

Lecture 15

Announcement: For test: Sec 1 sit on left (when facing BB).
Sec 2 sit on Right (.....).

Last time: - Binomial, geometric, and negative binomial

Random Variables.

- All involve independent trials.

Today: Hypergeometric Random Variable.

Suppose you have:

- N objects
- m of those are "successes"
- $N-m$ are "failures."

If we choose a random sample of n objects
(without replacement!!), then if $X \equiv$

$X = \#$ of successes in a sample of size n ,

then X is a hypergeometric random variable

We have

$$P(X=i) = \frac{\binom{m}{i} \binom{N-m}{n-i}}{\binom{N}{n}}$$

where i satisfies

$$n - (N - m) \leq i \leq \min\{n, m\}.$$

— see the textbook for verification of the following facts. Let X be hypergeometric with parameters

N, n, m . Then

$$E[X] = n \frac{m}{N}$$

$$\text{Var}(X) = n \frac{m}{N} \left(1 - \frac{m}{N}\right) \left(1 - \frac{n-1}{N-1}\right).$$

Setting $\frac{m}{N} = p$, we can write

$$E[X] = np$$

$$\text{Var}(X) = np(1-p) \left(1 - \frac{n-1}{N-1}\right).$$

Now, if $N \gg n$, $\left(1 - \frac{n-1}{N-1}\right) \sim 1$
and so $\text{Var}(X) \sim np(1-p)$, $E[X] = np$.

Idea: If n is small compared to N , then $N - n \sim N$
and so even though the sample is without replacement,
the probabilities do not change much (between trials) and
so $X \sim \text{Bin}(n, p)$, where $p = \frac{m}{N}$.

Ex: Suppose an urn contains 10 balls: 5 red, and 5 blue.

Suppose we choose a sample of size 4.

What is the prob. of choosing 2 red balls?

$N=10$, $m=5$, $n=4$. $X = \# \text{ of reds in a sample of } 4$.

$$P(X=2) = \frac{\binom{5}{2} \binom{10-5}{4-2}}{\binom{10}{4}}$$

$$= \frac{\binom{5}{2}^2}{\binom{10}{4}} = \frac{10}{21}.$$

so $p = \frac{m}{2m} = \frac{1}{2}$.

Suppose m is the # of red balls and $N=2m$. If $n=4$,

$$\text{Then } P(X=2) = \frac{\binom{m}{2} \binom{2m-m}{4-2}}{\binom{2m}{4}}$$

$$= \frac{3(m^2 - m)}{8m^2 - 16m + 6}$$

Letting $m \rightarrow \infty$ gives

$$\lim_{m \rightarrow \infty} \frac{3(m^2 - m)}{8m^2 - 16m + 6} = \frac{3}{8}$$

$$= \binom{4}{2} \left(\frac{1}{2}\right)^2 \left(\frac{1}{2}\right)^2 = P(\text{Bin}(4, \frac{1}{2}) = 2).$$

as expected

Zeta Distribution.

A Random var X is said to have the Zeta dist. if

$$P(X=k) = \frac{C}{k^{\alpha+1}} \quad k=1, 2, 3, \dots$$

C, α constants.. X is called "Zeta" since

$$\begin{aligned} 1 &= \sum_{k=1}^{\infty} \frac{C}{k^{\alpha+1}} = C \sum_{k=1}^{\infty} \frac{1}{k^{\alpha+1}} \\ &= C \underbrace{\zeta(\alpha+1)}_{\zeta \text{ function.}} \end{aligned}$$

If X is a Zeta RV with param α ,

Then $E[X] = \sum_{k=1}^{\infty} \frac{kC}{k^{\alpha+1}} = C \sum_{k=1}^{\infty} \frac{1}{k^{\alpha}} = \frac{\zeta(\alpha)}{\zeta(\alpha+1)}.$

Notice that $\zeta(1) = 1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \dots$ is

not convergent, so we assume $\alpha \geq 2$.

If $\alpha \geq 3$, then

$$\begin{aligned} \text{Var}(X) &= E[X^2] - E[X]^2 = \frac{1}{\zeta(\alpha+1)} \sum_{k=1}^{\infty} \frac{k^2}{k^{\alpha+1}} - \frac{\zeta(\alpha)^2}{\zeta(\alpha+1)^2} \\ &= \frac{\zeta(\alpha-1)}{\zeta(\alpha+1)} - \frac{\zeta(\alpha)^2}{\zeta(\alpha+1)^2}. \end{aligned}$$

Expected Value of Sums of RVs

Let $X: S \rightarrow \mathbb{R}$ be a DRV.

Prop: $E[X] = \sum_{s \in S} X(s) p(s).$

Proof: Suppose that $\text{range}(X) = \{x_1, x_2, x_3, \dots\}.$

Let $S_i = \{s \in S: X(s) = x_i\}.$ Then

$$\begin{aligned} E[X] &= \sum_{i=1}^{\infty} x_i P(X=x_i) \\ &= \sum_{i=1}^{\infty} x_i P(S_i) \\ &= \sum_{i=1}^{\infty} x_i \left(\sum_{s \in S_i} P(s) \right) \\ &= \sum_{i=1}^{\infty} \sum_{s \in S_i} x_i P(s). \\ &= \sum_{i=1}^{\infty} \sum_{s \in S_i} X(s) P(s) \\ &= \sum_{s \in S} X(s) P(s). \end{aligned}$$

} since $\{S_i: i=1,2,\dots\}$
is exhaustive, and
mutually exclusive.

□

Corollary: Let X_1, \dots, X_n be ^{discrete} random variables,

with $X_j: S \rightarrow \mathbb{R}$ $j=1,2,\dots,n$. Then

$X := X_1 + X_2 + \dots + X_n$ is a DRV, $X: S \rightarrow \mathbb{R}$ and

$$E[X] = \sum_{i=1}^n E[X_i].$$

Pf:
$$E[X] = E[X_1 + \dots + X_n]$$

$$= \sum_{s \in S} (X_1(s) + \dots + X_n(s)) p(s)$$

by the previous proposition.

So
$$E[X] = \sum_{s \in S} X_1(s)p(s) + \dots + X_n(s)p(s).$$

$$= \sum_{s \in S} X_1(s)p(s) + \dots + \sum_{s \in S} X_n(s)p(s).$$

$$= E[X_1] + \dots + E[X_n].$$