

Common Distributions

~~Functions of random variables,
Expectation~~

~~Putting a value on random variables~~

How we model the real world of randomness

Michael Psenka

Review: functions of random variables

$$X = (\Omega, \mathcal{P})$$

↑

$x \in \Omega$

e.g. 3 indep. Bernoulli f.v.
each w/ p.

$$\Omega = \{0,1\} \times \{0,1\} \times \{0,1\} \quad (\text{e.g. } (0,1,0) \in \Omega)$$

$$\mathcal{P} = \mathcal{P}(e_1) \mathcal{P}(e_2) \mathcal{P}(e_3)$$

$$f(x) = x_1 + x_2 + x_3$$

$$\Omega' = \{0,1,2,3\}$$

$$f(x) := \boxed{(\Omega', \mathcal{P}')} \quad \mathcal{P}'$$

$$\mathcal{P}'(f(x)=1) := \mathcal{P}(\text{exactly 1 heads})$$

$$= \sum_{x: f(x)=1} \mathcal{P}(x) = 3p(1-p)^2$$

$(1, 0, 0)$, $(0, 1, 0)$, $(0, 0, 1)$
 $p(1-p)(1-p)$

Review: Expectation

Definition. *The expectation of a numeric random variable $X = (\Omega, \mathbb{P})$ is given by the following:*

$$E[X] := \sum_{x \in \Omega} x \mathbb{P}(X = x).$$

Ex. Bernoulli, $p = 0.5$, $E[X] = 0 \cdot 0.5 + 1 \cdot 0.5 = 0.5$

Expectation of (functions of) random variables

$$E[f(x)] = \sum_{i \in \Omega'} i P(f(x)=i)$$

$$= \sum_{i \in \Omega'} \left(\sum_{x | f(x)=i} i P(x) \right) \quad i = f(x)$$

$$= \sum_{x \in \Omega} f(x) P(x)$$

(double sum reindex, see expectation equality proof)

e.g. $E(x^2) = \sum x^2 P(x)$

" $f(x) = x^2$ "

($E(x+y)$)
 $E(xy)$
 $E(x^2)$
 $f: (x,y) \rightarrow \mathbb{R}$
 all "expectations of functions"

Linearity of expectation

Theorem. Let X, Y be two (numeric) random variables, and their joint distribution given by $\mathbb{P}(X = x, Y = y)$. The expectation is a “linear operator”: that is,

1. $E[X + Y] = E[X] + E[Y]$,
2. $E[cX] = cE[X]$ for any fixed $c \in \mathbb{R}$.

$$E[X+Y] = \sum (x+y)P(X=x, Y=y)$$

$(x, y) \rightarrow \mathbb{R}$ $f(x, y) = x + y$. See y'day slides for rest of proof

Some important distributions

Review: Bernoulli

$$\Omega = \{0, 1\} \quad P(X=0) = (1-p), \quad P(X=1) = p$$

$(P(\{0\}))$

Bernoulli(p)

E.g. sending signal, $P(\text{received}) = p$

Review: Binomial

n independent Bernoulli(p), denote x_1, x_2, \dots, x_n

$$P\left(\sum_{i=1}^n x_i = k\right) = \sum_{x \in \Omega \mid \sum x_i = k} P(x) = \sum_{x \in \Omega \mid \sum x_i = k} p^k (1-p)^{n-k}$$

Binomial(n, p)

$$\boxed{\binom{n}{k} p^k (1-p)^{n-k}}$$

Geometric

$$P(X=k) = (1-p)^{k-1} p$$

e.g. $P(X=3) = (1-p)(1-p)p$

$f: \{0,1\}^{\mathbb{N}} \rightarrow \mathbb{N}$ (not necessary)

$$P(\Omega) = \sum_{i=1}^{\infty} (1-p)^{i-1} p = p \left(\sum_{i=1}^{\infty} (1-p)^{i-1} \right)$$

$$= p \left(\sum_{i=0}^{\infty} (1-p)^i \right)$$

$$= p \frac{1}{1-(1-p)} = p \left(\frac{1}{p} \right) = 1 \quad \checkmark$$

Geometric(p)

$E(\text{Geometric}(p))?$

Expectation equality

Let X be a random variable. If its sample space $\Omega \subset \mathbb{N}$, the following equality holds:

$0: P(0)$

$\sum i P(i)$

$P(1)$

$P(2) + P(2)$

e.g. $P(X=0) = 1$
 $P(\text{otherwise}) = 0$

$$E[X] = \sum_{i=1}^{\infty} P(X \geq i).$$

$$\boxed{P(X \geq 0) = 1}$$

$$E[X] = 0 \cdot 1 + 1 \cdot 0 + 2 \cdot 0, \dots = 0$$

$$= 1 + 0 + 0 + \dots + 0 = 0$$

NOTE: this was all scratchwork to determine in lecture if it's $i=0$ or $i=1$ on the RHS

Expectation equality

$$E[X] = \sum_{i=1}^{\infty} i P(X=i)$$

$$= \sum_{i=1}^{\infty} \left(\sum_{j=1}^i P(X=i) \right)$$

$$\stackrel{?}{=} \sum_{i=1}^{\infty} \left(\sum_{j=i}^{\infty} P(X=i) \right)$$

$$= \sum_{j=1}^{\infty} P(X \geq j) \quad \square$$

$$\textcircled{1} \sum_{i=0}^{\infty} f(i) \quad \text{or} \quad \sum_{i \in \mathbb{N}} f(i)$$

$$\sum_{i=0}^{\infty} \sum_{j=0}^{\infty} f(i, j) \quad \text{or} \quad \sum_{(i, j) \in \mathbb{N} \times \mathbb{N}} f(i, j)$$

$$\sum_{j=0}^{\infty} \sum_{i=0}^{\infty} f(i, j)$$

$$\left(\sum_{i=0}^{\infty} \sum_{j=0}^i \right)$$

$$\textcircled{3} \quad \text{or} \quad \sum_{(i, j) \in A}$$

$$A = \{(i, j) \in \mathbb{N} \times \mathbb{N} \mid j < i\}$$

To switch the order of a sum like $\sum_{i=1}^{\infty} \sum_{j=1}^i f(i,j)$:

① Write $\sum_{i=1}^{\infty} \sum_{j=1}^i f(i,j) = \sum_{(i,j) \in A} f(i,j)$ for the appropriate index set $A \subset \mathbb{N} \times \mathbb{N}$. Here, it's $A = \{(i,j) \mid i \geq j\}$.

② Put j term on outside, determine limits (here, it's $\sum_{j=1}^{\infty}$)

③ For each j , determine which i make the pair $(i,j) \in A$: here, it's $\sum_{i=j}^{\infty}$

④ Put it all together to get $\sum_{j=1}^{\infty} \sum_{i=j}^{\infty} f(i,j)$

Geometric revisited

$$X = \text{Geom}(p)$$

$$E(X) = \sum_{i=1}^{\infty} P(X \geq i)$$

$$= \sum_{i=1}^{\infty} (1-p)^{i-1}$$

$$= \sum_{i=0}^{\infty} (1-p)^i = \frac{1}{1-(1-p)}$$
$$= \frac{1}{p}$$

eg $p = 0.5$

$$E[X] = 2$$

$p = 0.1$

$$E[X] = 10$$

$p = 0.9$

$$E[X] = \frac{10}{9}$$

where $P(X \geq i) = (1-p)^{i-1}$ since this is exactly the event of failing the first $i-1$ times (we don't care what happens after)

Poisson

$$P(X=k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

$$\Omega = \{0, 1, 2, 3, 4, \dots\}$$

$\lambda :=$ "Poisson parameter" Poisson(λ)

$$E[X] = \lambda \quad \checkmark$$

valid dist?

$$e^x := \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

$$P(X \in \Omega) = \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} e^{-\lambda} = e^{-\lambda} \sum_{k=0}^{\infty} \frac{\lambda^k}{k!} = e^{-\lambda} e^{\lambda} = 1$$

Poisson

$$E[X] = \sum_{k=0}^{\infty} k \frac{\lambda^k}{k!} e^{-\lambda}$$

$$= \sum_{k=1}^{\infty} k \frac{\lambda^k}{k!} e^{-\lambda}$$

$$= \lambda e^{-\lambda} \left(\sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{k!} k \right)$$

$$= \lambda e^{-\lambda} \left(\sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} \right)$$

$$= \lambda e^{-\lambda} \left(\sum_{k=0}^{\infty} \frac{\lambda^k}{k!} \right) = \lambda e^{-\lambda} e^{\lambda} = \boxed{\lambda} \quad \square$$

Sum of Poisson Suppose $X = \text{Poisson}(\lambda_1)$, $Y = \text{Poisson}(\lambda_2)$, and X, Y independent. Then $X+Y \sim \text{Poisson}(\lambda_1 + \lambda_2)$.

Pf Writing out all possibilities where $X+Y=k$ yields the following: (here, " \sim " means "has same distribution as")

$$\begin{aligned}
 P(X+Y=k) &= \sum_{i=0}^k P(X=i, Y=k-i) = \sum_{i=0}^k P(X=i) P(Y=k-i) \\
 &= \sum_{i=0}^k \frac{\lambda_1^i}{i!} e^{-\lambda_1} \frac{\lambda_2^{k-i}}{(k-i)!} e^{-\lambda_2}
 \end{aligned}$$

(factoring out constant factors) $= e^{-\lambda_1 - \lambda_2} \left(\frac{1}{k!} \right) \sum_{i=0}^k \frac{k!}{i!(k-i)!} \lambda_1^i \lambda_2^{k-i}$
 & additional $k!$ term

(binomial theorem) $= \left[e^{-\lambda_1 - \lambda_2} \frac{(\lambda_1 + \lambda_2)^k}{k!} \right]$

Poisson as limit of binomial

$\lim_{n \rightarrow \infty} \text{Binomial}(n, \frac{\lambda}{n}) \sim \text{Poisson}(\lambda)$. Note intuitively: for $n=1$, this is akin to e.g. exactly at the 1 hr mark, checking if a call came or not. $n=2$ checks twice (every half hour), where in each interval it's now half as likely to get a call, and so on until you've divided up the hour arbitrarily small and get Poisson, which models "number of heads" in a continuous time interval.

PF: See notes, well articulated there and would be messy here. Main identity needed is $e^x = \lim_{n \rightarrow \infty} (1 + \frac{x}{n})^n$, which is proved in Rudin's Principles of Mathematical Analysis for those curious.