

DISCRETE RANDOM VARIABLES

The Poisson distribution

Poisson random variables

- Some random variables arise from counting – typically the occurrence of some phenomenon during a fixed amount of time.
- Examples:
 - Number of phone calls during one hour to a poison control hotline
 - Number of live births at a hospital in a week
 - Number of traffic accidents at an intersection in a month
- It doesn't have to be time (but it often is):
 - Number of weeds in a section of farmland
 - Number of lesions in a brain region

Derivation of the Poisson distribution

- We are counting the number of phone calls received in an hour.
- Let's break the hour up into many tiny intervals (think: intervals so small it's hard to conceive of there being more than one call per interval). (one second?)
- Suppose that for any given tiny interval: $P(\text{one phone call}) = p$.
- So in that tiny interval, $P(\text{zero phone calls}) = 1 - p$.
- (This then implies that for any tiny interval $P(\text{more than one phone call}) = 0...$)

- Suppose that phone calls coming are independent of each other.
- If Y is the total number of calls in an hour, and if there are n tiny intervals, then:

$$Y \sim \text{Binom}(n, p)$$

- Expected number of calls in the hour is np (right?).
- What if I cut each tiny interval in half? (Number of intervals is $n' = 2n$)
- Expected number of calls in the hour is still $np = n'p' = (2n)(\frac{p}{2})$.
- Let λ be the expected number of calls in the hour. (This should not depend on how tiny the intervals are.)
- So for any choice of $n, p = \lambda/n$.

- Since Y is binomial,
- $P(Y = y) = \binom{n}{y} p^y (1 - p)^{n-y}$
- Let's think about the tiny intervals getting ever more tiny...

$$P(Y=y) = \binom{n}{y} p^y (1-p)^{n-y}$$

$$= \frac{n!}{y! (n-y)!} p^y (1-p)^{n-y}$$

$$np = \lambda$$

$$p = \frac{\lambda}{n}$$

FACT Appendix A.6

$$\lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n = e^{-\lambda}$$

$$\lim_{n \rightarrow \infty} P(Y=y) = \lim_{n \rightarrow \infty} \frac{\overbrace{n(n-1) \dots (n-y+1)}^{y \text{ terms}}}{y!} \left(\frac{\lambda}{n}\right)^y \left(1 - \frac{\lambda}{n}\right)^{n-y}$$

$$\lim_{n \rightarrow \infty}$$

$$\frac{n}{y}$$

$$\frac{n-1}{y-1}$$

$$\frac{n-2}{y-2}$$

$$\dots$$

$$\frac{n-y+1}{1}$$

$$\frac{\lambda^y}{n^y}$$

$$\left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-y}$$

$$\lim_{n \rightarrow \infty}$$

$$\frac{n}{y \cdot n}$$

$$\frac{n-1}{(y-1) \cdot n}$$

$$\frac{n-2}{(y-2) \cdot n}$$

$$\dots \frac{n-y+1}{1 \cdot n}$$

$$\lambda^y$$

$$\left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-y}$$

$$e^{-\lambda}$$

(FACT)

$$\rightarrow 1$$

$$= \frac{\lambda^y}{y!} e^{-\lambda}$$

Poisson distribution

- The Poisson distribution can be defined as the limit of a binomial distribution.
- A random variable Y has a POISSON DISTRIBUTION with rate parameter λ probabilities given by

$$P(Y = y) = \frac{\lambda^y}{y!} e^{-\lambda}, \quad y = 0, 1, 2, \dots$$

- Check: Is this a valid distribution function?

$$\sum_{y=0}^{\infty} P(Y = y) = \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} e^{-\lambda} = e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} = e^{-\lambda} e^{\lambda} = 1$$

(See Appendix A1.11)

Example 9

Poisson

$$\lambda = 2.6$$

- Accidents occur at a workplace at a rate of 2.6 per month. What is the probability that there will be 4 accidents next month?

counting

model is $Y \sim \text{Poisson}(2.6)$

$$P(Y = 4) = \frac{(2.6)^4}{4!} e^{-2.6} = 0.113$$

Example 9 (continued)

- What is the probability there are between 3 and 6 accidents next month?

$$P(3 \leq Y \leq 6) = P(Y=3) + P(Y=4) + P(Y=5) + P(Y=6)$$

- See Appendix 3 Table 3 or use ppois() in R

$$P(3 \leq Y \leq 6) = P(Y \leq 6) - P(Y \leq 2)$$

$P(Y \leq 6)$:

```
> ppois(q=6, lambda=2.4)  
[1] 0.9884059
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$P(Y \leq 2)$:

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> ppois(q=2, lambda=2.4)  
[1] 0.5697087
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$$P(3 \leq Y \leq 6) = 0.988 - 0.5697 = 0.418$$

Mean and variance of the Poisson distribution

- If Y is a Poisson random variable with rate parameter λ then

$$E[Y] = \lambda;$$

$$Var(Y) = \lambda$$

$$\begin{aligned}
 E[Y] &= \sum_{y=0}^{\infty} y P(Y=y) \quad \text{def. of exp.} \\
 &= \sum_{y=1}^{\infty} y \frac{\lambda^y}{y!} e^{-\lambda} = \sum_{y=1}^{\infty} \frac{\lambda^y}{(y-1)!} e^{-\lambda} \\
 &= \sum_{x=0}^{\infty} \frac{\lambda^{x+1}}{x!} e^{-\lambda} = \lambda \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} e^{-\lambda} \\
 &= \lambda
 \end{aligned}$$

change of variables
 $x = y - 1 \quad y = x + 1$
 $y = 1 \rightarrow x = 0$
 $"y = \infty" \rightarrow "x = \infty"$

Some notes about the Poisson distribution

- If $Y_1 \sim \text{Poisson}(\lambda_1)$ and $Y_2 \sim \text{Poisson}(\lambda_2)$ and Y_1 and Y_2 are independent, then

$$Y_1 + Y_2 \sim \text{Poisson}(\lambda_1 + \lambda_2)$$

- If λ is the rate of occurrences for one time period, then the distribution of the number of occurrences for m time periods is $\text{Poisson}(m\lambda)$.
- Note that the above requires that the rate be the same for all time periods.
- Just because you are counting some occurrences doesn't necessarily mean the random variable has a Poisson distribution!