

MATH 451/551

Chapter 4. Common Discrete Distributions

4.3 Geometric Distribution

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Mean



$X \sim \text{Geo}(p) \Rightarrow \mathcal{X} = \{0, 1, 2, \dots\} \Rightarrow X$ is discrete.

$$f(x) = p(1-p)^x$$

$$E(X) = \sum_{\mathcal{X}} x f(x) = \sum_{x=0}^{\infty} x p(1-p)^x = p(1-p) + \underset{\uparrow}{2} p(1-p)^2 + \underset{\uparrow}{3} p(1-p)^3 + \dots$$

$$= \underbrace{p(1-p)} + p(1-p)^2 + p(1-p)^3 + p(1-p)^4 + \dots + p \frac{(1-p)}{1-(1-p)} = 1-p$$

$$p(1-p)^2 + p(1-p)^3 + p(1-p)^4 + \dots + p(1-p) \frac{(1-p)}{1-(1-p)} = (1-p)^2$$

$$p(1-p)^3 + p(1-p)^4 + \dots + p(1-p)^2 \frac{1-p}{1-(1-p)} = (1-p)^3$$

$$p(1-p)^4 + \dots + (1-p)^4 + \dots$$

$$= (1-p) + (1-p)^2 + (1-p)^3 + \dots = \frac{(1-p)}{1-(1-p)} = \frac{1-p}{p}$$

Variance



$$V(X) = E(X^2) - \{E(X)\}^2$$

$$E(X^2) = \sum_{x=0}^{\infty} x^2 p(1-p)^x = p(1-p) + \underline{4p(1-p)^2} + 9p(1-p)^3 + \dots$$

$$= p(1-p) + p(1-p)^2 + p(1-p)^3 + \dots + \frac{p(1-p) \cdot 1}{1-(1-p)} = 1-p$$

$$+ 3p(1-p)^2 + 3p(1-p)^3 + \dots + \frac{3p(1-p)^2 \cdot 1}{1-(1-p)} = 3(1-p)^2$$

$$+ 5p(1-p)^3 + \dots + \frac{5p(1-p)^3 \cdot 1}{1-(1-p)} = 5(1-p)^3 \dots$$

$$= (1-p) + (1-p)^2 + (1-p)^3 + \dots + \frac{1-p}{1-(1-p)} = \frac{1-p}{p}$$

$$+ 2(1-p)^2 + 2(1-p)^3 + \dots + \frac{2(1-p)^2}{1-(1-p)} = \frac{2(1-p)^2}{p}$$

$$+ 2(1-p)^3 + \dots + \frac{2(1-p)^3}{1-(1-p)} = \frac{2(1-p)^3}{p} + \dots$$

$$= \frac{2\{(1-p) + (1-p)^2 + (1-p)^3 + \dots\} - (1-p)}{p}$$

$$= \frac{2 \frac{1-p}{1-(1-p)} - (1-p)}{p} = \frac{2(1-p) - p(1-p)}{p^2} = \frac{(1-p)(2-p)}{p^2}$$

$$\begin{aligned}
 V(X) &= E(X^2) - \{E(X)\}^2 \\
 &= \frac{(1-p)(2-p)}{p^2} - \frac{(1-p)^2}{p^2} = \frac{(1-p)(2-p-1+p)}{p^2} \\
 &= \frac{1-p}{p^2}
 \end{aligned}$$

$$p(1-p) + 2p(1-p)^2 + p(1-p)^3 + \dots$$

$$S$$

$$\begin{aligned}
 S - (1-p)S &= 1-p \\
 S &= \frac{1-p}{p}
 \end{aligned}$$

$$p(1-p)^2 + 2p(1-p)^3 + \dots$$

$$(1-p)S$$

$$\underbrace{p(1-p) + p(1-p)^2 + p(1-p)^3 + \dots}_{= \frac{p(1-p)}{p} = (1-p)}$$

$$1 + 2 + 2^2 + 2^3 + \dots = \frac{1}{1-2}$$

$$\underbrace{1 + 2 \cdot 2 + 3 \cdot 2^2 + 4 \cdot 2^3 + \dots}_{= \frac{1}{(1-2)^2}}$$

MGF



$$M(t) = E(e^{tx}) = \sum_{x=0}^{\infty} e^{tx} p(1-p)^x$$

$$= p \sum_{x=0}^{\infty} \{e^t(1-p)\}^x$$

$$= p [1 + e^t(1-p) + \{e^t(1-p)\}^2 + \dots]$$

$$= p \frac{1}{1 - e^t(1-p)} = \frac{p}{1 - (1-p)e^t}, \quad (1-p)e^t < 1$$

$$\begin{aligned} E(X) &= \left. \frac{dM(t)}{dt} \right|_{t=0} = \left. \frac{d}{dt} \frac{p}{1 - (1-p)e^t} \right|_{t=0} = \left. \frac{+p(1-p)e^t}{\{1 - (1-p)e^t\}^2} \right|_{t=0} \\ &= \frac{p(1-p)}{(1-1+p)^2} = \frac{1-p}{p} \end{aligned}$$

Skewness



$$E \left\{ \left(\frac{x - \mu}{\sigma} \right)^3 \right\} = \frac{2 - p}{\sqrt{1 - p}}$$

Kurtosis



$$E \left\{ \left(\frac{x - \mu}{\sigma} \right)^4 \right\} = \frac{\mu^2 - 9\mu + 9}{1 - \mu}$$

Memoryless Property



Memoryless Property

For $X \sim \text{Geo}(p)$ and any two nonnegative integers x and y ,

$$P(X \geq x + y | X \geq x) = P(X \geq y).$$

Interpretation

Consider a sequence of repeated, mutually independent, and identically distributed Bernoulli trials and a random variable X that is the number of failures before the first success. If you know that X is greater than or equal to x , then the distribution of the **remaining** number of Bernoulli trials before the first success has the same distribution as if the original x trials had never occurred. This interpretation is consistent with intuition: the previous history of the sequence of Bernoulli trials has no effect on the outcomes of future Bernoulli trials.

$$\text{LHS: } P(\underbrace{X \geq x+y}_A \mid \underbrace{X \geq x}_B) = \frac{P(X \geq x+y \cap X \geq x)}{P(X \geq x)}$$

$$= \frac{P(X \geq x+y)}{P(X \geq x)}$$

$$\text{RHS: } P(x \geq y)$$

$$= 1 - F_X(y-1)$$

$$= (1-p)^y$$

$$= \frac{1 - P(X < x+y)}{1 - P(X < x)}$$

$$= \frac{1 - P(X \leq x+y-1)}{1 - P(X \leq x-1)} = \frac{1 - F_X(x+y-1)}{1 - F_X(x-1)}$$

$$= \frac{1 - \left\{ 1 - (1-p)^{x+y-1} \right\}}{1 - \left\{ 1 - (1-p)^{x-1} \right\}} = \frac{(1-p)^{x+y}}{(1-p)^x}$$

$$= (1-p)^y$$

Example 1



Example 1 $P(\text{"double six"}) = \frac{1}{6} \times \frac{1}{6} = \frac{1}{36}$

Roll a pair of fair dice repeatedly until a "double six" appears. Let X be the number of rolls prior to the appearance of the first double six. Find $f(x)$, $E(X)$, $V(X)$, and $P(X < 24)$.

$$X \sim \text{Geo}\left(\frac{1}{36}\right)$$

$$f(x) = \frac{1}{36} \left(\frac{35}{36}\right)^x, \quad x = 0, 1, 2, \dots$$

$$E(X) = \frac{1-p}{p} = \frac{35/36}{1/36} = 35$$

$$V(X) = \frac{1-p}{p^2} = \frac{35/36}{(1/36)^2} = 1260$$

$$P(X < 24) = P(X \leq 23) = F(23) = 1 - \left(1 - \frac{1}{36}\right)^{24} \approx 0.4914$$

R Functions

Function	Returned Value
<u>dgeom</u> (<u>x</u> , <u>p</u>)	calculates the <u>probability mass function</u> $f(x)$
<u>pgeom</u> (<u>x</u> , <u>p</u>)	calculates the <u>cumulative mass function</u> $F(x)$
<u>qgeom</u> (<u>u</u> , <u>p</u>)	calculates the percentile (quantile) $F^{-1}(u)$
<u>rgeom</u> (<u>m</u> , <u>p</u>)	generates m random variates

Alternative Definition

$X = \#$ failures before
1st success

$X = \#$ trials to get 1st
success



Alternative Definition

The geometric distribution can also be parameterized as the trial number of the first success.

- **PMF:** A discrete random variable X with PMF

$$f(x) = p(1 - p)^{x-1}, \quad x = 1, 2, \dots$$

for $0 < p < 1$ is a *Geometric*(p) random variable.

- This shifts the distribution one unit to the right.

$$\mu = \frac{1}{p} \quad \text{and} \quad \sigma^2 = \frac{1 - p}{p^2}.$$

Example 2



Example 2

How many tosses of a pair of fair dice are necessary to be 99% certain that a double six will appear?

Thank You



THANK YOU!