CaffeLink

MNIST example

RequirementandInitialization

This example shows how to use CaffeLink to build, train and test LeNet. It relies on examples from Caffe, which can be found here.

Except for successfully build Caffe and CaffeLink, for this example MNIST data are also needed. Those can be obtained using script from Caffe:

```
cd $CAFFE_ROOT
./data/mnist/get_mnist.sh
./examples/mnist/create mnist.sh
```

As is shown in MNIST example on Caffe web. This document is only concerned with CaffeLink, so for details regarding net design please refer to Caffe.

Caffe is a command line tool and it would be inefficient to somehow transmit all its output to *Mathematica*. So it is highly recommended to launch *Mathematica* session with terminal. CaffeLink also relies on stoud.

InitializeCaffeLink

The first think to do is load CaffeLink module and initialize it with path to caffe.proto protobuffer definition of all parameters Caffe uses.

```
In[27]:= Needs["CaffeLink`", FileNameJoin[{NotebookDirectory[], "..", "caffeLink.m"}]];
InitCaffeLinkModule["fel-lin/caffe/src/caffe/proto/caffe.proto"];
```

Before training, testing etc. CaffeLink library must be initialized. That is: choose data type (double or float), computing mode (GPU or CPU) and device ID for GPU mode.

```
initCaffeLink[True, True, 0]
```

Mode can be switched later, data type not.

Net

Building

A net is represented by its proto file. Path to or content of LeNet proto lenet_train_test.prototxt could be directly passed to CaffeLink, but it would be no fun. Functions NewNet, AddLayer, SetNetParam and SetLayerParam allow to gradually build the net. GetNetParamString then produces protobuffer string which is used to create net in CaffeLink.

This path should point to your mnist data set. Absolute path is recommended.

```
mnistPath = "/home/kerhy/fel-lin/caffe/examples/mnist/";
```

BuildLeNet

```
net = NewNet["LeNet"];
li = 0;
li++;
net = AddLayer[net, "DATA", "mnist"];
net = SetLayerParam[net, li, "top", {"data", "label"}];
net = SetLayerParam[net, li, "source", StringJoin[mnistPath, "mnist_train_lmdb"]];
net = SetLayerParam[net, li, "backend", "LMDB"];
net = SetLayerParam[net, li, "batch_size", 64];
net = SetLayerParam[net, li, "scale", 0.00390625];
net = SetLayerParam[net, li, "include.phase", "TRAIN"];
```

```
li++;
net = AddLayer[net, "DATA", "mnist"];
net = SetLayerParam[net, li, "top", {"data", "label"}];
net = SetLayerParam[net, li, "source", StringJoin[mnistPath, "mnist_train_lmdb"]];
net = SetLayerParam[net, li, "backend", "LMDB"];
net = SetLayerParam[net, li, "batch_size", 64];
net = SetLayerParam[net, li, "scale", 0.00390625];
net = SetLayerParam[net, li, "include.phase", "TEST"];
li++;
net = AddLayer[net, "CONVOLUTION", "conv1"];
net = SetLayerParam[net, li, "bottom", "data"];
net = SetLayerParam[net, li, "top", "conv1"];
net = SetLayerParam[net, li, "blobs lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 20];
net = SetLayerParam[net, li, "kernel_size", 5];
net = SetLayerParam[net, li, "stride", 1];
net = SetLayerParam[net, li, "weight_filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "POOLING", "pool1"];
net = SetLayerParam[net, li, "bottom", "conv1"];
net = SetLayerParam[net, li, "top", "pool1"];
net = SetLayerParam[net, li, "pool", "MAX"];
net = SetLayerParam[net, li, "kernel_size", 2];
net = SetLayerParam[net, li, "stride", 2];
```

```
li++;
net = AddLayer[net, "CONVOLUTION", "conv2"];
net = SetLayerParam[net, li, "bottom", "pool1"];
net = SetLayerParam[net, li, "top", "conv2"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 50];
net = SetLayerParam[net, li, "kernel_size", 5];
net = SetLayerParam[net, li, "stride", 1];
net = SetLayerParam[net, li, "weight filler.type", "xavier"];
net = SetLayerParam[net, li, "bias filler.type", "constant"];
li++;
net = AddLayer[net, "POOLING", "pool2"];
net = SetLayerParam[net, li, "bottom", "conv2"];
net = SetLayerParam[net, li, "top", "pool2"];
net = SetLayerParam[net, li, "pool", "MAX"];
net = SetLayerParam[net, li, "kernel_size", 2];
net = SetLayerParam[net, li, "stride", 2];
li++;
net = AddLayer[net, "INNER PRODUCT", "ip1"];
net = SetLayerParam[net, li, "bottom", "pool2"];
net = SetLayerParam[net, li, "top", "ip1"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 500];
net = SetLayerParam[net, li, "weight_filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "RELU", "relu1"];
net = SetLayerParam[net, li, "bottom", "ip1"];
net = SetLayerParam[net, li, "top", "ip1"];
```

```
li++;
net = AddLayer[net, "INNER_PRODUCT", "ip2"];
net = SetLayerParam[net, li, "bottom", "ip1"];
net = SetLayerParam[net, li, "top", "ip2"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 10];
net = SetLayerParam[net, li, "weight_filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "ACCURACY", "accuracy"];
net = SetLayerParam[net, li, "bottom", {"ip2", "label"}];
net = SetLayerParam[net, li, "top", "accuracy"];
net = SetLayerParam[net, li, "include.phase", "TEST"];
li++;
net = AddLayer[net, "SOFTMAX LOSS", "loss"];
net = SetLayerParam[net, li, "bottom", {"ip2", "label"}];
net = SetLayerParam[net, li, "top", "loss"];
Resulting string is equal to content of lenet_train_test.prototxt.
netParam = GetNetParamString[net]
name: "LeNet"
layers {
  top: "data"
  top: "label"
  name: "mnist"
  type: DATA
  include {
    phase: TRAIN
  transform_param {
    scale: 0.00390625
  }
  data_param {
    source: "/home/kerhy/fel-lin/caffe/examples/mnist_train_lmdb"
    batch_size: 64
```

```
backend: LMDB
}
layers {
  top: "data"
  top: "label"
  name: "mnist"
  type: DATA
  include {
   phase: TEST
  transform_param {
    scale: 0.00390625
  }
  data_param {
    source: "/home/kerhy/fel-lin/caffe/examples/mnist/mnist_train_lmdb"
   batch_size: 64
   backend: LMDB
}
layers {
  bottom: "data"
  top: "conv1"
  name: "conv1"
  type: CONVOLUTION
  blobs_lr: 1
  blobs_lr: 2
  convolution_param {
   num_output: 20
   kernel_size: 5
    stride: 1
    weight_filler {
      type: "xavier"
   bias_filler {
```

```
type: "constant"
  }
layers {
  bottom: "conv1"
  top: "pool1"
  name: "pool1"
  type: POOLING
  pooling_param {
   pool: MAX
   kernel_size: 2
   stride: 2
  }
layers {
 bottom: "pool1"
  top: "conv2"
  name: "conv2"
  type: CONVOLUTION
 blobs_lr: 1
  blobs_lr: 2
  convolution_param {
   num_output: 50
   kernel_size: 5
   stride: 1
   weight_filler {
      type: "xavier"
   bias_filler {
      type: "constant"
  }
layers {
```

```
bottom: "conv2"
  top: "pool2"
  name: "pool2"
  type: POOLING
 pooling_param {
   pool: MAX
   kernel_size: 2
   stride: 2
  }
layers {
  bottom: "pool2"
  top: "ip1"
 name: "ip1"
  type: INNER_PRODUCT
  blobs_lr: 1
 blobs_lr: 2
  inner_product_param {
   num_output: 500
   weight_filler {
     type: "xavier"
   bias_filler {
      type: "constant"
  }
layers {
 bottom: "ip1"
  top: "ip1"
 name: "relu1"
  type: RELU
}
layers {
  bottom: "ip1"
```

```
top: "ip2"
  name: "ip2"
  type: INNER_PRODUCT
  blobs_lr: 1
  blobs_lr: 2
  inner_product_param {
    num_output: 10
    weight_filler {
      type: "xavier"
   bias_filler {
      type: "constant"
  }
layers {
 bottom: "ip2"
 bottom: "label"
  top: "accuracy"
  name: "accuracy"
  type: ACCURACY
  include {
    phase: TEST
  }
}
layers {
 bottom: "ip2"
 bottom: "label"
  top: "loss"
 name: "loss"
  type: SOFTMAX_LOSS
```

For now, CaffeLink supports net definition in solver only via path to protobuffer file, so it must be saved.

```
Export[StringJoin[NotebookDirectory[], "lenet.proto"], netParam, "Text"]
/home/kerhy/fel-lin/mthmt/lenet.proto
```

Notes:

Repeating parameters can be set with array:

```
net=SetLayerParam[net,li,"top",{"data","label"}];
resulting in:
 top: "data"
 top: "label"
```

Parameters with the same name but in different block must be specified with the block name as well:

```
net=SetLayerParam[net,li,"weight_filler.type","xavier"];
net=SetLayerParam[net,li,"bias_filler.type","constant"];
gives:
 weight_filler {
   type: "xavier"
bias_filler {
    type: "constant"
```

This holds even if only one of these is used.

Training

Before training a solver must be defined. This can be done in a similar way to building net, or an existing file can be used.

Definesolver

```
solver = NewSolver["SGD"];
solver =
  SetSolverParam[solver, "net", StringJoin[NotebookDirectory[], "lenet.proto"]];
solver = SetSolverParam[solver, "test_iter", 100];
solver = SetSolverParam[solver, "test_interval", 500];
solver = SetSolverParam[solver, "base_lr", 0.01];
solver = SetSolverParam[solver, "momentum", 0.9];
solver = SetSolverParam[solver, "weight_decay", 0.0005];
solver = SetSolverParam[solver, "lr_policy", "inv"];
solver = SetSolverParam[solver, "gamma", 0.0001];
solver = SetSolverParam[solver, "power", 0.75];
solver = SetSolverParam[solver, "display", 100];
solver = SetSolverParam[solver, "max_iter", 1000];
solver = SetSolverParam[solver, "snapshot", 1000];
solver = SetSolverParam[solver, "snapshot prefix",
   StringJoin[NotebookDirectory[], "snaps/lenet"]];
solver = SetSolverParam[solver, "solver_mode", "GPU"];
solveParam = GetNetParamString[solver]
```

```
net: "/home/kerhy/fel-lin/mthmt/lenet.proto"
test_iter: 100
test_interval: 500
base_lr: 0.01
display: 100
max_iter: 1000
lr_policy: "inv"
gamma: 0.0001
power: 0.75
momentum: 0.9
weight_decay: 0.0005
snapshot: 1000
snapshot_prefix: "/home/kerhy/fel-lin/mthmt/snaps/lenet"
solver_mode: GPU
solver_type: SGD
```

Notes

Create snapshot directory:

CreateDirectory[StringJoin[NotebookDirectory[], "snaps/"]]

■ Either double check relative paths or simply use absolute. All directories must exist. Be careful especially with source path in DATA layers. Current working directory can be printed in console with printWorkingPath.

printWorkingPath[]

TrainLeNet

New net can be trained with functions trainNetString and trainNetFile. The first expects solver parameters in a string, the second expects path pointing to a file with parameters. Both do the same. Training progress is printed to console.

trainNetString[solveParam]

Testing

Net testing requires a net and net parameters, referred by Caffe as deploy proto. The main difference between deploy and training proto is in input definition, the rest can be the same.

DeployLeNet

```
net = NewNet["LeNet - deploy"];
net = SetNetParam[net, "input", "data"];
net = SetNetParam[net, "input_dim", {10, 1, 28, 28}];
li = 0;
li++;
net = AddLayer[net, "CONVOLUTION", "conv1"];
net = SetLayerParam[net, li, "bottom", "data"];
net = SetLayerParam[net, li, "top", "conv1"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 20];
net = SetLayerParam[net, li, "kernel_size", 5];
net = SetLayerParam[net, li, "stride", 1];
net = SetLayerParam[net, li, "weight filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "POOLING", "pool1"];
net = SetLayerParam[net, li, "bottom", "conv1"];
net = SetLayerParam[net, li, "top", "pool1"];
net = SetLayerParam[net, li, "pool", "MAX"];
net = SetLayerParam[net, li, "kernel_size", 2];
net = SetLayerParam[net, li, "stride", 2];
```

```
li++;
net = AddLayer[net, "CONVOLUTION", "conv2"];
net = SetLayerParam[net, li, "bottom", "pool1"];
net = SetLayerParam[net, li, "top", "conv2"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 50];
net = SetLayerParam[net, li, "kernel_size", 5];
net = SetLayerParam[net, li, "stride", 1];
net = SetLayerParam[net, li, "weight filler.type", "xavier"];
net = SetLayerParam[net, li, "bias filler.type", "constant"];
li++;
net = AddLayer[net, "POOLING", "pool2"];
net = SetLayerParam[net, li, "bottom", "conv2"];
net = SetLayerParam[net, li, "top", "pool2"];
net = SetLayerParam[net, li, "pool", "MAX"];
net = SetLayerParam[net, li, "kernel_size", 2];
net = SetLayerParam[net, li, "stride", 2];
li++;
net = AddLayer[net, "INNER PRODUCT", "ip1"];
net = SetLayerParam[net, li, "bottom", "pool2"];
net = SetLayerParam[net, li, "top", "ip1"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 500];
net = SetLayerParam[net, li, "weight_filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "RELU", "relu1"];
net = SetLayerParam[net, li, "bottom", "ip1"];
net = SetLayerParam[net, li, "top", "ip1"];
```

```
li++;
net = AddLayer[net, "INNER_PRODUCT", "ip2"];
net = SetLayerParam[net, li, "bottom", "ip1"];
net = SetLayerParam[net, li, "top", "ip2"];
net = SetLayerParam[net, li, "blobs_lr", {1, 2}];
net = SetLayerParam[net, li, "num_output", 10];
net = SetLayerParam[net, li, "weight_filler.type", "xavier"];
net = SetLayerParam[net, li, "bias_filler.type", "constant"];
li++;
net = AddLayer[net, "SOFTMAX", "prob"];
net = SetLayerParam[net, li, "bottom", "ip2"];
net = SetLayerParam[net, li, "top", "prob"];
netDeploy = GetNetParamString[net]
name: "LeNet - deploy"
input: "data"
input_dim: 10
input_dim: 1
input_dim: 28
input_dim: 28
layers {
  bottom: "data"
  top: "conv1"
  name: "conv1"
  type: CONVOLUTION
  blobs_lr: 1
  blobs_lr: 2
  convolution_param {
    num_output: 20
    kernel_size: 5
    stride: 1
    weight_filler {
      type: "xavier"
    bias_filler {
```

```
type: "constant"
  }
layers {
  bottom: "conv1"
  top: "pool1"
  name: "pool1"
  type: POOLING
  pooling_param {
   pool: MAX
   kernel_size: 2
   stride: 2
  }
layers {
 bottom: "pool1"
  top: "conv2"
  name: "conv2"
  type: CONVOLUTION
 blobs_lr: 1
  blobs_lr: 2
  convolution_param {
   num_output: 50
   kernel_size: 5
   stride: 1
   weight_filler {
      type: "xavier"
   bias_filler {
      type: "constant"
  }
layers {
```

```
bottom: "conv2"
  top: "pool2"
  name: "pool2"
  type: POOLING
 pooling_param {
   pool: MAX
   kernel_size: 2
   stride: 2
  }
layers {
  bottom: "pool2"
  top: "ip1"
 name: "ip1"
  type: INNER_PRODUCT
  blobs_lr: 1
 blobs_lr: 2
  inner_product_param {
   num_output: 500
   weight_filler {
     type: "xavier"
   bias_filler {
      type: "constant"
  }
layers {
 bottom: "ip1"
  top: "ip1"
 name: "relu1"
  type: RELU
}
layers {
  bottom: "ip1"
```

```
top: "ip2"
  name: "ip2"
  type: INNER_PRODUCT
  blobs_lr: 1
  blobs_lr: 2
  inner_product_param {
    num_output: 10
    weight_filler {
      type: "xavier"
    bias_filler {
      type: "constant"
  }
layers {
 bottom: "ip2"
  top: "prob"
  name: "prob"
  type: SOFTMAX
```

Test LeNet

■ Tested net must be prepared at first. Caffe allocates memory and creates net topology based on given deploy proto. Preparation is done either from string with prepareNetString or from file path USing prepareNetFile.

```
prepareNetString[netDeploy]
```

Net info can be printed (to console) with printNetInfo.

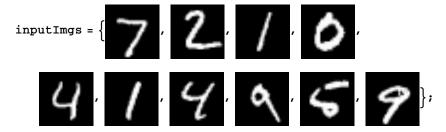
printNetInfo[]

After preparation, any net data with the same topology can be loaded.

```
loadNet[StringJoin[NotebookDirectory[], "snaps/lenet_iter_1000.caffemodel"]]
```

That also allows extracting learned parameters (weights, filters, etc.).

• Insertion of data to be tested is necessary only if there is no DATA layer in the deploy proto.



input = Flatten[Map[ImageData[#] &, inputImgs, 1]];

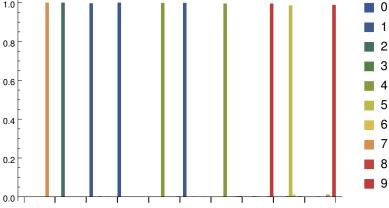
setInput[input]

■ This runs the test.

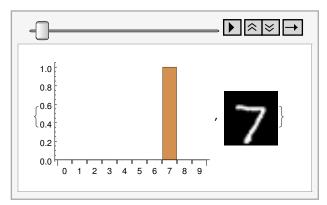
testNet[]

Result can be obtained from top blob of the last layer.

```
result = getTopBlob[7, 0];
res = ArrayReshape[result, {10, 10}];
leg = \{0, 1, 2, 3, 4, 5, 6, 7, 8, 9\};
BarChart[res, ChartLegends → leg, ChartStyle → "DarkRainbow"]
resList = Map[BarChart[#, ChartStyle → "DarkRainbow", ChartLabels → leg] &, res, 1];
resList = ArrayReshape[Riffle[resList, inputImgs], {10, 2}];
1.0
                                                    0
```



ListAnimate[resList, 0.25]



Fine tuning

For fine tuning, functions trainNetWeightsString or trainNetWeightsFile can be used. These two functions takes solver parameters from string or file and start training a new net with weights initialized from given net data.

```
trainNetWeightsString[solveParam,
StringJoin[NotebookDirectory[], "snaps/lenet_iter_1000.caffemodel"]]
```

Extracting eatures and net surgery

Functions getParamBlob, getTopBlob and getBottomBlob allow inspection of all blobs in net. In combination with setParamBlob weights can be easily copied to another layer or to completely different net.

As shown in test section, this loads net, inserts input and tests it.

```
prepareNetString[netDeploy]
loadNet[StringJoin[NotebookDirectory[], "snaps/lenet_iter_10000.caffemodel"]]
setInput[input]
testNet[]
printNetInfo[]
```

Surgery

For example, copying filters from layer conv1 to conv2 can be done like this:

```
filters = getParamBlob[0, 0]; (* 0th layer: conv1,
0th par. blob - filters, 20x1x5x5 *)
filters2 = getParamBlob[2, 0]; (* 2th layer: conv2,
0th par. blob - filters, 50x20x5x5 *)
filters2[[1;; Length[filters]]] = filters;
setParamBlob[filters2, 2, 0];
```

Now each of 20 channels in the first filter of conv2 is the same as one of 20 filters in conv1.

That could of course breaks the whole net. Anyway, altered net can be exported back to Caffe format with exportNet.

```
exportNet[StringJoin[NotebookDirectory[], "snaps/lenet_edited.caffemodel"]]
```

Visualization

In short: get parameters, separate and reshape them and then render.

```
■ Layer conv1 - filters
```

```
par = getParamBlob[0, 0]; (* 0th layer: conv1, 0th par. blob - filters *)
filters = ArrayReshape[par, {20, 5, 5}]; (* conv1 has 20 filters, 5x5 *)
fImgs = Map[Image[#, Magnification → 5] &, filters, 1]
```

Mapping image data to range 0,1 might be useful.

```
filters = Map[(#-Min[#]) / (Max[#] - Min[#]) &, filters, 1];
fImgs = Map[Image[#, Magnification \rightarrow 5] &, filters, 1]
■ Layer conv1 - features
in = getBottomBlob[0, 0];
out = getTopBlob[0, 0];
```

This shows one input image and result after convolution (20 filters).

```
imgi = 1; (* test image index, 1 - 10 *)
inImg = ArrayReshape[in, {10, 28, 28}][[imgi]];
outImgs = ArrayReshape[out, {10, 20, 24, 24}][[imgi]];
{Image[inImg, Magnification → 2],
Map[Image[#, Magnification → 2] &, outImgs, 1]}
·フ·フ·フ·フ·フ·フ·フ·フ·フ·
  フ·フ·ラ·ラ·ラ·フ·フ·フ·
The same in range 0, 1:
outImgs = Map[(#-Min[#]) / (Max[#]-Min[#]) &, outImgs, 1];
{Image[inImg, Magnification \rightarrow 2],
Map[Image[#, Magnification → 2] &, outImgs, 1]}
  Layer conv2 - filters
par = getParamBlob[2, 0]; (* 2th layer: conv2, 0th par. blob - filters *)
This show 20 channels of the first filter.
```

```
filters = ArrayReshape[par, {20, 5, 5}];
(* conv2 has 50 filterst with 20 channels each, 5x5 *)
Map[Image[#, Magnification → 5] &, filters, 1]
filters = Map[(#-Min[#]) / (Max[#]-Min[#]) &, filters, 1];
Map[Image[#, Magnification → 5] &, filters, 1]
}
Layer conv2 - features
in = getBottomBlob[2, 0];
out = getTopBlob[2, 0];
This shows one input image and result after convolution (50 filters).
imgi = 1; (* test image index, 1 - 10 *)
imgfi = 3; (* conv1 filter index, 1 - 20 *)
inImg = ArrayReshape[in, {10, 20, 12, 12}][[imgi, imgfi]];
outImgs = ArrayReshape[out, {10, 50, 8, 8}][[imgi]];
{Image[inImg, Magnification \rightarrow 4],
Map[Image[#, Magnification → 4] &, outImgs, 1]}
          7227755557725
 9579岁33岁后9582
 为世世之不为与别言的多有话
 }}
```

The same in range 0, 1:

```
outImgs = Map[(#-Min[#]) / (Max[#] - Min[#]) &, outImgs, 1];
inImg = Map[(#-Min[#]) / (Max[#]-Min[#]) &, {inImg}, 1][[1]];
{Image[inImg, Magnification \rightarrow 4],
Map[Image[#, Magnification \rightarrow 4] &, outImgs, 1]
    , {
 5. 7. 7. 3. 3. 3. 3. 3. 5. 7. 7. 7. 3. 5.
 ラ·フ·フ·フ·ラ·シ·ラ·シ·フ·
```