Machine Learning HW2 TA Hours

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Announcement

- Strong baseline released !!
- Kaggle deadline postponed to 03/23/2019 11:59:59 (GMT+8)
- Github deadline postponed to 03/24/2019 23:59:59 (GMT+8)

Probabilistic generative model

Outline

- Load data & Normalization
- Posterior probability
- Predict

Load data & Normalization

```
def read(self,name,path):
    with open(path,newline = '') as csvfile:
        rows = np.array(list(csv.reader(csvfile))[1:] ,dtype = float)
    if name == 'x train':
        self.mean = np.mean(rows,axis = 0).reshape(1,-1)
        self.std = np.std(rows,axis = 0).reshape(1,-1)
        self.theta = np.ones((rows.shape[1] + 1,1).dtype = float)
        for i in range(rows.shape[0]):
            rows[i,:] = (rows[i,:] - self.mean) / self.std

elif name == 'X test':
        for i in range(rows.shape[0]):
            rows[i,:] = (rows[i,:] - self.mean) / self.std

self.data[name] = rows
```

$$Z = rac{X - \mu}{\sigma}$$

Posterior probability

If $P(C_1|x) > 0.5$ \blacksquare x belongs to class 1

$$P(C_1|x) = \frac{P(x|C_1)P(C_1)}{P(x|C_1)P(C_1) + P(x|C_2)P(C_2)}$$

$$f_{\mu,\Sigma}(x) = \frac{1}{(2\pi)^{D/2}} \frac{1}{|\Sigma|^{1/2}} exp\left\{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)\right\}$$

$$\mu^* = \frac{1}{79} \sum_{n=1}^{79} x^n$$
 average

$$\mu^* = \frac{1}{79} \sum_{n=1}^{79} x^n \qquad \qquad \Sigma^* = \frac{1}{79} \sum_{n=1}^{79} (x^n - \mu^*) (x^n - \mu^*)^T$$
average

Posterior probability

```
class 0 id = []
class 1 id = []
for i in range(self.data['Y train'].shape[0]):
    if self.data['Y train'][i][0] == 0:
        class 0 id.append(i)
    else:
        class 1 id.append(i)
class 0 = self.data['X train'][class 0 id]
class 1 = self.data['X train'][class 1 id]
```

```
n = class 0.shape[1]
cov \theta = np.zeros((n,n))
cov 1 = np.zeros((n,n))
for i in range(class 0.shape[0]):
for i in range(class 1.shape[0]):
```

```
average
```

```
mean 0 = np.mean(class 0,axis = 0)
mean 1 = np.mean(class 1,axis = 0)
```

$$\Sigma^* = \frac{1}{79} \sum_{n=1}^{79} (x^n - \mu^*) (x^n - \mu^*)^T$$

```
cov 0 += np.dot(np.transpose([class 0[i] - mean 0]), [(class 0[i] - mean 0)]) / class 0.shape[0]
    cov 1 += np.dot(np.transpose([class 1[i] - mean 1]), [(class 1[i] - mean 1)]) / class 1.shape[0]
cov = (cov \ 0*class \ 0.shape[0] + cov \ 1*class \ 1.shape[0]) / (class \ 0.shape[0] + class \ 1.shape[0])
```

Posterior probability $f_{\mu,\Sigma}(x) = \frac{1}{(2\pi)^{D/2}} \frac{1}{|\Sigma|^{1/2}} exp \left\{ -\frac{1}{2} (x - \mu)^T \Sigma^{-1} (x - \mu) \right\}$

$$\begin{split} P(C_1|x) &= \frac{P(x|C_1)P(C_1)}{P(x|C_1)P(C_1) + P(x|C_2)P(C_2)} \\ P(C_1|x) &= \sigma(z) \\ z &= (\mu^1 - \mu^2)^T \Sigma^{-1} x - \frac{1}{2} (\mu^1)^T \Sigma^{-1} \mu^1 + \frac{1}{2} (\mu^2)^T \Sigma^{-1} \mu^2 + \ln \frac{N_1}{N_2} \\ \pmb{w^T} & b \\ \text{self.w = np.transpose(((mean_0 - mean_1)).dot(inv(cov)))} \\ \text{self.b = (-0.5)* (mean_0).dot(inv(cov)).dot(mean_0) \setminus +0.5* (mean_1).dot(inv(cov)).dot(mean_1) \setminus +0.5* (mean_1).dot(inv(cov)).do$$

+ np.log(float(class 0.shape[0]) / class 1.shape[0])

Predict

```
def func(self,x):
    arr = np.empty([x.shape[0],1],dtype=float)
    for i in range(x.shape[0]):
        z = x[i,:].dot(self.w) + self.b
        z *= (-1)
        arr[i][0] = 1 / (1 + np.exp(z))
    return np.clip(arr, 1e-8, 1-(1e-8))
```

```
def predict(self,x):
    ans = np.ones([x.shape[0],1],dtype=int)
    for i in range(x.shape[0]):
        if x[i] > 0.5:
            ans[i] = 0;
    return ans
```

$$P(C_{1}|x) = \sigma(z)$$

$$z = (\mu^{1} - \mu^{2})^{T} \Sigma^{-1} x - \frac{1}{2} (\mu^{1})^{T} \Sigma^{-1} \mu^{1} + \frac{1}{2} (\mu^{2})^{T} \Sigma^{-1} \mu^{2} + \ln \frac{N_{1}}{N_{2}}$$

$$b$$

Logistic regression

Outline

- Load data
- Tips in ML training
- Logictic regression
- Gradient descent

Load data

```
import sys

X_train_fpath = sys.argv[1]

Y_train_fpath = sys.argv[2]

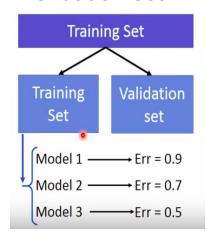
X_test_fpath = sys.argv[3]

output_fpath = sys.argv[4]
```

```
In [3]: X_train = np.genfromtxt(X_train_fpath, delimiter=',', skip_header=1)
Y_train = np.genfromtxt(Y_train_fpath, delimiter=',', skip_header=1)
```

Tips in ML training

Validation set





Normalization

$$z_i = \frac{x_i - \min(x)}{\max(x) - \min(x)}$$

$$Z = \frac{X - \mu}{\sigma}$$

Tips in ML training

- One-hot encoding
- Discretization

Education		
	2	
	1	
	1	
	2	
	3	

fnlwgt	
226802	
89814	
336951	
160323	
103497	

(0 = others, 1 = graduate school, 2 = university, 3 = high school)

others	graduate school	university	high school
0	0	1	0
0	1	0	0
0	1	0	0
0	0	1	0
0	0	0	1

0~100k	100k~200k	200k~300k	300k~400k
0	0	1	0
1	0	0	0
0	0	0	1
0	1	0	0
0	1	0	0

Logictic regression

$$f_{w,b}(x) = \sigma\left(\sum_{i} w_{i} x_{i} + b\right)$$

Output: between 0 and 1

to compute the sigmoid of the input. Use np.clip to avoid overflow. The smallest representable positive number is

- >> np.finfo(np.float32).eps
- >> 1.1920929e-07

 $L(f) = \sum_{n} C(f(x^n), \hat{y}^n)$

Hence, we choose to clip at 1e-6 and 1-1e-6.

get_prob given weight and bias, find out the model predict the probability to output 1 $\sum_n -(\hat{y}^n - f_{w,b}(x^n))x_i^n$

infer if the probability > 0.5, then output 1, or else output 0.

$$\sum_n -(\hat{y}^n - f_{w,b}(x^n))x_i^n$$

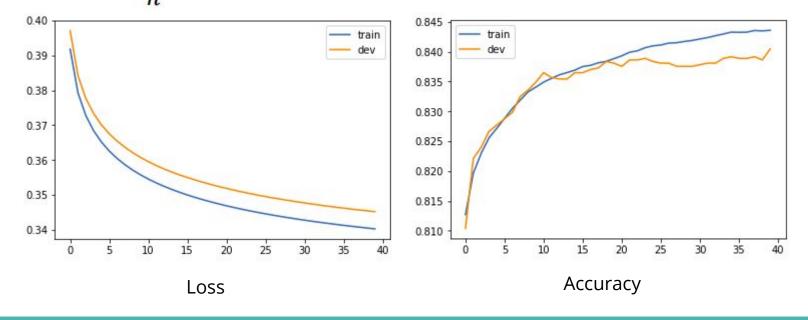
cross entropy: compute the cross-entropy between the model output and the true label.

compute loss to compute the loss function L(w) with input X, Y and w

With math derivation, the gradient of the cross entropy is $\sum_{n} -(\hat{y}^{n} - f_{w,b}(x^{n}))x_{i}^{n}$ gradient

Gradiend descent

$$w_i \leftarrow w_i - \eta \sum_n -\left(\hat{y}^n - f_{w,b}(x^n)\right) x_i^n$$



Shell script example

```
report.pdf
                              $1: raw data (train.csv) $2: test data (test.csv)
hw2 best.sh
                              $3: provided train feature (X train.csv) $4: provided train label (Y train.csv)
hw2_generative.py
                              $5: provided test feature (X test.csv) $6: prediction.csv
hw2 generative.sh
hw2_logistic.py
                                                x = np.genfromtxt(sys.argv[1], delimiter=',')
hw2 logistic.sh
                                                s = np.genfromtxt(sys.argv[2], delimiter=',')
                                                 s = np.genfromtxt(sys.argv[3], delimiter=',')
                                                 f = open(sys.argv[4], "w")
     #!/bin/bash
     python3 hw2_logistic.py $3 $4 $5 $6
```