

# Chapter 4 part 3

## Discrete Random Variables

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MATH 241

# Outline

- 1 Poisson distribution
- 2 Geometric distribution and Negative Binomial distribution

# Count the number of ...



In beer brewing, cultures of yeast are kept alive in jars of fluid before being put into the mash.

- It's critical to control the amount of yeast used.
- Number of yeast cells in a fluid sample can be seen under a microscopes.
- Yeast cells are constantly multiplying and dividing.
- A famous statistician, Wiliam Sealy Gosset (aka "Student"), who worked for the Guinness Brewing Compnay in early 1900's, modeled the counts of yeast cells using the *Poisson distribution*.

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The Poisson distribution was used in 1898 to count the number of soldiers in the Prussian Army who died accidentally from horse kicks

# Poisson distribution

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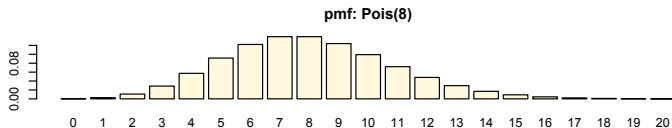
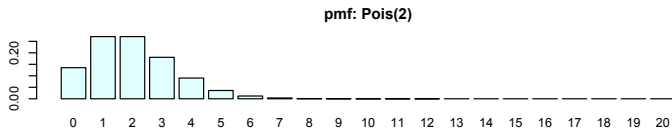
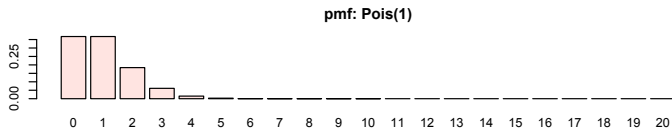
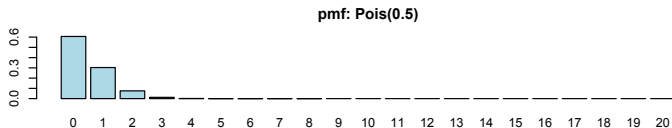
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# Pmf of Poisson distribution



# Properties of Poisson distribution $p(k) = e^{-\lambda} \frac{\lambda^k}{k!}$

- Well-defined (validness of pmf): non-negative, and

$$\sum_{k=0}^{\infty} p(k) = 1$$

- Taylor Series

$$f(x) = f(x_0) + \frac{x - x_0}{1!} f'(x_0) + \frac{(x - x_0)^2}{2!} f''(x_0) + \dots$$

- Use Taylor series to verify that the Poisson distribution is well-defined (required)

$$e^{\lambda} = 1 + \lambda + \frac{\lambda^2}{2!} + \dots + \frac{\lambda^k}{k!} + \dots$$

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- Mean (**required**) textbook page 137

$$E[X] = \lambda$$

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$$E[X] = \lambda$$

- Variance (**required**) textbook page 138

$$Var[X] = \lambda$$

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$$X \sim P(\lambda), E[X] = 1.73 \implies \lambda = 1.73$$

$$\begin{aligned} P(X \leq 1) &= p(0) + p(1) \\ &= e^{-\lambda} \frac{\lambda^0}{0!} + e^{-\lambda} \frac{\lambda^1}{1!} \\ &= 0.1773 + 0.3067 = 0.4840 \end{aligned}$$

# Use Poisson to approximate Binomial distribution

Let  $X \sim \text{Bin}(n, p)$ . If

- $p$ : small
- $n$ : large
- $\lambda = np$ : of moderate size

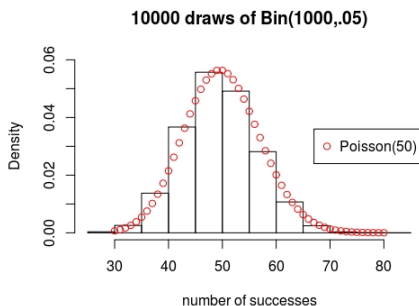
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# Recap

Poisson distribution  $X \sim P(\lambda)$

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

- mean  $\mu = \lambda$
- variance  $\sigma^2 = \lambda$
- Approximate Binomial distribution with small  $p$ , large  $n$ , moderate  $np$

$$P(np) \approx \text{Bin}(n, p)$$

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## Geometric distribution

A gambler plays at a roulette table and always bet on red until he wins...

In each round, his chance of winning is  $18/38 = 0.47$ .

Let  $X$  denote the number of rounds he plays.

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- Note the difference between Geometric distribution and Binomial distribution! Eg: whether the total number of trials is fixed.

# Properties of Geometric distribution $p(k) = (1 - p)^{k-1}p$

- Well-defined (validness of pmf): non-negative,  $\sum_{k=1}^{\infty} p(k) = 1$   
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- Mean (**not required**) Textbook page 148 for derivation

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- Variance (**not required**) Textbook page 148 for derivation

$$Var[X] = \frac{1 - p}{p^2}$$



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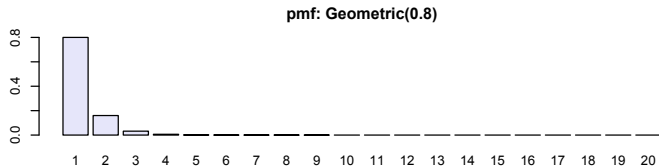
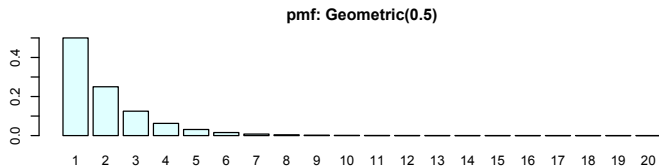
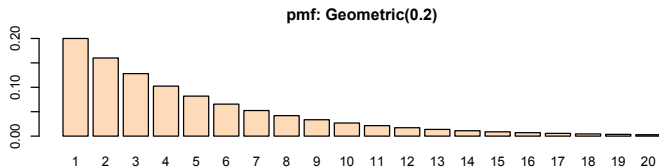
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- Geometric random variable is memoryless.

$$\begin{aligned} P(X > n + k \mid X > n) &= \frac{P(X > n + k)}{P(X > n)} \\ &= \frac{(1 - p)^{n+k}}{(1 - p)^n} = (1 - p)^k = P(X > k) \end{aligned}$$

# Pmf of Geometric distribution



# Negative Binomial distribution

## Definition

Denote random variable  $X$  that takes value in  $\{1, 2, \dots\}$  as having a *Negative Binomial distribution* with parameter  $p \in (0, 1)$  if its pmf is

$$X \sim NB(r, p) \iff p(k) = \binom{k-1}{r-1} (1-p)^{k-r} p^r, \quad k = r, r+1, \dots$$

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- Well-defined (validness of pmf). (not required)
- Connection between Negative Binomial and Geometric distributions

$$X \sim \text{Geometric}(p) \iff X \sim \text{NB}(1, p)$$

---

*Note:* Proof of  $\sum_{k=r}^{\infty} p(k) = 1$  for Negative Binomial distribution is not required.

# Properties of Neg Binom $p(k) = \binom{k-1}{r-1} (1-p)^{k-r} p^r$

- Mean (**not required**)

$$E[X] = \frac{r}{p}$$

Textbook page 150 for derivation

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- Variance (**not required**)

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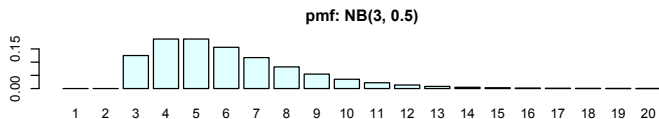
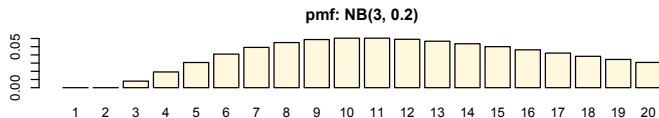
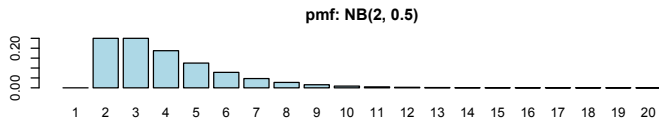
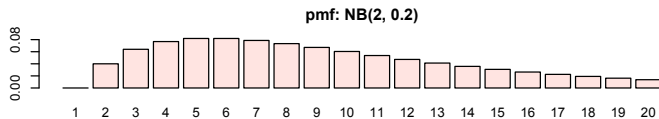
$$Var[X] = \frac{r(1-p)}{p^2}$$

Textbook page 151 for derivation

- Recall that for Geometric( $p$ ),

$$\mu = \frac{1}{p}, \sigma^2 = \frac{1-p}{p^2}$$

# Pmf of Negative Binomial distribution



# Recap

$X$ : the number of trials performed until we get  $r$  success, where  $p$  is the probability of success on each trial.

$$p(k) = \binom{k-1}{r-1} (1-p)^{k-r} p^r, \quad k = r, r+1, \dots$$

- Negative Binomial distribution  $X \sim \text{NB}(r, p)$
- Mean  $\mu = \frac{r}{p}$ , variance  $\sigma^2 = \frac{r(1-p)}{p^2}$ .
- If  $r = 1$ , Geometric distribution  $X \sim \text{NB}(1, p) = \text{Geometric}(p)$
- Geometric distribution is memoryless.

# Review: discrete distributions

Name	Range	pmf $p(x)$	mean	variance
Ber( $p$ )	$\{0, 1\}$	$p^x(1-p)^{1-x}$	$p$	$p(1-p)$



# Review: discrete distributions

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NegBin( $r, p$ )	$\{r, r+1, \dots\}$	$\binom{x-1}{r-1} (1-p)^{x-r} p^r$	$\frac{r}{p}$	$\frac{r(1-p)}{p^2}$