

Lecture Notes 12

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Today we talk about the discrete distributions related to Bernoulli distribution.

1 Big picture

Bernoulli distribution is a single event with two possible outcome: yes/no. The probability is p for the yes result and $(1 - p)$ for the no. Intuitively, we can view a Bernoulli distribution as an indicator to identify whether an event has happened.

What if we want to consider more than one event?

Imagine the following scenario, there is a large population containing N elements and M of them are label as *type-I* and the rest $N - M$ are labeled as *type-II*. Now, as a statistician, we want to draw some inference about the population, but we have only limited access to the population, say k samples. What can we know from the experiment?

Basically, we can categorize the above scenario with two different properties:

- Draw *with replacement* or *without replacement*.
- Draw *fix number of samples*, or keep drawing *until a certain event happens*?

With these two factors, we can extend Bernoulli distribution into the following three discrete distribution:

	Replacement	Draw	Goal
Hypergeometric	Without	k times	Number of yes
Binomial	With	k times	Number of yes
Negative Binomial	With	Wait until r yes	Number of no

2 Hypergeometric distribution

2.1 Definition

Hypergeometric distribution describes the probability of the number of *yes* result under k samples **without replacement**. The density function consists of three parameters: (N, M, k) and the pdf is

$$f(x|N, M, k) = \frac{\binom{M}{x} \binom{N-M}{k-x}}{\binom{N}{k}} \mathbf{1}_{(\max(0, k-(N-M)), \min(M, k))}(x)$$

Here, we discuss the meaning of each term:

- $\binom{N}{k}$ in the denominator is the number of possible k samples outcome.
- $\binom{M}{x}$ in the numerator is the number of possible combinations of k yes instances.
- $\binom{N-M}{M-x}$ in the numerator is the number of possible combinations of $x - k$ no instances.

2.2 Basic properties

Here, we list the mean and variance of hypergeometric distribution and discuss the idea of reparametrize techniques.

- $\mathbb{E}[X|N, M, k] = k \frac{M}{N}$
- $\text{var}[X|N, M, k] = k \frac{M}{N} \frac{N-M}{N} \frac{N-k}{N-1}$

In the following, we are going to prove the above results via reparametrize techniques and factorial moment. **Proof:**

- The mean of $X \sim \text{Hypergeometric}(N, M, k)$

$$\begin{aligned}
\mathbb{E}[X|N, M, k] &= \sum_{x=\max(0, k-(N-M))}^{\min(M, k)} x \frac{\binom{M}{x} \binom{N-M}{k-x}}{\binom{N}{k}} = \sum_{x=\max(0, k-(N-M))}^{\min(M, k)} M \frac{\binom{M-1}{x-1} \binom{N-M}{k-x}}{\binom{N}{k}} \\
&= \sum_{x=\max(0, k-(N-M))}^{\min(M, k)} M \frac{\binom{M-1}{x-1} \binom{N-(M-1)}{(k-1)-(x-1)}}{\binom{N-1}{k-1} \times \frac{N}{k}} \\
&= k \frac{M}{N} \sum_x \frac{\binom{M-1}{x-1} \binom{N-(M-1)}{(k-1)-(x-1)}}{\binom{N-1}{k-1}} = k \frac{M}{N}
\end{aligned}$$

- The variance of $X \sim \text{Hypergeometric}(N, M, k)$

$$\text{var}[X|N, M, k] = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = \mathbb{E}[X(X-1)] + \mathbb{E}[X] - \mathbb{E}[X]^2$$

As we know $\mathbb{E}[X]$, it suffices to find $\mathbb{E}[X(X-1)]$. The trick that computing the expectation of $X(X-1)$ instead of that of X^2 is called *factorial moment*, which is computation-friendly when having lots of binomial terms. As a result,

$$\mathbb{E}[X(X-1)]$$

2.3 Approximation

Hypergeometric distribution can be approximated by binomial distribution as $N, M \rightarrow \infty$ and $\frac{M}{N} \rightarrow p$. Intuitively, the population size grows to infinite and thus each draw has negligible influence to the population, which makes the whole process to be identical. As the drawing process is uniform,

the process becomes independent. Thus, we can view it as a binomial. The derivation is simple:

$$\begin{aligned}
P(x|N, M, k) &= \frac{\binom{M}{x} \binom{N-M}{k-x}}{\binom{N}{k}} \approx \frac{\frac{M^x}{x!} \frac{(N-M)^{k-x}}{(k-x)!}}{\frac{N^k}{k!}} \\
&= \frac{k!}{x!(k-x)!} \frac{M^x (N-M)^{k-x}}{N^k} \\
\left(\frac{M}{N} = p\right) &= \binom{k}{x} p^x (1-p)^{k-x}
\end{aligned}$$

3 Binomial distribution

3.1 Definition

The binomial distribution describe the probability of the number of *yes* results with a fixed number of i.i.d. drawing **with replacement**. The density function consists of two parameters: (N, p) and the pdf is

$$f(x|N, p) = \binom{N}{x} p^x (1-p)^{N-x} \mathbf{1}_{0,1,\dots,N}(x)$$

Note that, the difference between the definitions of hypergeometric distribution and binomial distribution is not only with/without replacement, the underlying mechanism of binomial distribution is not a fix finite sample space as hypergeometric. For example, the number of drawing can be unbounded, or the *yes* probability should not be necessarily a rational number.

3.2 Basic properties

Suppose $X \sim \text{Binomial}(N, p)$, the following is the mean and variance of X :

- $\mathbb{E}[X|N, p] = Np$
- $\text{var}[X|N, p] = Np(1-p)$

4 Negative binomial distribution

4.1 Definition

The negative binomial distribution describes the probability of the number of *no* instances before certain number of *yes* results in a sequence of i.i.d. drawing. Formally speaking, for a negative binomial distribution with parameters: (p, r) where p is the probability of *yes* and r is the number of *yes* instances we are waiting for, the pdf is

$$f(x|p, r) = \binom{n-1}{r-1} p^r (1-p)^x \mathbf{1}_{0,1,\dots}(x)$$

4.2 Basic properties

Suppose $X \sim \text{Negative Binomial}(p, r)$

- $\mathbb{E}[X|p, r] = \frac{pr}{1-p}$
- $\text{var}[X|p, r] = \frac{pr}{(1-p)^2}$
- When $r = 1$, it is called *geometric* distribution.
- The drawing process is memoryless. For example, the distribution of number of *no* will remain the same as we conditioned on the number of *no* instances before.
- As we let $p \rightarrow 1$, the *yes* result will tend to happen and some how the distribution will converge to Poisson distribution similarly to binomial distribution. (Detail discussion next time)