Machine Learning Assignment 49

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Problem 1

(a) Given that $X \sim p(x)$, where p(x) is a continuous distribution, prove that for any real number a we have E[aX] = aE[X].

Since a is a constant, we can literally just take it out of the integral.

$$E[aX] = \int_0^1 (ax) * P(x) dx$$
$$= a \int_0^1 x * P(x) dx$$
$$= aE[X]$$

(b) Given that $X_1, X_2 \sim p(x)$, where p(x) is a continuous distribution, prove that $E[X_1 + X_2] = E[X_1] + E[X_2]$.

$$E[X_1 + X_2] = \int_0^1 (x_1 + x_2) * p(x) dx$$

$$= \int_0^1 x_1 * p(x) + x_2 * p(x) dx$$

$$= \int_0^1 x_1 * p(x) dx + \int_0^1 x_2 * p(x) dx$$

$$= E[X_1] + E[X_2]$$

(c) Given that $X \sim p(x)$ where p(x) is a continuous probability distribution, prove the identity $Var[X] = E[X^2] - E[X]^2$.

$$Var[X] = E[(X - \overline{X})^{2}]dx$$

$$= \int_{0}^{1} (x - E[X])^{2} * p(x)dx$$

$$= \int_{0}^{1} (x^{2} - 2xE[X] + E[X]^{2}) * p(x)dx$$

$$= \int_{0}^{1} x^{2} * p(x)dx + E[X] \int_{0}^{1} (E[X] - 2x) * p(x)dx$$

$$= E[X^{2}] + E[X](E[X] \int_{0}^{1} p(x)dx - 2 \int_{0}^{1} x * p(x)dx$$

$$= E[X^{2}] + E[X](E[X] - 2E[X])$$

$$= E[X^{2}] - E[X]^{2}$$

(d) Use bisection search to estimate $\sqrt{5}$ to 4 decimal places by hand, showing your work at each step of the way. See problem 5-2 for a refresher on bisection search.

Start out with interval [2, 3] since $2^2 = 4$ and $3^2 = 9$

The function with $\sqrt{5}$ as a root can just be $f(x) = x^2 - 5$

Middle is 2.5. f(2.5) = 1.25. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.25. f(2.25) = 0.0625. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.125. f(2.125) = -0.484375. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.1875. f(2.1875) = -0.21484375. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.21875. f(2.21875) = -0.0771484375. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.234375. f(2.234375) = -0.007568359375. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.2421875. f(2.2421875) = 0.02740478515625. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.23828125. f(2.23828125) = 0.0099029541015625. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.236328125. f(2.236328125) = 0.001163482666015625. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.2353515625. f(2.2353515625) = -0.0032033920288085938. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.23583984375. f(2.23583984375) = -0.001020193099975586. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.236083984375. f(2.236083984375) = 7.158517837524414e-05. It is less than 0 so the guess is too low. We must lower the bounds

Middle is 2.2359619140625. f(2.2359619140625) = -0.00047431886196136475. It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.23602294921875. f(2.23602294921875) = -0.00020137056708335876.

It is greater than 0 so the guess is too high. We must raise the bounds

Middle is 2.236053466796875. f(2.236053466796875) = -6.489362567663193e

05. It is greater than 0 so the guess is too high. We must raise the bounds

Now, f(x) rounded to 4 decimal places is 0. Therefore, the most accurate 4 decimals we have is:

$$\sqrt{5} \approx 2.2361$$

(e) Use "merge sort" to sort the list [4,8,7,7,4,2,3,1]. Do the problem by hand and show your work at each step of the way. See problem 23-3 for a refresher on merge sort.

input list: [4,8,7,7,4,2,3,1] break it in half: [4,8,7,7] 4,2,3,1] sort the two halves input list: [4,8,7,7] break it in half: [4,8] [7,7] 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] 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 [3, 1]

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

1 Problem 2

(a) Other people make 1000 successful trips through the wormhole with no failures. What is the likelihood function for k given these 1000 successful trips?

data = [1000 successful]

 $L(k|[1000successful]) = k^{1000}$

(b) What is the posterior distribution for k given these 1000 successful trips? (This is the same as just normalizing the likelihood function).

$$\begin{split} 1 &= \int_0^1 c * k^{1000} \mathrm{d}k \\ &= c \left. \frac{k^{1001}}{1001} \right|_0^1 \\ c &= 1001 \\ P(k|1000 \, \mathrm{successes}) &= 1001 k^{1000} \\ P(0.99|100 \, \mathrm{successes}) &= \sim 0.043 \end{split}$$

(c) Assuming that you will use the wormhole 500 times per year, what is the posterior probability that the risk of disappearing forever into the wormhole is no more than your risk of dying from a car crash in a given year (1 in 10000)? In other words, what is $P(1k^{500} \le 110000|1000successes)$

$$1 - k^{500} \le \frac{1}{10000}$$

$$k^{500} \ge \frac{9999}{10000}$$

$$k \ge \sqrt[500]{\frac{9999}{10000}}$$

$$P(k \ge \sqrt[500]{\frac{9999}{10000}}) = 1 - \int_0^k 1001 k^{1000} dk$$

$$= 1 - k^{1001} \Big|_0^{500} \sqrt{\frac{9999}{10000}}$$

$$= \sim 0.000200$$

1.1 Part 2

(a) Looking at the given posterior distribution, how many successes have you counted?

5000, because k is raised to that power, implying k had to have happened 5000 times.

(b) Suppose you observe 2000 more successes. What is the posterior distribution now?

Using similar reasoning to part 1-b, we can see that the new posterior distribution will be:

$$P(k|7000 \text{ successes}) = 7001k^{7000}$$

$$P(0.999|7000 \text{ successes}) = \sim 6.362$$

1.2 Part 3

(a) Suppose you observe 2000 more successes. Fill in the blanks: likelihood = k^{2000}

$$p(2000successes|k=0.999)=\sim 0.1351$$

prior * likelihood =

$$5001k^{2000+5000}$$

after re-normalizing the prior * likelihood function, we get: posterior distribution = $7001k^{7000}$

(with the re-normalization constant being $\frac{7001}{5001}$) The reason why this is the same is that the posterior distribution is propor-

The reason why this is the same is that the posterior distribution is proportional to k^{7000} , and so is the likelihood function for P(7000successes|k). Since both likelihood functions are equal, normalizing each will give the exact same results.

1.3 Part 4

(a) Assuming that you will use the wormhole 500 times per year, how many more people do you need to observe successfully come out of the wormhole to be 99% sure the risk of disappearing forever into the wormhole is no more than your risk of dying from a car crash in a given year (1 in 10000)?

$$\mathcal{L} = k^{5000+n}$$

Through similar reasoning in Part 3a, we have:

$$P(5000 + N) = (5001 + n)k^{5000+n}$$

And to check if this function is right,

$$a = \sqrt[500]{\frac{9999}{10000}}$$

$$0.99 = P(k \ge a)|5000 + Nsuccesses)$$

$$= 1 - \int_0^a (5001 + n)k^{5000+n} dk$$

$$= 1 - k^{5001+n}|_0^a$$

$$0.01 = (\sqrt[500]{\frac{9999}{10000}})^{5001+n}$$

$$n = \frac{\ln 0.01}{\ln \sqrt[500]{\frac{9999}{10000}}} - 5001$$

$$= \sim 23019699$$