

UNIVERSITY OF TORONTO Neural Networks for Machine Learning

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Test your PA3 with images of numbers





9 vote(s) Hi, in this thread I want to show you a way to test the model created in PA3 with images of numbers. I think that you would enjoy more this programming assignment if you can see that it really works.



First of all we have to modify the function a3 to save the model. It's very easy, just add this line at the end of the function.

```
function a3(wd_coefficient, n_hid, n_iters, learning_rate, mo
mentum_multiplier, do_early_stopping, mini_batch_size)
.
.
.
save -mat7-binary model.mat model; %save the model
end
```

Ok, now create a file called a3Test.m and paste this code.

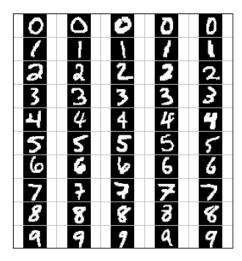
```
function a3Test(imageName)
a3{\rm Test} gets the name of the image as an input and displays t
%probabilities of each number
%imageName name of the image to load
   close all; %close all figures
%Load the model
   load model; %load the model of the network
%Load the image and transform it
   number=imread(imageName); %read the image file
   imshow(number); %show the image
   movegui('northwest');
   number=im2double(number); %transform the format from 8b
it to double
   number = number';
                            %this is just to adapt to USPS da
tabase
   data=number(:);
                            %finally transform the matrix int
o a vector
%Compute the probabilities
   probabilities = computeProbabilities (model, data);
%Show the probabilities in a graph
 figure;
 numbers = [ 0 1 2 3 4 5 6 7 8 9];
 bar(numbers, probabilities);
 movegui('northeast');
end
function ret = logistic(input)
 ret = 1 . / (1 + exp(-input));
function ret = log_sum_exp_over_rows(a)
 \mbox{\ensuremath{\$}} This computes \log\left(\text{sum}\left(\exp\left(a\right),\ 1\right)\right) in a numerically stable
 maxs_small = max(a, [], 1);
 maxs_big = repmat(maxs_small, [size(a, 1), 1]);
```

```
ret = log(sum(exp(a - maxs big), l)) + maxs small;
function ret = computeProbabilities(model, data, wd coefficie
 % model.input to hid is a matrix of size (n hid, 256)
 % model.hid_to_class is a matrix of size (10,256)
 % data.inputs is a matrix of size (256, < number of data case
 % data.outputs is a matrix of size (10,<number of data case
 % first, do the forward pass, i.e. calculate a variety of r
 hid in = model.input to hid * data; % input to the hidden u
nits, i.e. before the logistic. size: (n hid,<number of data
 hid out = logistic(hid in); % output of the hidden units, i
.e. after the logistic. size: (n hid,<number of data cases>)
 class in = model.hid to class * hid out; % input to the com
ponents of the softmax. size: (10, <number of data cases>)
 class normalizer = log sum exp over rows(class in); % log(s
um(exp)) is what we subtract to get normalized log class prob
abilities. size: (1,<number of data cases>)
 log_class_prob = class_in - repmat(class_normalizer, [size(
class_in, 1), 1]); % log of probability of each class. size:
(10, <number of data cases>)
 class_out = exp(log_class_prob); % probability of each clas
s. Each column (i.e. each case) sums to 1. size: (10, \leqnumber
of data cases>)
 ret = class out;
end
%Theta is a column vector that holds the weights
%Model contains two matrix (,) with the weights
function ret = theta_to_model(theta)
 n_{hid} = size(theta, 1) / (256+10);
 ret.input_to_hid = transpose(reshape(theta(1: 256*n_hid), 2
56, n hid));
 ret.hid to class = reshape(theta(256 * n hid + 1 : size(the
ta,1)), n_hid, 10).';
end
function ret = model_to_theta(model)
 input_to_hid_transpose = transpose(model.input_to_hid);
 hid_to_class_transpose = transpose(model.hid_to_class);
 ret = [input_to_hid_transpose(:); hid_to_class_transpose(:)
];
end
```

Great, now we need 16x16 greyscale images to test the code. You can use a program like Photoshop as I do in the next video o you can get images from the dataset writing something like this:

```
imwrite((reshape(data.training.inputs(:,8)',16,16)'),'number7
.jpg')
```

The images will look like these ones:



You can watch how it works in this video: http://youtu.be/OPfL4SJkjno



Comments

0

Very nice. I just linked to it.

John Rood

on Sun 4 Nov 2012 9:10 PM CET

0

It is great to see how vulnerable this feed forward neural network to variances. I draw a vertical line to the right, instead of the middle, of the screen and it is recognized as 3 with .88 probability `)

by the way add the following code to see the probabilities on the plot:

```
for i = 1:length(numbers)
    text(i-1,probabilities(i)+0.04,num2str(probabi
lities(i),2),'Color','r','HorizontalAlignment','ce
nter')
end
```

and thanks!

Özgür Aydoğan

on Thu 15 Nov 2012 5:10 PM CET

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Very nice Guillermo, thanks! It always helps a lot to get a more concrete feel for what all these numbers, gradients, matrices, etc, are actually doing. Visualization is an important part of Neural Network research.

Tijmen Tieleman (Staff) on Sun 4 Nov 2012 6:17 PM CET

Comments

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Thank you:)

[Delete] Guillermo Barbadillo on Thu 8 Nov 2012 5:00 PM CET

-1

Tijmen, i am hoping that someone, Prof Hinton?, you?, both?, others as well?, will *speak to this issue*. In the introductory video for this course, iirc, Prof Hinton talks about what the neural network [or George Bush?] "sees" and what the human mind sees [literally?].

I would be interested to hear how (human) visualization its into the picture of research on neural networks.

Hmmm. Please allow me a technical detail. Jean Dieudonné famously mentions, in his *Traité d'Analyse*, that he has deliberately "refrained from introducing even a single diagram" into the book, because mathematics is supposed to proceed axiomatically (logically), not intuitively. [Paraphrase from memory, but i can find the exact quote if need be.]

I am wondering about, among other things, whether this comment has anything to do with the study of neural networks. Ultimately, if we are doing something like "the mind" does, doesn't that mean we need to avoid any kind of *causa sui* -like argument?

So we shouldn't have to use visualization to simulate vision ... ?

And, vice versa, if you will, what does it mean if we do rely on visualization to come to understand our subject matter? Does the subject matter help us to understand visualization ... and vision? Certainly i would say the subject aims to do this, if you will ...?

John Rood

on Thu 8 Nov 2012 5:58 PM CET

0

typo: "its" should be "fits" in the 2nd paragraph

also: http://en.wikipedia.org/wiki/Causa sui

John Rood

on Thu 8 Nov 2012 6:02 PM CET

2

Hi John, visualization is a big part of NN research. Most NN researchers couldn't care less about the analogies with human brains (Hinton is the biggest exception to that rule - he famously pursues this research to with the goal of finding out how the brain works), but that doesn't take away the fact that NN's are hard to inspect, and any images can be hugely helpful.

Tijmen Tieleman (Staff) on Thu 8 Nov 2012 6:19 PM CET

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