# Deep learning & applications

Practice#1-0

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## Task: binary classification using logistic regression (loss = binary 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

**Step 1**. Generate 10000(=m) train samples:

```
 x1\_train=[], x_2\_train=[], y\_train=[]  for i in range(m):  x1\_train.append(random.uniform(-10, 10))   x2\_train.append(random.uniform(-10, 10))  if  x1\_train[-1] + x2\_train[-1] > 0:   y\_train.append(1)  else:  y\_train.append(0)
```

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

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#### Pseudo code

**Step 2**. Initialize  $W = [w_1, w_2]$  and b (small random values? large random values? zeros?)

**Step 3.** Determine the learning rate  $\alpha$  (let's tart with 1/100)

**Step 4**. Update  $W = [w_1, w_2]$ , b with 'm' samples for 5000 (=K) iterations: #K updates with the grad descent

**Step 3-1.** Print *W*, *b* every 500 iterations

**Step 3-2.** Calculate the cost on the 'm' train samples!

**Step 3-3.** Assume that predicted output is 1 if  $\tilde{y} > 0.5$ , other wise 0. Then, print accuracy for the 'm' train samples! (display the number of correctly predicted outputs/m\*100)