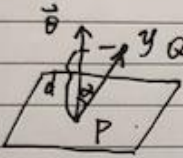


1. Linear Algebra and Probability Review

(a)  $d = |\vec{PA}| \cdot \cos \alpha = \frac{|\vec{\theta}| \cdot |\vec{PA}| \cdot \cos \alpha}{|\vec{\theta}|} = \frac{\vec{\theta} \cdot \vec{PA}}{|\vec{\theta}|}$

$$= \frac{(y - y_P) \vec{\theta}}{|\vec{\theta}|} = \frac{y \vec{\theta} - y_P \vec{\theta}}{|\vec{\theta}|}$$

$$= \frac{y \vec{\theta} + \theta_0}{|\vec{\theta}|} \quad (y_P \text{ is on the hyperplane, so } \vec{\theta} \cdot y_P + \theta_0 = 0)$$

#

(b) $P(X=x) = \frac{\alpha^x e^{-\alpha}}{x!}$, $P(Y=y) = \frac{\beta^y e^{-\beta}}{y!}$, for all $x, y > 0$

$$P(Z=x+y) = \sum P(x) \cdot P(z-x) = \sum \frac{\alpha^x e^{-\alpha}}{x!} \cdot \frac{\beta^{z-x} e^{-\beta}}{(z-x)!}$$

$$= e^{-\alpha-\beta} \sum \frac{\alpha^x \beta^{z-x}}{x! (z-x)!} = e^{-\alpha-\beta} \sum \frac{C_z^x \cdot \alpha^x \beta^{z-x}}{z!}$$

$$= \frac{(\alpha+\beta)^z \cdot e^{-(\alpha+\beta)}}{z!}$$

$z = x+y$ is also Poisson, and the rate $\gamma = \alpha + \beta$

2. Python and Theano

(a) python: 3.6.5

theano: 1.0.2 + 2.g0449c8b

(a) Original Code Result:

```
[-0.57392068  1.35757059  0.01527565 -1.88288076]
```

(b) Eexact solution:

```
In [13]: XTX = np.dot(X.T, X)
         inverseXTX = np.linalg.inv(XTX)
         XXX = np.dot(inverseXTX, X.T)
         theta = np.dot(XXX, Y)
         print(theta)

[-0.57392068  1.35757059  0.01527565 -1.88288076]
```

(c) Sklearn:

```
In [1]: import pandas as pd
        df = pd.read_csv('linear.csv', header=None)
        X = df.drop(0, axis=1)
        y = df[[0]]

In [2]: from sklearn.linear_model import LinearRegression
        regression_model = LinearRegression()
        regression_model.fit(X, y)

Out[2]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=1, normalize=False)

In [3]: for idx, col_name in enumerate(X.columns):
        print("The coefficient for {} is {}".format(col_name, regression_model.coef_[0][idx]))

        intercept = regression_model.intercept_[0]
        print("The intercept for our model is {}".format(intercept))

The coefficient for 1 is -0.5739206829247852
The coefficient for 2 is 1.3575705905387387
The coefficient for 3 is 0.01527565450724058
The coefficient for 4 is 0.0
The intercept for our model is -1.882880764360151
```

(d) SGD

```
%matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import random

data = np.genfromtxt('linear.csv', delimiter=',')
X = data[:,1:]
```

```
Y = data[:,0]
# print(type(X))
```

In [42]:

```
def mini_batch(X, y):
    loc = random.randint(1, 44)
    result_X = X[loc:loc + 5, :]
    result_Y = Y[loc:loc + 5]
    return result_X, result_Y
```

```
test_X, test_y = mini_batch(X, Y)
# print(test_X, test_y)
```

In [20]:

```
import theano
import theano.tensor as T
```

```
x = T.matrix(name='x')      # feature matrix
y = T.vector(name='y')      # response vector
w = theano.shared(np.zeros((4,1)),name='w')    # model parameteres
```

```
risk = T.sum((T.dot(x,w).T - y)**2)/10      # empirical risk
grad_risk = T.grad(risk, wrt=w)            # gradient of the risk
```

In [41]:

```
n_steps = 1000
lost_list = []
w_dict = {}
for i in range(n_steps):
    for_x, for_y = mini_batch(X, Y)
    train_model = theano.function(inputs=[],
                                   outputs=risk,
                                   updates=[(w, w-(1/(i+1))*grad_risk)],
                                   givens={x:for_x, y:for_y})
```

```
#     print(train_model())
#     print(w.get_value())
    lost_list.append(train_model().item(0))
    w_dict[lost_list[-1]] = w.get_value()
```

```
print(w.get_value())
print(lost_list[-1])
```

```
lost_list.sort()
print(w_dict[lost_list[0]])
print(lost_list[0])
[[-0.57992765]
 [ 1.35436854]
 [ 0.01884709]
 [-1.88086824]]
```

0.006610305112215585
[[-0.5673264]
[1.34984663]
[0.017143]
[-1.87757628]]
0.0007559927287628303