

$$= k \lambda^k y^{k-1} e^{-(\lambda y)^k} \mathbb{1}_{y \geq 0} = k \lambda (\lambda y)^{k-1} e^{-(\lambda y)^k} \mathbb{1}_{y \geq 0} \text{Weib}(\lambda, k)$$

Note Weibull(1, λ) = (1) λ (1) y^{0} , $\mathbb{1}_{y \geq 0} = \lambda e^{-\lambda y} \mathbb{1}_{y \geq 0} = \text{Exp}(\lambda)$

k is really cool... this is the main property:

$$k=1 \quad P(Y \geq y+c | Y \geq c) = P(Y \geq y). \quad \text{eg } Y=3, c=14$$

Memorylessness.

$$P(Y \geq 17 | Y \geq 14) = P(Y \geq 3)$$

$k > 1$ $P(Y \geq y+c | Y \geq c) < P(Y \geq y)$ \Rightarrow survival less likely as time goes on

$k < 1$ $P(Y \geq y+c | Y \geq c) > P(Y \geq y)$ \Rightarrow survival more likely as time goes on.

prove this in Hw.

order statistics (p160 in the book)

Let X_1, X_2, \dots, X_n be a collection of continuous rvs and let

$X_{(1)}, X_{(2)}, \dots, X_{(n)}$ be their "order statistics" defined as:

$$X_{(1)} = \min \{X_1, \dots, X_n\}$$

$$X_{(n)} = \max \{X_1, \dots, X_n\}$$

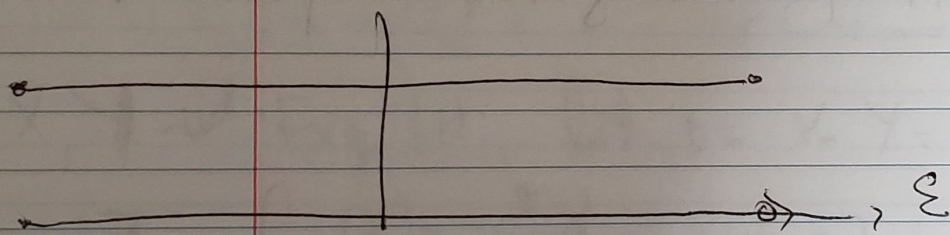
$$X_{(k)} = k^{\text{th}} \text{ Largest } \{X_1, \dots, X_n\}$$

$$R = X_{(n)} - X_{(j)} \text{ "Range"}$$

We want to find the Cdf and pdf of the order statistics. we'll start by looking at the Cdf of

Monday October 19th 2016

1774: First Law of errors. Imagine you're trying to measure something a constant quantity x , but your measurements have random error, epsilon, so your measurement M is a random variable looking like: $M = x + \text{epsilon}$. So what is a good model for the error (epsilon)? It makes sense for $E(\text{epsilon}) = 0$ and symmetric. How about: $\sim N(-1, 1)$



It also makes sense for larger errors (in magnitude) to be less probable than smaller errors. $\Rightarrow \forall \epsilon > 0, f'(\epsilon) < 0$
 $\forall \epsilon > 0, f''(\epsilon) = f'(\epsilon) \Rightarrow$ solve $f(\epsilon) = C e^{-\frac{1}{2}\epsilon^2} \Rightarrow \text{Gaussian (2D)}$

$X \sim \text{Exp}(1) = e^{-x} \mathbb{1}_{x \geq 0}$. Let $Y = \frac{1}{\lambda} X^{\frac{1}{k}} = g(X)$ for $\lambda, k > 0$
 $Y \sim f_Y(Y) = ?$ Inverse function $\lambda Y = X^{\frac{1}{k}} \Leftrightarrow X = (\lambda Y)^k \Leftrightarrow Y = g^{-1}(X)$

$$\left| \frac{d}{dy} [g^{-1}(y)] \right| = \left| \frac{d}{dy} [\lambda^k y^k] \right| = |k \lambda^k y^{k-1}| = k \lambda^k y^{k-1}$$

$$f_Y(y) = f_X(g^{-1}(y)) \left| \frac{d}{dy} [g^{-1}(y)] \right| = e^{-(\lambda y)^k} \lambda^k y^{k-1} \cdot k \lambda^k y^{k-1}$$