Lab 1 – The Basics of Python and Pytorch

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March 18, 2021

This lab aims to help the students refresh the basics of python, particularly, NumPy. You may use python functions to answer these questions, e.g., np.sum, torch.sigmoid, etc. But all the algorithm should be implemented by yourself unless otherwise specified, e.g., you should write a dataset class instead of using pytorch Dataset class in Problem 13. Please write the report, including the codes, and screen-shot the results, and send to Yikai, by 17:00 April 8th, 2021.

1. Write a Python function that takes a list and returns a new list with unique elements of the first list.

   e.g.,

   Input:[1, 2, 3, 3, 3, 3, 4, 5].

   Output: [1, 2, 3, 4, 5].

2. Write a Python function that checks whether a passed string is palindrome or not. A palindrome is a word, phrase, or sequence that reads the same backward as forward. For example, both "madam" and "nurses run" are palindromes.

3. Write a NumPy program to find the real and imaginary parts of an array of complex numbers.

   e.g.,

   Input: array [1.00000000+0.j 0.70710678+0.70710678j]

   Output: array [[1, 0], [0.70710678, 0.70710678]]

4. Write a Python program to add two binary numbers.

   e.g.,

   Input : ('11', '1')

   Output : 100

5. You are given two non-empty linked lists representing two non-negative integers. The digits are stored in reverse order and each of their nodes contain a single digit. Add the two numbers and return it as a linked list. You may assume the two numbers do not contain any leading zero, except the number 0 itself.

   e.g.,

   Input: (2 -> 4 -> 3) + (5 -> 6 -> 4)

   Output: 7 -> 0 -> 8

   Explanation: 342 + 465 = 807.

   Linked list is defined as follow

```
# Definition for singly-linked list.
# class ListNode:
    # def __init__(self, x):
        # self.val = x
        # self.next = None
```

6. Implement quick sort
7. Implement shell sort
8. Implement linear regression model and use autograd to optimize it by Pytorch.
9. Implement logistic regression model and use autograd to optimize it by Pytorch.
10. Implement linear SVM model for binary classification task and use autograd to optimize it by Pytorch.

    Hint: you may use the loss of $\sum \max\left[0, 1 - y\left(wx + b\right)\right]$
11. Add a Frobenius norm penalty for the weight w in your SVM model by two different ways: (1) use a pytorch function to calculate the norm; (2) implement the code by yourself.

    Hint: Frobenius norm of a matrix A is $\|A\|_F = \left(\sum_{i=1}^{n}\sum_{j=1}^{m}|a_{ij}|^2\right)^{\frac{1}{2}}$.
12. Download CIFAR-10 dataset[1] and visualize some of its images.
13. Write a dataset class for loading CIFAR-10. Make sure it could be transferred to Pytorch *Dataloader*. The class should meet the following requirements: (1) Inherit pytorch's *DataSet* class; (2) Load the image file and save in proper way; (3) Override *__getitem__* and *__len__* methods.

    Hint: If you find this part a little hard, check the official code[2] and make sure you understand each part.
14. Run one epoch for loading CIFAR-10 with Pytorch *Dataloader* and test the loading time of different *batch_size* (1, 4, 64, 1024), different *num_workers* (0,1,4,16), and whether use *pin_memory* or not.
15. Calculate the mean and std of CIFAR-10' training set within each RGB channel.
16. Numpy exercises
    - Consider a random 10x2 matrix representing cartesian coordinates, convert them to polar coordinates.
    - Create a 2D array subclass such that Z[i, j] == Z[j, i].
    - Consider 2 sets of points P0, P1 describing lines (2d) and a set of points P, how to compute distance from each point j (P[j]) to each line i (P0[i],P1[i])?
17. Bilinear Interpolation

    Please implement the bilinear interpolation algorithm using python. Check this for an introduction to bilinear interpolation.

    Test samples:

[1] https://www.cs.toronto.edu/~kriz/cifar.html
[2] https://pytorch.org/docs/stable/_modules/torchvision/datasets/cifar.html#CIFAR10

A =

((110, 120, 130),

(210, 220, 230),

(310, 320, 330))

BilinearInterpolation(A, (1, 1)) == 110

BilinearInterpolation(A, (2.5, 2.5)) == 275

18. Cartesian product

    Given an arbitrary number of vectors, build the cartesian product (every combinations of every item).

    e.g. [1, 2, 3], [4, 5], [6, 7] ==> [[1 4 6] [1 4 7] [1 5 6] [1 5 7] [2 4 6] [2 4 7] [2 5 6] [2 5 7] [3 4 6] [3 4 7] [3 5 6] [3 5 7]]

19. Extracting a subpart of an array

    Consider an arbitrary array, write a function that extract a subpart with a fixed shape and centered on a given element (pad with a *fill* value when necessary)

    e.g.

    In:

    >> Z = np.random.randint(0, 10, (5, 5))

    >> shape = (4, 4)

    >> fill = 0

    >> position = (1,1)

    >> Z

    [[3 6 8 5 9]

    [4 9 0 0 9]

    [6 1 4 0 8]

    [9 1 2 0 9]

    [4 1 7 5 0]]

    Out:

    [[0 0 0 0]

    [0 3 6 8]

    [0 4 9 0]

    [0 6 1 4]]

20. Matrix operations

    Please implement following matrix (just 2D) operations without numpy:

    - add
    - subtract
    - scalar multiply

- multiply
- identity
- transpose
- inverse

Test samples:

In:

>> matrix_a = [[12, 10], [3, 9]]

>> matrix_b = [[3, 4], [7, 4]]

>> matrix_c = [[11, 12, 13, 14], [21, 22, 23, 24], [31, 32, 33, 34], [41, 42, 43, 44]]

>> matrix_d = [[3, 0, 2], [2, 0, -2], [0, 1, 1]]

Out:

add(matrix_a, matrix_b) == [[15, 14], [10, 13]]

subtract(matrix_a, matrix_b) == [[9, 6], [-4, 5]]

scalar_multiply(matrix_b, 3) == [[9, 12], [21, 12]]

multiply(matrix_a, matrix_b) == [[106, 88], [72, 48]]

identity(3) == [[1, 0, 0], [0, 1, 0], [0, 0, 1]]

transpose(matrix_c) == [[11, 21, 31, 41], [12, 22, 32, 42], [13, 23, 33, 43], [14, 24, 34, 44]]

inverse(matrix_d) == [[0.2, 0.2, 0.0], [-0.2, 0.30000000000000004, 1.0], [0.2, -0.30000000000000004, 0.0]]

21. Greatest common divisor

    Find the greatest common divisor(gcd) of two integers.

    Test samples:

    - GCD(3, 5) = 1
    - GCD(6, 3) = 3
    - GCD(-2, 6) = 2
    - GCD(0, 3) = 3

22. Find all consecutive positive number sequences whose sum is N

    e.g. 18+19...+22 = 9+10+...+16 = 100

    Find all consecutive positive number sequences whose sum is 1000, and report your results.

23. Password checking

    A website requires the users to input username and password to register. Write a program to check the validity of password input by users. Following are the criteria for checking the password:

    - At least 1 letter between [a-z]
    - At least 1 number between [0-9]
    - At least 1 letter between [A-Z]

- At least 1 character from [$#@]
- Minimum length of transaction password: 6
- Maximum length of transaction password: 12

Your program should accept a sequence of comma separated passwords and will check them according to the above criteria. Passwords that match the criteria are to be printed, each separated by a comma.

e.g.

If the following passwords are given as input to the program: ABd1234@1,a F1#,2w3E*,2We3345

Then, the output of the program should be: ABd1234@1

24. Use pytorch to solve three simple UCI(http://archive.ics.uci.edu/ml/index.php) feature based problems(iris, adult, wine), and report the accuracy of the validation. You can leverage logistic regiression or SVM to tackle these problems.
iris:http://archive.ics.uci.edu/ml/machine-learning-databases/iris/
adult:http://archive.ics.uci.edu/ml/machine-learning-databases/adult/
wine:http://archive.ics.uci.edu/ml/datasets/Wine

25. Use Cats and Dogs dataset to train a DL model in https://www.kaggle.com/tongpython/cat-and-dog/code, provide the training code and accuracy of the validation.
Welcome to discuss any beneficial improvements of the model components or training strategies.