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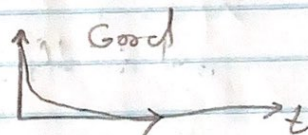
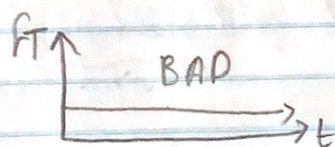
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Mixture and compound distributions

eg.  $\frac{1}{3}$  of the time, you get bad traffic and your download speed are  $T \sim \text{EXP}(1/20)$  i.e.

$E(T) = 20$  and  $\frac{2}{3}$  of the time you have good

Internet traffic and your download speeds are  $T \sim \text{EXP}(1/5)$  i.e.  $E(T) = 5$ . What is the distribution of  $T$  "overall"



Let  $X \sim \text{Bern}(2/3)$ , a RV modeling traffic.

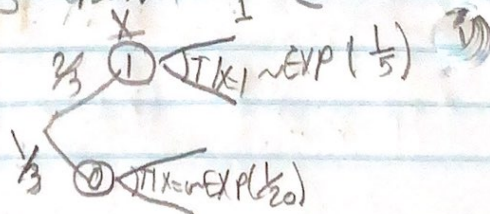
If  $X=1$ , then we have good traffic and if  $X=0$ ,

we have bad traffic. So now we have  $T|X=1 \sim \text{EXP}(1/5)$

and  $T|X=0 \sim \text{EXP}(1/20)$ . Now we essentially use

marginalization to get  $T$  "unconditional"

(meaning overall). First let's draw a tree





Marginalization:

$$h(x) = \int_{\mathcal{R}} h(x, y) dy \quad \text{or} \quad d(x) = \sum_y d(x, y)$$

$$F_T(t) = \sum_{X \in \text{supp}(X)} f_{T|X}(t|X) = \sum_{X \in \text{supp}(X)} f_{T|X}(t|X) P_X(X) = \sum_{X \in \{0,1\}} f_{T|X}(t|X) P_X(X)$$

$$= f_{T|X}(t|0) P_X(0) + f_{T|X}(t|1) P_X(1) = \frac{1}{20} e^{-\frac{1}{20}t} \frac{1}{3} + \frac{1}{5} e^{-\frac{1}{5}t} \frac{2}{3}$$

This was our first "mixture model" where generally  $Y|X$  is the model and  $X$  is the mixing distribution.

If the download took  $t = 25$  s, what is the probability you had bad traffic? Let's find the distribution of traffic

conditional on  $t = 25$  s.

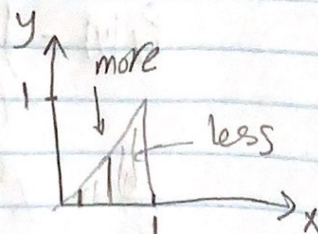
$$P_{X|T}(X, t) = \frac{f_{T|X}(t|X) P_X(X)}{F_T(t)} = \text{Bern}(\underbrace{?}_{X|T \sim})$$

$$P_{X|T}(1|t) = \frac{f_{T|X}(t|1) P_X(1)}{F_T(t)} = \frac{\frac{1}{5} e^{-\frac{1}{5}t} \cdot \frac{2}{3}}{\frac{1}{20} e^{-\frac{1}{20}t} \cdot \frac{1}{3} + \frac{1}{5} e^{-\frac{1}{5}t} \cdot \frac{2}{3}}$$

$$P_{X|T}(0|25) = 1 - P_{X|T}(1|25) = 1 - 0.581 = 0.842.$$

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

$$\begin{array}{cc} X & Y \\ \cancel{X} & \cancel{Y} \\ \cancel{U(0,1)} & \cancel{U(0,1)} \end{array}$$





$$f_y(y) = \int_{\text{supp}(X)} f_{y|x}(y|x) f_x(x) dx$$

pl 56-57, Let  $y|x \sim \text{Poisson}(x)$  and  $x \sim \text{Gamma}(\alpha, \beta)$ ,  $y \sim ?$

$$\begin{array}{cc} x & y \\ \swarrow & \searrow \\ \text{Gamma}(\alpha, \beta) & \text{Poisson}(x) \end{array}$$

$$p_y(y) = \int_{\text{supp}(X)} p_{y|x}(y|x) = \int_{\text{supp}(X)} p_{y|x}(y|x) f_x(x) dx = \int_0^\infty \frac{x^y e^{-x}}{y!} \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} dx \quad 1_{y \in \{0,1,2,\dots\}}$$

$$\frac{\beta^\alpha}{\Gamma(\alpha)} \int_0^\infty x^{\alpha+y-1} e^{-(\beta+1)x} dx = \frac{\beta^\alpha}{\Gamma(\alpha)} \frac{1}{y!} 1_{y \in \mathbb{N}_0} \int_0^\infty x^{y+\alpha-1} e^{-(\beta+1)x} dx \quad \text{Leq.}$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \frac{1}{y!} \frac{\Gamma(y+\alpha)}{(\beta+1)^{y+\alpha}} 1_{y \in \mathbb{N}_0} = \frac{\beta^\alpha}{(\beta+1)^\alpha} \frac{\Gamma(\alpha)}{\Gamma(\alpha+y)} \frac{1}{y!} = \frac{\beta^\alpha}{(\beta+1)^\alpha} \frac{\Gamma(\alpha)}{\Gamma(\alpha+y)} \frac{1}{y!} \quad \text{HW} \quad \propto \text{EX} + \text{Neg Bin}(\alpha, \frac{\beta}{\beta+1})$$

This is a "more flexible" count distribution than the Poisson.

$y|x \sim \text{Bin}(n, x)$ ,  $n$  fixed,  $x \sim \text{Beta}(\alpha, \beta)$ ,  $y \sim ?$

This is analogue to the problem above because binomial

is also a count distribution with a fixed upper bound  $n$ .

$$\begin{aligned} p_y(y) &= \int_{\text{supp}(X)} p_{y|x}(y|x) dx = \int_{\text{supp}(X)} p_{y|x}(y|x) f_x(x) dx = \\ &= \int_0^1 \binom{n}{y} x^y (1-x)^{n-y} \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx \quad 1_{y \in \{0, \dots, n\}} \end{aligned}$$



$$\begin{aligned}
 &= \frac{\binom{n}{y}}{B(\alpha, \beta)} \mathbb{1}_{y \in \{0, \dots, n\}} \int_0^1 x^{y+\alpha-1} (1-x)^{B+n-y-1} dx \\
 &= \frac{B(y+\alpha, B+n-y)}{B(\alpha, \beta)} \binom{n}{y} \mathbb{1}_{y \in \{0, \dots, n\}} = \text{BetaBinomial}(n, \alpha, \beta)
 \end{aligned}$$

H.W.  $Y|X \sim X \sim \text{EXP}(X), X \sim \text{Gamma}(\alpha, \beta) \Rightarrow Y \sim \text{Lomax}(\beta, \alpha)$

Midterm II ↑

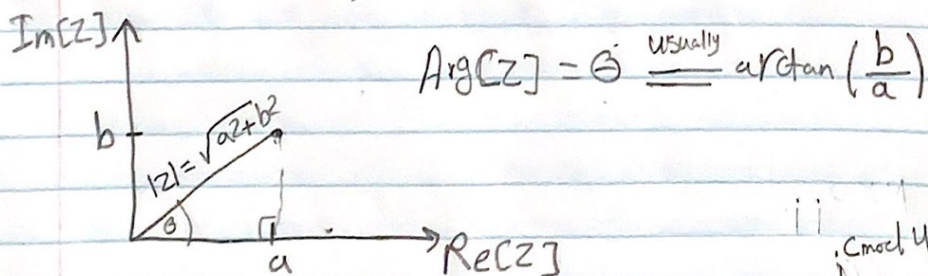
Final ↓

Moment Generating functions (mgf's) and characteristic functions (chf's). We first need to review imaginary #'s and high school trig.

$a, b \in \mathbb{R}, Z := a + bi \in \mathbb{C}$ , the complex #'s,  $i = \sqrt{-1}$

Let  $\text{Re}[Z] := a$ , the "real" component of the imaginary #  $Z$ .

Let  $\text{Im}[Z] := b$ , the "imaginary" component of the imaginary #  $Z$ .



$i^2 = -1, i^3 = i^2 i = -i, i^4 = i^2 i^2 = 1, i^c = i^{\text{clock 4 system}}$

$$e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!} = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \frac{x^4}{4!} + \frac{x^5}{5!} + \dots$$

$$e^{itx} = 1 + itx - \frac{t^2 x^2}{2!} - \frac{it^3 x^3}{3!} + \frac{t^4 x^4}{4!} + \frac{it^5 x^5}{5!} + \dots$$

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots$$

$$\cos(x) = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots$$

$$i \sin(tx) = itx - i \frac{t^3 x^3}{3!} + i \frac{t^5 x^5}{5!} - \dots$$

$$\Rightarrow e^{itx} = i \sin(tx) + \cos(tx) \Rightarrow e^{i\theta} = i \sin(\theta) + \cos(\theta)$$

$$\stackrel{\theta=\pi}{\Rightarrow} e^{i\pi} = -1 \Rightarrow e^{i\pi} + 1 = 0 \quad \text{Euler's formula}$$