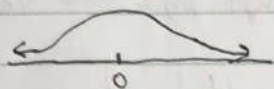


Wednesday October 14 2020

Lecture 1.1



$$X \sim \text{logistic}(0,1) = \frac{e^x}{(1+e^x)^2} \approx N(0,1) \text{ both with sticker tails}$$

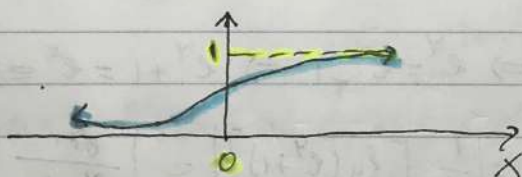
$$E[X] = 0, SD[X] = \frac{\pi}{\sqrt{3}} \approx 1.8 > 1$$

under the shift and scale $\text{sigma} > 0$

$$Y = \mu + \sigma X \sim f_Y(y) = f_X\left(\frac{y-\mu}{\sigma}\right) \frac{1}{|\sigma|} = \frac{e^{\frac{y-\mu}{\sigma}}}{\sigma(1+e^{\frac{y-\mu}{\sigma}})^2} = \text{Logistic}(\mu, \sigma)$$

Why is this called the "logistic distribution"? there's a famous function called the "logistic function". It has three parameters: L (maximum value), k (steepness), μ (center) and it is:

$$l(x) := \frac{L}{1+e^{-k(x-\mu)}} \stackrel{L=1, k=1, \mu=0}{=} \frac{1}{1+e^{-x}} \cdot \frac{e^x}{e^x} = \frac{e^x}{e^x+1} \text{ (standard logistic function).}$$



$$X \sim \text{logistic}(0,1)$$

$$F_X(x) = \int_{-\infty}^x f_X(t) dt = \int_{-\infty}^x \frac{e^t}{(1+e^t)^2} dt = \int_{-\infty}^x \frac{1+e^x}{u^2} \frac{du}{1+e^x} = [-u^{-1}]_1^{1+e^x} = 1 - \frac{1}{1+e^x} = \frac{e^x}{1+e^x}$$

$$\text{Let } u = 1+e^t \Rightarrow e^t = u-1 \Rightarrow \frac{du}{dt} = e^t \Rightarrow dt = \frac{e^{-t}}{1+e^t} du \Rightarrow t=-\infty \Rightarrow u=1, t=x \Rightarrow u=1+e^x$$

The "quantile" q or "percentile" $100q$ for r.v. X is defined as the minimum x s.t. $q \leq P(X \leq x) = F(x) \Leftrightarrow F(x) \geq q$. It is denoted $Q[X, q]$ where Q is the "quantile operator" (Not the upper incomplete regularized gamma function).

When $q=0.5$, the quantile has a special name, the "median", $\text{Med}[X] := Q[X, q]$. Here's an example:

$$X \sim U(\{2, 4, 6, \dots, 20\}) = \frac{1}{10} \mathbb{1}_{x \in \dots}$$

x	$p(x)$	$F(x)$
2	0.1	0.1
4	0.1	0.2
6		0.3
8		0.4
10		0.5
12		0.6
14		0.7
16		0.8
18		0.9
20		1

$$Q[X, 30\%] = 6$$

$$\text{med}[X] = 10$$

$$Q[X, 80\%] = 16$$

$$Q[X, 85\%] = 18 = Q[X, 0.9]$$

However, if X is a continuous r.v. with "continuous support" e.g. $[0, 10]$, $[0, \text{infinity})$, all real numbers, etc and not something like $[0, 1]$ union $[2, 3]$. In the latter case, $F(x)$ is flat between $[1, 2]$ which means it's not invertible. In the former case, $F(x)$ is invertible.

$Q[X, q] = F_x^{-1}(q)$, and the inverse CDF is called appropriately, the "quantile function".

$$X \sim \text{Exp}(\lambda) := \lambda e^{-\lambda x} \mathbb{1}_{x \geq 0} \Rightarrow F_X(x) = 1 - e^{-\lambda x} = q \Rightarrow 1 - q = e^{-\lambda x}$$

$$\Rightarrow \ln(1 - q) = -\lambda x \Rightarrow x = -\frac{1}{\lambda} \ln(1 - q) = \frac{1}{\lambda} \ln\left(\frac{1}{1 - q}\right) = F_X^{-1}(q)$$

$$\text{Med}[X] = \frac{\ln(2)}{\lambda} = F_X^{-1}(0.5)$$

Quantile functions are not usually available in closed form since CDF's aren't even usually available in closed form e.g.

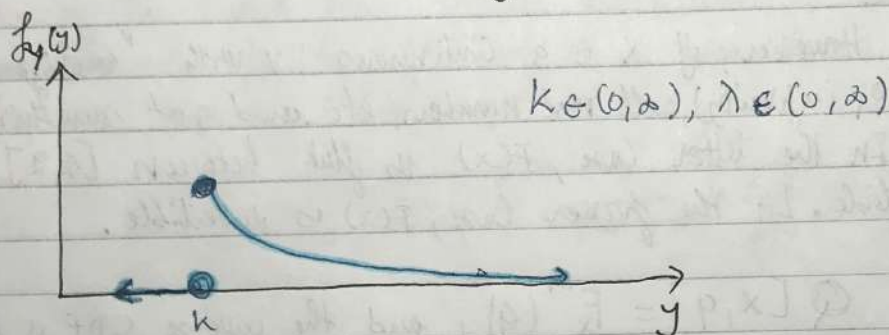
$$X \sim \text{Erlang}(k, \lambda) \Rightarrow F_X(x) = P(k, \lambda x)$$

$\text{Med}[X] = x$ s.t. $P(k, \lambda x) = 0.5$. Need a Computer Solver.

Let $X \sim \text{exp}(\lambda) = \lambda e^{-\lambda x} \mathbb{1}_{x \geq 0}$, $Y = g(X) = ke^X \sim f_Y(Y) = ?$

$$Y = ke^X \Rightarrow \frac{Y}{k} = e^X \Rightarrow X = \ln\left(\frac{Y}{k}\right) = \ln(Y) - \ln(k) = g^{-1}(Y)$$

$$\begin{aligned} \left| \frac{d}{dy} [g^{-1}(y)] \right| &= \left| \frac{1}{y} \right| = \frac{1}{|y|} & \ln\left(\frac{y}{k}\right)^{-\lambda} & \quad y \in [k, \infty) \\ & & \ln(y) \in [\ln(k), \infty) & \\ f_Y(y) &= f_X(\ln(\frac{y}{k})) \frac{1}{|y|} = \frac{\lambda}{|y|} e^{-\lambda \ln(\frac{y}{k})} \mathbb{1}_{\ln(y) - \ln(k) \in [0, \infty)} \\ &= \frac{\lambda}{y} \left(\frac{y}{k}\right)^{-\lambda} \mathbb{1}_{y \in [k, \infty)} = \text{Pareto I}(k, \lambda) \end{aligned}$$



$$F_Y(y) = \int_k^y \frac{\lambda}{t^{\lambda+1}} dt = \frac{\lambda}{k^\lambda} \left[-\frac{1}{\lambda t^\lambda} \right]_k^y = k^\lambda \left(\frac{1}{k^\lambda} - \frac{1}{y^\lambda} \right) = 1 - \left(\frac{k}{y} \right)^\lambda$$

$$\Rightarrow F_Y^{-1}(q) = k (1-q)^{-\frac{1}{\lambda}} \quad \text{"typically test question"}$$

This distribution was discovered by **Vilfredo Pareto**, an Italian economist in 1896 when he observed that 20% of the richest Italians owned 80% of the land (ie the wealth). This is known as the "Pareto Principle" and it corresponds to the Pareto (1, 1.61) distribution.

Further, the Pareto distribution is a waiting time / survival time model. It's used for [see wikipedia if you're interested]. Wealth, music talent, number of patents, ...

$X, Y \stackrel{iid}{\sim} \text{Exp}(1) = e^{-x} \mathbb{1}_{x \in [0, \infty)}$, $D = X - Y = X + \widetilde{(-Y)} \sim f_D(d) = ?$

$$f_D(d) = \int_{\text{supp}[x]} f_x^{\text{old}}(x) f_z^{\text{old}}(d-x) \mathbb{1}_{d-x \in \text{supp}(z)} dx = \int_0^\infty e^{-x} e^{d-x} \mathbb{1}_{d-x \in (-\infty, 0]} dx$$

$x-d \in [0, \infty)$
 $x \in [d, \infty)$

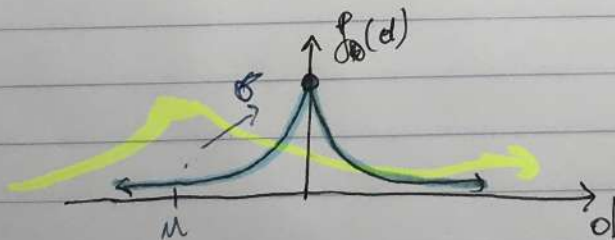
$$= e^d \int_0^\infty e^{-2x} \mathbb{1}_{x \in [d, \infty)} dx = e^d \begin{cases} \int_d^\infty e^{-2x} dx & \text{if } d \geq 0 \\ \int_0^\infty e^{-2x} dx & \text{if } d < 0 \end{cases}$$

$$= e^d \begin{cases} \left[-\frac{1}{2} e^{-2x} \right]_d^\infty & \text{if } d \geq 0 \\ \left[-\frac{1}{2} e^{-2x} \right]_0^\infty & \text{if } d < 0 \end{cases} = \frac{1}{2} e^d \begin{cases} \left[-e^{-2x} \right]_d^\infty & \text{if } d \geq 0 \\ \left[-e^{-2x} \right]_0^\infty & \text{if } d < 0 \end{cases}$$

$$= \frac{1}{2} e^d \begin{cases} e^{-2d} & \text{if } d \geq 0 \\ 1 & \text{if } d < 0 \end{cases} = \frac{1}{2} \begin{cases} e^{-d} & \text{if } d \geq 0 \\ e^d & \text{if } d < 0 \end{cases}$$

$$= \frac{1}{2} e^{-|d|}$$

↑ Laplace(0, 1)
std Laplace distribution
AKA "double exponential"



$$X = \mu + \sigma D \sim \text{Laplace}(\mu, \sigma) := \frac{1}{2\sigma} e^{-\frac{|x-\mu|}{\sigma}} \quad \sigma > 0$$