

Assignment-3

Python

Q1. (Base Conversion) [15 marks]

Note:- Please take inputs as command line arguments

You might be familiar with converting a given number in binary (base, $b = 2$) into its decimal equivalent. This question requires you to generalise it to any base b , and the program should be written in Python.

Given a string of a fractional number N_b in base b , convert it into its decimal equivalent N_D . You need to check if the input is a valid number and get rid of leading zeros from the input.

$$\begin{aligned} -999999999 &\leq N_D \leq 999999999, N_D \in \mathbb{R} \\ 2 &\leq b \leq 36 \end{aligned}$$

Example-

For $N_b = \text{HELLO.PY}$	&	$b = 35;$	output, $N_D = 26137359.742041$.
For $N_b = 00101.101$	&	$b = 2;$	output, $N_D = 5.625$.
For $N_b = \text{GJDGXR}$	&	$b = 36;$	output, $N_D = 999999999$.
For $N_b = -4G$	&	$b = 17;$	output, $N_D = -84$.
For $N_b = \text{HELLO.PY}$	&	$b = 10;$	output, "Invalid Input".

Important Note: You are **NOT** allowed to use built in functions like `int()` etc.

Q2. (Intro to ML) [15 marks]

In this question, we will code a 1-D linear regression problem. We will provide pseudo-code here, the students are required to transform it into a python code. Use numpy library for array/vector/matrices etc.

The data files are `train.csv` and `test.csv`.

File structure:

The structure of both files are similar. `Train.csv` has $n_{\text{train}} = 10^4$ rows and `test.csv` has $n_{\text{test}} = 10^3$ rows, each row corresponding to one data point. Each row has two values separated by comma. The first value is feature and second value is label of the data point.

Note:- z^T : transpose of z

Example-

If one row of train.txt is :

4,7

Then feature, $x = 4$ and label, $y = 7$.

Pseudo-code

Step-1: **[1.5 marks]**

- Read files train.csv
- Create vector - X_train (dim - $n_train \times 1$) and vector - y_train (dim - $n_train \times 1$)
- Add a column to X_train so that its dimension becomes $n_train \times 2$. First column of X_train should be all 1 and 2nd column is the same as before adding extra column.

Example -

```
X_train = [  
    2  
    3  
    4  
    ]
```

```
New X_train = [  
    1  2  
    1  3  
    1  4  
    ]
```

Step-2: **[0.5 marks]**

Generate a 2-D vector w (dim: 2×1) initialised randomly with floating point numbers.

Step-3: **[1.5 marks]**

Plot y vs x using matplotlib where x is the feature and y is the label read from the file train.csv.

Consider $x' = [1 \ x]^T$ (prepending 1 to x to generate 2-dimensional x -vector, which is nothing but a row of X_train transposed)

On the same figure plot the line $w^T x'$ vs x .

Your figure should have a dot corresponding to each datapoint (x,y) and a straight line on the plot corresponding to $w^T x'$.

Step-4: **[2.5 marks]**

Set $w_direct = (X_train^T * X_train)^{-1} * X_train^T * y_train$

X_{train} is the $n_{\text{train}} \times 2$ matrix defined earlier and y_{train} is the corresponding label vector.

Plot y vs x using matplotlib where x is the feature and y is the label read from the file train.csv.

Consider $x' = [1 \ x]^T$ (prepending 1 to x to generate 2-dimensional x -vector)

On the same figure plot the line $w_{\text{direct}}^T x'$ vs x .

Your figure should have a dot corresponding to each datapoint (x,y) and a straight line on the plot corresponding to $w_{\text{direct}}^T x'$.

Step-5: (Training) [4.5 marks]

w - 2-dim vector initialised earlier (step-2)

Loop: for $\text{nepoch} = 1$ to N (N is the number of pass through the data (~ 2), play with it to find best fit)

 Loop : for $j = 1$ to n_{train}

$(x,y) \leftarrow j\text{th row of train.csv}$

$x' \leftarrow [1, x]^T$

$w \leftarrow w - \text{eta} * (w^T x' - y) * x'$ ($\text{eta} = 0.00000001$ students can change this value)

 If $j \% 100 == 0$

 Then plot y vs x as earlier and use current value of w to plot $w^T x'$ vs x

Step-6: [0.5 marks]

Finally redraw the plot as earlier with latest value of w .

Note:- Don't use test.csv for training

Step-7: (Evaluation) [4 marks]

- Read files test.csv
- Create vector - X_{test} (dim - $n_{\text{test}} \times 1$) and vector - y_{test} (dim - $n_{\text{test}} \times 1$)
- Add a column to X_{test} so that its dimension becomes $n_{\text{test}} \times 2$. First column of X_{test} should be all 1 and 2nd column is the same as before adding extra column.
- Let $y_{\text{pred1}} = X_{\text{test}} * w$ (w is the final value after doing step 5)
- Calculate root mean squared error between y_{pred1} and y_{test} .
- Let $y_{\text{pred2}} = X_{\text{test}} * w_{\text{direct}}$
- Calculate root mean squared error between y_{pred2} and y_{test} .

Derivation of Updates (optional reading)

In the loop of step 5 we did the following

$$w \leftarrow w - \eta (w^T x' - y) x'$$

The objective of the above step is to reduce the least squared error.

Let $\text{error} = 0.5(w^T x' - y)^2$

So, derivative w.r.t. w gives: $(w^T x' - y) x'$

We need to descend down the gradient to reach the minima, so we subtract η times the above derivative from w to gradually reach minima. We set η to small value because otherwise, w may overshoot the minima. w_{direct} can also be computed on the same line by setting derivative to zero (Google it!).