Deep Learning CS583B: Midterm Exam

October 23, 2022

Due: 5:30 PM, 24 October, 2020	
Student name:	
Student ID:	
Student email address:	

• Read these instructions carefully

Instructor: Jia Xu

- Fill-in your personal info as indicated above.
- There are 10 pages (including this one) and 10 questions.
- Fill in your answers with a clear handwritting. The answer will not be accepted if the scanned form of the writting is not easy to read.
- You can also direct edit PDF and fill in your answer.
- Submit your answer sheets on Cancas by 5:30 PM on the 24th.

good luck!

Suppose that you are training a Speech Recognition System on 100 million sentences. Your goal is to improve the accuracy of the translation system. Your small team of experts and engineers has 1 day only to improve the system. which of the following task/tasks would you?

	1. Preprocessing
	2. Change of the training data
	3. Implement a new training algorithm
	4. Add new features
	5. Change pruning threshold in search
	Justify your answer:
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You are given X-ray images of patients where malignant tumors have been labeled at the pixel level (for each X-ray, you have another image where each pixel is labeled 1 for tumor and -1 for non tumor). Describe how you could use deep learning to solve the problem of detecting malignant tumors. specify which type of model you would use, and how you would use the problem of the lates of the problem of the lates of the problem.
go about setting up the dataset to train it.

Specify how convolution of a given image is computed. Assume the input is an image size $H \times W$ with C channels, the kernel K has size $N \times M$, the stride is $T \times S$, the opperformed is in fact cross-correlation (as usual in convolutional neural networks) as						
O output channels are computed. Spell out the computations for both padding schemes.	SAME and VALID					

Consider a convolution layer. The input consists of 6 feature maps of size 20×20 . The output consists of 8 feature maps, and the filters are of size 5×5 . The convolution is done with a stride of 2 and zero padding, so the output feature maps are of size 10×10 . For both parts, you can leave your expression as a product of integers; you do not need to actually compute the product. You do not need to show your work, but doing so can help you receive partial credit.

1.	Determine the number of weights in this convolution layer.					
2.	Now suppose we made this a fully connected layer, but where the number of input and output units are kept the same as in the network described above. Determine the number of weights in this layer.					

• For a fully-connected deep network with one hidden layer, increasing the number of hidden units should have what effect on bias and variance? Explain briefly.

• You are solving the binary classification task of classifying images as cat vs. non-cat. You design a CNN with a single output neuron. Let the output of this neuron be z. The final output of your network, \hat{y} is given by: $\hat{y} = \delta(ReLU(z))$. You classify all inputs with a final value $\hat{y} \geq 0.5$ as cat images. What problem are you going to encounter?

- A convolutional neural network has an input image of 28×28 , and the filter is 3×3 . The stride is 2. What is the output dimension of the 2nd layer?
- What is the max pooling result of a 2×2 pool size on the below input?

20	55	101	102
32	11	103	104
55	55	10	20
55	55	30	40

• What is the average pooling result of a 2×2 pool size on the above input?

 $\bullet\,$ Name a possible solution for the vanishing of the gradients and explain.

Consider the following linear auto-encoder with 1 input and 1 output: $\tilde{x} = w_2 w_1 x$, trained with the squared reconstruction error: $L(W) = \frac{1}{P} \sum_{i=1}^{P} \frac{1}{2} (x^i - w_2 w_1 x^i)^2$ The scalar training samples have variance 1.

$$L(W) = \frac{1}{P} \sum_{i=1}^{P} \frac{1}{2} (x^i - w_2 w_1 x^i)^2$$

- (a) What is the set of solutions (with 0 loss)?
- (b) Does the loss have a saddle point? Where?

A neural network has been encrypted on a device. You can access neither its architecture nor the values of its parameters. Is it possible to create an adversarial example to attack this network? Explain why.

You want to perform a regression task with the following dataset: $x^{(i)} \in R$ and $y^{(i)} \in R$, $i=1,\cdots,m$ are the ith example and output in the dataset, respectively. Denote the prediction for example i by $f(x^{(i)})$. Remember that for a given loss L we minimize the following cost function

$$J = \frac{1}{m} \sum_{i=1}^{m} L(f(x^{(i)}), y^{(i)}).$$

 $J = \frac{1}{m} \sum_{i=1}^{m} L(f(x^{(i)}), y^{(i)}).$ In this part we are deciding between using loss 1 and loss 2, given by:

$$L_1(f(x^{(i)}), y^{(i)}) = |y^{(i)} - f(x^{(i)})|,$$

$$L_2(f(x^{(i)}), y^{(i)}) = (y^{(i)} - f(x^{(i)})^2.$$

(a) Draw $L_1(x,0)$ and $L_2(x,0)$ versus $x \in R$ on the same plot.

(b) An outlier is a datapoint which is very different from other datapoints of the same class. Based on your plots, which method do you think works better when there is a large number of outliers in your dataset? Hint: Contributions of outliers to gradient calculations should be as small as possible.

(c) "Using L_2 loss forces the weights of the network to end up small." Do you agree with this statement? Why/Why not?

You want to perform a classification task. You are hesitant between two choices: Approach A and Approach B. The only difference between these two approaches is the loss function that is minimized.

Assume that $x^{(i)} \in R$ and $y^{(i)} \in \{1, -1\}$, i = 1, ..., m are the ith example and output label in the dataset, respectively. $f(x^{(i)})$ denotes the output of the classifier for the ith example. Recall that for a given loss L you minimize the following cost function:

$$J = \frac{1}{m} \sum_{i=1}^{m} L(f(x^{(i)}), y^{(i)}).$$

As we mentioned, the only difference between approach A and approach B is the choice

$$L_A(f(x^{(i)}), y^{(i)}) = \max\{0, 1 - y^{(i)}f(x^{(i)})\},\tag{1}$$

$$L_B(f(x^{(i)}), y^{(i)}) = \log_2^{(1 + exp(-y^{(i)}f(x^{(i)})))}.$$
 (2)

- (i) Rewrite L_B in terms of the sigmoid function.
- (ii) You are given an example with $y^{(i)} = -1$. What value of $f(x^{(i)})$ will minimize L_B ?
- (iii) You are given an example with $y^{(i)} = -1$. What is the greatest value of $f(x^{(i)})$ that will minimize L_A ?