

HW3

1~5 bacea
6~10 babbb
11~15 eebdc
16~20 cbabd

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1 $0.1^2(1 - \frac{114}{N}) \geq 0.006 \Rightarrow N \geq 30 \Rightarrow \text{Choose (b)}$

2 Let $\hat{y} = \text{proj}_{\text{col}(X)} y$, \hat{y} is in the column space of X .
Hence, $\hat{y} = Xw$.

We must can find an $\hat{y} \in \text{column } X$ such that $y - \hat{y}$ is orthogonal to column space of X . Therefore, $X^T(\hat{y} - y) = 0 \Rightarrow X^T(Xw - y) = 0 \Rightarrow X^T X w = X^T y$. There exist at least one solution for this equation. $\Rightarrow \text{Choose (a)}$.

3 Let $X = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ $H = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \left(\begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 & 2 \end{bmatrix} = \frac{1}{5} \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$

Let's look at (c), X will become $X' = \begin{bmatrix} 1 \\ 2 \times \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

$X' = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ $H' = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \left(\begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right)^{-1} \begin{bmatrix} 1 & 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \neq H$

Choose (c)

4 ① $\Pr(|v - \theta| > \epsilon) \leq 2 \exp(-2\epsilon^2 N)$ for all $N \in \mathbb{N}$ and $\epsilon > 0$.
Hoeffding's Inequality. $\Rightarrow \text{True}$

② Likelihood = $f(\theta) = \theta^k (1-\theta)^{N-k}$, where k is the number of flipping result being head.
 $f'(\theta) = \frac{k}{\theta} \theta^k (1-\theta)^{N-k} + \frac{k-N}{1-\theta} \theta^k (1-\theta)^{N-k}$

$= \left(\frac{k}{\theta} - \frac{N-k}{1-\theta} \right) f(\theta) = 0$

$\Rightarrow \frac{k}{\theta} = \frac{N-k}{1-\theta} \Rightarrow \theta = \frac{k}{N} = v \Rightarrow v$ maximize likelihood in $[0, 1]$
 $\Rightarrow \text{True}$

③ $E_{in}'(\hat{y}) = \frac{1}{N} \sum_{n=1}^N 2(\hat{y} - y_n) = 2\left(\hat{y} - \sum_{n=1}^N \frac{y_n}{N}\right) = 2(\hat{y} - v) = 0$

$\Rightarrow \hat{y} = v \Rightarrow v$ minimize $E_{in}(\hat{y})$ over all $\hat{y} \in \mathbb{R}$. $\Rightarrow \text{True}$

④ $-\nabla E_{in}(\hat{y}) = -2(\hat{y} - v) \Rightarrow -2(\hat{y} - v)|_{\hat{y}=0} = 2v$

$\Rightarrow 2v$ is the negative gradient direction $-\nabla E_{in}(\hat{y})$ at $\hat{y}=0$.

$\Rightarrow \text{True} \Rightarrow 4$ statements are true. $\Rightarrow \text{Choose (e)}$

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5 Uniform distribution $[0, \hat{\theta}]$.

If $\hat{\theta} \geq \max(y_1, y_2, \dots, y_n)$, p.d.f. $f_Y(y) = \frac{1}{\hat{\theta}}$

$$P(Y_1 = y_1, Y_2 = y_2, \dots, Y_n = y_n) = f_Y(y_1) \times f_Y(y_2) \times \dots \times f_Y(y_n) = \left(\frac{1}{\hat{\theta}}\right)^n \Rightarrow (a)$$

6 Let's look at (b), $\text{err}(w, x, y) = \max(0, -y w^T x)$

$$\text{If } y \neq \text{sign}(w^T x) \Rightarrow y w^T x > 0 \Rightarrow -\nabla \text{err}(w, x, y) = y x$$

$$\text{If } y \neq \text{sign}(w^T x) \Rightarrow y w^T x < 0 \Rightarrow -\nabla \text{err}(w, x, y) = 0$$

The error function is consistent with the update rule.

The answer is (b).

$$7 \quad -\nabla \text{err}_{\text{exp}}(w, x_n, y_n) = -\frac{\partial \text{err}_{\text{exp}}(w, x_n, y_n)}{\partial w_i} = y_n x_n \exp(-y_n w^T x_n)$$

\Rightarrow Choose (a)

$$8 \quad E(w) \approx E(u) + b_E(u)^T V + \frac{1}{2} V^T A_E(u) V, \quad [V = w - u]$$

$$\nabla E(w) = \frac{\partial E(w)}{\partial V} = b_E(u) + A_E(u) V = 0 \Rightarrow V = -(A_E(u))^{-1} b_E(u)$$

\Rightarrow Choose (b)

9 According to the slide of lecture 9

$$\nabla E_{\text{in}}(w) = \frac{1}{N} (2Aw - 2b)$$

$$\nabla(\nabla E_{\text{in}}(w)) = \nabla E(w) \times (\nabla E_{\text{in}}(w))^T = \nabla \left(\frac{2}{N} (w^T A^T - b^T) \right)$$

$$= \frac{2}{N} A^T = \frac{2}{N} X^T X \Rightarrow \text{Choose (b)}$$

$$10 \quad \frac{\partial \text{err}(w, x, y)}{\partial w_{ik}} = -\frac{1}{h_y(x)} \times \frac{1}{\left(\sum_{i=1}^K e^{w_i^T x}\right)^2} \times \left[[y=k] \frac{\sum_{i=1}^K e^{w_i^T x}}{e^{w_i^T x}} - x_i e^{w_i^T x} e^{w_i^T x} \right]$$

$$= \frac{-1}{h_y(x)} \left[[y=k] x_i h_k(x) - x_i h_k(x) h_y(x) \right]$$

$$= (h_k(x) - [y=k]) x_i \Rightarrow \text{Choose (b)}$$

$$11 \quad P(y=1|x) = \frac{e^{w_1^* T x}}{e^{w_1^* T x} + e^{w_2^* T x}} = 1 - p \quad P(y=2|x) = \frac{e^{w_2^* T x}}{e^{w_1^* T x} + e^{w_2^* T x}} = p$$

\parallel $P(y'=1|x)$ \parallel $P(y'=1|x)$

$$\theta(w^T x) = \frac{1}{1 + e^{-w^T x}} = p = \frac{1}{1 + e^{(w_1^* - w_2^*)^T x}}$$

$$\Rightarrow W = w_2^* - w_1^*$$

$$13 \quad \left(\bigcup_{k=1}^d H_k \right) \Rightarrow 2dN \text{ separations}$$

These separations have to shatter 2^N results

$$2^N \leq 2dN \Rightarrow N \leq \log_2 2dN = 1 + \log_2 d + \log_2 N \leq 1 + \frac{N}{2} + \log_2 d$$

when $N \geq 4$

$$\Rightarrow \frac{N}{2} \leq 1 + \log_2 d \Rightarrow N \leq 2(1 + \log_2 d) \Rightarrow \text{Choose (b)}$$

14. Please use “python hw3_14.py hw3_train.txt” to execute.

```
import sys
import numpy as np
from numpy.linalg import inv
from numpy.linalg import multi_dot
in_filename = sys.argv[1]
fin = open(in_filename, 'r')
data = np.loadtxt(fin)
padding = np.full((1000, 1), 1)
fin.close()
data = np.append(padding, data, axis = 1)

x = data[0:1000 , :11]
y = data[0:1000 , 11:12]
xt = x.transpose()
xr = np.dot(xt, x)
xin = inv(xr)
wlin = multi_dot([xin, xt, y])
reg = np.dot(x, wlin)
err = reg - y
error = np.vdot(err, err)/1000
print(error)
```

15. Please use “python hw3_15.py hw3_train.txt” to execute.

```
import sys
import numpy as np
import random
import math
from numpy.linalg import inv
from numpy.linalg import multi_dot
in_filename = sys.argv[1]
fin = open(in_filename, 'r')
data = np.loadtxt(fin)
padding = np.full((1000, 1), 1)
fin.close()
data = np.append(padding, data, axis = 1)
x = data[0:1000 , :11]
y = data[0:1000 , 11:12]
xt = x.transpose()
```

```

xr = np.dot(xt, x)
xin = inv(xr)
Wlin = multi_dot([xin, xt, y])
reg = np.dot(x, Wlin)
err = reg - y
error = np.vdot(err, err)/1000
w = np.zeros((1, 11))
def SGDRegression(w, x, y):
    j = 0
    e = 1
    while e > 1.01 * error:
        k = random.randint(0, 999)
        xn = x[k, :11]
        yn = y[k]
        check = np.vdot(w, xn)
        w = w + 0.001 * 2 * (yn - check) * xn
        j = j+1
        r = np.dot(x, w.transpose())
        er = r - y
        e = np.vdot(er, er)/1000
    return j
iternum = 0
for i in range(1000):
    num = SGDRegression(w, x, y)
    iternum = iternum + num
print(iternum/1000)

```

16. Please use “python hw3_16.py hw3_train.txt” to execute.

```

import sys
import numpy as np
import random
import math
from numpy.linalg import inv
from numpy.linalg import multi_dot
in_filename = sys.argv[1]
fin = open(in_filename, 'r')
data = np.loadtxt(fin)
padding = np.full((1000, 1), 1)
fin.close()

```

```

data = np.append(padding, data, axis = 1)
x = data[0:1000 , :11]
y = data[0:1000 , 11:12]
w = np.zeros((1, 11))
def SGDLogistic(w, x, y):
    e = 0
    for i in range(500):
        k = random.randint(0, 999)
        xn = x[k, :11]
        yn = y[k]
        check = np.vdot(w, xn)
        w = w + 0.001 * yn * xn * (1 / (1 + math.exp(yn * check)))
    for i in range(1000):
        xn = x[i, :11]
        yn = y[i]
        e = e + np.log(((1 + math.exp(-yn * np.vdot(w, xn)))))
    return e/1000
aver = 0
for i in range(1000):
    num = SGDLogistic(w, x, y)
    aver = aver + num
print(aver/1000)

```

17. Please use “python hw3_17.py hw3_train.txt” to execute.

```

import sys
import numpy as np
import random
import math
from numpy.linalg import inv
from numpy.linalg import multi_dot
in_filename = sys.argv[1]
fin = open(in_filename, 'r')
data = np.loadtxt(fin)
padding = np.full((1000, 1), 1)
fin.close()
data = np.append(padding, data, axis = 1)
x = data[0:1000 , :11]
y = data[0:1000 , 11:12]
xt = x.transpose()

```

```

xr = np.dot(xt, x)
xin = inv(xr)
Wlin = multi_dot([xin, xt, y])
w = Wlin.transpose()
def SGDLogistic(w, x, y):
    e = 0
    for i in range(500):
        k = random.randint(0, 999)
        xn = x[k, :11]
        yn = y[k]
        check = np.vdot(w, xn)
        w = w + 0.001 * yn * xn * (1 / (1 + math.exp(yn * check)))
    for i in range(1000):
        xn = x[i, :11]
        yn = y[i]
        e = e + np.log(((1 + math.exp(-yn * np.vdot(w, xn)))))
    return e/1000
aver = 0
for i in range(1000):
    num = SGDLogistic(w, x, y)
    aver = aver + num
print(aver/1000)

```

18. Please use “python hw3_18.py hw3_train.txt hw3_test.txt” to execute.

```

import sys
import numpy as np
from numpy.linalg import inv
from numpy.linalg import multi_dot
in_filename1 = sys.argv[1]
in_filename2 = sys.argv[2]
fin = open(in_filename1, 'r')
data = np.loadtxt(fin)
padding1 = np.full((1000, 1), 1)
padding2 = np.full((3000, 1), 1)
fin.close()
fin = open(in_filename2, 'r')
test = np.loadtxt(fin)
data = np.append(padding1, data, axis = 1)
test = np.append(padding2, test, axis = 1)

```

```

x = data[0:1000 , :11]
y = data[0:1000 , 11:12]
xt = x.transpose()
xr = np.dot(xt, x)
xin = inv(xr)
Wlin = multi_dot([xin, xt, y])
xtest = test[0:3000 , :11]
ytest = test[0:3000 , 11:12]
reg1 = np.dot(x, Wlin)
reg2 = np.dot(xtest, Wlin)
err1 = 0
err2 = 0
for i in range(1000):
    if reg1[i][0] * y[i][0] < 0:
        err1 = err1 + 1
for i in range(3000):
    if reg2[i][0] * ytest[i][0] < 0:
        err2 = err2 + 1
print(abs(err1-err2/3)/1000)
fin.close()

```

19. Please use “python hw3_19.py hw3_train.txt hw3_test.txt” to execute.

```

import sys
import numpy as np
from numpy.linalg import inv
from numpy.linalg import multi_dot
def Qtransform(A, Q):
    a = A[0:A.shape[0], 0:1]
    b = A[0:A.shape[0], 1:11]
    c = A[0:A.shape[0], 11:12]
    for i in range(Q):
        paste = np.power(b, i+1)
        a = np.concatenate((a, paste), axis=1)
    a = np.concatenate((a, c), axis=1)
    return a
in_filename1 = sys.argv[1]
in_filename2 = sys.argv[2]
fin = open(in_filename1, 'r')

```



```

data = np.loadtxt(fin)
padding1 = np.full((1000, 1), 1)
padding2 = np.full((3000, 1), 1)
fin.close()
fin = open(in_filename2, 'r')
test = np.loadtxt(fin)
data = np.append(padding1, data, axis = 1)
test = np.append(padding2, test, axis = 1)
data = Qtransform(data, 3)
test = Qtransform(test, 3)
x = data[0:1000 , 0:31]
y = data[0:1000 , 31:32]
xt = x.transpose()
xr = np.dot(xt, x)
xin = inv(xr)
Wlin = multi_dot([xin, xt, y])
xtest = test[0:3000 , 0:31]
ytest = test[0:3000 , 31:32]
reg1 = np.dot(x, Wlin)
reg2 = np.dot(xtest, Wlin)
err1 = 0
err2 = 0
for i in range(1000):
    if reg1[i][0] * y[i][0] < 0:
        err1 = err1 + 1
for i in range(3000):
    if reg2[i][0] * ytest[i][0] < 0:
        err2 = err2 + 1
print(abs(err1-err2/3)/1000)
fin.close()

```

20. Please use “python hw3_20.py hw3_train.txt hw3_test.txt” to execute.

```

import sys
import numpy as np
from numpy.linalg import inv
from numpy.linalg import multi_dot
def Qtransform(A, Q):
    a = A[0:A.shape[0], 0:1]
    b = A[0:A.shape[0], 1:11]

```

```

    c = A[0:A.shape[0], 11:12]
    for i in range(Q):
        paste = np.power(b, i+1)
        a = np.concatenate((a, paste), axis=1)
    a = np.concatenate((a, c), axis=1)
    return a
in_filename1 = sys.argv[1]
in_filename2 = sys.argv[2]
fin = open(in_filename1, 'r')
data = np.loadtxt(fin)
padding1 = np.full((1000, 1), 1)
padding2 = np.full((3000, 1), 1)
fin.close()
fin = open(in_filename2, 'r')
test = np.loadtxt(fin)
data = np.append(padding1, data, axis = 1)
test = np.append(padding2, test, axis = 1)
data = Qtransform(data, 10)
test = Qtransform(test, 10)
x = data[0:1000 , 0:101]
y = data[0:1000 , 101:102]
xt = x.transpose()
xr = np.dot(xt, x)
xin = inv(xr)
Wlin = multi_dot([xin, xt, y])
xtest = test[0:3000 , 0:101]
ytest = test[0:3000 , 101:102]
reg1 = np.dot(x, Wlin)
reg2 = np.dot(xtest, Wlin)
err1 = 0
err2 = 0
for i in range(1000):
    if reg1[i][0] * y[i][0] < 0:
        err1 = err1 + 1
for i in range(3000):
    if reg2[i][0] * ytest[i][0] < 0:
        err2 = err2 + 1
print(abs(err1-err2/3)/1000)

```

```
fin.close()
```