

# Homework 0

Amy Pitts

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## 1 Instructions

*No deliverables*

## 2 Course Setup

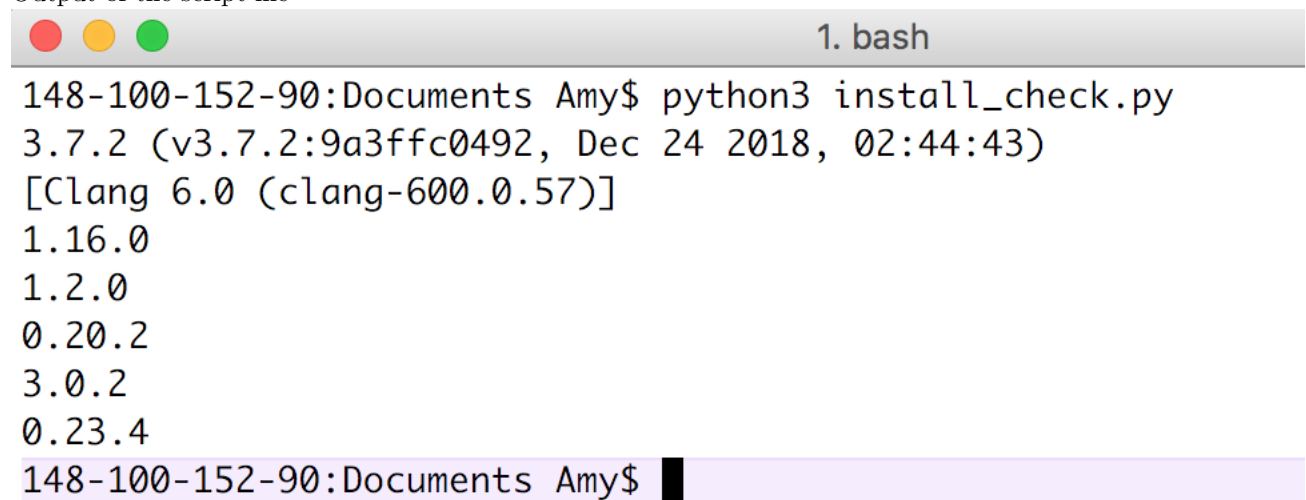
### 2.1 Python

After downloading all the required packages running this script shows that everything is install and their versions.

```
import sys
import numpy
import scipy
import sklearn
import matplotlib
import pandas

print (sys.version) #prints python version
print (numpy.__version__) #prints numpy s version
print (scipy.__version__) #print scipy s version
print (sklearn.__version__) #print sklearn's version
print (matplotlib.__version__) #print matplotlib version
print (pandas.__version__) #prints pandas version
```

Output of the script file



```
1. bash
148-100-152-90:Documents Amy$ python3 install_check.py
3.7.2 (v3.7.2:9a3ffc0492, Dec 24 2018, 02:44:43)
[Clang 6.0 (clang-600.0.57)]
1.16.0
1.2.0
0.20.2
3.0.2
0.23.4
148-100-152-90:Documents Amy$
```

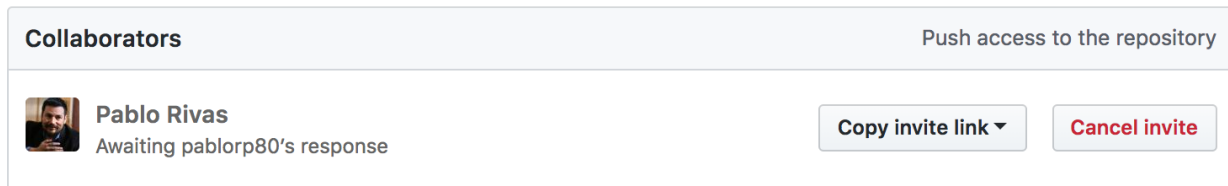
## 2.2 GitHub Class Repository

- Github name and class repository:



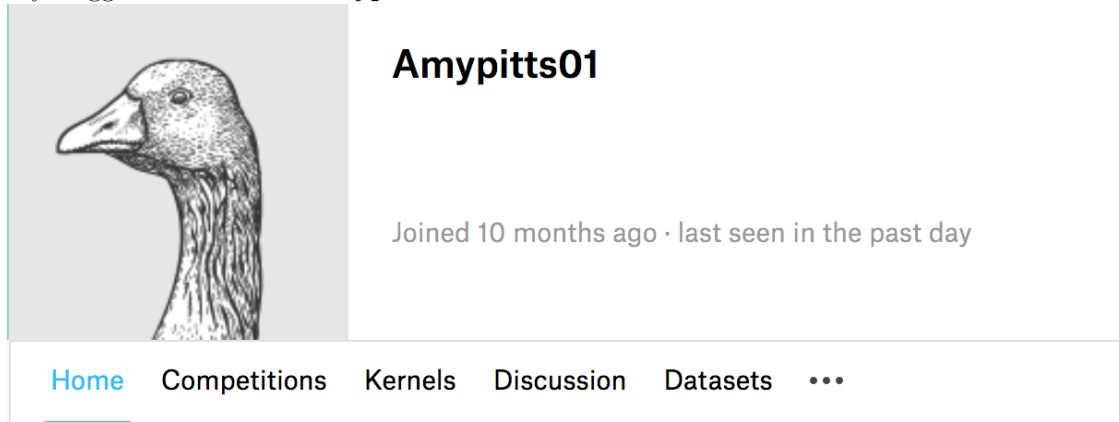
- Link to class repository: <https://github.com/amypitts01/data440/invitations>

- Proof that pablorp80 was asked to be a collaborator:



## 2.3 Kaggle Account

My Kaggle user name is: Amypitts01



## 3 Mathematics Requirements

*No deliverables*

## 4 Problems

1. For the function  $g(x) = -3x^2 + 24x - 30$ , find the value for  $x$  that maximizes  $g(x)$ .

**Solution:** To maximize this function the first step is to take the derivative such that  $g'(x) = -6x + 24$ . Setting the derivative equal to 0 to find the critical values we have:

$$\begin{aligned} -6x + 24 &= 0 \\ -6(x - 4) &= 0 \\ x - 4 &= 0 \\ x &= 4. \end{aligned}$$

The maximum occurs when the derivative is positive left of the critical value and negative right of the critical value. Since there is only one critical value we do not need to compare to other critical values to find the

biggest  $y$  value. To show that 4 is in fact a max we see that:

Left:  $g'(0) = -6(0) + 24 = 24$  *positive*

Right:  $g'(10) = -6(10) + 24 = -36$  *negative*

Therefore,  $x = 4$  maximizes the function.

2. Consider the following function:

$$f(x) = 3x_0^3 - 2x_0x_1^2 + 4x_1 - 8$$

what are the partial derivatives of  $f(x)$  with respect to  $x_0$  and  $x_1$ .

**Solution:**

$$\frac{\partial f}{\partial x_0} = 9x_0^2 - 2x_1^2$$

$$\frac{\partial f}{\partial x_1} = -4x_0x_1 + 4$$

3. Consider the matrix  $A = \begin{bmatrix} 1 & 4 & -3 \\ 2 & -1 & 3 \end{bmatrix}$  and  $B = \begin{bmatrix} -2 & 0 & 5 \\ 0 & -1 & 4 \end{bmatrix}$  then answer the following and verify your answers in Python:

(a) can you multiply the two matrices? elaborate on your answer.

- **Solution:** No you can not multiply the two because  $A$  does not have the same number of columns and  $B$ 's rows. They are both  $(2, 3)$  matrices and to be able to multiply  $A$  would need to be  $(2, 3)$  and  $B$  dims  $(3, 2)$  or  $A$  could be  $(3, 2)$  and  $B$   $(2, 3)$ .

(b) multiply  $A^T$  and  $B$  and give its *rank*.

- **Solution:** First we need to find the transposes of  $A$ . This results in  $A^T = \begin{bmatrix} 1 & 2 \\ 4 & -1 \\ -3 & 3 \end{bmatrix}$ . Now we

$$\text{can multiply } A^T B. \text{ This gives us } A^T B = \begin{bmatrix} 1 & 2 \\ 4 & -1 \\ -3 & 3 \end{bmatrix} \begin{bmatrix} -2 & 0 & 5 \\ 0 & -1 & 4 \end{bmatrix}$$

$$= \begin{bmatrix} (1)(-2) + (2)(0) & (1)(0) + (2)(-1) & (1)(5) + (2)(4) \\ (4)(-2) + (-1)(0) & (4)(0) + (-1)(-1) & (4)(5) + (-1)(4) \\ (-3)(-2) + (3)(0) & (-3)(0) + (3)(-1) & (-3)(5) + (3)(4) \end{bmatrix}$$

$$= \begin{bmatrix} -2 & -2 & 13 \\ -8 & 1 & 16 \\ 6 & -3 & -3 \end{bmatrix}.$$

- **Solution:** Next to find the rank we need to row reduce until we reach row echelon form. Then we have  $\begin{bmatrix} -2 & -2 & 13 \\ -8 & 1 & 16 \\ 6 & -3 & -3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & \frac{-13}{2} \\ -8 & 1 & 16 \\ 6 & -3 & -3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & \frac{-13}{2} \\ 0 & 9 & -36 \\ 0 & -9 & 36 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & \frac{-13}{2} \\ 0 & 1 & -4 \\ 0 & 0 & 0 \end{bmatrix}$ . The row echelon of  $A^T B$  shows two independent row vectors therefore the **rank is 2**.

(c) (extra credit) let  $c = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}$  be a new matrix; what is the result of  $AB^T + C^{-1}$ ?

**Solution:** Here is the script I ran to check the matrices.

```
import numpy as np

#The two arrays given in question three
```

```
A = np.array([[1,4,-3],[2,-1,3]])
B = np.array([[-2,0,5],[0,-1,4]])
```

*#Question 3.a*

```
A.dot(B) #This shows that we can not multiply the two arrays
```

*#Question 3.b*

```
A.T.dot(B) #this is just the transpose
```

```
np.linalg.matrix_rank(A.T.dot(B)) #this is the rank
```

Python answer for 3.a

```
148-100-152-90:hw0 Amy$ python3 matrix_checks.py
```

```
Traceback (most recent call last):
```

```
File "matrix_checks.py", line 8, in <module>
```

```
A.dot(B) #This shows that we can not multiply the two arrays
```

```
ValueError: shapes (2,3) and (2,3) not aligned: 3 (dim 1) != 2 (dim 0)
```

Python answer for 3.b

```
>>> A.T.dot(B)
```

```
array([[ -2,  -2, 13],
       [ -8,   1, 16],
       [  6,  -3, -3]])
```

```
>>> np.linalg.matrix_rank(A.T.dot(B))
```

```
2
```

4. Give the mathematical definitions of the simple Gaussian, multivariate Gaussian, Bernoulli, binomial, and exponential distributions.

**Solutions:**

**Gaussian:** This is also known as the normal distribution and has a pdf of  $f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\}$  and its notation is  $N(\mu, \sigma^2)$ .

**Multivariate Gaussian:** This distribution has a pdf of  $f_X(x) = (2\pi)^{-\frac{k}{2}} |\Sigma|^{-\frac{1}{2}} e^{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)}$  and its notation is  $MVN(\mu, \Sigma)$ .

**Bernoulli:** This distribution has a pdf of  $f_X(x) = p^x(1-p)^{1-x}$  with notation  $\text{Bern}(p)$ .

**binomial:** This distribution has a pdf of  $f_X(x) = \binom{n}{k} p^x(1-p)^{n-x}$  with notation  $\text{Bin}(n, p)$ .

**exponential distributions:** This distribution has a pdf of  $f(x; \lambda) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases}$ .

5. (extra credit) What is the relationship between the Bernoulli and binomial.

**Solution:** The Bernoulli distribution is made up of Bernoulli trials where an event occurs  $p$  or the event does not occur  $1-p$ . There are only two options: success or failure. When you have  $n$  independent Bernoulli trials, those successes and failures are represented in a Binomial Distribution.

6. Suppose that random variable  $X \sim N(2, 3)$ . What is its expected value?

**Solution:**  $N(2, 3)$  is a normal distribution; it has a  $E[x] = \mu$ . Since  $N(\mu, \sigma^2)$ , we have that in our case  $E[x] = 2$ .

7. An *euclidean projection* of a  $d$ -dimensional point  $y \in \mathbb{R}^d$  to a set  $\mathcal{Z}$  is given by the following optimization problem:

$$x^* = \arg \min_x \|x - y\|_2^2, \text{ subject to } x \in \mathcal{Z}$$

where  $\mathcal{Z}$  is the feasible set,  $\|\cdot\|_2$  is the  $l_2$ -norm (euclidean) of a vector, and  $x^* \in \mathbb{R}^d$  is the projected vector.

(a) What is  $x^*$  if  $y = 1.1$  and  $\mathcal{Z} = \mathbb{N}$ ?

- **Solution:** the  $\arg \min_x \|x - y\|_2^2$  is the value of  $x$  for which  $\|x - y\|_2^2$  attains it's min value. Since  $x \in \mathcal{Z} = \mathbb{N}$  we can use trial and errors to find the min.

First lets start with  $x = 0$  then we have:  $\|0 - 1.1\|_2^2 = \sqrt{(0 - 1.1) \cdot (0 - 1.1)}^2 = 1.21$ .

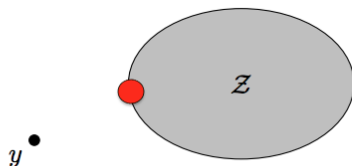
Now looking at  $x = 1$  we have:  $\|1 - 1.1\|_2^2 = \sqrt{(1 - 1.1) \cdot (1 - 1.1)}^2 = 0.01$ .

Finally looking at  $x = 2$  we have  $\|2 - 1.1\|_2^2 = \sqrt{(2 - 1.1) \cdot (2 - 1.1)}^2 = .81$ .

Because the natural numbers only increase from here the smallest value is going to be made when  $x = 1$  obtaining a  $x^* = 0.01$ .

(b) Locate  $x^*$  in the following picture:

The red dot should be the closest point on the perimeter of the  $\mathcal{Z}$ .



**Solution:**

8. Suppose that random variable  $Y$  has distribution:

$$p(Y = y) = \begin{cases} e^{-y} & y \geq 0 \\ 0 & otherwise \end{cases}$$

(a) Verify that  $\int_{y=-\infty}^{\infty} p(Y = y)dy = 1$

(b) What is  $\mu_y = E[Y] = \int_{y=-\infty}^{\infty} p(Y = y)ydy$ ; that is, the expected value of  $Y$ ?

(c) What is  $\sigma^2 = Var[Y] = \int_{y=-\infty}^{\infty} p(Y = y)(y - \mu_y)^2dy$ ; that is, the variance of  $Y$ ?

(d) What is  $E[Y|Y \geq 10]$ ; that is, the expected value of  $Y$ , given that (or conditioned on)  $Y \geq 10$ ?

**8.a Solution:**

$$\begin{aligned} \int_{y=-\infty}^{\infty} p(Y = y)dy &= 1 \\ \int_{-\infty}^0 0dy + \int_0^{\infty} e^{-y}dy &= 1 \\ 0 - e^{-y}|_0^{\infty} &= 1 \\ 1 &= 1. \end{aligned}$$

**8.b Solution:**

$$\begin{aligned} \mu_y = E[Y] &= \int_{y=-\infty}^{\infty} p(Y = y)ydy \\ &= \int_{-\infty}^0 0dy + \int_0^{\infty} ye^{-y}dy \\ &= 0 + \int_0^{\infty} ye^{-y}dy \end{aligned}$$

Now we need to do integration by parts. Using  $u = y$ ,  $du = 1dy$ ,  $v = -e^{-y}$  and  $dv = e^{-y}$  we have:

$$\begin{aligned}\mu_y &= \int_0^\infty ye^{-y}dy \\ &= -ye^{-y}|_0^\infty + \int_0^\infty e^{-y}dy \\ &= 0 - e^{-y}|_0^\infty \\ &= 1.\end{aligned}$$

### 8.c Solution:

$$\begin{aligned}\sigma^2 = Var[Y] &= \int_{y=-\infty}^\infty p(Y=y)(y-\mu_y)^2dy \\ &= \int_{-\infty}^0 0(y-\mu_y)^2dy + \int_0^\infty e^{-y}(y-\mu_y)^2dy \\ &= \int_0^\infty e^{-y}(y^2 - 2y\mu_y + \mu_y^2)dy \\ &= \int_0^\infty e^{-y}y^2dy - \int_0^\infty 2e^{-y}y\mu_ydy + \int_0^\infty e^{-y}\mu_y^2dy\end{aligned}$$

Now looking at each integral individually we have

$$\begin{aligned}\int_0^\infty e^{-y}y^2dy &= -y^2e^{-y}|_0^\infty - 2\int_0^\infty e^{-y}ydy \\ &= 0 + 2E[Y] \\ &= 2.\end{aligned}$$

$$\begin{aligned}-\int_0^\infty 2e^{-y}y\mu_ydy &= -\int_0^\infty 2e^{-y}y\mu_ydy \\ &= -2\mu_y\int_0^\infty e^{-y}ydy \\ &= -2\mu_yE[Y] \\ &= -2\mu_y.\end{aligned}$$

$$\int_0^\infty e^{-y}\mu_y^2dy = \mu_y^2\int_0^\infty e^{-y}dy = \mu_y^2.$$

Putting this all together we have that

$$Var[Y] = \int_0^\infty e^{-y}y^2dy - \int_0^\infty 2e^{-y}y\mu_ydy + \int_0^\infty e^{-y}\mu_y^2dy = 2 - 2\mu_y + \mu_y^2.$$

### 8.d Solution:

$$\begin{aligned}E[Y|Y \geq 10] &= \int_{10}^\infty p(Y=y)ydy \\ &= \int_{10}^\infty ye^{-y}dy\end{aligned}$$

Now we need to do integration by parts. Using  $u = y$ ,  $du = 1dy$ ,  $v = -e^{-y}$  and  $dv = e^{-y}$  we have:

$$\begin{aligned}\int_{10}^{\infty} ye^{-y}dy &= -ye^{-y}|_{10}^{\infty} + \int_{10}^{\infty} e^{-y}dy \\&= 0 + 10e^{-10} - e^{-y}|_{10}^{\infty} \\&= 10e^{-10} + e^{-10} \\&= 11e^{-10}.\end{aligned}$$