

# Special Discrete Random Variables (Ch. 5.1)

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# Outline

Special Discrete  
Random Variables  
(Ch. 5.1)

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Binomial  
Distribution

Geometric  
Distribution

Poisson  
Distribution

Binomial Distribution

Geometric Distribution

Poisson Distribution

# Purpose of the binomial random variable

- ▶ A  $\text{Bin}(n, p)$  random variable counts the number of successes in  $n$  success-failure trials that:
  - ▶ are independent of one another.
  - ▶ each succeed with probability  $p$ .
- ▶ Examples:
  - ▶ Number of conforming hexamine pellets in a batch of  $n = 50$  total pellets made from a pelletizing machine.
  - ▶ Number of runs of the same chemical process with percent yield above 80%, given that you run the process a total of  $n = 1000$  times.
  - ▶ Number of rivets that fail in a boiler of  $n = 25$  rivets within 3 years of operation. (Note; “success” doesn’t always have to be good.)

# The Binomial Distribution

- ▶  $X \sim \text{Binomial}(n, p)$  – i.e.,  $X$  is distributed as a binomial random variable with parameters  $n$  and  $p$  ( $0 < p < 1$ ) if:

$$f_X(x) = \begin{cases} \binom{n}{x} p^x (1-p)^{n-x} & x = 0, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

where:

- ▶  $\binom{n}{x} = \frac{n!}{x!(n-x)!}$ , read “ $n$  choose  $x$ ”
- ▶  $n! = n \cdot (n-1) \cdot \dots \cdot 2 \cdot 1$ , the factorial function.
- ▶  $E(X) = np$
- ▶  $\text{Var}(X) = np(1-p)$

# The Binomial Distribution

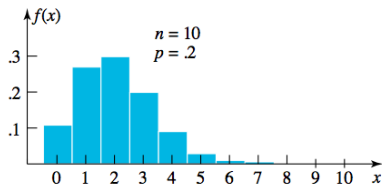
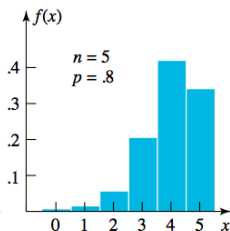
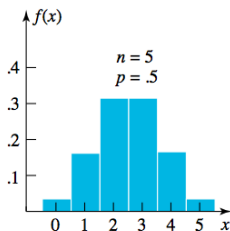
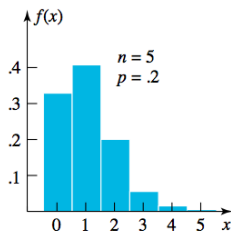
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## Example: machine with 10 components



- ▶ Suppose you have a machine with 10 **independent** components in series. The machine only works if all the components work.
- ▶ Each component succeeds with probability  $p = 0.95$  and fails with probability  $1 - p = 0.05$ .
- ▶ Let  $Y$  be the number of components that succeed in a given run of the machine. Then:

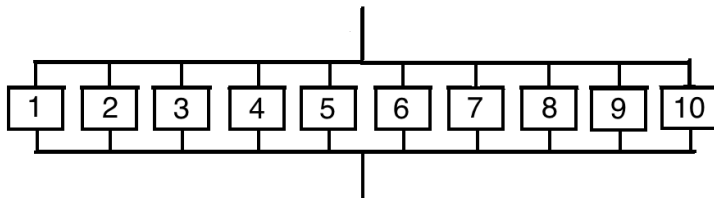
$$Y \sim \text{Binomial}(n = 10, p = 0.95)$$

## Example: machine with 10 components

$$\begin{aligned}P(\text{machine succeeds}) &= P(Y = 10) \\&= \binom{10}{10} p^{10} (1 - p)^{10-10} \\&= p^{10} \\&= 0.95^{10} \\&= 0.5987\end{aligned}$$

- This machine isn't very reliable.

## Example: machine with 10 components



- ▶ What if I arrange these 10 components in parallel? This machine succeeds if at least 9 of the components succeed.
- ▶ What is the probability that the new machine succeeds?



## Example: machine with 10 components

$$\begin{aligned}P(\text{improved machine succeeds}) &= P(Y \geq 9) \\&= P(Y = 9) + P(Y = 10) \\&= \binom{10}{9} p^9 (1 - p) + \binom{10}{10} p^{10} (1 - p)^{10-10} \\&= (10) \cdot 0.95^9 \cdot 0.05 + (1) \cdot 0.95^{10} \\&= 0.9139\end{aligned}$$

- By allowing just one component to fail, we made this machine far more reliable.

## Example: machine with 10 components

- If we allow up to 2 components to fail:

$$\begin{aligned} &P(\text{improved machine succeeds}) \\ &= P(Y \geq 8) \\ &= P(Y = 8) + P(Y = 9) + P(Y = 10) \\ &= \binom{10}{8} p^8 (1-p)^{10-8} + \binom{10}{9} p^9 (1-p) + \binom{10}{10} p^{10} (1-p)^{10-10} \\ &= \frac{10!}{(10-8)!8!} \cdot 0.95^8 \cdot 0.05^2 + (10) \cdot 0.95^9 \cdot 0.05 + (1) \cdot 0.95^{10} \\ &= 0.9885 \end{aligned}$$

## Example: machine with 10 components

- ▶  $E(Y) = np = 10 \cdot 0.95 = 9.5$ . So the number of components to fail per run on average is 9.5.
- ▶  $Var(Y) = np(1 - p) = 10 \cdot 0.95 \cdot (1 - 0.95) = 0.475$ .
- ▶  $SD(Y) = \sqrt{Var(Y)} = \sqrt{np(1 - p)} = 0.689$ .

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# Uses of the $X \sim \text{Geom}(p)$

- ▶ For an indefinitely-long sequence of independent, success-failure trials, each with  $P(\text{success}) = p$ ,  $X$  is the number of trials it takes to get a success.
- ▶ Examples:
  - ▶ Number of rolls of a fair die until you land a 5.
  - ▶ Number of shipments of raw material you get until you get a defective one.
  - ▶ The number of enemy aircraft that fly close before one flies into friendly airspace.
  - ▶ Number hexamine pellets you make before you make one that does not conform.
  - ▶ Number of buses that come before yours.

# Geometric random variables

- ▶  $X \sim \text{Geometric}(p)$  – that is,  $X$  has a geometric distribution with parameter  $p$  ( $0 < p < 1$ ) – if its pmf is:

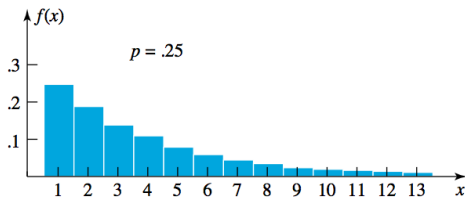
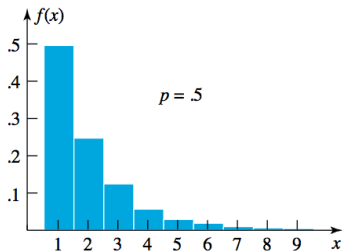
$$f_X(x) = \begin{cases} p(1-p)^{x-1} & x = 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

and its cdf is:

$$F_X(x) = \begin{cases} 1 - (1-p)^x & x = 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

- ▶  $E(X) = \frac{1}{p}$
- ▶  $\text{Var}(X) = \frac{1-p}{p^2}$

# A look at the $\text{Geom}(p)$ distribution



## Example: shorts in NiCad batteries

- ▶ An experimental program was successful in reducing the percentage of manufactured NiCad cells with internal shorts to around 1%.
- ▶ Let  $T$  be the test number at which the first short is discovered. Then,  $T \sim \text{Geom}(p)$ .

$$\begin{aligned}P(\text{1st or 2nd cell tested is has the 1st short}) &= P(T = 1 \text{ or } T = 2) \\&= f(1) + f(2) \\&= p + p(1 - p) \\&= 0.01 + 0.01(1 - 0.01) \\&= 0.02\end{aligned}$$

$$\begin{aligned}P(\text{at least 50 cells tested w/o finding a short}) &= P(T > 50) \\&= 1 - P(T \leq 50) \\&= 1 - F(50) \\&= 1 - (1 - (1 - p)^x) \\&= (1 - p)^x \\&= (1 - 0.01)^{50} \\&= 0.61\end{aligned}$$



## Example: shorts in NiCad batteries

$$E(T) = \frac{1}{p} = \frac{1}{0.01}$$

= 100 tests for the first short to appear, on avg.

$$SD(T) = \sqrt{Var(T)} = \sqrt{\frac{1-p}{p^2}}$$

$$= \sqrt{\frac{1-0.01}{0.01^2}} = 99.5 \text{ tested batteries}$$

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# Meaning of the Poisson distribution

- ▶ A  $\text{Poisson}(\lambda)$  random variable counts the number of occurrences that happen over a fixed interval of time or space.
- ▶ These occurrences must:
  - ▶ be independent
  - ▶ be sequential in time (no two occurrences at once)
  - ▶ occur at the same constant rate,  $\lambda$ .
- ▶  $\lambda$ , the **rate parameter**, is the expected number of occurrences in the specified interval of time or space.

# Poisson random variables

- ▶  $X \sim \text{Poisson}(\lambda)$  – that is,  $X$  has a poisson distribution with parameter  $\lambda > 0$  – if its pmf is:

$$f_X(x) = \begin{cases} \frac{e^{-\lambda} \lambda^x}{x!} & x = 0, 1, 2, 3, \dots \\ 0 & \text{otherwise} \end{cases}$$

- ▶  $E(X) = \lambda$
- ▶  $\text{Var}(X) = \lambda$

# A look at the Poisson distribution

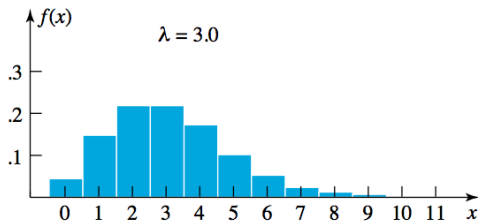
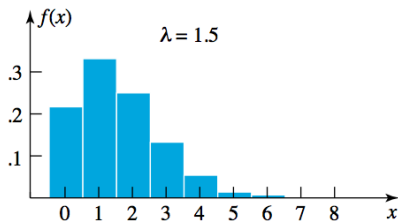
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# Examples

- ▶  $Y$  is the number of shark attacks off the coast of CA next year.  $\lambda = 100$  attacks per year.
- ▶  $Z$  is the number of shark attacks off the coast of CA next month.  $\lambda = 100/12 = 8.3333$  attacks per month
- ▶  $N$  is the number of  $\beta$  particles emitted from a small bar of plutonium, registered by a Geiger counter, in a minute.  $\lambda = 459.21$  particles/minute.
- ▶  $J$  is the number of particles per three minutes.  $\lambda = ?$

$$\begin{aligned}\lambda &= \frac{459.21 \text{ (units particle)}}{1 \text{ (unit minute)}} \cdot \frac{3 \text{ (units minute)}}{1 \text{ (unit of 3 minutes)}} \\ &= \frac{1377.63 \text{ (units particle)}}{1 \text{ (unit of 3 minutes)}} = 1377.62 \text{ particles per 3 minutes}\end{aligned}$$

## Example: Rutherford/Geiger experiment

- ▶ Rutherford and Geiger measured the number of  $\alpha$  particles detected near a small bar of plutonium for 8-minute periods.
- ▶ The average number of particles per 8 minutes was  $\lambda = 3.87$  particles / 8 min.
- ▶ Let  $S \sim \text{Poisson}(\lambda)$ , the number of particles detected in the next 8 minutes.

$$f(s) = \begin{cases} \frac{e^{-3.87}(3.87)^s}{s!} & s = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

$P(\text{at least 4 particles recorded})$

$$= P(S \geq 4)$$

$$= f(4) + f(5) + f(6) + \dots$$

$$= 1 - f(0) - f(1) - f(2) - f(3)$$

$$= 1 - \frac{e^{-3.87}(3.87)^0}{0!} - \frac{e^{-3.87}(3.87)^1}{1!} \\ - \frac{e^{-3.87}(3.87)^2}{2!} - \frac{e^{-3.87}(3.87)^3}{3!}$$

$$= 0.54$$

## Example: arrival at a university library

- ▶ Some students' data indicate that between 12:00 and 12:10 P.M. on Monday through Wednesday, an average of around 125 students entered a library at Iowa State University library.
- ▶ Let  $M$  be the number of students entering the ISU library between 12:00 and 12:01 PM next Tuesday.
- ▶ Model  $M \sim \text{Poisson}(\lambda)$ .
- ▶ Having observed 125 students enter between 12:00 and 12:10 PM last Tuesday, we might choose:

$$\begin{aligned}\lambda &= \frac{125 \text{ (units of student)}}{1 \text{ (unit of 10 minutes)}} \cdot \frac{1 \text{ (unit of 10 minutes)}}{10 \text{ (units of minute)}} \\ &= \frac{12.5 \text{ (units of student)}}{1 \text{ (unit minute)}} = 12.5 \text{ students per minute}\end{aligned}$$



## Example: arrival at a university library

- Under this model, the probability that between 10 and 15 students arrive at the library between 12:00 and 12:01 PM is:

$$\begin{aligned}P(10 \leq M \leq 15) &= f(10) + f(11) + f(12) + f(13) + f(14) + f(15) \\&= \frac{e^{-12.5}(12.5)^{10}}{10!} + \frac{e^{-12.5}(12.5)^{11}}{11!} + \frac{e^{-12.5}(12.5)^{12}}{12!} \\&\quad + \frac{e^{-12.5}(12.5)^{13}}{13!} + \frac{e^{-12.5}(12.5)^{14}}{14!} + \frac{e^{-12.5}(12.5)^{15}}{15!} \\&= 0.60\end{aligned}$$

## Example: shark attacks

- ▶ Let  $X$  be the number of unprovoked shark attacks that will occur off the coast of Florida next year.
- ▶ Model  $X \sim \text{Poisson}(\lambda)$ .
- ▶ From the shark data at <http://www.flmnh.ufl.edu/fish/sharks/statistics/FLactivity.htm>, 246 unprovoked shark attacks occurred from 2000 to 2009.
- ▶ Hence, I calculate:

$$\begin{aligned}\lambda &= \frac{246 \text{ (units attack)}}{1 \text{ (unit of 10 years)}} \cdot \frac{1 \text{ (unit of 10 years)}}{10 \text{ (units year)}} \\ &= \frac{24.6 \text{ (units attack)}}{1 \text{ (unit year)}} = 24.6 \text{ attacks per year}\end{aligned}$$

## Example: shark attacks

$$P(\text{no attacks next year}) = f(0) = e^{-24.6} \cdot \frac{24.6^0}{0!}$$
$$\approx 2.07 \times 10^{-11}$$

$$P(\text{at least 5 attacks}) = 1 - P(\text{at most 4 attacks})$$
$$= 1 - F(4)$$
$$= 1 - f(0) - f(1) - f(2) - f(3) - f(4)$$
$$= 1 - e^{-24.6} \frac{24.6^0}{0!} - e^{-24.6} \frac{24.6^1}{1!} - e^{-24.6} \frac{24.6^2}{2!}$$
$$- e^{-24.6} \frac{24.6^3}{3!} - e^{-24.6} \frac{24.6^4}{4!}$$
$$\approx 0.9999996$$

$$P(\text{more than 30 attacks}) = 1 - P(30 \text{ or less attacks})$$
$$= 1 - e^{-24.6} \sum_{i=0}^{30} \frac{24.6^i}{i!} = 1 - e^{-24.6} \cdot 4.251 \times 10^{10}$$
$$\approx 0.1193$$

## Example: shark attacks

- ▶ Now, let  $Y$  be the total number of shark attacks in Florida during the next 4 months.
- ▶ Let  $Y \sim \text{Poisson}(\theta)$ , where  $\theta$  is the true shark attack rate per 4 months:

$$\begin{aligned}\theta &= \frac{24.6 \text{ (units attack)}}{1 \text{ (unit year)}} \cdot \frac{1/3 \text{ (unit year)}}{1 \text{ (unit of 4 months)}} \\ &= \frac{8.2 \text{ (units attack)}}{1 \text{ (unit of 4 months)}} = 8.2 \text{ attacks per 4 months}\end{aligned}$$

## Example: shark attacks

$$P(\text{no attacks in next 4 months}) = f(0) = e^{-8.2} \cdot \frac{8.2^0}{0!} \\ \approx 0.000275$$

$$P(\text{at least 5 attacks}) = 1 - P(\text{at most 4 attacks}) \\ = 1 - F(4) \\ = 1 - f(0) - f(1) - f(2) - f(3) - f(4) \\ = 1 - e^{-8.2} \frac{8.2^0}{0!} - e^{-8.2} \frac{8.2^1}{1!} - e^{-8.2} \frac{8.2^2}{2!} \\ - e^{-8.2} \frac{8.2^3}{3!} - e^{-8.2} \frac{8.2^4}{4!} \\ \approx 0.9113$$

$$P(\text{more than 30 attacks}) = 1 - P(30 \text{ or less attacks}) \\ = 1 - e^{-8.2} \sum_{i=0}^{30} \frac{8.2^i}{i!} = 1 - e^{-8.2} \cdot 4.251 \times 10^{10} \\ \approx 9.53 \times 10^{-10}$$