

Sec 3.1 - continued.

⑤ Trinomial Dist. (special case of multinomial when $k=3$).

$$\text{Let } X_1 = X \quad X_2 = Y \quad X_3 = n - X - Y$$

Joint pmf of $X + Y$:

$$p(x, y) = \frac{n!}{x! y! (n-x-y)!} p_1^x p_2^y p_3^{n-x-y}$$

$$\text{where } \begin{matrix} x+y \leq n \\ p_1 + p_2 + p_3 = 1 \end{matrix} \quad \begin{matrix} x, y \text{ nonneg} \\ \text{integers.} \end{matrix}$$

Conditional pmf of Y given $X=x$ is

$$P_{\text{all}}(y|x) = \begin{cases} \frac{(n-x)!}{y! (n-x-y)!} \left(\frac{p_2}{1-p_1}\right)^y \left(\frac{p_3}{1-p_1}\right)^{n-x-y} \\ \text{for } y=0, 1, \dots, n-x \\ 0, \text{ o.w.} \end{cases}$$

Binom pmf.

$$\binom{n}{x} p^x (1-p)^{n-x}$$

This implies that $Y | X=x \sim \text{Bin}(n-x, \frac{p_2}{1-p_1})$

$$\text{and } E[Y | x] = (n-x) \left(\frac{p_2}{1-p_1}\right)$$

$$E[X | y] = (n-y) \left(\frac{p_1}{1-p_2}\right).$$

②

We can also see this relationship through the correlation coefficient

$$\rho = -\sqrt{\frac{p_1 p_2}{(1-p_1)(1-p_2)}} = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

⑥ Geometric Distribution.

Let $Y = \#$ of failures before the 1st success.

ex) S ← 0 failures $y=0$
 FS ← 1 failure $y=1$
 FFS ← 2 failures $y=2$

PMF

$$P_Y(y) = \begin{cases} p(1-p)^y, & y=0, 1, 2, \dots \\ 0, & \text{o.w.} \end{cases}$$

Assume independent trials.

ⓧ What if we want exactly r successes?
 (stop when obtain r^{th} success).

No fixed n .

⑦ Negative Binomial Dist.

$Y = \#$ of failures before the r^{th} success

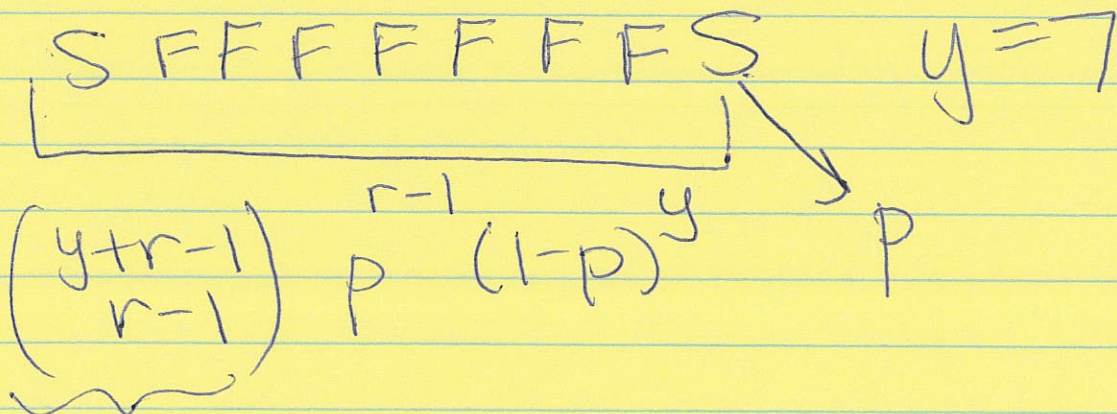
(# trials required to produce exactly r successes = $Y + r$)

$r =$ fixed positive integer.

ex) $r = 2$. (want 2 successes total
last trial = 2nd success)

SS	← 0 failures	$y = 0$
SFS	← 1 failure	$y = 1$

FSS $y = 1$



ways to arrange failures.

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pmf of Y is

$$P_Y(y) = \begin{cases} \binom{y+r-1}{r-1} p^r (1-p)^y & \text{for } y=0,1,2,\dots \\ 0 & \text{o.w.} \end{cases}$$

MGF

$$M_Y(t) = p^r [1 - (1-p)e^t]^{-r}$$

for $t < -\log(1-p)$

↑
ln

In Stat 381:

Neg Binom $X = \#$ trials until k^{th} success.

$$P_X(x) = \begin{cases} \binom{x-1}{k-1} p^k (1-p)^{x-k} & x=k, k+1, \dots \\ 0 & \text{o.w.} \end{cases}$$

Geom $X = \#$ trials until 1st success

$$P_X(x) = \begin{cases} (1-p)^{x-1} p & x=1,2,\dots \\ 0 & \text{o.w.} \end{cases}$$

Sec 3.2 - Poisson Dist.

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Recall a special series.

$$1 + m + \frac{m^2}{2!} + \frac{m^3}{3!} + \dots = \sum_{x=0}^{\infty} \frac{m^x}{x!}$$

→ converges for all m

→ converges to e

def) Poisson Dist.

$$P_X(x) = \begin{cases} \frac{\mu^x e^{-\mu}}{x!}, & x=0, 1, 2, \dots \\ 0, & \text{o.w.} \end{cases}$$

where $\mu > 0$

μ = parameter → it sets the distribution → determines what it looks like.

$$X \sim \text{Pois}(\mu)$$

Models # of events happening in a fixed period of time.

$$M(t) = e^{\mu(e^t - 1)} \quad \text{for } -\infty < t < \infty$$

$$E[X] = \mu = \text{Var}[X]$$

Purpose of μ : this is a parameter that describes the intensity.

Probability Calc.

$$P[X = k] = p_X(k)$$

$$P[X \leq k] = F_X(k)$$

$$= P[X = k] + P[X = k - 1] + \dots + P[X = 0]$$

$$P[X \geq k] = 1 - P[X < k]$$

$$= 1 - P[X \leq k - 1]$$

ex) $X \sim \text{Pois}(2)$ μ

$$\text{PMF: } p(x) = \begin{cases} \frac{2^x e^{-2}}{x!}, & x = 0, 1, 2, \dots \\ 0, & \text{o.w.} \end{cases}$$

$$E[X] = \text{Var}[X] = 2$$

$$\mu = \cancel{0} = 2$$

$$P[1 \leq X] = P[X \geq 1]$$

$$= 1 - P[X < 1]$$

$$= 1 - P[X \leq 0]$$

$$= 1 - P[X = 0]$$

$$= 1 - P_X(0)$$

$$= 1 - \frac{2^0 \cdot e^{-2}}{0!}$$

$$= \cancel{e} \cdot 1 - e^{-2}$$

$$= 1 - 0.135$$

ex) $P[X \leq 4] = 0.947$

$$P[X = 4] = P[X \leq 4] - P[X \leq 3]$$

$$= 0.947 - 0.857$$

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ex) $M_X(t) = e^{4(e^t - 1)}$

$\mu = 4$

$$M_X(t) = e^{\mu(e^t - 1)}$$
$$X \sim \text{Pois}(4)$$

Poisson Process

Setup:

→ Let $g(x, \omega)$ denote the prob. of x changes in each interval of length ω .

ex) 3 claims in 2 days
 $g(3, 2)$

define $g(0, 0) = 1$

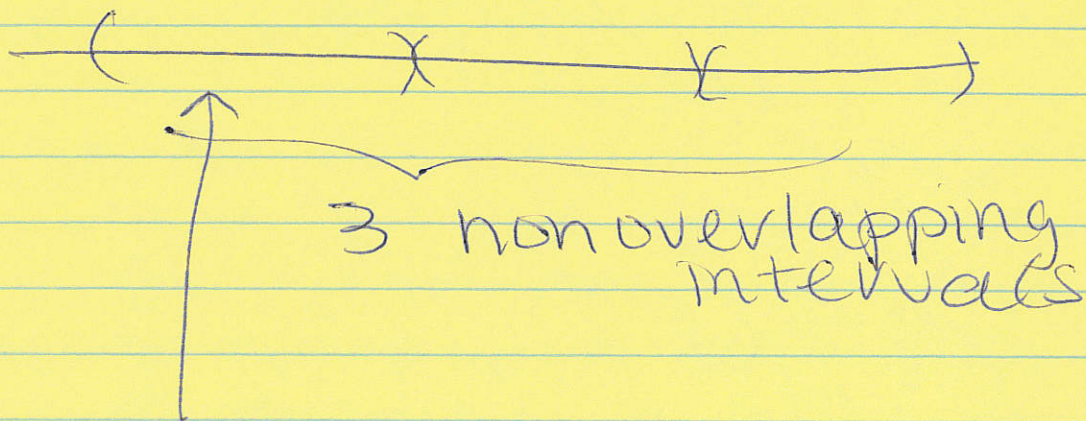
Poisson.

Postulates

- ① the prob. of 1 change in an interval h is equal to the length of interval + a little bit.

(2) the prob. of 2 or more changes in an interval h is negligible.

(3) The numbers of changes in non-overlapping intervals are independent.



of changes in interval 1

does not impact # of changes in any other interval.

In summary:

Take an interval of length w . Let the # of changes in an interval of length 1 be denoted by λ .

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then the # of changes in w (length of time interested in) is given by

$$\lambda w$$

corresponding parameter for Poisson dist.

ex) 5 accidents / hr.

$$\lambda = 5 \quad w = 1 \text{ exactly.}$$

Prob there are ^{exactly} 4 accidents in 1 hr.

$$\mu = 5 \quad P[X=4]$$

$$X \sim \text{Pois}(5).$$

Prob there are ^{exactly} 4 accidents in 2 hours.

$$\lambda = 5 \quad w = 2$$

$$\lambda w = 10 \quad Y \sim \text{Pois}(10)$$

$$P[Y = 4]$$

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Thm Suppose X_1, \dots, X_n are indep RVs.

Take $X_i \sim \text{Pois}(\mu_i)$

then $Y = \sum_{i=1}^n X_i \sim \text{Pois}\left(\sum_{i=1}^n \mu_i\right)$