

$$Y = \frac{X_1}{X_1 + X_2} \quad \text{where } X_1 \sim \text{Gamma}(\alpha, \lambda), X_2 \sim \text{Gamma}(\beta, \lambda)$$

$$\text{Supp}(Y) = \text{if } X_1 \approx 0, X_2 \approx \infty \Rightarrow Y \approx 0 = [0, 1]$$

$$\text{if } X_1 \approx \infty, X_2 \approx 0 \Rightarrow Y \approx 1$$

$$\text{if de both about the same} \Rightarrow Y \approx \frac{1}{2}$$

It can't be > 1

since $X_1 > X_1 + X_2$

is not possible since $X_2 > 0$.

Verify de independence, change of variables formula, we found

$$f_Y(y) = \int_0^\infty f_{X_1}(y y_2) f_{X_2}(y_2(1-y)) y_2 dy_2$$

$|y_2| = y_2$ since always positive

$$= \int_0^\infty \frac{\lambda^\alpha (y y_2)^{\alpha-1} e^{-\lambda y y_2}}{\Gamma(\alpha)} \frac{\lambda^\beta (y_2(1-y))^{\beta-1} e^{-\lambda y_2(1-y)}}{\Gamma(\beta)} y_2 dy_2$$

$$= \frac{\lambda^{\alpha+\beta} y^{\alpha-1} (1-y)^{\beta-1}}{\Gamma(\alpha) \Gamma(\beta)} \int_0^\infty y_2^{\alpha+\beta-1} e^{-\lambda y y_2 - \lambda y_2(1-y)} dy_2$$

$-\lambda y y_2 - \lambda y_2(1-y)$
 $= -\lambda (y y_2 + y_2(1-y))$
 $= -\lambda y_2 (y + 1 - y)$
 $= -\lambda y_2$

Let $u = \lambda y_2 \Rightarrow \frac{du}{dy_2} = \lambda \Rightarrow dy_2 = \frac{1}{\lambda} du \Rightarrow y_2 = \frac{u}{\lambda}$

$$= \frac{\lambda^{\alpha+\beta} y_1^{\alpha-1} (1-y_1)^{\beta-1}}{\Gamma(\alpha)\Gamma(\beta)} \int_0^{\infty} \underbrace{\left(\frac{y}{\lambda}\right)^{\alpha+\beta-1} e^{-y} \frac{1}{\lambda} dy}_{\frac{y^{\alpha+\beta-1}}{\lambda^{\alpha+\beta-1}} e^{-y} \frac{1}{\lambda}}$$

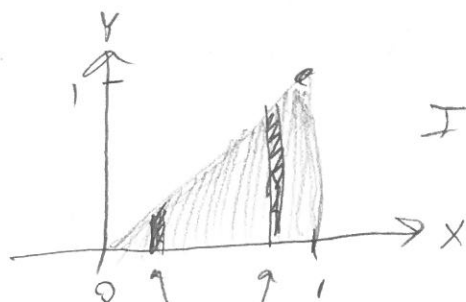
$$\underbrace{\frac{1}{\lambda^{\alpha+\beta}} \int_0^{\infty} y^{\alpha+\beta-1} e^{-y} dy}_{\Gamma(\alpha+\beta)}$$

$$= \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} y_1^{\alpha-1} (1-y_1)^{\beta-1} = \text{Beta}(\alpha, \beta)$$

$$\underbrace{\frac{1}{\Gamma(\alpha+\beta)}}_{\text{Beta}(\alpha, \beta)}$$

p155 cond. density... More practice with conditional densities...

let $X \sim U(0,1)$, $Y|X=x \sim U(0,x)$



Is this uniform completely throughout the Δ ?

No

the mass

has not

to equal since $X \sim U(0,1)$

What does $f_Y(y)$ look like?

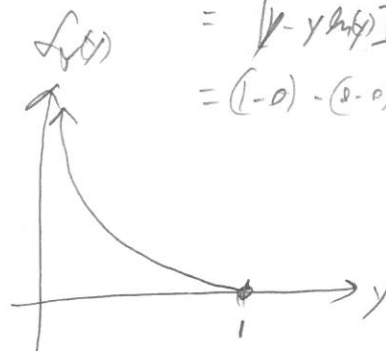
$$\text{Supp}(Y) = [0,1]$$

$$f_Y(y) = \int_{\text{Supp}(X)} f_{X|Y}(x,y) dx = \int_{\text{Supp}(X)} f_{Y|X}(y,x) f_X(x) dx$$

$$= \int_{\mathbb{R}} \frac{1}{x} \mathbb{1}_{\substack{y \in [0,x] \\ 0 \leq y \leq x \\ y \leq x \\ x \geq y}} (1) \mathbb{1}_{x \in [0,1]} dx$$

$$= \int_y^1 \frac{1}{x} dx = \ln(x) \Big|_y^1 = -\ln(y)$$

$$\begin{aligned} \int_0^1 f_Y(y) dy &= 1 = - \int_0^1 \ln(y) dy \\ &= -[y \ln(y) - y]_0^1 \\ &= [y - y \ln(y)]_0^1 \\ &= (1-0) - (0-0) = 1 \checkmark \end{aligned}$$



What is $f_Y(y)$? The marginal density.

How about this. A download either takes on average 10 min with network traffic or on avg. 20 min with network traffic. Network traffic occurs w.p. $\frac{2}{3}$.

$$Y \sim \begin{cases} \text{Exp}(\frac{1}{10}) & \text{w.p. } \frac{1}{3} \\ \text{Exp}(\frac{1}{20}) & \text{w.p. } \frac{2}{3} \end{cases}$$

A sampler may describe this as follows:

$$\text{let } X = \mathbb{1}_{\text{network traffic}} = \text{Bern}\left(\frac{2}{3}\right) = \left(\frac{2}{3}\right)^X \left(\frac{1}{3}\right)^{1-X}$$

$$Y|X \sim \text{Exp}\left(\left(\frac{1}{20}\right)^X \left(\frac{1}{10}\right)^{1-X}\right) = \left(\frac{1}{20}\right)^X \left(\frac{1}{10}\right)^{1-X} e^{-\left(\frac{1}{20}\right)^X \left(\frac{1}{10}\right)^{1-X} y}$$

How long does a download take? Traffic is measured... this...

We want the unconditional prob, $f_Y(y)$. Note: $\text{Supp}(Y) = (0, \infty)$

In general

$$f_Y(y) = \int_{\text{supp}(X)} f_{X,Y}(x,y) dx$$

Here, X is discrete ... so...

$$= \sum_{x \in \text{supp}(X)} f_{X,Y}(x,y) dx$$

$$= \sum_{x \in \text{supp}(X)} f_{Y|X}(y,x) P_X(x)$$

$$= \sum_{x \in \{0,1\}} \left(\frac{1}{20}\right)^x \left(\frac{1}{10}\right)^{1-x} e^{-\frac{1}{20}x} \left(\frac{1}{10}\right)^{1-x} y \left(\frac{2}{3}\right)^x \left(\frac{1}{3}\right)^{1-x}$$

$$= \frac{2}{3} \left(\frac{1}{20} e^{-\frac{1}{20}y} \right) + \frac{1}{3} \left(\frac{1}{10} e^{-\frac{1}{10}y} \right)$$

OR,

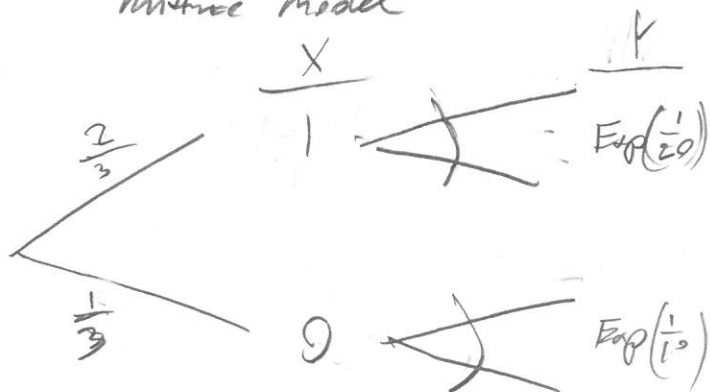
$$= \frac{2}{3} \text{Exp}\left(\frac{1}{20}\right) + \frac{1}{3} \text{Exp}\left(\frac{1}{10}\right)$$

mixture distribution

or

mixture model

tree shows all



if Y is discrete: mixture distr.

Ab

If the download took 25 min, what is the prob there was 4x the traffic?

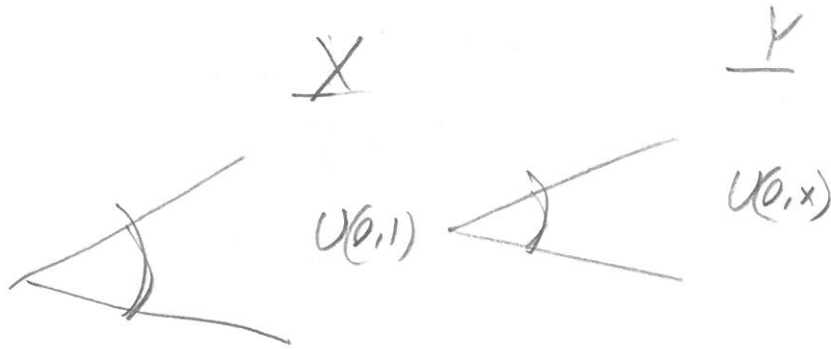
$$P(X=1 | Y=25 \text{ min}) = ? \quad \text{"Flip the tree"}$$

$$P_{X|Y}(x,y) = \frac{f_{X,Y}(x,y)}{f_Y(y)} = \frac{f_{Y|X}(y,x) P_X(x)}{f_Y(y)}$$

$$P_{X|Y}(1,25) = \frac{\frac{1}{20} e^{-\frac{1}{20}(25)} \left(\frac{2}{3}\right)}{\frac{1}{3} \left(\frac{1}{10} e^{-\frac{1}{10}(25)}\right) + \frac{2}{3} \left(\frac{1}{20} e^{-\frac{1}{20}(25)}\right)} \approx 90\%$$

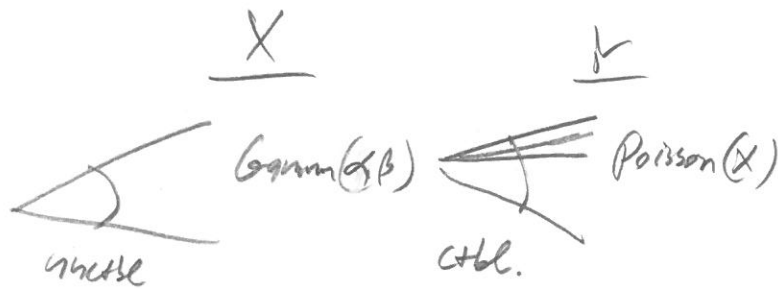
1.21825 29.6826

Tree for first situation



Here, since Y is a mixture of ctbl. may values from X it's called a "Compound distr." Mixture distr. is for at most ctbl. may elements. If X cont. \Rightarrow Compound distr.

ex 9. Accidents $Y \sim \text{Poisson}(\lambda)$ but λ is not the same for all drivers. It is drawn from a $\text{Gamma}(\alpha, \beta)$



$\text{Supp}(Y) = \mathbb{N}$ still...
Since all values of λ are valid

Compound distr. since X is a cont. r.v.

$$\begin{aligned}
 P_Y(y) &= \int_{\text{Supp}(X)} P_{Y|X}(y, x) f_X(x) dx = \int_0^{\infty} \frac{e^{-x} x^y}{y!} \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)} dx \\
 &= \frac{\beta^\alpha}{y! \Gamma(\alpha)} \int_0^{\infty} x^{y+\alpha-1} e^{-(\beta+1)x} dx \\
 &= \frac{\beta^\alpha}{y! \Gamma(\alpha)} \int_0^{\infty} \left(\frac{u}{\beta+1}\right)^{y+\alpha-1} e^{-u} \frac{1}{\beta+1} du \quad \text{let } u = (\beta+1)x \Rightarrow x = \frac{u}{\beta+1} \\
 &\quad \frac{du}{dx} = \beta+1 \Rightarrow dx = \frac{1}{\beta+1} du \\
 &= \frac{\beta^\alpha}{y! \Gamma(\alpha)} \int_0^{\infty} \left(\frac{u}{\beta+1}\right)^{y+\alpha-1} e^{-u} \frac{1}{\beta+1} du = \frac{\beta^\alpha}{y! \Gamma(\alpha) (\beta+1)^{y+\alpha}} \int_0^{\infty} u^{y+\alpha-1} e^{-u} du = \frac{\beta^\alpha}{y! \Gamma(\alpha) (\beta+1)^{y+\alpha}} \Gamma(y+\alpha)
 \end{aligned}$$

$$\text{let } k = \alpha, p = \frac{\beta}{1+\beta} \Rightarrow 1-p = \frac{1}{1+\beta}$$

$$= \frac{\Gamma(y+k)}{\Gamma(k)\Gamma(y-1)} p^k (1-p)^y = \text{ExtNegBin}(p, k)$$

If $k \in \mathbb{N}$

$$= \binom{y+k-1}{k} p^k (1-p)^y = \text{NegBin}(p, k)$$

Poisson - count data

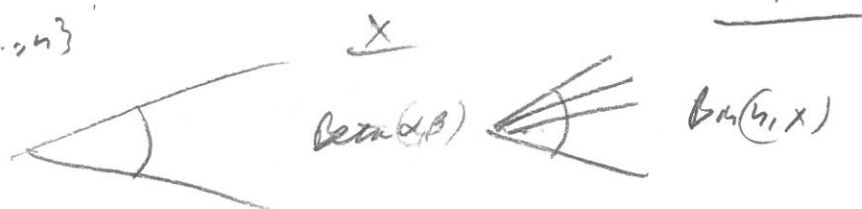
Neg Bin - also count data but non overdispersed

Another Example of this.

$$Y|X \sim \text{Bin}(n, x), \quad X \sim \text{Beta}(\alpha, \beta)$$

$$\{y_p(x)\} = \{0, \dots, n\}$$

still!



compound distr.

$$P_Y(y) = \int_{\{y_p(x)\}} P_{Y|X}(y|x) f_X(x) dx = \int_0^1 \binom{n}{y} x^y (1-x)^{n-y} \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx$$

$$= \frac{\binom{n}{y}}{B(\alpha, \beta)} \int_0^1 x^{y+\alpha-1} (1-x)^{n-y+\beta-1} dx = \frac{\binom{n}{y}}{B(\alpha, \beta)} B(y+\alpha, n-y+\beta) = \text{Beta Binomial}(n, \alpha, \beta)$$

"overdispersed binomial"