MGM-Assignment-2

February 2024

1 Theory

- 1. Find the conjugates of the following:
 - Negative Entropy. $f(x) = x \log x$
 - Inverse. f(x) = 1/x
- 2. (5 points) Choose $f(x) = x \log x (x+1) \log(x+1)$, then show that

$$D_f(p || q) = 2 \text{ JSD } (p, q) - \log(4).$$

Recall that the Jensen-Shannon Divergence (JSD) is given by

$$JSD(p,q) = \frac{1}{2}D_{KL}\left(p \parallel \frac{p+q}{2}\right) + \frac{1}{2}D_{KL}\left(q \parallel \frac{p+q}{2}\right).$$

3. (5 points) Show that the convex conjugate of f above is

$$f^*(t) = -\log(1 - e^t).$$

4. (2 marks) If f is 1-Lipschitz, then show that

$$\|\nabla f(x)\| < 1$$
, for all $x \in \text{dom}(f)$.

5. (2 marks) Consider $A \subset \mathbb{R}^n$. If A is closed and bounded, then it is compact. Show that for all 1-Lipschitz functions $f: A \to \mathbb{R}$, we must have a function f^* such that the following holds:

$$|f^*(x) - f^*(y)| = ||x - y||.$$

You are allowed to guess a function.

- 6. $(2 \text{ marks})(\epsilon \delta \text{ definition})$ A function is defined to continuous if for every $\epsilon > 0$, there exists a δ such that whenever $d(x,y) < \delta$, then $d(f(x),f(y)) < \epsilon$. Here d is any valid metric. Prove or disprove that any L Lipschitz continuous function is continuous.
- 7. (3 marks) We define two vectors x and y to be orthogonal if $x^Ty=1$. Given any unit vector v, show that v and $(I-vv^T)v$ are orthogonal. Given a unit vector $v=[1/\sqrt{2},1/\sqrt{2}]$ in \mathbb{R}^2 , write down the component of vector [1,2] along v and its orthogonal vector $(I-vv^T)v$. Plot the vector v, and the projected components.
- 8. (4 points) Consider the following optimization problem

minimize
$$x^T x$$

subject to $Ax = b$, $A \in \mathbb{R}^{p \times n}$

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- Find the Lagrangian function
- Find the dual function
- Check whether the dual function is concave
- Check whether the dual function is lower bound to p^*
- Does Slater condition hold?
- 9. (3 points) Show that the distance function defined as d(x,y) = ||x-y||, where $x,y \in \mathbb{R}^n$ is a convex function. Can you think of a subset A of \mathbb{R}^n , where the distance function is not necessarily convex? [Hint: if the distance function is convex then there exists (x^*, y^*) for which, d(x,y) is minimum.] Can this uniqueness be violated on some special subset of \mathbb{R}^n ? If the subset A is convex, is it necessary that d(x,y) is convex function?

2 MLP in Python

Instructions:

- Copy the following python code in your google colab: https://colab.research.google.com/drive/1fEi5TAZHjCFZrSsHUURYlakdlbrsVhlg?usp=sharing
- Run the examples shown in this colab notebook.

Answer the following:

1. Generate a tensor data from

using appropriate function, for example, torch.tensor as shown in colab notebook, and store this tensor in mytensor variable.

- 2. Report the device. Function used?
- 3. Report the shape of tensor. Function used?
- 4. Report total number of elements. Function used?
- 5. Print third slice of this tensor. Function used?
- 6. Print top right elements of your tensor. Function used?
- 7. Create a tensor of all ones and all zeros, but same shape as your tensor mytensor. Function used?
- 8. Create a tensor sampled from standard normal. Function used? How you would create this tensor on GPU?
- 9. Subtract all entries of your tensor mytensor by 10, and multiply by 4.
- 10. Compute mean and standard deviation of your tensor.
- 11. The linear layer nn.linear(m,n) takes in matrix of dimension $\ell \times m$ and outputs matrix of dimension $\ell \times n$. We want to perform the following operation:

$$y = Ax + b,$$
 $A \in \mathbb{R}^{m \times n}, b, y \in \mathbb{R}^n.$

Set m = 5, n = 3, create A, b random. Show the output y using the linear layer.

- 12. Create a normal random tensor of dimension $3 \times 4 \times 3$. Apply ReLU activation on this.
- 13. Apply batch normalization on this tensor you just created.
- 14. Create a sequential container that implement: linear layer that takes in input vector of length 10, and produces output vector of length 5, followed by a ReLU activation, followed by linear layer of input dimension (guess?) and output dimension 3, followed by batch normalization, followed by ReLU.
- 15. Create a random batch: 100×10 then add one to this. Then use the same loss function as in colab notebook. Run one step of ADAM optimizer and show the loss. Do 3 more iterations, then print the loss.