

## Meeting 12

Laplace first published this in 1774 calling it the "first law of errors"

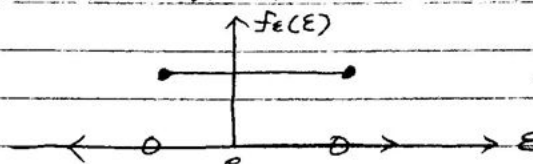
His context was measurement. When you measure a quantity  $V$ , you measure it with error, epsilon, so that your measurement is:

$$M = V + \text{epsilon}$$

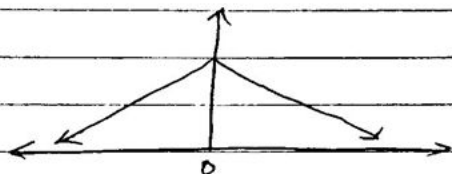
What makes a good distribution for the error, epsilon?

The expectation should be zero and should be symmetric.

How about



→ This is not very good.  
It should have the property that the probability of error should decrease with its magnitude.



Also, why should it stop at some maximum magnitude?

Another good property is that the density should be decreasing in magnitude of error

Laplace assumed for all positive errors that  $f_{\epsilon}''(\epsilon) = f_{\epsilon}'(\epsilon)$

$$\Rightarrow f_{\epsilon}(\epsilon) = ce^{-d\epsilon} \Rightarrow \epsilon \sim \text{Laplace}(0, 1)$$

$$X \sim \text{Exp}(1) = e^{-x} \mathbb{I}_{x \geq 0}, \quad Y = g(X) = \frac{1}{\lambda} X^{\frac{1}{K}} \quad \text{with } K, \lambda > 0$$

$$Y = \frac{1}{\lambda} X^{\frac{1}{K}} \Rightarrow \lambda Y = X^{\frac{1}{K}} \Rightarrow X = (\lambda Y)^K = \lambda^K Y^K = g^{-1}(Y)$$

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

$$f_Y(y) = f_X(g^{-1}(x)) \left| \frac{d}{dy} [g^{-1}(y)] \right| = e^{-(\lambda y)^K} \mathbb{I}_{\lambda^K y^K \geq 0} \cdot K \lambda^K y^{K-1}$$

$$= K \lambda^K y^{K-1} e^{-(\lambda y)^K} \mathbb{I}_{y \geq 0}$$

$$= K \lambda (\lambda y)^{K-1} e^{-(\lambda y)^K} \mathbb{I}_{y \geq 0} = \text{Weibull}(K, \lambda) \quad \underline{1951}$$

This is a very famous waiting time / survival rv model and is used e.g. in insurance companies to price life insurance

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

Property of Weibull r.v's under different values of  $k$ :

$$k=1, P(Y \geq y+c | Y \geq c) = P(Y \geq y)$$

this equality is called "memorylessness"

$$k > 1, P(Y \geq y+c | Y \geq c) < P(Y \geq y) \quad \text{e.g. } \left[ \begin{array}{l} \text{old lifespan of human,} \\ \text{waiting for bus} \end{array} \right.$$

$$k < 1, P(Y \geq y+c | Y \geq c) > P(Y \geq y) \quad \text{e.g. startup company lifespan}$$

Order Statistics (p.160) Let  $X_1, X_2, \dots, X_n$  be a collection of continuous r.v.'s.

Let the "order statistics" be the r.v.'s:

$$\left\{ \begin{array}{l} X_{(1)} := \min \{X_1, X_2, \dots, X_n\} \\ \vdots \\ X_{(k)} := k^{\text{th}} \text{ largest of } X_1, \dots, X_n \\ \vdots \\ X_{(n)} := \max \{X_1, \dots, X_n\} \\ R := X_{(n)} - X_{(1)} \text{ range} \end{array} \right. \quad \begin{array}{l} \text{e.g. } X_1=9, X_2=2, X_3=12, X_4=7 \\ X_{(1)}=2, X_{(2)}=7, X_{(3)}=9, \\ X_{(4)}=12 \\ r=12-2=10 \end{array}$$

We want to find both the CDF and PDF of the  $k$ th order statistic.

We'll build this up in stages. The first thing we'll do is find the CDF and PDF of the maximum.

$$\begin{aligned} F_{X_{(n)}}(x) &= P(X_{(n)} \leq x) = P(X_1 \leq x, X_2 \leq x, \dots, X_n \leq x) \\ &\stackrel{\text{iid}}{=} P(X_1 \leq x) \dots P(X_n \leq x) \\ &= \prod_{i=1}^n F_{X_i}(x) \stackrel{\text{iid}}{=} [F_X(x)]^n = F_X(x)^n \\ f_{X_{(n)}}(x) &\stackrel{\text{iid}}{=} \frac{d}{dx} [F_X(x)]^n = \frac{d}{dx} [F_X(x)^n] = n f_X(x) F_X(x)^{n-1} \end{aligned}$$

Next thing we'll do is to find the CDF and PDF of the minimum

$$\begin{aligned} F_{X_{(1)}}(x) &= P(X_{(1)} \leq x) = 1 - P(X_{(1)} > x) \\ &= 1 - P(X_1 > x, X_2 > x, \dots, X_n > x) \\ &\stackrel{\text{iid}}{=} 1 - \prod_{i=1}^n (1 - F_{X_i}(x)) \stackrel{\text{iid}}{=} 1 - (1 - F_X(x))^n \\ f_{X_{(1)}}(x) &\stackrel{\text{iid}}{=} \frac{d}{dx} [1 - (1 - F_X(x))^n] = n f_X(x) (1 - F_X(x))^{n-1} \end{aligned}$$

Next we'll do is assume  $n=10$  and derive the  $k=4$ th order statistic's CDF and PDF. First, let's find the prob. that the first four #'s

are less than  $x$  and the last 6 numbers are greater than  $x$ .

$$= P(X_1 \leq x, \dots, X_4 \leq x, X_5 > x, \dots, X_{10} > x)$$

$$\stackrel{\text{iid}}{=} \prod_{i=1}^4 F_{X_i}(x) \prod_{i=5}^{10} (1 - F_{X_i}(x)) \stackrel{\text{iid}}{=} F_X(x)^4 (1 - F_X(x))^6$$

Let's find the probability any 4 of the 10 are below  $x$  and the remaining are below  $x$ . Let  $S$  be a subset of size 4 of the index set

$$\begin{aligned} \{1, 2, \dots, 10\} &= \sum_{\text{all } S} P(X_{S_1} \leq x, \dots, X_{S_4} \leq x, X_{S_5} > x, \dots, \\ &X_{S_6} > x) \stackrel{\text{iid}}{=} \sum_{\text{all } S} \prod_{i=1}^4 F_{X_{S_i}} \prod_{i=1}^6 (1 - F_{X_{S_i}}(x)) \\ &\stackrel{\text{iid}}{=} \sum_{\text{all } S} F(x)^4 (1 - F(x))^6 \\ &= \binom{10}{4} F(x)^4 (1 - F(x))^6 \end{aligned}$$

Now let's derive the CDF for the  $k=4^{\text{th}}$  order statistic.

$$F_{X_{(4)}}(x) = P(\underbrace{X_{(4)} \leq x}_{\text{event}}) = P(\text{a subset of 4 } X_i\text{'s} \leq x \text{ and the remaining 6 are } > x)$$

$$+ P(\text{a subset of 5 } X_i\text{'s} \leq x \text{ and the remaining 5 are } > x)$$

$$+ \dots + P(\text{all 10 } X_i\text{'s are less than/equal to } x)$$

$$\stackrel{\text{iid}}{=} \sum_{j=4}^{10} \binom{10}{j} F(x)^j (1 - F(x))^{10-j}$$

For iid continuous rv's  $X_1, \dots, X_n$ , the CDF and PDF for the  $k^{\text{th}}$  order stat. is:  $F_{X_{(k)}}(x) = \sum_{j=k}^n \binom{n}{j} F(x)^j (1 - F(x))^{n-j}$

$$\begin{aligned} f_{X_{(k)}}(x) &= \frac{d}{dx} \left[ \sum_{j=k}^n \binom{n}{j} F(x)^j (1 - F(x))^{n-j} \right] \\ &= \sum_{j=k}^n \binom{n}{j} \underbrace{\frac{d}{dx} F(x)^j}_{u'} \underbrace{(1 - F(x))^{n-j}}_v \end{aligned}$$

$$u' = j f(x) F(x)^{j-1}, \quad v' = -(n-j) f(x) (1 - F(x))^{n-j-1}$$

$$\frac{d}{dx} [uv] = uv' + u'v.$$