

* Announcement

① Quiz 4 : Thu (10/1) 9:00 AM PT
~ Fri (10/2) 9:00 AM PT

② Midterm review material today

After today's lecture

→ HW6 Q3, Q4

Ch 5.1 ~ 5.3

Joint distribution table

→ Marginal distribution, Find probability of joint event
Independence, Find Expectation etc. of X and Y .

Method of indicator

- Expectation for Binomial (n, p) , $HG(N, G, n)$

- $E(X)$ for a random count X . Set up indicators and find p .

"Think of what distribution
indicators follow"

STAT 88: Lecture 15

Contents

Section 5.4: Unbiased Estimators

Section 5.5: Conditional Expectation

Section 5.6: Expectation by Conditioning

Last time

Unbiased estimators

< Binom (n, p)
prior (μ)

Probability distributions often have parameters that we wish to estimate. An estimator is a random variable and there is uncertainty what you will get. With an unbiased estimator, on average the estimator will be correct.

Ex Suppose the population has population mean μ , i.e. any sample X from the population has mean $E(X) = \mu$. Let X_1, \dots, X_n be a SRS from the population. We use the sample mean \bar{X} as an estimator of the population mean μ . Sample mean is always unbiased since

$$E(\bar{X}) = E\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n} \sum_{i=1}^n E(X_i) = \mu.$$

Ex Suppose the population consists of zeros and ones. Then the population mean p is the population proportion of ones. If X_1, \dots, X_n are i.i.d. samples from the population, the sample mean \bar{X} is the sample proportion of ones in your sample. By unbiasedness of the sample mean, we have

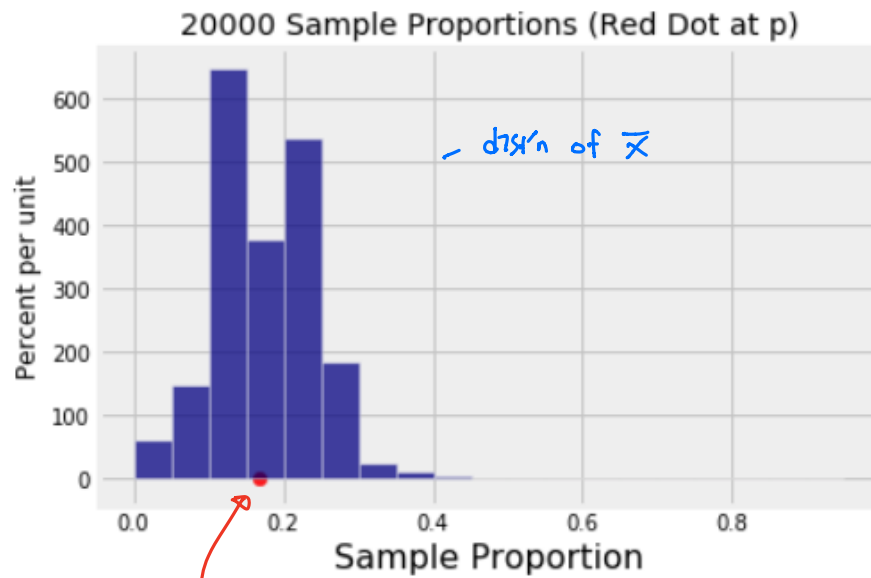
$$E(\bar{X}) = p. \quad \begin{array}{l} X_i \sim \text{Ber}(p) \\ S = X_1 + \dots + X_n \sim \text{Binomial}(n, p) \\ \bar{X} = S/n \quad \rightarrow E(S) = np \end{array}$$

The sampling distribution of sample proportions from the 20,000 repeated experiments:

$n = 30$

$p = 0.1667$ pop. mean

Average of observed sample proportions = 0.1664



$p =$ population proportion

$= E(\bar{x})$

↑

Unbiasedness of sample mean

$\begin{bmatrix} 0 & 0 & 0 & 1 & 1 \end{bmatrix}$

$$E(X) = 0 \cdot \frac{3}{5} + 1 \cdot \frac{2}{5}$$

$$= \frac{2}{5}$$

$$= p$$

\bar{x} ↑ Estimate

Warm up: (Exercise 5.7.11) Let X be the number of cars owned by a Cal student. Here is the distribution of X .

number of cars	0	1	2
probability	2θ	θ	$1 - 3\theta$

$$E(X) = \text{pop. mean} = 2 - 5 \cdot \underbrace{\theta}_{\text{parameter}}$$

(a) Find $E(X)$ (as a function of θ).

(b) Let X_1, \dots, X_n be the number of cars owned by n randomly picked students. Use \bar{X} to find an unbiased estimator of θ .

$$\begin{aligned} \text{(a)} \quad E(X) &= 0 \cdot 2\theta + 1 \cdot \theta + 2 \cdot (1 - 3\theta) \quad \checkmark \quad E(X) = \sum_{\text{all } x} x \cdot P(X=x) \\ &= 2 - 5\theta \end{aligned}$$

$$\text{(b)} \quad E(\bar{X}) = 2 - 5\theta \quad (\text{sample mean unbiased for pop. mean})$$

$$\Rightarrow E(\bar{X}) - 2 = -5\theta$$

$$\Rightarrow \frac{E(\bar{X}) - 2}{-5} = \theta$$

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$\begin{aligned} &\Rightarrow E\left(\frac{2 - \bar{X}}{5}\right) \\ &\quad \underbrace{\frac{2 - \bar{X}}{5}}_{T} \end{aligned}$$

$$E(aX + b) = aE(X) + b$$

$$\Rightarrow T = \frac{2 - \bar{X}}{5}, \quad E(T) = \theta$$

$$X_1, \dots, X_n \rightarrow \bar{X}$$

$\rightarrow T$: estimator for θ
parameter

$$E(aX + b) = aE(X) + b$$

$$\text{WTS: } \frac{E(\bar{X}) - 2}{-5} = E\left(\frac{2 - \bar{X}}{5}\right)$$

$$\frac{E(\bar{X}) - 2}{-5} = -\frac{1}{5}E(\bar{X}) + \frac{2}{5}$$

$$\rightarrow = E\left(-\frac{1}{5}\bar{X} + \frac{2}{5}\right)$$

$$a = -\frac{1}{5}$$

$$b = \frac{2}{5}$$

$$= E\left(\frac{2 - \bar{X}}{5}\right)$$

5.4. Unbiased Estimators (Continued)

Estimating the largest possible value

$$\mu = \sum_{all\ x} x \times P(X=x) = \sum_{i=1}^N i \cdot \frac{1}{N} = \frac{N+1}{2} \quad \text{parameter}$$

Let $X_1, X_2, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Uniform}\{1, 2, \dots, N\}$ for some fixed but unknown N . To estimate N , there are two possible estimators we can come up with:

1. $M = \max\{X_1, \dots, X_n\}$. Note that this is a biased estimator. $E(M) < N$ $\left(\begin{array}{l} N=2 \\ n=2 \\ M=1 \text{ or } 2 \end{array} \right) \sim$
2. We know that the population mean is $\mu = (N+1)/2$ and thus $E(\bar{X}) = (N+1)/2$ since it is unbiased. Then what is an estimator T such that

$$E(T) = N?$$

$$T = 2\bar{X} - 1$$

$$2E(\bar{X}) = N+1$$

$$\Rightarrow 2E(\bar{X}) - 1 = N$$

$$\Rightarrow E(2\bar{X} - 1) = N$$

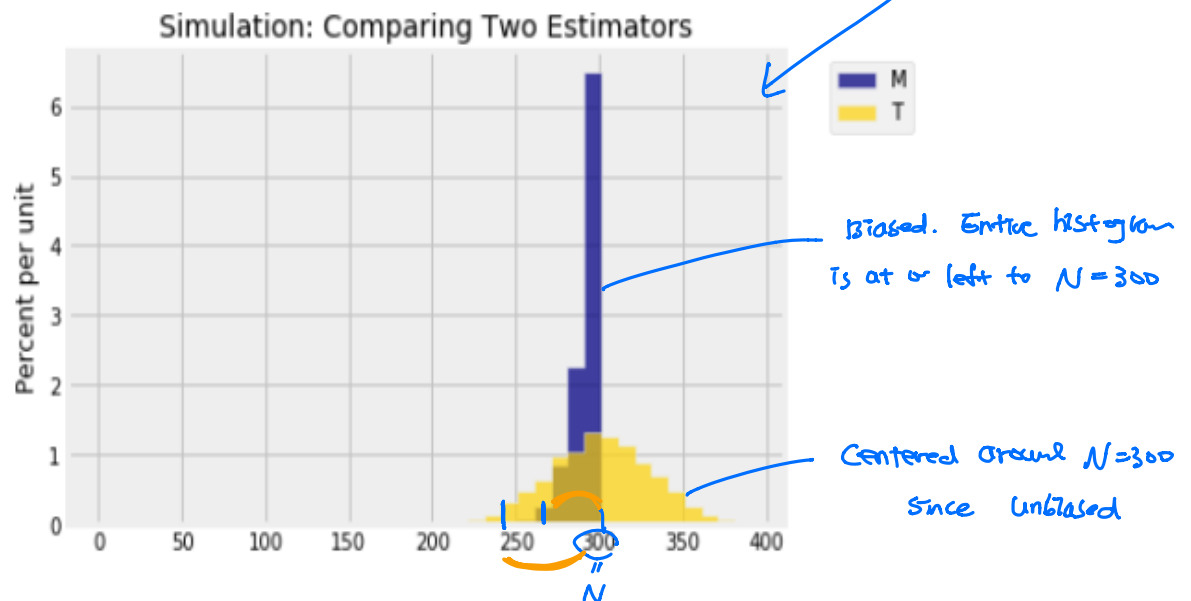
Unbiasedness

Lets look at sampling distribution of (1) $M = \max(X_1, \dots, X_n)$ and (2) $T = 2\bar{X} - 1$.

$N = 300$ parameter
 $n = 30$ #Samples
 5000 repetitions

Exp 1 $X_1, \dots, X_{30} \rightarrow M^{(1)}, T^{(1)}$
 Exp 2 $"$
 \vdots
 Exp 5000 $X_1, \dots, X_{30} \rightarrow M^{(5000)}, T^{(5000)}$

$\sim \text{Unif}\{1, 2, \dots, 300\}$



The histograms show that both estimators have pros and cons.

M - Pros: small spread of values; Cons: biased.

T - Pros: unbiased; Cons: big spread of values.

Unbiasedness is a good property, but so is low variability. Bias-variance tradeoff

$$\text{Risk} = \underbrace{(\text{Bias})^2}_{\substack{\text{A measure of} \\ \text{how good your estimator is}}} + \underbrace{(\text{Var})}_{\substack{\text{"spread of histogram"}}$$

5.5. Conditional Expectation

Let's first review how to find expectation of a joint distribution. A joint distribution for two random variables X and S is given below:

	$X = 1$	$X = 2$	$X = 3$
$S = 2$	0.0625	0	0
$S = 3$	0.125	0.125	0
$S = 4$	0.0625	0.25	0.0625
$S = 5$	0	0.125	0.125
$S = 6$	0	0	0.0625

$0.25 = P(S=3)$

The marginal distribution of S is given by summing along the rows:

s	2	3	4	5	6
$P(S = s)$	0.0625	0.25	0.375	0.25	0.0625

Conditional Distribution Suppose someone runs the experiment and tells you that $S = 3$. Given this information, what is the distribution of X ?

$$P(X = 1|S = 3) = \frac{P(X = 1, S = 3)}{P(S = 3)} = \frac{0.125}{0.25} = 0.5.$$

Similarly we can get $P(X = 2|S = 3) = 0.5$ and $P(X = 3|S = 3) = 0$.

(conditional) distribution of X given $S=3$

If X and S are two random variables on the same outcome space, then for a fixed value s of S , the conditional distribution of X given $S = s$ is

- the set of all possible values of X under the condition that $S = s$, and
- all the corresponding conditional probabilities $P(X = x|S = s)$.

The distribution of X changes depending on the given value of S :

	$X=1$	$X=2$	$X=3$
Conditional Dist'n of X given $S=3$	0.5	0.5	0

	$X=1$	$X=2$	$X=3$
Conditional Dist'n of X given $S=4$	0.667	0.667	0.667

Conditional Expectation The expectation of X , also called the unconditional expectation of X , is easy to see from the distribution table:

ordinary expectation
= $E(X)$

x	1	2	3
$P(X = x)$	0.25	0.5	0.25

$$E(X) = 2 = 1 \cdot 0.25 + 2 \cdot 0.5 + 3 \cdot 0.25$$

Given that S has the value s , the conditional distribution of X is just an ordinary distribution and thus has an expectation. This is called the conditional expectation of X given $S = s$ and is denoted $E(X|S = s)$.

$$E(X|S = 3) = 1 \cdot 0.5 + 2 \cdot 0.5 + 3 \cdot 0 = 1.5$$

Generally, $E(X|S=s) = \sum_{\text{all } x} x P(X=x|S=s)$

Unconditional Expectation
 $E(X) = \sum_{\text{all } x} x P(X=x)$

What is relationship between expectation and conditional expectation?

$$E(X) = \sum_{\text{all } x} xP(X = x) = \sum_{\text{all } x} \sum_{\text{all } s} xP(X = x, S = s).$$

By multiplication rule,

$$P(X = x, S = s) = P(X = x|S = s)P(S = s).$$

So

$$E(X) = \sum_{\text{all } s} \sum_{\text{all } x} xP(X = x, S = s) = \sum_{\text{all } s} \underbrace{\sum_{\text{all } x} xP(X = x|S = s)}_{=E(X|S=s)} P(S = s).$$

Therefore

$$E(X) = \sum_{\text{all } s} E(X|S = s)P(S = s).$$

Important: $E(X|S = s)$ is a function of s . For example,

s	2	3	4	5	6
$E(X S = s)$	1	1.5	2	2.5	3

Marginal Distribution of S

s	2	3	4	5	6
$P(S = s)$	0.0625	0.25	0.375	0.25	0.0625

$$\begin{aligned}
 E(X) &= \sum_{\text{all } s} E(X|S=s) P(S=s) \\
 &= 1 \cdot 0.0625 + 1.5 \cdot 0.25 + 2 \cdot 0.375 + 2.5 \cdot 0.25 + 3 \cdot 0.0625 \\
 &= 2
 \end{aligned}$$

5.6. Expectation by Conditioning

To find expectation of one random variable, it sometimes helps to condition on another random variable.

Time to Reach Campus A student has two routes to campus. Each route has a random duration. The student prefers Route A because its expected duration is 15 minutes compared to 20 minutes for Route B. However, on 10% of her trips she is forced to take Route B because of road work on Route A. On the remaining 90% of the days she takes Route A.

What is the expected duration of her trip on a random day?

Example: (Exercise 5.7.13) A survey organization in a town is studying families with children. Among the families that have children, the distribution of the number of children is as follows.

Number of Children n	1	2	3	4	5
Proportion with n Children	0.2	0.4	0.2	0.15	0.05

Suppose each child has chance 0.51 of being male, independently of all other children. What is the expected number of male children in a family picked at random from those that have children?

Example: You flip a fair coin N times where N is a random variable $N \sim \text{Poisson}(5)$.
What is the expected number of heads you will get?