

# CMPS 142 Machine Learning

## Spring 2018, Homework #0

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### Question 1

X is Normal, mean=  $\mu$ , variance=  $\sigma^2$ .

The Probability Density Function (PDF) is

$$\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

### Question 2

Probability for heads on a coin is  $\lambda$ .

#### A

The probability of getting a head is  $1/2$ .

So, the probability for getting a head on the  $(k+1)$ -th toss is

$$\frac{1}{2}^{(k+1)}$$

#### B

The expected number of failures before first success is

$$E(X) = \frac{(1-p)}{p} = \frac{1-0.5}{0.5} = 1$$

Therefore, 2 tosses are expected to get the first head.

### Question 3

A

$\text{Var}(X) = E[(X - E[X])^2]$ . Prove  $\text{Var}(X) = E[X^2] - E[X]^2$

$$\begin{aligned} E[(X - E[X])^2] &= E[X^2 - 2XE[X] + E[X]^2] \\ &= E[X^2] - 2E[XE[X]] + E[E[X]^2] \\ &= E[X^2] - 2(E[X]E[X]) + E[X]^2 \\ &= E[X^2] - 2(E[X]^2) + E[X]^2 \\ &= E[X^2] - E[X]^2 \end{aligned}$$

B

$\text{Var}(X) = E[X^2] - E[X]^2$ , so  $1 - 0^2 = 1$ .

$Y = a + bX$ ,  $\text{Var}(Y) = ?$

$$\begin{aligned} \text{Var}(Y) &= E[((a + bX) - E[a + bX])^2] \\ &= E[((a + bX) - (a - b\mu))^2] \\ &= E[a^2(X - \mu)^2] \\ &= a^2 E[(X - \mu)^2] \end{aligned}$$

$\text{Var}(Y) = a^2 \text{Var}(X)$ , by prior problem.

### Question 4

A

$$\begin{aligned} \frac{\partial f}{\partial x} &= 6x - y - 11 \\ \frac{\partial f}{\partial y} &= 2y - x \end{aligned}$$

## B

Find critical point of  $f(x,y)$ .

$$6x - y - 11 = 0$$

$$2y - x = 0$$

Solve for  $y$  from the first equation

$$y = 6x - 11$$

Substitute  $y$  into the second equation

$$2(6x - 11) - x = 0 \Rightarrow 12x - 22 - x = 0 \Rightarrow 11x - 22 = 0 \Rightarrow x = 2$$

Substitute  $x$  back into the second equation

$$y = 6 \cdot 2 - 11 = 1$$

Now that we have the  $x,y$  pair, use the 2nd Derivative Test to determine what kind of critical point this is.

$$D = \frac{\partial^2 f}{\partial^2 x} = 6$$

$$z = 6(2, 1) = 6$$

$D > 0$ , and  $z > 0$ , which means it is a relative minimum. It also happens to be a global minimum.

## Question 5

### A

It is known that if the second derivative of  $f(x)$  is  $\geq 0$  then it is convex.

$$f'(x) = 2x, f''(x) = 2$$

Which is  $\geq 0$ , therefore  $x^2$  is convex.

## B

$$x^T(\lambda A + (1 - \lambda)B)x = \lambda x^T A x + (1 - \lambda)x^T B x \geq 0$$

$$x^T(\lambda A x + (1 - \lambda)B x) \geq 0$$

$$x^T x(\lambda A + (1 - \lambda)B) \geq 0$$

$$\lambda A + (1 - \lambda)B \geq 0$$

So,  $x^T A x$  is convex.

## Question 6

Machine learning is a hot topic in the computing world, and I know pretty much nothing about it. When one of my friends asked me what the difference between ML and AI was I wasn't able to tell them. So, I want to know the difference.

I also want to be able to use the basics of it and more easily learn about it since it seems such a versatile tool.