

# STAT 2593

## Lecture 014 - The Hypergeometric and Negative Binomial Distributions

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## The Hypergeometric and Negative Binomial Distributions

## Learning Objectives

1. Understand the hypergeometric distribution, its use cases, and its properties.
2. Understand the negative binomial distribution, its use cases, and its properties.



## The Hypergeometric Distribution

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- ▶ The random variable  $X$  of the number of successes then follows  $X \sim \text{HyperGeo}(N, M, n)$ .
  - ▶ We have  $E[X] = n\frac{M}{N}$ ,  $\text{var}(X) = \frac{N-n}{N-1}n\frac{M}{N}\left(1 - \frac{M}{N}\right)$ , and

$$p(x) = \frac{\binom{M}{x} \binom{N-M}{n-x}}{\binom{N}{n}}.$$

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- ▶ This is almost identical to the Binomial distribution, with an extra multiplicative term.
  - ▶ This term is known as the **finite population correction factor**.
  - ▶ As  $N \rightarrow \infty$ , then this tends towards 1, which demonstrates why the binomial can approximate sampling without replacement.

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- ▶ The random variable  $X$  of the counts of the number of failures before the  $r$ th success follows  $X \sim \text{NegBin}(r, p)$ .
  - ▶ We have  $E[X] = \frac{r(1-p)}{p}$ ,  $\text{var}(X) = \frac{r(1-p)}{p^2}$ , and

$$p(x) = \binom{x + r - 1}{r - 1} p^r (1 - p)^x.$$

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- ▶ If we take  $r = 1$  then the negative binomial is just the geometric distribution.
  - ▶ Very often we care about  $r = 1$ , which is why it gets a special name and is treated separately.
- ▶ Like the geometric distribution, sometimes the random variable  $X$  will instead count the total number of trials, including successes.

## Summary

- ▶ The hypergeometric distribution characterizes the number of successes when sampling *without* replacement.
- ▶ The hypergeometric distribution relies on the assumptions of fixed, finite total, with a fixed number of successes, and a fixed number of trials.
- ▶ The negative binomial distribution counts the number of failures required to achieve a certain number of successes.
- ▶ The negative binomial distribution relies on the same assumptions regarding bernoulli trials as we have seen.
- ▶ Both distributions have closed form PMFs, expectations, and variances.