

Lecture 7 (Ch 1)

We have played around with lots of distributions. It's time to derive one. Here is the derivation of Binomial:

Consider N objects (population), where

Each object is 1 (Head, Girl, ...) or 0 (Tail, Boy, ...)

population is Bernoulli with param. π

Suppose the proportion of 1's in the pop. is known = π .

Now, select n (e.g. 3) of the objects (with replacement) = sample and note the value of each object.

Repeat many many times (e.g. 10^8)

Q What long-run proportion (of the 10^8) will be 1,1,1? 1,1,0? Etc.

Note: I'm not asking for the prop. of 1's in each sample.

I'm asking for the prop., out of 10^8 trials, that we get 1,1,1? Etc.

Make sure you understand prop. & prob.

This is the probability of 1,1,1; or what the book calls "long-run prop."

		$X = \# \text{ of } 1\text{'s}$	
pr. of 1,1,1	$= \pi \cdot \pi \cdot \pi$	3	
1,1,0	$= \pi \cdot \pi \cdot (1-\pi)$	2	} (3)
1,0,1	$= \pi (1-\pi) \pi$	2	
0,1,1	$= (1-\pi) \pi \pi$	2	
Etc.			
0,0,0	$= (1-\pi) (1-\pi) (1-\pi)$	0	

$$\begin{aligned}
 P(X=3) &= 1 \cdot \pi^3 & \frac{3!}{3!(3-3)!} \\
 P(X=2) &= 3 \pi^2 (1-\pi) & \frac{3!}{2!(3-2)!} \\
 P(X=1) &= 3 (1-\pi)^2 \pi & \frac{3!}{2!(3-2)!} \\
 P(X=0) &= 1 (1-\pi)^3 & \frac{3!}{0!(3-0)!}
 \end{aligned}$$

$$P(X=x) = \frac{3!}{x!(3-x)!} \pi^x (1-\pi)^{3-x}$$

$x = 0, 1, 2, 3$

$$P(X=x) = \frac{n!}{x!(n-x)!} \pi^x (1-\pi)^{n-x} \quad (\text{Table II})$$

$x = 0, 1, 2, \dots, n = \# \text{ of } 1\text{'s out of } n$

This is the mass function, $p(x)$, of a binomial variable x .

Because we derived the above expression using proportions, it follows that $\sum_{x=0}^n p(x) = \sum_{x=0}^n \text{prop}(x) = 1$.

Some Comments:

⇒ Recall the connection between coin tosses and sampling:

prob. of getting x heads out of n tosses of a coin
or 1 toss of n coins

The prob. of getting x boys out of a sample of size n .

⇒ What's π ? Don't confuse it with $P(X=x)$ [Important!]

For the coin example, it's the prob. of getting a H on one toss.

In the other example, it's the prob. of drawing a boy,

i.e. the proportion of boys in the pop.

Note: The prop. of 1's in each sample of size n , does not show up in discussions of Binomial.

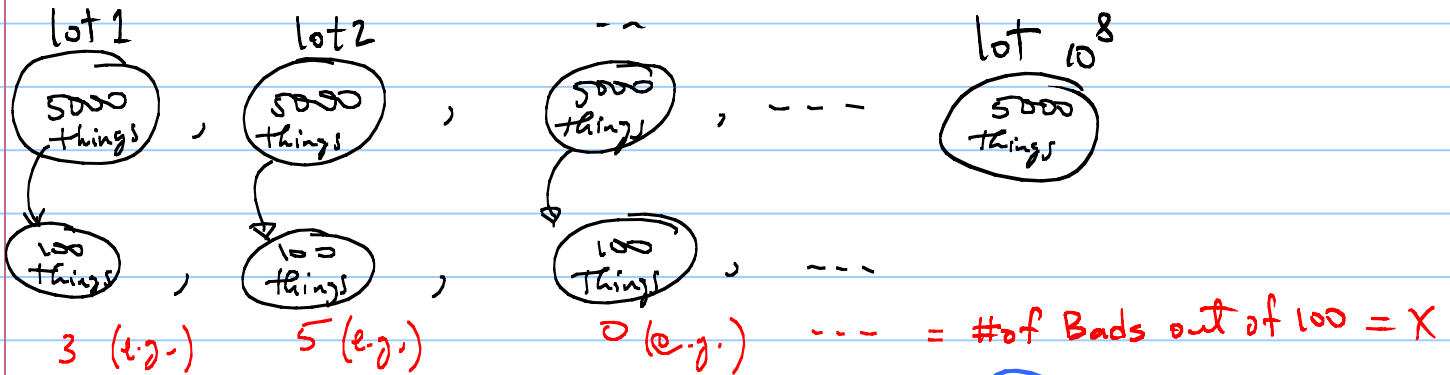
⇒ What's n in $\text{Binom}(n, \pi)$

A comment about phrases like "take a sample of size x from distribution y ."

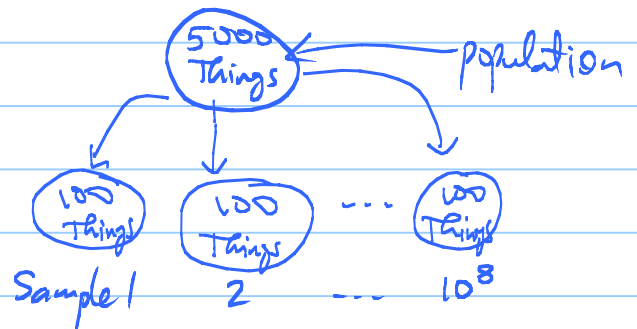
Statements like that are unambiguous when we talk about a distribution like $N(\mu, \sigma)$. For example, a sample of size 3 from $N(\mu=1, \sigma=2)$ is 3 (real) numbers, each between $-\infty$ and $+\infty$, most likely around 1, and with typical spread around 2.

But things get confusing when we talk about "take a sample of size 3 from $\text{Binom}(n, \pi)$." In that case, the sample size is NOT the n in $\text{Binom}(n, \pi)$. Read the last sentence again! The n in $\text{Binom}(n, \pi)$ *may* be interpreted as a "sample size," but that sample (of size n) would be taken from a Bernoulli(π), NOT Binomial. This is evident in the derivation of the binomial; there, when we took a "sample of size 3", the sample was taken from a Bernoulli distribution with parameter π - that's why each sample of size 3 consisted of a sequence of 0s and 1s. By contrast, a sample of size 3 from $\text{Binom}(n, \pi)$ would be a sequence of 3 integers, each between 0 and n . Once again, when we talk about taking a sample of some size from a binomial distribution, the size of the sample is NOT the n parameter of the Binomial.

Example 1.23 (p.56), using binomial.



Assume the lots are identical, i.e. the company manufacturing the 5000 Things is extremely consistent. Then, the picture looks like this:



Q What proportion of these 10^8 lots will have $X=0, 1, \dots, 100$?

G = Good
B = Bad

Sample = $\{G, G, \dots, G\}$
100

Sample = $\{B, B, \dots, B\}$
100

Suppose we know the prop. of Bads in the pop. = $0.5\% = .005 = \pi$

Then $P(X=x) = \binom{100}{x} \pi^x (1-\pi)^{100-x}$

prop. of lots with $X=0$: $\binom{100}{0} \pi^0 (1-\pi)^{100} = .6058$
 $= 1$: $\binom{100}{1} \pi^1 (1-\pi)^{100-1} = .3044$
 $= 2$: $\binom{100}{2} \pi^2 (1-\pi)^{100-2} = .0757$
 $= 3$: Etc. $\binom{100}{3} \pi^3 (1-\pi)^{100-3} = .0124$
 $\vdots$

Lab.

Important Interpretation

In the long-run we expect $\left\{ \begin{array}{l} \sim 60\% \text{ of the lots to be all good.} \\ \sim 30\% \text{ " " " to have 1 bad out of 100.} \\ \sim 7\% \text{ " " " " 2 bads " " " } \\ \text{(i.e. 7\% of the lots to be 2\% defective)} \end{array} \right.$

Now, for large n (a common situation), The $n!$ gets nasty. Also, sometimes π is small. So, consider This limit:

$n \rightarrow \infty$, $\pi \rightarrow 0$ [ie. rare events] while $n\pi = \text{const.} \equiv \lambda$.

$$\binom{n}{x} \pi^x (1-\pi)^{n-x} \xrightarrow{\text{approx.}} \frac{e^{-\lambda} \lambda^x}{x!} = p(x) \text{ of Poisson with param } \lambda. \quad (\text{Table III})$$

No $n!$, No π . Just λ (average of something, next!)

In The last example, since we know n & π , we can compute λ :

$$\lambda = n \cdot \pi = 1000 \cdot 0.005 = 0.5$$

$$\text{Then prop. } (x=0) \approx \frac{e^{-\lambda} \lambda^0}{0!} = e^{-0.5} = \boxed{0.6065} \quad \begin{matrix} \text{exact answer} \\ .6058 \end{matrix}$$

Similarly for $\text{prop}(x=1, 2, 3), \dots \approx \text{exact answers from binomial.}$

Important: Although we derived Poisson (λ) as The limit of Binom (n, π) as $n \rightarrow \infty$, $\pi \rightarrow 0$ with $\lambda = n \cdot \pi$, it turns out that certain random variables follow The Poisson distribution quite independently of Binomial. Said differently, in some problems we can use Poisson (λ) as long as λ itself is known, in which case we don't even need to know n and π (The Binom params).

In short, if you know n, π , use Binom (n, π) -

if you know λ (but not n, π), use Poisson (λ).

But, there is a more fundamental way of knowing which dist to use. Poisson r.v.'s have a certain defining characteristic:

Examples of data that follow the poisson distr:

- # of bombs dropped over London per block.
- # of potholes per unit length of roads.
- # of crashes (cars, planes, buildings) per year.
- # of people arriving at a website per unit time. = X

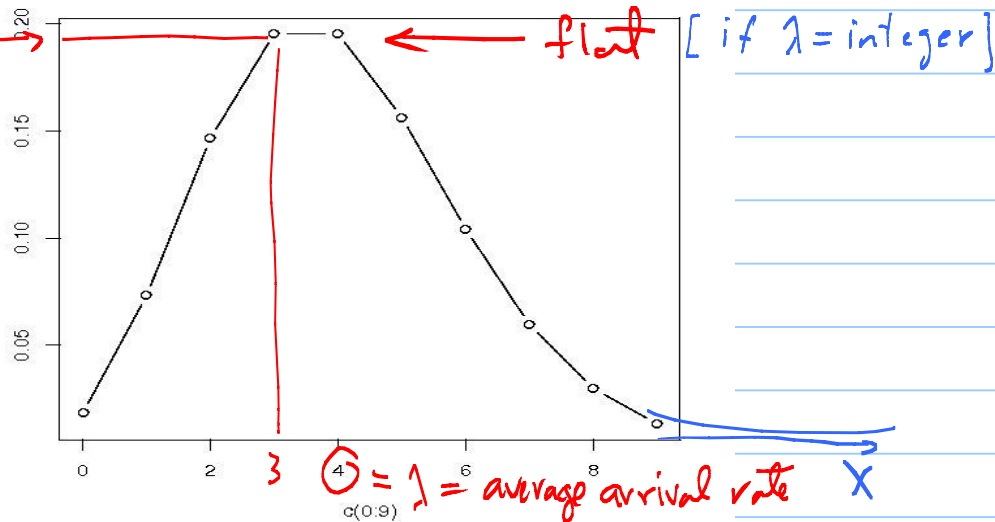
∴ Any r.v. that fits the template "# per unit ..." = Poisson.
And in that case λ = average # per unit ...

(Eg.) An avg. of 4 people arrive at a website per hour.
What's the prob. that 3 people arrive per hour?

Assume X = poisson with $\lambda = 4$ people/hr.

$$P(X=3) = e^{-4} 4^3 / 3!$$

$$= 0.19 \rightarrow$$



Lab
poisson
with $\lambda=4$
for $x=0,1,\dots,9$

Make sure you are able to recognize the situations that lend themselves to the application of Binomial, Poisson, etc.

hw_lect7_1

Consider a Bernoulli dist with parameter p , (i.e. a population consisting of two types of objects denoted $x = 0, 1$, with the proportion of 1s in the population given by p). Take samples of size $n=3$.

- What's the probability that the minimum of the three numbers is 1?
- What's the probability that the minimum of the three numbers is 0?

Hint: repeat the derivation of the binomial distribution, i.e. writing out all possibilities, etc. but with X (i.e., the number of heads out of n) replaced with Min (i.e., the minimum of the three numbers).

hw_lect7-2

For a period of 10 hours, we observe the number of cars that went through a stop sign (without stopping), per hour. Here is my **data**: 2, 2, 3, 2, 4, 2, 0, 1, 3, 2. what's the prob that, for a random hour, all cars will stop at the stop sign? Note: What we are given here is data, and so, we can only **approximate** the distribution parameter(s); for now, go ahead and approximate.

hw-lect7-3

In The prev. hw problem, you have to approximate the param. of a distr. based on some observed data. In some problems, however, the value of the param. can be obtained from knowledge of the probs themselves. For example, suppose $x \sim \text{Pois}(\lambda)$.

- If $\text{pr}(x=0)$ is known, what's the value of λ ?
- If the ratio $\frac{\text{pr}(x=2)}{\text{pr}(x=1)}$ is known, what's the value of λ ?
- Suppose $\text{pr}(x=1)$ is known. Then $\text{pr}(x=1) = \lambda e^{-\lambda}$.
Plot (by hand) $\lambda e^{-\lambda}$ as a function λ , where $0 < \lambda < \infty$.
Clearly mark the maximum value of $\lambda e^{-\lambda}$; call it P .
- Finally, suppose in a problem involving some random variable x , whose distr. is not known, we find $\text{pr}(x=1) > P$. What does that say about using Poisson distr. to describe x ?