

from last time:

$$\frac{\partial f(x, w)}{\partial w} = -\lambda f(x, w) + \lambda f(x-1, w) \quad \text{for } x \geq 1$$

Stat. 501

11-6-18

①

$x=1$ case:

$$\frac{\partial f(1, w)}{\partial w} = -\lambda f(1, w) + \lambda \underbrace{f(0, w)}_{e^{-\lambda w}}$$

$$e^{\lambda w} \frac{\partial f(1, w)}{\partial w} = -\lambda e^{\lambda w} f(1, w) + \lambda$$

$$e^{\lambda w} \frac{\partial f(1, w)}{\partial w} + \lambda e^{\lambda w} f(1, w) = \lambda$$

$$\frac{d}{dw} [e^{\lambda w} f(1, w)] = \lambda$$

②

$$e^{\lambda w} f(1, w) = \lambda w + c$$

$$f(1, w) = \lambda w e^{-\lambda w} + c e^{-\lambda w}$$

$$\text{Use } f(1, 0) = 0$$

$$\therefore \underset{0}{f(1, 0)} = 0 + c$$

$$\underline{c=0}$$

$$\text{So } \underline{f(1, w) = \lambda w e^{-\lambda w}}$$

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 $k=2$ case:

$$\frac{\partial f(2, w)}{\partial w} = -\lambda f(2, w) + \underbrace{\lambda f(1, w)}_{\lambda w e^{-\lambda w}}$$

$$e^{\lambda w} \frac{\partial f(2, w)}{\partial w} + \lambda e^{\lambda w} f(2, w) = \lambda^2 w$$

$$\frac{d}{dw} [e^{\lambda w} f(2, w)] = \lambda^2 w$$

$$e^{\lambda w} f(2, w) = \lambda^2 \frac{w^2}{2} + C$$

$$f(2, w) = \frac{\lambda^2 w^2}{2} e^{-\lambda w} + C e^{-\lambda w}$$

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Use $f(2, 0) = 0$

$$0 = f(2, 0) = 0 + C \quad \therefore C = 0$$

$$f(2, w) = \frac{\lambda^2 w^2}{2} e^{-\lambda w}$$

Summary: $f(0, w) = e^{-\lambda w}$

$$f(1, w) = \lambda w e^{-\lambda w}$$

$$f(2, w) = \frac{(\lambda w)^2}{2} e^{-\lambda w}$$

$$\vdots$$

$$f(k, w) = \frac{(\lambda w)^k}{k!} e^{-\lambda w}$$

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You usually see this as

$$f_X(x) = \frac{(\lambda t)^x e^{-\lambda t}}{x!}$$

$$\text{or } f_X(x) = \frac{\mu^x e^{-\mu}}{x!} \quad \mu = 0, 1, 2, \dots$$

This is the Poisson Distribution

$$\begin{aligned} \text{check } 1 &\stackrel{?}{=} \sum_{x=0}^{\infty} \frac{\mu^x e^{-\mu}}{x!} = e^{-\mu} \underbrace{\left(1 + \mu + \frac{\mu^2}{2!} + \dots\right)}_{e^{\mu}} \\ &= 1 \quad \checkmark \end{aligned}$$

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$$M_X(t) = E[e^{tX}]$$

$$= \sum_{x=0}^{\infty} e^{tx} \frac{\mu^x e^{-\mu}}{x!}$$

$$= e^{-\mu} \sum_{x=0}^{\infty} \frac{(\mu e^t)^x}{x!}$$

$$= e^{-\mu} e^{\mu e^t} = e^{\mu(e^t - 1)}$$

$$M'_X(t) = e^{\mu(e^t - 1)} \mu e^t$$

$$M_x''(t) = \mu \left[e^{\mu(e^t-1)} e^t + e^t e^{\mu(e^t-1)} \mu e^t \right]$$

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$$E[X] = M_x'(0) = \mu$$

$$E[X^2] = M_x''(0) = \mu[1 + \mu] = \mu + \mu^2$$

$$\sigma^2 = \mu + \mu^2 - \mu^2 = \mu$$

Start with Bino(n, p)

(8)

let $n \rightarrow \infty$

$p \rightarrow 0$ (hold np constant)

μ

$$p(x) = \binom{n}{x} p^x q^{n-x}$$

$$p = \frac{\mu}{n}$$

$$= \frac{n!}{x!(n-x)!} \frac{\mu^x}{n^x} \frac{(1 - \frac{\mu}{n})^n}{(1 - \frac{\mu}{n})^x}$$

$$= \frac{\mu^x}{x!} \frac{n(n-1) \cdots (n-x+1)}{n \cdot n \cdots n} \frac{(1 - \frac{\mu}{n})^n}{(1 - \frac{\mu}{n})^x}$$

$$\lim_{n \rightarrow \infty} p(x) = \lim_{n \rightarrow \infty} \frac{\mu^x}{x!} 1 \left(1 - \frac{1}{n}\right) \dots \left(1 - \frac{x-1}{n}\right) \frac{(1 - \frac{\mu}{n})^n}{(1 - \frac{\mu}{n})^x} \quad (9)$$

$$= \frac{\mu^x}{x!} e^{-\mu}$$

$$\text{Let } y = \left(1 - \frac{\mu}{n}\right)^n$$

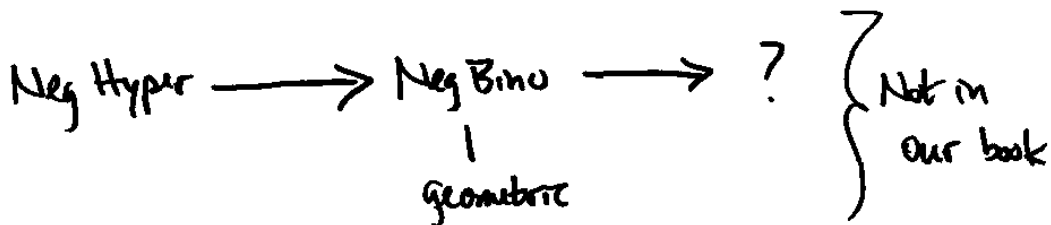
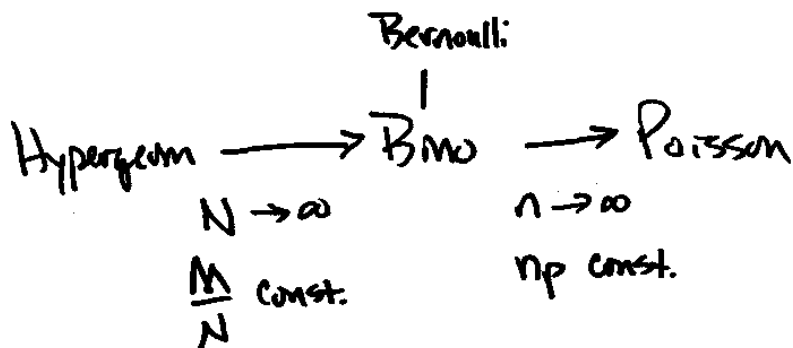
$$\ln y = n \ln \left(1 - \frac{\mu}{n}\right)$$

$$= \frac{\ln \left(1 - \frac{\mu}{n}\right)}{\frac{1}{n}}$$

$$\lim_{n \rightarrow \infty} \ln y = \lim_{n \rightarrow \infty} \frac{\frac{1}{n} \cdot \frac{\mu}{n^2}}{-\frac{1}{n^2}}$$

$$= -\mu$$

$$\therefore \lim_{n \rightarrow \infty} y = e^{-\mu}$$



(11)

Numerical example : Population of 10,000 people

Take a sample of 1000, without replacement

In the population, 500 have blue eyes.

Find the prob. of getting exactly 45 people
in the sample with blue eyes.

Hyper($N = 10000, M = 500, K = 1000$)

$$P(X=45) = \frac{\binom{500}{45} \binom{9500}{955}}{\binom{10000}{1000}} = .047136$$

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Do the Binomial approximation

Bino($n = 1000, p = .05$)

$$P(X=45) = \binom{1000}{45} (.05)^{45} (.95)^{955} = .046281$$

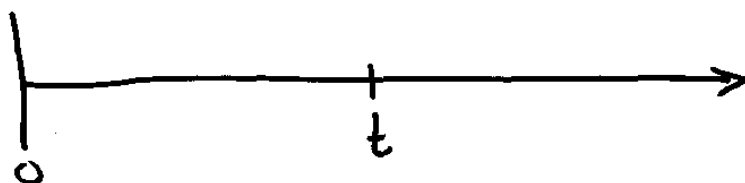
And the Poisson approximation

Poisson($\mu = 50$)

$$P(X=45) = \frac{50^{45}}{45!} e^{-50} = .045826$$

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Suppose we have a Poisson process



Let T be the waiting time until the r^{th} occurrence.

$$\begin{aligned} \text{Find } P(T > t) &= P(\text{there were at most } r-1 \text{ occurrences} \\ &\quad \text{in } (0, t)) \\ &= P(X \leq r-1), \end{aligned}$$

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where X has a Poisson distribution

$$\text{with } \mu = \lambda t$$

↑
expected # occurrences per unit time

Special case of $r=1$:

T is the waiting time until the 1st occurrence

$$\begin{aligned} P(T > t) &= P(X \leq 0) = P(X=0) \\ &= \frac{(\lambda t)^0 e^{-\lambda t}}{0!} = e^{-\lambda t} \end{aligned}$$

$$F_T(t) = P(T \leq t) = 1 - P(T > t)$$

$$= 1 - e^{-\lambda t}$$

(15)

$$f_T(t) = \frac{d}{dt} F_T(t) = \lambda e^{-\lambda t} \quad t \geq 0$$

This is the exponential density function

Define the gamma function:

$$\Gamma(x) = \int_0^{\infty} t^{x-1} e^{-t} dt$$

Properties:

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$$\Gamma(1) = \int_0^{\infty} t^0 e^{-t} dt = -e^{-t} \Big|_0^{\infty}$$

$$= 0 - (-1)$$

$$= 1$$

$$\Gamma(x) = \int_0^{\infty} t^{x-1} e^{-t} dt \quad \left| \begin{array}{l} \text{let } u = t^{x-1} \quad du = (x-1)t^{x-2} dt \\ du = (x-1)t^{x-2} dt \quad v = -e^{-t} \end{array} \right.$$

$$= -t^{x-1} e^{-t} \Big|_0^{\infty} + \int_0^{\infty} (x-1)t^{x-2} e^{-t} dt$$

$$= 0 + (\alpha-1) \int_0^{\infty} t^{\alpha-2} e^{-t} dt$$

\uparrow
 by L'H,
 for $\alpha \geq 1$

$$\therefore \Gamma(\alpha) = (\alpha-1) \Gamma(\alpha-1)$$

$$\text{So } \Gamma(1) = 1$$

$$\Gamma(2) = 1 \cdot \Gamma(1) = 1$$

$$\Gamma(3) = 2 \Gamma(2) = 2$$

$$\Gamma(4) = 3 \Gamma(3) = 6$$

Pattern:

$$\Gamma(n) = (n-1)!$$

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