobs_lr: 2.
25
       weight_decay: 1.
26
       weight_decay: 0.
27
     bottom: "data"
     top: "conv1"
29
```

Convolution Architecture For Feature Extraction

- Research & Engineering
 - Key part of our publication code
 - State-of-the-art models
 - Blazing fast, and it has unit tests!



R-CNN, Girshick CVPR14.



Cuda ni	umber of devices: 2
Setting	g to use device Ø
Current	t device id: 0
] Running 10 tests from 2 test cases.
C-seeb] Global test environment set-up.
Entern] 5 tests from InnerProductLayerTest/0, where TypeParam = float
E RUN	InnerProductLayerTest/0.TestSetUp
E014-0:	OK] InnerProductLayerTest/0.TestSetUp (493 ms)
E RUN	InnerProductLayerTest/0.TestCPU
Eatz as	OK] InnerProductLayerTest/0.TestCPU (103 ms)
E RUN	InnerProductLayerTest/0.TestGPU
	OK] InnerProductLayerTest/0.TestGPU (0 ms)
E RUN	InnerProductLayerTest/0.TestCPUGradient
E014-00	OK] InnerProductLayerTest/0.TestCPUGradient (1492 ms)
E RUN	InnerProductLayerTest/0.TestGPUGradient
East as	OK] InnerProductLayerTest/0.TestGPUGradient (217 ms)
Latara] 5 tests from InnerProductLayerTest/0 (2305 ms total)

Convolution Architecture For Feature Extraction

• An active research and development community

527 commits	★ Star 232 ³ / ₂ Fork 11	3	
February 12 2014 - Marcl	า 12 2014		Period: 1 month -
Overview			
46 Active Pull Requests		62 Active Issues	
î 23 Merged Pull Requests	23 Proposed Pull Requests	Closed Issues	① 25 New Issues

23 authors have pushed 511 commits to all branches, excluding merges. On master, 137 files have changed and there have been 7,376 additions and 1,436,167 deletions.



Caffe and cuda-convnet

Caffe

- C++/CUDA deep learning and vision
 - library of layers
 - fast, general-purpose for ILSVRC, PASCAL, your data
- An active research and development community: public on GitHub
- Seamless GPU/CPU switch
- Model schemas
 - Define the model
 - Configure solver
 - Finetuning
- Wrappers for Python and MATLAB

cuda-convnet

- C++/CUDA deep learning and vision
 - library of layers
 - highly-optimized for given case: image, kernel, and batch size.
- Static codebase, no community contributions: last update Jul 17, 2012
- GPU only
- Model schemas
 - Define the model
 - Write and run solver command
 - No finetuning
- No wrappers: monolithic

Why not live caffeine-free?

- It's all about speed.
- cuda-convnet and DeCAF are awesome
 - but cuda-convnet is inflexible
 - and DeCAF is too slow
- Caffe is fast
 - with CPU: 2x speedup over DeCAF
 - with GPU: 10x speedup (under C++)
- Forward pass of a single image takes 2.5ms
 - Caffe reference ImageNet model with ~60 million parameters
 - (when in batch mode)
 - (~20ms in CPU mode)

A Caffe Net



Blob: all your data, derivatives, and parameters.

- example input blob (256 images, RGB, height, width)
 ImageNet training batches: 256 x 3 x 227 x 227
- example convolutional parameter blob
 - 128 filters with 96 input channels:128 x 96 x 3 x 3

A Layer

- The layer is the fundamental unit of computation.
- Caffe nets are composed of layers as defined in model schema.



A Layer defines...

- Forward: given input, computes the output. —
- Backward: given the gradient w.r.t. the output, compute the gradient w.r.t. the input and its internal parameters.
- Setup: how to initialize the layer.



Definition of a Net

Model schema are defined as Protocol Buffers:

```
message NetParameter {
  optional string name = 1;
  repeated LayerConnection layers = 2;
  repeated string input = 3;
  repeated int32 input_dim = 4;
}
```

schema definition at /src/caffe/proto/caffe.proto Protocol Buffer documentation:

https://developers.google.com/protocol-buffers/ docs/overview

name: "linear regressor" input: "data" input dim: 1 input dim: 3 input dim: 28 input dim: 28 layers { layer { name: "ip" type: "innerproduct" num output: 10 bottom: "data" top: "prediction" }

Definition of a Net

```
name: "mnist-small"
# data layer for input
layers {
 layer {
  name: "mnist"
  type: "data"
  source: "data/mnist-train-leveldb"
  batchsize: 64
  scale: 0.00390625
 top: "data"
 top: "label"
# linear classifier by inner product
layers {
 layer {
  name: "ip"
  type: "innerproduct"
  num output: 10
  weight filler {
   type: "xavier"
 bottom: "data"
 top: "ip"
}
```

```
# softmax loss for training
# takes classifier output and labels
layers {
    layer {
        name: "prob"
        type: "softmax_loss"
    }
    bottom: "ip"
    bottom: "label"
}
```



How about ImageNet?

- It's another network definition... only this time a state-of-the-art model
- See caffe/models/imagenet.prototxt



Finetuning

- Once you have a model—like caffe_reference_imagenet_model you can solve many problems.
- Where training from scratch can fail for lack of sufficient data, finetuning can succeed.
- Rename the layers you need to change...
- ...and continue training.
- No coding needed.

. . .

layer {

}

name: "fc8" type: "innerproduct" num_output: 1000 layer {

}

name: "**fc8-t**" type: "innerproduct" num_output: **397**

. . .

A Few Practical Questions

- What's the shortest path to features?
 - Swap deep features into your pipeline without tears via the Caffe Reference ImageNet model.
 - Any layer can be extracted.
 - Prototype with Python and MATLAB wrappers.
- Do I have to train from scratch for every problem?
 - Not at all! Finetune learned models to new data and tasks.
 - Define a new model and solver.
 - Call ./finetune_net new_solver old_model # then get a cup of coffee
- What do I do with my own loss, special operation, or data format?
 - Well, this is trickier but doable.
 - Code the layers needed.
 - Define the model and carry on.

Questions!

Check out <u>caffe.berkeleyvision.org</u>, the Github repository <u>https://github.com/BVLC/caffe</u>, and our issue tracker <u>https://github.com/BVLC/caffe/issues</u> (but search before posting).

Try our examples and tutorials!