

# Homework #0 Solution A

Spring 2020, CSE 446/546: Machine Learning

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Due: 4/8/20 11:59 PM

A: 37 points. B: 3 points

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## Probability and Statistics

### A.1 Solution:

From the problem, let's set  $T$ :=Tested positive and  $D$ :=Have disease. And we can know that  $P(T | D) = 0.99$  and  $P(D) = \frac{1}{10000}$ . Now we need  $P(D | T)$ . By Bayes Rule, we have:

$$P(D | T) = \frac{P(D \cap T)}{P(T)} = \frac{P(T | D)P(D)}{P(T | D)P(D) + P(T | \text{no}D)P(\text{no}D)} = \frac{0.99 \frac{1}{10000}}{0.99 \frac{1}{10000} + 0.01 \frac{9999}{10000}} = \frac{1}{102}.$$

### A.2 Solution:

a. By Law of Total Expectation,

$$\begin{aligned} \text{Cov}(X, Y) &= E[E[(X - E[X])(Y - E[Y]) | X]] \\ &= E[(X - E[X])(E[Y | X] - E[Y] | X)] \\ &= E[(X - E[X])(X - E[X])] \\ &= E[(X - E[X])^2] \end{aligned}$$

b. if  $X$  and  $Y$  are independent, then:

$$\begin{aligned} &E[(X - E[X])(Y - E[Y])] \\ &= E[X - E[X]]E[Y - E[Y]] \\ &= 0 \end{aligned}$$

### A.3 Solution:

a. For the cdf of  $Z$ ,  $H(Z) = P(Z \leq z) = P(X+Y \leq z)$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{z-x} f(x)g(y)dydx$$

To get the density of  $Z$  we need to differentiate this with respect to  $Z$ .

The only  $z$  dependence is in the upper limit of the inside integral.

$$\text{so } h(z) = \frac{\partial H}{\partial z} = \int_{-\infty}^{\infty} \frac{\partial \int_{-\infty}^{z-x} f(x)g(y)dy}{\partial z} dx$$

$$\text{finally we get } h(z) = \int_{-\infty}^{\infty} f(x)g(z-x)dx$$

- b. According to our solution above,  
 $f(z) = z$  for  $0 \leq z \leq 1$ ;  
 $f(z) = 2-z$  for  $1 \leq z \leq 2$ ;  
 $f(z) = 0$  otherwise.

**A.4 Solution:**

$$a = \frac{1}{\sigma}, b = -\frac{\mu}{\sigma} \quad (1)$$

**A.5 Solution:** From CLM,

$$mean = 0, variance = \sigma^2 \quad (2)$$

**A.6 Solution:**

- a.  $E[\hat{F}_n(x)] = \frac{n \cdot E[F]}{n}$   
 $= E[F] = F(x)$
- b. Notice that  $1\{X \leq x\}$  is a Bernoulli random variable with  
 $P[1\{X \leq x\}] = F(x)$  so we can see  $\hat{F}_n(x)$  is just a mean of  $n$  Bernoulli variables  
Hence the variance is in a form with  $\frac{p(1-p)}{n}$  so it's  $\frac{F(x)(1-F(x))}{n}$
- c. Set  $G(x) = x \cdot (1-x)$ , for  $0 \leq x \leq 1$ ,  $G(x)$  has the maximum value at  $x = 0.5$ , and  $G(x) = 0.25$ .  
So the maximum of  $E[(\hat{F}_n(x) - F(x))^2]$  is 0.25, and we can conclude that  $E[(\hat{F}_n(x) - F(x))^2] \leq \frac{1}{4n}$ .

## Linear Algebra and Vector Calculus

**A.7 Solution:**

- a.  $\text{rank}(A) = 2$  and  $\text{rank}(B) = 2$

b.  $\left\{ \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} \right\}.$

**A.8 Solution:**

a.  $A \cdot c = \begin{bmatrix} 0 & 2 & 4 \\ 2 & 4 & 2 \\ 3 & 3 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 6 \\ 8 \\ 7 \end{bmatrix}$

- b. set  $x = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$  and from  $Ax=b$ , we can get a set of linear equations.

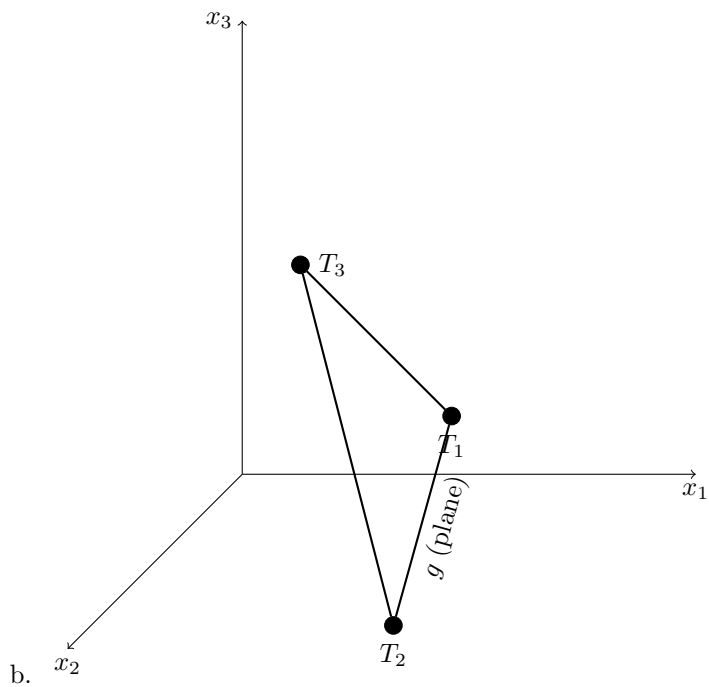
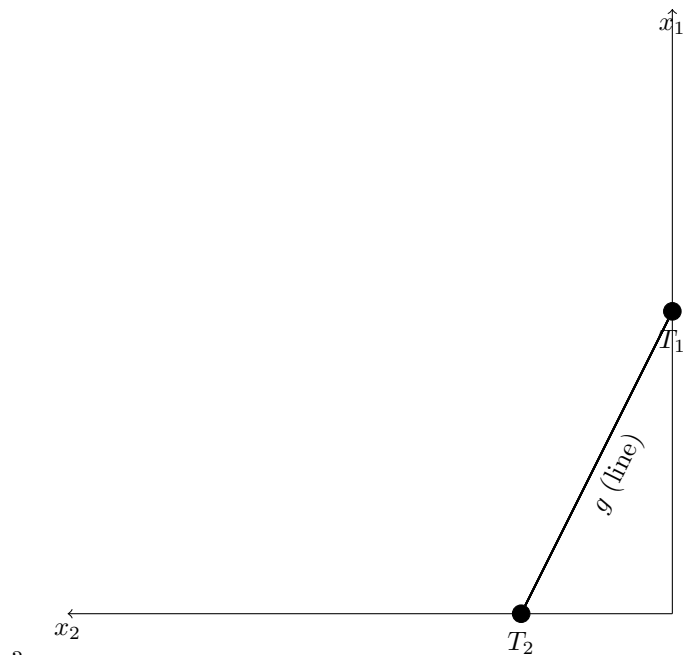
$$\begin{cases} 2b + 4c = -2 \\ (1) \\ 2a + 4b + 2c = -2 \\ (2) \\ 3a + 3b + c = -4 \\ (3) \end{cases}$$

if we use  $3 \cdot (2) - 2 \cdot (3)$ ,

we can get  $6b+4c=2$ , and we combine this and (1)

$$b = 1, c = -1 \text{ and so } a = -2 \text{ and finally } x = \begin{bmatrix} -2 \\ 1 \\ -1 \end{bmatrix}$$

**A.9 Solution:**



c. according to the hint, the minimized distance is

$$\|x_0 - \tilde{x}_0\| = \left| \frac{w^T \cdot x_0 + b}{\|w\|} \right|$$

**A.10 Solution:**

- a.  $f(x, y) = \sum_{i,j=1}^n x_i A_{i,j} x_j + \sum_{i,j=1}^n y_i B_{i,j} x_j + c.$
- b.  $\nabla_x f(x, y) = (A + A^T)x + B^T y$
- c.  $\nabla_y f(x, y) = B y$

#### A.11 Solution:

the solutions to both questions are included in this screenshot:

```
print(A_inv)
print(A_inv.dot(b))
print(np.dot(A, c))

[[ 0.125 -0.625  0.75 ]
 [-0.25  0.75 -0.5 ]
 [ 0.375 -0.375  0.25 ]]
[[-2.]
 [ 1.]
 [-1.]]
[[6]
 [8]
 [7]]
```

Listing 1: A.11 code

```
#!/usr/bin/env python
# coding: utf-8

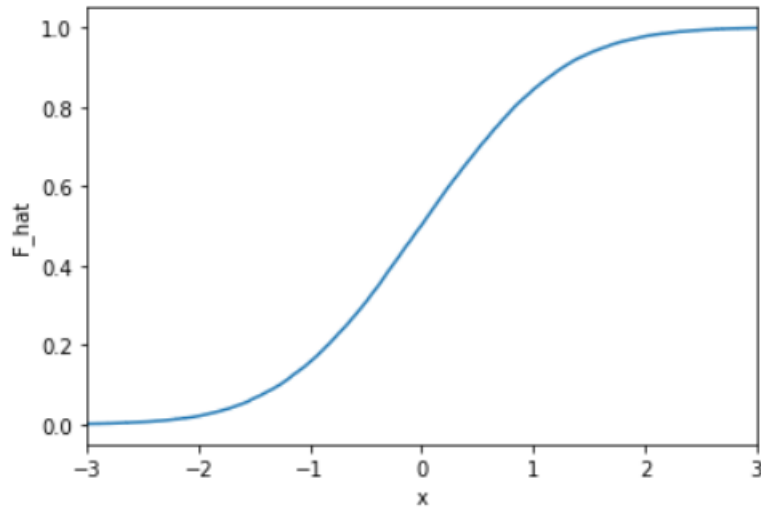
# In [1]:

import numpy as np
A = ([0, 2, 4],
      [2, 4, 2],
      [3, 3, 1]);
b = ([-2], [-2], [-4]);
c = ([1], [1], [1]);
A_inv = np.linalg.inv(A)
print(A_inv)
print(A_inv.dot(b))
print(np.dot(A, c))

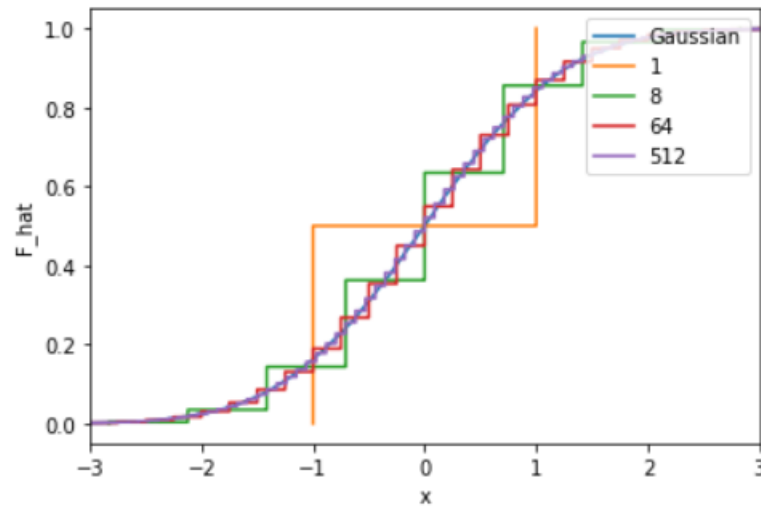
# In [ ]:
```

#### A.12 Solution:

- a. Since  $\frac{1}{4n} = 0.0025^2 = \frac{1}{160000}$ , we can get that  $n = \frac{1}{40000}$  and the plot:



b. The plot is as follows:



Listing 2: A.12 code

```
#!/usr/bin/env python
# coding: utf-8
```

```
# In[13]:
```

```
import numpy as np
n = 40000
Z=np.random.randn(n);
import matplotlib.pyplot as plt
plt.step(sorted(Z), np.arange(1,n+1)/float(n))
plt.xlim(-3,3)
plt.xlabel("x")
plt.ylabel("F_hat")
```

```
Y_1 = np.sum(np.sign(np.random.randn(n, 1))*np.sqrt(1./1), axis=1)
```

```

Y_8 = np.sum(np.sign(np.random.randn(n, 8))*np.sqrt(1./8), axis=1)
Y_64 = np.sum(np.sign(np.random.randn(n, 64))*np.sqrt(1./64), axis=1)
Y_512 = np.sum(np.sign(np.random.randn(n, 512))*np.sqrt(1./512), axis=1)
plt.step(sorted(Y_1), np.arange(1,n+1)/float(n))
plt.step(sorted(Y_8), np.arange(1,n+1)/float(n))
plt.step(sorted(Y_64), np.arange(1,n+1)/float(n))
plt.step(sorted(Y_512), np.arange(1,n+1)/float(n))
plt.legend(["Gaussian", "1", "8", "64", "512"], loc="upper_right")

```

*# In[ ]:*