

# Machine Learning Assignment 49

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## 49-1

(A)  $E[a \cdot X] = a \cdot E[X]$

$$\begin{aligned} E[a \cdot X] &= \int_{-\infty}^{\infty} (a \cdot x) \cdot p(x) dx \\ &= a \cdot \int_{-\infty}^{\infty} x \cdot p(x) dx \\ &= a \cdot E[X] \end{aligned}$$

(B)  $E[X_1 + X_2] = E[X_1] + E[X_2]$

$$\begin{aligned} E[X_1 + X_2] &= \int_{-\infty}^{\infty} (X_1 + X_2) \cdot p(x) dx \\ &= \int_{-\infty}^{\infty} X_1 \cdot p(x) + X_2 \cdot p(x) dx \\ &= \int_{-\infty}^{\infty} X_1 \cdot p(x) + \int_{-\infty}^{\infty} X_2 \cdot p(x) dx \\ &= E[X_1] + E[X_2] \end{aligned}$$

(C)  $Var[X] = E[X^2] - E[X]^2$

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

(D) Bisection Search of  $\sqrt{5}$  to 4 decimal places.

Because  $\sqrt{4} = 2$  and  $\sqrt{9} = 3$  and since  $4 < 5 < 9$  the initial bounds are  $[2, 3]$ , and the to find the  $\sqrt{5}$  we have to make a function such that  $f(\sqrt{5}) = 0$  and the function that fits this is  $f(x) = x^2 - 5$ , so we compute,

Bounds	Mid-Point	F(Mid-Point)	Estimation
$[2, 2.5]$	2.5	1.25	2.25
$[2, 2.25]$	2.25	0.0625	2.125
$[2.125, 2.25]$	2.125	-0.484375	2.1875
$[2.1875, 2.25]$	2.1875	-0.21484375	2.21875
$[2.21875, 2.25]$	2.21875	-0.0771484375	2.234375
$[2.234375, 2.25]$	2.234375	-0.007568359375	2.2421875
$[2.234375, 2.2421875]$	2.2421875	0.02740478515625	2.23828125
$[2.234375, 2.23828125]$	2.23828125	0.0099029541015625	2.236328125
$[2.234375, 2.236328125]$	2.236328125	0.001163482666015625	2.2353515625
$[2.2353515625, 2.236328125]$	2.2353515625	-0.0032033920288085938	2.23583984375
$[2.23583984375, 2.236328125]$	2.23583984375	-0.001020193099975586	2.236083984375

We know the  $\sqrt{5} = 2.23606798$  Using the final bounds the estimation 2.2360... and the first 4 decimal places of my Bisection search are the same as what we want. therefore I am done.

(E) Merge sort of  $[4, 8, 7, 7, 4, 2, 3, 1]$

Input list:  $[4, 8, 7, 7, 4, 2, 3, 1]$

Break it in half:  $[4, 8, 7, 7]$ ,  $[4, 2, 3, 1]$

Use Merge Sort recursively to sort the two halves

Input list:  $[4, 8, 7, 7]$

Break it in half:  $[4, 8]$ ,  $[7, 7]$

Use Merge Sort recursively to sort the two halves

Input list:  $[4, 8]$

Break it in half:  $[4]$ ,  $[8]$

The two halves have only one element each, so they are already sorted

So we can combine them to get  $[4, 8]$

Input list:  $[7, 7]$

Break it in half:  $[7]$ ,  $[7]$

The two halves have only one element each, so they are already sorted

So we can combine them to get  $[7, 7]$

Now we have two sorted lists  $[4, 8]$  and  $[7, 7]$

So we can combine them to get  $[4, 7, 7, 8]$

Input list:  $[4, 2, 3, 1]$

Break it in half:  $[4, 2]$ ,  $[3, 1]$

Use Merge Sort recursively to sort the two halves

Input list:  $[4, 2]$

Break it in half:  $[4]$ ,  $[2]$

The two halves have only one element each, so they are already sorted

So we can combine them to get  $[2, 4]$

Input list:  $[3, 1]$

Break it in half:  $[3]$ ,  $[1]$

The two halves have only one element each, so they are already sorted

So we can combine them to get  $[1, 3]$

Now we have two sorted lists  $[2, 4]$  and  $[1, 3]$

So we can combine them to get  $[1, 2, 3, 4]$

Now we have two sorted lists  $[4, 7, 7, 8]$  and  $[1, 2, 3, 4]$

So we can combine them to get  $[1, 2, 3, 4, 4, 7, 7, 8]$

## 49-2

### Part 1 Bayesian inference with a uniform prior

(A)  $\mathcal{L}(k|1000 \text{ successes})$

$$\mathcal{L}(k|1000 \text{ successes}) = k \cdot k \cdot k \dots 996 \text{ times} \dots \cdot k = k^{1000}$$

(B) Normalized  $\mathcal{L}(k|1000 \text{ successes})$

$$\begin{aligned} \mathcal{L}(k|1000 \text{ successes}) &= \int_{-\infty}^{\infty} c \cdot \mathcal{L}(k|1000 \text{ successes}) dk \\ &= \int_{-\infty}^{\infty} c \cdot k^{1000} dk \\ &= \int_0^1 c \cdot k^{1000} dk \\ &= c \cdot \left( \frac{(1)^{1001}}{1001} - \frac{(0)^{1001}}{1001} \right) \\ &\Rightarrow c \cdot \frac{1}{1001} = 1 \\ &\Rightarrow c = 1001 \end{aligned}$$

(C)  $P(1 - k^{500} \leq \frac{1}{10000} | 1000 \text{ successes})$

$$\begin{aligned} P(1 - k^{500} \leq \frac{1}{10000} | 1000 \text{ successes}) &= \int_{0.99999979999}^1 1001 k^{1000} dk \\ &= k^{1001} \Big|_{0.99999979999}^1 \\ &= 1 - 0.999799810011 \\ &= 0.000200189989 \end{aligned}$$

### Part 2 Updating by inspecting the posterior

(A) Number of successes for the posterior distribution  $P(k|? \text{ successes}) = 5001k^{5000}$  also the likelihood function is  $\mathcal{L}(k|? \text{ successes})$  so using this,

$$\begin{aligned} \mathcal{L}(k|? \text{ successes}) &= k \cdot k \cdot \dots ? \text{ times} \cdot k \\ &= k^? \\ &= k^{5000} \\ &\Rightarrow ? = 5000 \end{aligned}$$

Therefore there are 5000 rolls.

(B) Normalized  $\mathcal{L}(k|7000 \text{ successes})$

$$\begin{aligned} \int_{-\infty}^{\infty} \mathcal{L}(k|7000 \text{ successes}) dk &= \int_{-\infty}^{\infty} c \cdot k \cdot k \cdot k \dots 6996 \text{ times} \dots \cdot k dk \\ &= \int_0^1 c \cdot k^{7000} dk \\ &= c \cdot \frac{k^{7001}}{7001} \Big|_0^1 \\ &\Rightarrow c = 7001 \end{aligned}$$

### Part 3 Bayesian updating

(A) Filling in the "2000 more successes" blanks

$$\text{prior distribution: } P(k) = 5001k^{5000}$$

$$\text{likelihood: } P(2000 \text{ more successes} | k) = 2001k^{2000}$$

$$\text{prior} \cdot \text{likelihood: } P(2000 \text{ more successes} | k) \cdot p(k) = 5001k^{5000} \cdot 2001k^{2000} = 10007001k^{7000}$$

$$\text{posterior distribution: } P(2000 \text{ more successes} | k) = \int_0^1 c \cdot 10007001k^{7000} dk \implies c \cdot \frac{10007001}{7001} = 1$$

$$\implies c = \frac{7001}{10007001} \implies P(2000 \text{ more successes} | k) = 7001k^{7000}$$

### Part 4 Inference

(A) Smallest  $N$  such that  $P(1 - k^{500} \leq \frac{1}{10000} | N \text{ successes}) = 0.99$

$$\begin{aligned} P(1 - k^{500} \leq \frac{1}{10000} | N \text{ successes}) &= \int_{0.99999979999}^1 (N+1)k^N dk \\ &= k^{N+1} \Big|_{0.99999979999}^1 \\ &= 1 - 0.99999979999^{N+1} \\ &\implies 1 - 0.99999979999^{N+1} = 0.99 \\ &\implies 0.99999979999^{N+1} = 0.01 \\ &\implies N + 1 \ln(0.99999979999) = \ln(0.01) \\ &\implies N = \frac{\ln(0.01)}{\ln(0.99999979999)} - 1 \\ &\implies N = 23024696.3916 \end{aligned}$$

The amount of more successes after 5000 are  $23024696.3916 - 5000 = 23019696.3916$