1774 - Pirst "law of errors". Imagine you're living to measure something, a constant quantity veriality, but your measurements have random error, E, so your measurement M is a rv looking like: M= V+ E. So what is a good model for the error (E)?

It makes sense for E[z]=0.

Med[z]=0 & symmetric.

1000

It also makes sense for larger errors (in magnitude) to be less probable than smaller errors. => \$\fomalle{t}_2 > 0 \, f'(\varepsilon) <0

 $V_{\varepsilon} > 0$ $f''(\varepsilon) = f'(\varepsilon) = 0$ $f(\varepsilon) = 0$ $f(\varepsilon) = 0$ $f''(\varepsilon) = 0$ Laplace $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ Laplace $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ Laplace $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ Laplace $f''(\varepsilon) = 0$ $f''(\varepsilon) = 0$ Laplace $f''(\varepsilon) =$

 $X \sim Exp(1) = e^{-x} I_{x>0}$. Let $Y = \frac{1}{x} X^{k} s.t. \lambda, k > 0$

 $\gamma \sim f_{\gamma}(y) = ?$ Inverse function $\lambda y = x^{k}$ => $x = \lambda^{k} y^{k} = g^{-1}(y)$

|dy[g'(y)] = |dy[x"y"] = |Kx"y"-1 = |Kx"y"-1 | = |Kxy"y"-1 |

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

= K > Y Y = - (29) 1 1 430

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= k \( (\lambda y) \) e (\lambda y) \( \lambda \) = Waiball (k, \( \lambda \))
Note Weiball (1, 2) = (1) 2 (1) 4 (1) 4 e (24) 1 430.
                     = > e - (xy) 1430
k is really cool _. this is the main property:
 e.g. Y=3, C=14 { P(Y>17.17 > 14) = P(Y>3)
                                         memorylessness
 K=1 P(Y>y+c | Y>c) = P(Y>y)
K>1 P(Y>Y+C | Y>C) < P(Y>y) - likely as time
kel P(Y>Y+C |Y>C) > P(Y=y) goes on.

kely os time
                                    goes on.
You will prove these facts on the HW.
Order statistics (p160 in the text book)
Let X,, Xon _ _ _ , Xn be a collection of continuous ru's and let the "order statistics"
the ru's: X(1), X(2), ---, X(n) defined as: X(1) := min. [x, x2, ---, xn]
Xiki := Kth largest of X1, --, Xn,
X (n) := max [ X, X2, .... Xn]
R := Xin - Xin Range.
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 $X_{(1)} = \min \{ X_{1}, \dots, X_{n} \}$ $X_{1} = 9, X_{2} = 2, X_{3} = 12, X_{n} = 7$ X(m) = max { X, - . . Xn} $\chi_{(1)} = 2$ $\chi_{(2)} = 7$, $\chi_{(3)} = 9$, $\chi_{(4)} = 12$ r=12-2=10. We want to find CDF and PDF of the kth order statistics. We'll start by looking of of the maximum $F_{X(n)}(x) = P(X_{(n)} \leq x) = P(X_1 \leq x \land X_2 \leq x \land \dots \land X_n \leq x)$ if indep $=\frac{n}{1}$ $P(x_i \leq x) = \frac{n}{11} F_{x_i(x)}$ $= \frac{d}{dx} \left[f_{\times (n)} \right] = \frac{d}{dx} \left[f_{\times (n)} \right] =$ $= n f_{X}(x) f_{X}(x)$ Let's now find the CDF $P_{X(i)}(x) = P(X(i) \leq x) = 1 - P(X(i) > x)$ = 1 - P(x, >x & x, >x & ... & xn >x) if indep = 1- P(x, >x) P(x, >x)....(p(xn>x) $= 1 - \frac{n}{n} \left(1 - f_{X_i(n)}\right)$

 $= n \int_{X} (x) \left(1 - \int_{X} (x) \right)^{n-1}$ Let's now find the CDF/PDF for the kth order statistic, XIK) Let's le let n=10, k=4. X, Ecd ... & X, Ecd X, & cd ... A X, o > c if indep in $p(X_i \le c)$ if $p(X > c) = \frac{4}{11} \int_{X_i(c)} \frac{10}{11} (1 - \int_{X_i(c)} \frac{10}{11}$ if iid Fx (c) (1- Fx(c)) any H Xis x & the other b xis >x $X_{s_i} \leq x_{s_i}, X_{s_i} \leq x_{s_i}, X_{s_i} > x_{s_i}$ over all subsets S 512e 4, 5°. S size 4, ST Size b

Size b

if f(d) f(x) f(xif indep

$$= \binom{10}{h} f_{X}(x)^{h} (1 - f_{X}(x))^{h}$$

$$F_{X_{(h_1)}}(x) = P(X_{(h_1)} \leq x) = P(h_1 \times l_1' \leq x, h_2 \times l_2' > x) + \dots$$

$$P(f_1 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_2 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_3 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_4 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_4 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_4 \times l_1' \leq x, f_2 \times l_2 > x) + \dots$$

$$P(f_4 \times l_1' \leq x, f_2 \times l_2 \times l_2 > x) + \dots$$

$$P(f_4 \times l_1' \leq x, f_2 \times l_2 \times$$

$$f_{X(x)} = \frac{d}{dx} \left[f_{X(x)} \right] = \frac{d}{dx} \left[f_{X(x)} \right] + \frac{d}{$$