

# Deep learning & applications

Practice#2

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## Task: *Quick training for binary classification using logistic regression (cross-entropy loss)*

**Input:** 2-dim vector,  $\mathbf{x} = \{x_1, x_2\}$

**Output:** label of the input,  $\mathbf{y} \in \{0,1\}$

**Pseudo code** #you can use numpy module!

**Step 1.** Generate 1000(=m) train samples, 100(=n) test samples:

```
x1_train=[], x2_train=[], y_train=[]
for i in range(m):
    x1_train.append(random.randint(-10, 10))
    x2_train.append(random.randint(-10, 10))
    if x1_train[-1] + x2_train[-1] > 0:
        y_train.append(1)
    else:
        y_train.append(0)
x1_test=[], x2_test=[], y_test=[] #generate 100 test samples!
```

**Step 2.** Update  $W = [w_1, w_2], b$  with 1000 samples for (100= $K$ ) iterations: # $K$  grad updates!

**Step 2-1.** print  $W = [w_1, w_2], b$  at each iteration

**Step 2-2.** calculate the cost with m train samples!

**Step 2-3.** calculate the cost with n test samples!

**Step 2-4.** print accuracy with m train samples! (display the number of correctly predicted outputs/1000\*100)

**Step 2-5.** print accuracy with n test samples! (display the number of correctly predicted outputs/100\*100)

# Report

- You need to submit a short report; (Due: 3/30, 3pm)
  - Format: studentid\_name.pdf
  - Should not be more than 5 pages
  - Should include
    - Time comparison (element-wise version vs. vectorized version,  $(m, K) = (1000, 100)$ )
    - Estimated unknown function parameters  $W$  &  $b$
    - Empirically determined (best) hyper parameter,  $\alpha$
    - Accuracy (fill in the blanks in the tables below and add them to the report)

	$(m, K) = (10, 100)$	$(m, K) = (100, 100)$	$(m, K) = (1000, 100)$
Accuracy (with train set)			
Accuracy (with test set)			

	$(m, K) = (100, 10)$	$(m, K) = (100, 100)$	$(m, K) = (100, 1000)$
Accuracy (with train set)			
Accuracy (with test set)			