



# B52 Nov 19 Lec 2 Notes

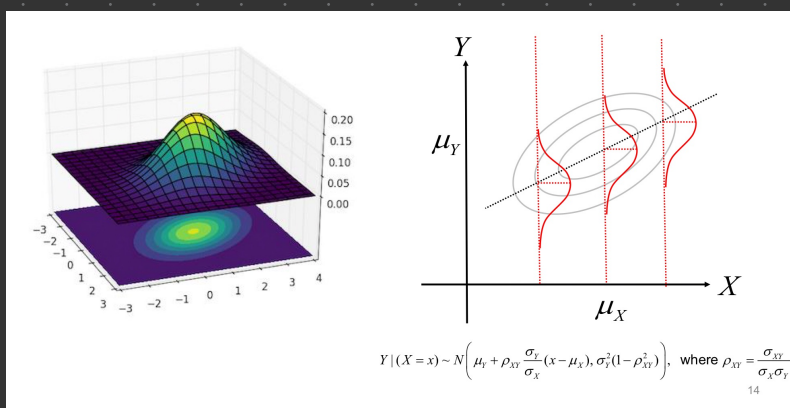
## Conditional distributions

$$\begin{bmatrix} X \\ Y \end{bmatrix} \sim N \left( \mu = \begin{bmatrix} \mu_X \\ \mu_Y \end{bmatrix}, \Sigma = \begin{bmatrix} \sigma_X^2 & \sigma_{XY} \\ \sigma_{XY} & \sigma_Y^2 \end{bmatrix} \right)$$

$$\Rightarrow Y|X=x \sim N \left( \mu_Y + \frac{\sigma_{XY}}{\sigma_X^2} (x - \mu_X), \sigma_Y^2 - \frac{\sigma_{XY}^2}{\sigma_X^2} \right)$$

Conditional mean of  $Y$  is linear function of given value of  $X=x$ .

Conditional variance of  $Y$  is constant &  $\leq$  unconditional variance.



## Ex 1: Linear regression

Assume  $X = \text{height}^2 (\text{m}^2)$  &  $Y = \text{weight (kg)}$  follows bivariate Normal with  $\mu_X = 3.1$ ,  $\sigma_X = 0.15$ ,  $\mu_Y = 70$ ,  $\sigma_Y = 5$ ,  $\rho_{XY} = 0.60$

- (i) What's best guess of person's weight, knowing nothing else?
- (ii) What's best guess of person's weight, if they are 1.6m tall?
- (iii) What's SD of previous guesses?

(i) 70 kg (marginal expected value of  $Y$ )

(ii)  $X = 1.6^2 = 2.56$

$$\begin{aligned} \text{Conditional mean of } Y &= \mu_Y + \rho \frac{\sigma_Y}{\sigma_X} (X - \mu_X) \\ &= 70 + 0.6 \frac{5}{0.15} (2.56 - 3.1) \\ &= 57.2 \text{ kg} \end{aligned}$$

(iii) For marginal of  $Y$ , SD: 5

$$\text{For conditional of } Y, \text{ SD: } \sqrt{\sigma_Y^2 (1 - \rho^2)} = \sqrt{5^2 (1 - 0.6^2)} = 4$$

## Poisson Distribution

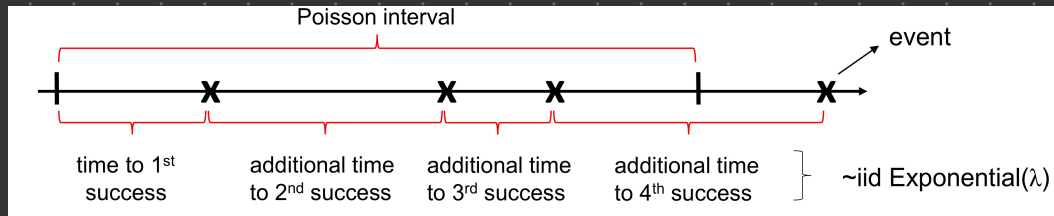
Model # of events over continuum, e.g. # goals in Hockey game.

↳ No fixed # of trials

↳ Thus we model space between events

Poisson RV counts # of successes in some continuous interval, e.g. time or space (length, area, volume)

If area to successes follows (independent)  $\text{Exp}(\lambda)$ , then a Poisson RV counts # of successes fitting in unit interval.



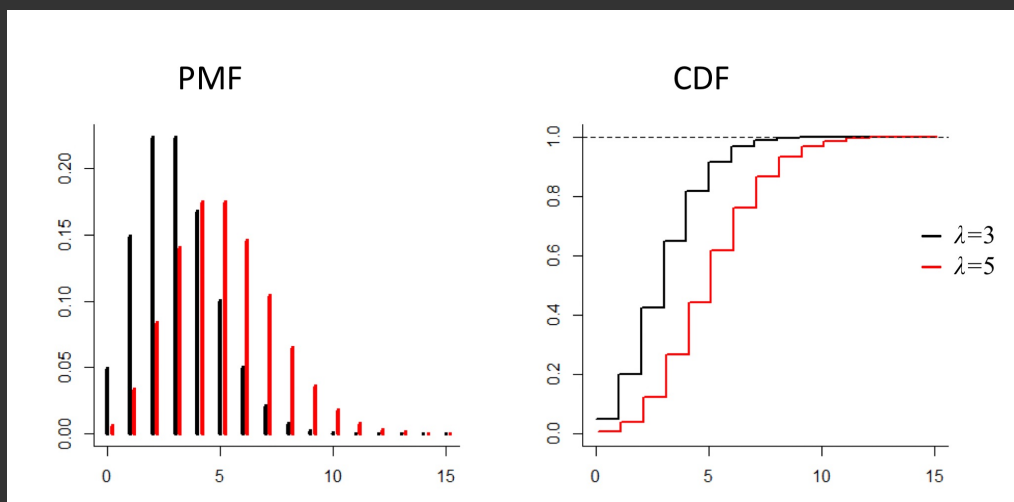
PMF:  $P(X=x) = \frac{\lambda^x}{x!} e^{-\lambda}$ ,  $x = 0, 1, 2, \dots$

Parameter  $\lambda > 0$  represents average # of successes over interval.

Denoted  $X \sim \text{Poisson}(\lambda)$

Verify this is a valid PMF:

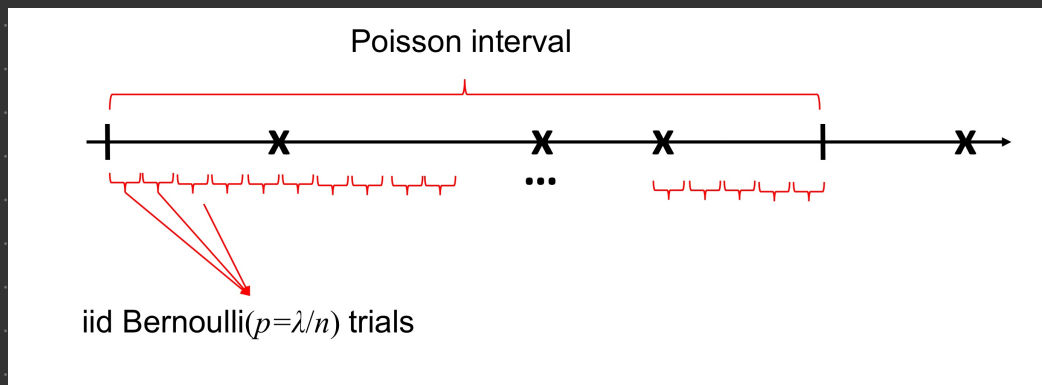
$$\sum_{x=0}^{\infty} P(X=x) = \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} e^{-\lambda} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{\lambda^x}{x!} = e^{-\lambda} e^{\lambda} = e^0 = 1$$



## Poisson Approximation to Binomial

$\text{Poisson}(\lambda)$  is approximated by  $\text{Binomial}(n, \lambda/n)$  as  $n \rightarrow \infty$

Think of dividing Poisson interval to #n pieces, each with  $p = \lambda/n$  probability of containing a single success.



### Binomial, Poisson & Normal

- ↳ Binomial ( $n, \lambda/n$ )  $\rightarrow$  Poisson ( $\lambda$ ) ,  $n \rightarrow \infty$
- ↳ Binomial ( $n, p$ )  $\rightarrow$  Normal ( $np, npq$ ) ,  $n \rightarrow \infty$
- ↳ Poisson ( $\lambda$ )  $\rightarrow$  Normal ( $\lambda, \lambda$ ) ,  $\lambda \rightarrow \infty$

