

## Hierarchical Models and Mixture Distributions

Example (Binomial-Poisson hierarchy) :-

- ① An insect lays a large number of eggs each surviving with prob  $p$ . on the average, how many eggs will survive?
- ② Large number of egg's laid is a random variable
- ③ Survival of each egg is a random variable.

Therefore:

$Y =$  number of egg's laid

$X =$  number of survivor's

$$\Rightarrow Y \sim \text{Poisson}(\lambda)$$

$$X|Y \sim \text{binomial}(Y, p)$$

$$X|Y \quad \text{is} \quad X|Y=y \sim \text{binomial}(y, p)$$

The advantage of hierarchy is that complicated processes may be modelled by a seq of relatively simple models placed in hierarchy

### Example 4.4.2

Therefore  $X = \text{number of survival's}$

$$\Rightarrow P(X=x) = \sum_{y=0}^{\infty} P(X=x, Y=y)$$

$$= \sum_{y=0}^{\infty} P(X=x | Y=y) P(Y=y)$$

$$= \sum_{y=0}^{\infty} \binom{y}{x} p^x (1-p)^{y-x} \cdot \frac{e^{-\lambda} \cdot \lambda^y}{y!}$$

$$= \sum_{y=0}^{\infty} \frac{y!}{(y-x)!x!} p^x \cdot (1-p)^{y-x} \cdot \frac{e^{-\lambda} \cdot \lambda^y}{y!}$$

for  $y=0, 1, \dots, x$  this is 0

$$= \frac{1}{x!} \left(\frac{p}{1-p}\right)^x \cdot e^{-\lambda} \sum_{y=x}^{\infty} \frac{((1-p)\lambda)^y}{(y-x)!}$$

$$= \frac{1}{x!} p^x \cdot e^{-\lambda} \cdot \lambda^x \sum_{y=x}^{\infty} \frac{((1-p)\lambda)^{y-x}}{(y-x)!}$$

$$= \frac{1}{x!} p^x \cdot e^{-\lambda} \cdot \lambda^x \cdot e^{(1-p)\lambda}$$

$$P(X=x) = \frac{(p\lambda)^x \cdot e^{-p\lambda}}{x!}$$

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

$$\Rightarrow E[X] = p\lambda$$

Theorem 4.4.3 :

if  $X$  and  $Y$  are any two r.v.,  
then

$$E[X] = E[E[X|Y]]$$

Proof:

$f(x,y)$  joint pdf of  $x,y$

$$E[x] = \int \int x f(x,y) dx dy$$

$$= \int \int x f(x|y) f(y) dx dy$$

$$= \int \left[ \int x f(x|y) dx \right] f(y) dy$$

$$= \int E[x|y] f(y) dy$$

$$= E[E[x|y]]$$

$$E_x[x] = E_y[E_{x|y}[x|y]]$$

in the previous problem

$$E_x[x] = E_y \left[ \underbrace{E_{x|y}[x|y]}_{\substack{\text{Binomial}(y,0) \\ \text{Poisson}(\lambda)}} \right]$$

$$= \mathbb{E}_y [yP] = \mathbb{E}_y [y] \cdot P \\ = \lambda P$$

$$\Rightarrow \mathbb{E}_x [x] = \lambda P$$

Mixture distribution = distribution arising from a hierarchical structure.

Definition:

A random variable  $x$  is said to have a mixture distribution if the distribution of  $x$  depends on a quantity that also has a distribution.

Theorem 4.4.7 (conditional variance identity)

for any two r.v.  $x$  and  $y$

$$\text{var}(x) = \mathbb{E} \left[ \text{var}(x|y) \right] \\ + \text{var} \left( \mathbb{E}[x|y] \right)$$

Proof:

$$\text{Var}(X) = \mathbb{E}_x \left[ (X - \mathbb{E}_x[X])^2 \right]$$

$$= \mathbb{E}_x \left[ (X - \mathbb{E}_{x|y}[X|y] + \mathbb{E}_{x|y}[X|y] - \mathbb{E}_x[X])^2 \right]$$

$$= \mathbb{E}_x \left[ (X - \mathbb{E}_{x|y}[X|y])^2 + (\mathbb{E}_{x|y}[X|y] - \mathbb{E}_x[X])^2 + 2(X - \mathbb{E}_{x|y}[X|y])(\mathbb{E}_{x|y}[X|y] - \mathbb{E}_x[X]) \right]$$

= Last term

$$\Rightarrow \mathbb{E} \left[ 2(X - \mathbb{E}_{x|y}[X|y])(\mathbb{E}_{x|y}[X|y] - \mathbb{E}_x[X]) \right]$$

$$\Rightarrow 2 \mathbb{E} \left[ X - \mathbb{E}_{x|y}[X|y] \right] \left[ \mathbb{E}[\mathbb{E}_{x|y}[X|y]] - \mathbb{E}_x[X] \right]$$

$$\Rightarrow 2 \mathbb{E} \left[ X - \mathbb{E}_{x|y}[X|y] \right] \left[ \mathbb{E}[X] - \mathbb{E}[X] \right]$$

$$\Rightarrow 0$$

$$\Rightarrow E_x \left[ (x - E_{x|y}(x|y))^2 + (E_{x|y}(x|y) - E_x(x))^2 \right]$$

$$\Rightarrow E_x \left[ (x - E_{x|y}(x|y))^2 \right]$$

$$+ E_x \left[ (E_{x|y}(x|y) - E_x(x))^2 \right]$$

$$\Rightarrow E \left[ \text{var}(x|y) \right] + \text{var}(E(x|y))$$