Advanced Machine Learning Methods Project CNN

Thomas Astad Sve

THOMAASV@STUD.NTNU.NO

Computer Science, Artificial Intelligence Norwegian University of Science & Technology University & Professional Studies UNIVERSITY OF CALIFORNIA, SAN DIEGO

AX004206@ACSMAIL.UCSD.EDU

Abstract

In this work, I use Convolutional Neural Networks trained on a GPU for classifying images in a tiny imagenet dataset. In particular I have used the python library lasagne [1] to built on the googlenet [2] architecture.

1. Introduction

I use Convolutional Neural Networks for image classifications on the tiny ImageNet dataset which is very similar to the original ImageNet [3] challenge.

2. Method

2.1 Model

The model used to build the classifier is

2.2 Dataset

3. Experiment

4. Results

Table 1: Shows results for running classification on Googlenet

# classes	# training	# validation	# testing	# epoch	test-loss	Accuracy
200	6903	1480	1480	500	10.9	39.30

5. Conclusion

References

[1] Sander Dieleman, Lasagne is a lightweight library to build and train neural networks in Theano. https://github.com/Lasagne/Lasagne

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- [2] C. Szegedy, W. Liu, Y. Jia, P. Sermanet, S. E. Reed, D. Anguelov, D. Erhan, V. Vanhoucke, A. Rabinovich. Going Deeper with Convolutions, 2014. http://arxiv.org/abs/1409.4842
- [3] J. Deng, W. Dong, R. Socher, L.-J. Li, K. Li, and L. FeiFei. ImageNet: A large-scale hierarchical image database. In CVPR, 2009

Code

main.py

```
import lasagne
import theano
import theano.tensor as T
import vgg16
import googlenet
import pickle
import time
from preprocess import load_dataset
def iterate_minibatches(inputs, targets, batchsize, shuffle=False):
      assert len(inputs) == len(targets) if shuffle:
             indices = np.arange(len(inputs))
      np.random.shuffle(indices)
for start_idx in range(0, len(inputs) - batchsize + 1, batchsize):
             if \quad {\tt shuffle}:
                   excerpt = indices[start_idx:start_idx + batchsize]
             else:
            excerpt = slice(start_idx, start_idx + batchsize)
yield inputs[excerpt], targets[excerpt]
\texttt{def train\_network} \, (\, \texttt{num\_epochs} \, , \, \, \, \texttt{X\_train} \, , \, \, \, \, \texttt{y\_train} \, , \, \, \, \, \texttt{X\_val} \, , \, \, \, \, \texttt{y\_val} \, , \, \, \, \, \texttt{train\_fn} \, , \, \, \, \, \, \texttt{val\_fn} \, \, ) \colon
      # We iterate over epochs:
for epoch in range(num_epochs):
             \# In each epoch, we do a full pass over the training data: train_err = 0
             train_batches = 0
             start_time = time.time() for batch in iterate_minibatches(X_train, y_train, 500, shuffle=True):
                   inputs, targets = batch
train_err += train_fn(inputs, targets)
                   train_batches += 1
            \# And a full pass over the validation data: val_err = 0
             val_acc = 0
             val_batches = 0
             for batch in iterate_minibatches(X_val, y_val, 500, shuffle=False):
                   inputs, targets = batch
err, acc = val.fn(inputs, targets)
val.err += err
val.acc += acc
                   val_batches += 1
            # Then we print the results for this epoch:
print("Epoch {} of {} took { :.3 f}s".format(
    epoch + 1, num_epochs, time.time() - start_time))
print(" training loss:\t\t { :.6 f}".format(train_err / train_batches))
print(" validation loss:\t\t { :.6 f}".format(val_err / val_batches))
print(" validation accuracy:\t\t { :.2 f} %".format(
    val_acc / val_batches * 100))
def test_network(X_test, y_test, val_fn):
       test\_err = 0
      test acc = 0
      test\_batches = 0
      for batch in iterate_minibatches(X_test, y_test, 500, shuffle=False):
            inputs , targets = batch
err , acc = val_fn(inputs , targets)
test_err += err
             test_acc += acc
             test_batches += 1
      def build_parameter_update(network, loss):
      # create parameter update expressions
params = lasagne.layers.get_all_params(network, trainable=True)
      {\tt updates = lasagne.updates.nesterov\_momentum(loss, params, learning\_rate = 0.01, }
                                                                       momentum = 0.9)
      return updates
def build_loss(network, target_var):
      # Create a loss expression for training, i.e., a scalar objective we want # to minimize (for our multi-class problem, it is the cross-entropy loss): prediction = lasagne.layers.get_output(network) loss = lasagne.objectives.categorical_crossentropy(prediction, target_var)
```

```
loss = loss.mean()
      return loss
def build_test_loss(network, target_var):
      # Create a loss expression for validation/testing. The crucial difference # here is that we do a deterministic forward pass through the network,
      # disabling dropout layers.

test_prediction = lasagne.layers.get_output(network, deterministic=True)

test_loss = lasagne.objectives.categorical_crossentropy(test_prediction,
                                                                                             target_var)
      test_loss = test_loss.mean()
      # As a bonus, also create an expression for the classification accuracy: test_acc = T.mean(T.eq(T.argmax(test_prediction, axis=1), target_var), dtype=theano.config.floatX)
      return test_loss . test_acc
def main(num_epochs = 500):
      print("Loading data...")
X_train, y_train, X_val, y_val, X_test, y_test = load_dataset()
      input_var = T.tensor4('inputs')
      target_var = T.ivector('targets')
      print("Building network...")
#network = vgg16.build_model(input_var)
      metwork = googlenet.build_model(input_var)
# Create a loss expression for training
loss = build_loss(network, target_var)
      # create parameter update expressions
      updates = build_parameter_update(network, loss)
      # Create a loss expression for validation/testing.
      test_loss, test_acc = build_test_loss(network, target_var)
      \# Compile a function performing a training step on a mini-batch (by giving \# the updates dictionary) and returning the corresponding training loss: print ("Setting training function...")
      train_fn = theano.function([input_var, target_var], loss, updates=updates)
      # Compile a second function computing the validation loss and accuracy:
print("Setting validation function for loss and accuracy...")
val_fn = theano.function([input_var, target_var], [test_loss, test_acc])
      \# Finally, launch the training loop.
      print("Starting training...")
train_network(num_epochs, X_train, y_train, X_val, y_val, train_fn, val_fn)
      \# After training, we compute and print the test error: print("Starting testing...") test_network(X_test, y_test, val_fn)
      np.savez('trained_alexnet_200.npz', *lasagne.layers.get_all_param_values(network))
      # And load them again later on like this:
# with np.load('model.npz') as f:
# param_values = [f['arr_%d' % i] for i in range(len(f.files))]
      # lasagne.layers.set_all_param_values(network, param_values)
if -name = "-main = ":
      main()
     googlenet.py
\frac{\phantom{a}}{\# \; BLVC} \; Googlenet \; , \; \; model \; from \; \; the \; \; paper \; :
```

```
# BLVC Googlenet, model from the paper:
# "Going Deeper with Convolutions"
# Original source:
# https://github.com/BVLC/caffe/tree/master/models/bvlc_googlenet
# License: unrestricted use

# Download pretrained weights from:
# https://s3.amazonaws.com/lasagne/recipes/pretrained/imagenet/blvc_googlenet.pkl

from lasagne.layers import InputLayer
from lasagne.layers import DenseLayer
from lasagne.layers import ConcatLayer
from lasagne.layers import NonlinearityLayer
from lasagne.layers import GlobalPoolLayer
from lasagne.layers.dnn import Conv2DDNNLayer as ConvLayer
from lasagne.layers.dnn import MaxPool2DDNNLayer as PoolLayerDNN
from lasagne.layers import MaxPool2DLayer as PoolLayer
```

```
from \ lasagne.\ layers \ import \ Local Response Normalization 2D Layer \ as \ LRN Layer
from lasagne.nonlinearities import softmax, linear
 \begin{array}{lll} \texttt{def build\_inception\_module(name, input\_layer, nfilters):} \\ \# \ nfilters: \ (\textit{pool\_proj}, \ 1x1, \ 3x3\_reduce, \ 3x3, \ 5x5\_reduce, \ 5x5) \end{array} 
     net = {}
     net = \{\}
net ['pool'] = PoolLayerDNN(input_layer, pool_size=3, stride=1, pad=1)
net ['pool_proj'] = ConvLayer(
    net ['pool'], nfilters [0], 1, flip_filters=False)
     net['1x1'] = ConvLayer(input_layer, nfilters[1], 1, flip_filters=False)
     net['3x3_reduce'] = ConvLayer(
   input_layer, nfilters[2], 1, flip_filters=False)
net['3x3'] = ConvLayer(
   net['3x3_reduce'], nfilters[3], 3, pad=1, flip_filters=False)
     net['5x5_reduce'] = ConvLayer(
    input_layer, nfilters[4], 1, flip_filters=False)
     net['5x5'] = ConvLayer(
  net['5x5-reduce'], nfilters[5], 5, pad=2, flip_filters=False)
     net['output'] = ConcatLayer([
          net['1x1'],
net['3x3'],
net['5x5'],
net['pool_proj'],
     return \ \{\,'\{\}/\{\}\,'.format(name,\ k)\colon v\ for\ k,\ v\ in\ net.items()\}
def build_model(input_var = None):
     net = {}
net ['input'] = InputLayer((None, 3, None, None), input_var = input_var)
     net['conv1/7x7.s2'] = ConvLayer(
net['input'], 64, 7, stride=2, pad=3, flip_filters=False)
net['pool1/3x3.s2'] = PoolLayer(
     net['pool1/3x3_s2'] = PoolLayer(
    net['conv1/7x7_s2'], pool_size=3, stride=2, ignore_border=False)
net['pool1/norm1'] = LRNLayer(net['pool1/3x3_s2'], alpha=0.00002, k=1)
net['conv2/3x3_reduce'] = ConvLayer(
    net['pool1/norm1'], 64, 1, flip_filters=False)
net['conv2/3x3'] = ConvLayer(
    net['conv2/3x3'] = ConvLayer(
    net['conv2/3x3_reduce'], 192, 3, pad=1, flip_filters=False)
net['conv2/norm2'] = LRNLayer(net['conv2/3x3'], alpha=0.00002, k=1)
net['pool2/3x3_s2'] = PoolLayer(
    net['conv2/norm2'], pool_size=3, stride=2, ignore_border=False)
     net.update(build_inception_module('inception_3a'
                                                     net['pool2/3x3_s2']
     [32, 64, 96, 128, 16, 32]))
net.update(build_inception_module('inception_3b',
                                                     net['inception_3a/output'
                                                     [64, 128, 128, 192, 32, 96]))
     net['pool3/3x3_s2'] = PoolLayer(
        net['inception_3b/output'], pool_size=3, stride=2, ignore_border=False)
     net.update(build_inception_module('inception_4a'
     net['inception_4d/output'],
[128, 256, 160, 320, 32, 128]))
     net['pool4/3x3_s2'] = PoolLayer(
        net['inception_4e/output'], pool_size=3, stride=2, ignore_border=False)
     net.update(build_inception_module('inception_5a'
                                                     net['pool4/3x3_s2'],
[128, 256, 160, 320, 32, 128]))
     [128, 384, 192, 384, 48, 128]))
     nonlinearity=linear)
```

preprocess.py

```
import numpy as np
import h5py
from random import shuffle
from math import floor
tiny_imagenet = "http://pages.ucsd.edu/~ztu/courses/tiny-imagenet-200.zip"
def crop_image(image, box):
      \# xmin, ymin, xmax, ymax
resized_image = image[int(box[0]):int(box[2]), int(box[1]):int(box[3])]
      return np.array(resized_image)
def load_dataset():
      with h5py.File('preprocessed_data/train_set.h5','r') as hf:
X = hf.get('X')
X_train = np.array(X, dtype=np.uint8)
            y = hf.get(',y')
            y_train = np.array(y, dtype=np.uint8)
      with h5py.File('preprocessed_data/val_set.h5','r') as hf:
    X = hf.get('X')
    X_val = np.array(X, dtype=np.uint8)
    y = hf.get('y')
            y_val = np.array(y, dtype=np.uint8)
      with h5py.File('preprocessed_data/test_set.h5','r') as hf:
    X = hf.get('X')
    X_test = np.array(X, dtype=np.uint8)
            y = hf.get('y')
            y_test = np.array(y, dtype=np.uint8)
      \tt return \ X\_train \ , \ y\_train \ , \ X\_val \ , \ y\_val \ , \ X\_test \ , \ y\_test
def save\_dataset(filename, X, y = None):
      if y != None:
    with h5py.File("preprocessed_data/" + filename, 'w') as hf:
                  hf.create_dataset('X', data=X)
hf.create_dataset('y', data=y)
            . with h5py.File("preprocessed_data/" + filename, 'w') as hf: hf.create_dataset('X', data=X)
def load_training_set(path, Image, wnids):
     import glob, os
owd = os.getcwd() # Get original path
      images = []
      y = \begin{bmatrix} 1 \end{bmatrix}
      \sum_{i=0}^{\infty} \cos x = []
      for class_id in wnids:
            class_id in wnids:
bbox_file = path + class_id + "/" + class_id + "_boxes.txt"
for line in open(bbox_file):
    words = line.split()
    #img = Image.open(path + class_id + "/images/" + words[0]).convert('L')
    img = Image.open(path + class_id + "/images/" + words[0])
                  image = np.array(img)
                  #image_cropped = crop_image(image, words[1:])
if image.ndim == 3:
                        bbox.append(words[1:])
                        #image = np.rollaxis(image) #Reshape image into columnvector image = np.rollaxis(image, 2)
                        images.append(image) # Append image to dataset
            y.append(i)
i = i + 1
            os.chdir(owd) # Reset to original path
     X = np.array(images, dtype=np.uint8)
      y = np.array(y)
bbox = np.array(bbox)
      return X, y, bbox
def load_val_set(path, Image):
    val_annotations = path + "val_annotations.txt"
    images_path = path + "images/"
      images = []
```

```
y = []
bbox = []
        for line in open(val_annotations):
               words = line.split()
image_file = words[0]
               \begin{array}{lll} \textit{Hindge-Inage-open(images-path + image-file).convert('L')} \ \# \ \textit{Read image and convert to grayscale} \\ \textit{img} = \textit{Image.open(images-path + image-file)} \end{array}
               image = np.array(img)
               \#image\_cropped = crop\_image(image, words[2:])
               \#image = np.ravel(image) \# Convert the image to a column vector \#print image\_file, image.shape if image.ndim == 3:
                       blox.append(words[1])
blox.append(words[2:])
image = np.rollaxis(image, 2)
                       images.append(image)
       X = np.array(images, dtype=np.uint8)
         = np.arrav(v)
       bbox = np.array(bbox)
       return X, y, bbox
def load_test_set(test_path):
       import glob, os
       owd = os.getcwd() # Get original path
       images = []
       for file in glob.glob("*.JPEG"): # For all images in folder
img = Image.open(file)
image = np.array(img)
               image = np.rollaxis(image, 2)
images.append(image)
        os.chdir(owd) # Reset to original path
       return np.array(images)
\label{eq:def_split_dataset} \begin{array}{lll} \texttt{def} & \texttt{split\_dataset} \, (X, \ y\,, \ \texttt{test\_size} \, = \, 0.2\,, \ \texttt{val} \, = \, \texttt{False} \,) \colon \end{array}
       data = zip(X, y)
shuffle(data)
       X, y = zip(*data)
        if val:
               split_point = int(floor(len(X)*(1 - test_size * 2)))
               split_point = int(floor(len(X)*[1 - test_size * 2)))
X_train, y_train = X[:split_point], y[:split_point]
X_test_val, y_test_val = X[split_point:], y[split_point:]
split_point = int(floor(len(X_test_val)*0.5))
X_test, y_test, X_val, y_val = X_test_val[:split_point], y_test_val[:split_point], \]
                                                                             X_test_val[split_point:], y_test_val[split_point:]
               return \ np.array(X\_train), \ np.array(y\_train), \ np.array(X\_test) \\ , \ np.array(y\_test), \ np.array(X\_val), \ np.array(y\_val)
               \label{eq:continuous_point} \begin{array}{l} \text{split_point} = \inf \left( \text{floor} \left( \text{len} \left( X \right) * \left( 1 - \text{test\_size} \right) \right) \right) \\ \text{return } X[: \text{split_point}] \,, \, \, y[: \text{split_point}] \,, \, \, X[[: \text{split_point}]] \,, \, \, y[[: \text{split_point}]] \,, \, \\ \end{array}
def\ generate\_dataset(num\_classes = 200, save = True):
       import Image print ("Generating dataset ...")
       train-path = "/home/thomas/data/dataset/tiny-imagenet-200/train/"
val.path = "/home/thomas/data/dataset/tiny-imagenet-200/val/"
test-path = "/home/thomas/data/dataset/tiny-imagenet-200/test/"
wnid_file = "/home/thomas/data/dataset/tiny-imagenet-200/wnids.txt"
        wnids = [line.strip() for line in open(wnid-file)]
       wnids = wnids[:num_classes]
print "Classes: ", len(wnids)
print "Loading training set"
X_train, y_train, train_box = load_training_set(train_path, Image, wnids)
        print "Loading validation set"
        if num_classes == 200:
               X, y, train_box = load_training_set(train_path, Image, wnids)
               X_train, y_train, X_test, y_test = split_dataset(X, y, test_size = 0.2)
X_val, y_val, val_box = load_val_set(val_path, Image)
       X, y, boxes = load_training_set(train_path, Image, wnids)
X_train, y_train, X_test, y_test, X_val, y_val = split_dataset(X, y, test_size = 0.15, val = True)
print "X_val shape: ", X_val.shape, " y_val shape: ", y_val.shape
print "X_train shape: ", X_train.shape, " y_train shape: ", y_train.shape
print "X_test shape: ", X_test.shape, " y_test shape: ", y_test.shape
               save_dataset("train_set.h5", X_train, y_train)
```

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```
save_dataset("val_set.h5", X_val, y_val)
save_dataset("test.set.h5", X_test, y_test)
print("Dataset saved")
else:
    return X_train, y_train, X_val, y_val, X_test, y_test

if -_name__ == "__main__":
    generate_dataset(20)
#X_train, y_train, X_val, y_val, X_test, y_test = load_dataset()
#print X_train.shape, y_train.shape
#print X_val, y_val
```