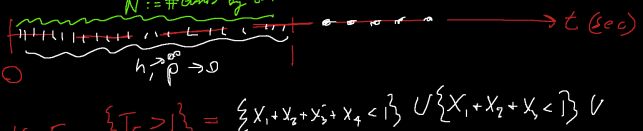


$T_k \sim \text{Erlang}(k, \lambda)$, $N \sim \text{Poisson}(\lambda)$ $X_1, X_2, \dots \stackrel{\text{iid}}{\sim} \text{Exp}(\lambda)$
 $P(T_k > t) = 1 - F_{T_k}(t) = Q(k, \lambda t)$ $F_N(x) = Q(x+1, \lambda)$
 $\Rightarrow 1 - F_{T_k}(t) = F_N(k-1)$ "Poisson Process"

 $k=5 \quad \{T_5 > 1\} = \{X_1 + X_2 + X_3 + X_4 < 1\} \cup \{X_1 + X_2 + X_3 < 1\} \cup \{X_1 + X_2 < 1\} \cup \{X_1 < 1\} \cup \{X_1 > 1\}$
 $1 - F_{T_5}(t) = F_N(4) = \{N=4\} \cup \{N=3\} \cup \{N=2\} \cup \{N=1\} \cup \{N=0\}$
 $P(T_5 > 1) = P(N \leq 4)$

$T \sim \text{Erlang}(k, \lambda) = \frac{\lambda^k e^{-\lambda t} t^{k-1}}{(k-1)!} \mathbb{1}_{t \geq 0} = \frac{\lambda^k e^{-\lambda t} t^{k-1}}{\Gamma(k)} \mathbb{1}_{t \geq 0}$ "Gamma"
 $k \in \mathbb{N}, \lambda \in (0, \infty)$

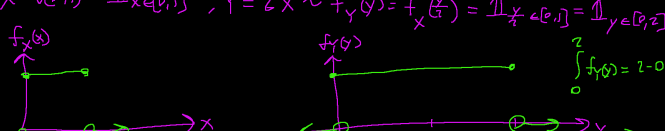
$T \sim \text{NegBin}(k, p) = \binom{k+t-1}{k-1} (1-p)^k p^k \mathbb{1}_{t \in \mathbb{N}_0} = \frac{\Gamma(k+t)}{\Gamma(k) t!} (1-p)^k p^k \mathbb{1}_{t \in \mathbb{N}_0}$
 $k \in \mathbb{N}, p \in (0, 1)$ "Extended Negative Binomial"

What if $k \in (0, \infty)$? Is the top PDF legal and the bottom PMF legal? YES
 $\int_0^\infty \frac{\lambda^k e^{-\lambda t} t^{k-1}}{\Gamma(k)} dt = 1$ and $\sum_{t=0}^\infty \frac{\Gamma(k+t)}{\Gamma(k) t!} (1-p)^k p^k = 1$

which means... these are rv's.
 $X \sim \text{Gamma}(\alpha, \beta) := \frac{\beta^\alpha}{\Gamma(\alpha)} t^{\alpha-1} e^{-\beta t} \mathbb{1}_{t \geq 0}$
 $X \sim \text{ExtNegBin}(k, p)$

Transformations for Discrete rv's
 $X \sim \text{Bern}(p)$, $Y = X+3 \sim \begin{cases} 3 & \text{w.p. } 1-p \\ 4 & \text{w.p. } p \end{cases} = p^{Y-3} (1-p)^{4-(Y-3)} \mathbb{1}_{Y \in \{3, 4\}}$
 $p^x (1-p)^{1-x} \mathbb{1}_{x \in \{0, 1\}}$
 If $Y = g(X) \sim P_Y(y) = P_X(g^{-1}(y))$ using the original PMF?
 $g^{-1}(y) = x$ Is this formula general? No. This is only the formula for g invertible. If g non-invertible...

$X \sim U(\{1, 2, \dots, 10\}) = \frac{1}{10} \mathbb{1}_{X \in \{1, 2, \dots, 10\}}$
 $Y = \min\{X, 3\} \sim \begin{cases} 1 & \text{w.p. } \frac{1}{10} \\ 2 & \text{w.p. } \frac{1}{10} \\ 3 & \text{w.p. } P(X=3) + P(X=4) + \dots + P(X=10) = \frac{8}{10} \end{cases}$
 $P_Y(y) = \sum_{\{x: g(x)=y\}} P_X(x) \stackrel{\text{if } g \text{ invertible for } S_{Y=y}[X]}{=} \sum_{\{x: x=g^{-1}(y)\}} P_X(x) = P_X(g^{-1}(y))$
 $X \sim \text{Bin}(n, p)$, $Y = X^2 \sim P_Y(y) = P_X(g^{-1}(y)) = P_X(\sqrt{y}) = \binom{n}{\sqrt{y}} p^{\sqrt{y}} (1-p)^{n-\sqrt{y}}$
 \Downarrow
 $x = \sqrt{y} = g^{-1}(y)$ $\mathbb{1}_{\sqrt{y} \in \{0, 1, \dots, n\}}$

Transformations for Continuous rv's $Y = g(X)$, X is continuous
 for invertible g , $f_Y(y) \stackrel{?}{=} f_X(g^{-1}(y))$ incorrect!
 $X \sim U(0, 1) = \mathbb{1}_{x \in [0, 1]}$, $Y = 2X \sim f_Y(y) = f_X(\frac{y}{2}) = \mathbb{1}_{\frac{y}{2} \in [0, 1]} = \mathbb{1}_{y \in [0, 2]}$


where'd we go wrong? The above formula was derived using probabilities. Densities are not probabilities! But... CDF's are probabilities! Strategy: let's derive the CDF of Y using the CDF of X. And then, like when we did convolutions, take the derivative to get the density for Y.

$F_Y(y) = P(Y \leq y) = P(g(X) \leq y) \stackrel{\text{if } g' > 0}{=} P(X \leq g^{-1}(y)) = F_X(g^{-1}(y))$
 $f_Y(y) = \frac{d}{dy} [F_Y(y)] = \frac{d}{dy} [F_X(g^{-1}(y))] = F'_X(g^{-1}(y)) \frac{d}{dy} [g^{-1}(y)]$
 $= f_X(g^{-1}(y)) \frac{d}{dy} [g^{-1}(y)]$
 $= f_X(g^{-1}(y)) \left| \frac{d}{dy} [g^{-1}(y)] \right|$
 $g' > 0 \Rightarrow g^{-1} > 0$
 if $g' < 0$
 $\Rightarrow P(X \geq g^{-1}(y)) = 1 - F_X(g^{-1}(y))$
 $f_Y(y) = \frac{d}{dy} [] = -\frac{d}{dy} [F_X(g^{-1}(y))] = f_X(g^{-1}(y)) \left(-\frac{d}{dy} [g^{-1}(y)] \right)$
 $\frac{d}{dy} [g^{-1}(y)] < 0$
 $\Rightarrow f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} [g^{-1}(y)] \right|$ for all g invertible

Let's derive some more rules! The most common invertible function is.... the straight line! $Y = aX + c \Rightarrow x = g^{-1}(y) = \frac{y-c}{a}$, $\left| \frac{d}{dy} [g^{-1}(y)] \right| = \frac{1}{|a|}$
 s.t. $a, c \in \mathbb{R}$

$f_Y(y) = f_X\left(\frac{y-c}{a}\right) \frac{1}{|a|}$ "shift and scale"
 if $c=0$ just a scale.... $Y = aX$
 $f_Y(y) = f_X\left(\frac{y}{a}\right) \frac{1}{|a|}$

if $a=1$ just a shift $Y = X + c$
 $f_Y(y) = f_X(y-c)$

$X \sim \text{Exp}(\lambda) = \lambda e^{-\lambda x} \mathbb{1}_{x \geq 0}$
 $Y = X + c = \lambda e^{-\lambda(y-c)} \mathbb{1}_{y \geq c}$

$X \sim \text{Exp}(1) = e^{-x} \mathbb{1}_{x \geq 0}$
 $Y = g(X) = -\ln\left(\frac{e^{-X}}{1-e^{-X}}\right) = \ln\left(\frac{1-e^{-X}}{e^{-X}}\right) = \ln(e^X - 1) = y$

$\Rightarrow e^y = e^x - 1 \Rightarrow e^y + 1 = e^x \Rightarrow x = \ln(e^y + 1) = g^{-1}(y)$

$\left| \frac{d}{dy} [g^{-1}(y)] \right| = \left| \frac{e^y}{e^y + 1} \right| = \frac{e^y}{e^y + 1}$ $e^{\ln(e^y+1)} = e^y + 1$ $y \in \mathbb{R}$ $e^y \geq 0$

$f_Y(y) = f_X(\ln(e^y + 1)) \frac{e^y}{e^y + 1} = e^{-\ln(e^y+1)} \frac{e^y}{e^y + 1} \stackrel{\text{if } \ln(e^y+1) \geq 0}{=} \frac{e^y}{e^y + 1}$

$= \frac{1}{e^y + 1} \frac{e^y}{e^y + 1} = \frac{e^y}{(e^y + 1)^2} = \text{Logistic}(0, 1)$

$\frac{e^{-y}}{(e^{-y} + 1)^2}$ 