

1. *Probability boot camp*

(a) *Prove Markov's inequality, $Pr[X \geq c] \leq E[X]/c$, with $c > 0$*

The formula for the probability of a continuous random variable X with probability density function $f(x)$ is

$$Pr[x_1 \leq X \leq x_2] = \int_{x_1}^{x_2} f(x)dx$$

And the formula for the expected value of a continuous random variable X with probability density function $f(x)$ is

$$E[x_1 \leq X \leq x_2] = \int_{x_1}^{x_2} xf(x)$$

so for $Pr[X \geq c]$ we have

$$Pr[X \geq c] = \int_c^{\infty} f(x)dx$$

and since X is a nonnegative random variable we have

$$E[X] = \int_0^{\infty} xf(x)dx$$

Notice that $0 < c \leq \infty$. This tells us that the bounds of $E[X]$ are greater than $Pr[X \geq c]$. We can break up the integral formed by $E[X]$ to create an inequality that will begin to look similar to the integral of $Pr[X \geq c]$.

$$\begin{aligned} E[X] &= \int_0^{\infty} xf(x)dx \\ &= \int_0^c xf(x)dx + \int_c^{\infty} xf(x)dx \\ &\geq \int_c^{\infty} xf(x)dx \end{aligned}$$

We can assume that $x \geq c$ because c is one of the bounds of the integral. This means we can substitute c for x .

$$\int_c^\infty xf(x)dx \geq \int_c^\infty cf(x)dx \geq c \int_c^\infty f(x)dx$$

We now have an equation for $E[X]$ that has $Pr[X \geq c]$.

$$E[X] \geq c \int_c^\infty f(x)dx = cPr[X \geq c]$$

Dividing both sides by c gives us

$$E[X]/c \geq Pr[X \geq c]$$

which is Markov's inequality. We have just shown that $Pr[X \geq c] \leq E[X]/c$, with $c > 0$ is true based on the probability and expected value of a continuous random variable.

- (b) *Prove Chebyshev's inequality*
- (c) *Show that for any discrete random variables X, X' , $E[X] = E[E[X|X']]$.*
- (d) *Prove by induction that $E[X_t] = 0$.*