## **Summary Sheet for Continuous Distributions**

## ECE 695/CS 590 –Fault-Tolerant Computer System Design School of Electrical and Computer Engineering Purdue University Spring 2024

Notation	Interpretation	Relations
$F_X(x)$	Pr(Random variable $X$ has a value $\leq x$ ) Called Cumulative Distribution Function, or simply, Distribution Function (cdf)	
$f_X(x)$	Probability density function (pdf) $\int_{-\infty}^{\infty} f(x) = 1$ $f(x) \ge 0, \text{ for all } x$	$F_X(x) = \int_{-\infty}^x f_X(t)dt$ $Pr(a < X \le b) = \int_a^b f_X(t)dt$ $= F_X(b) - F_X(a)$
R(t)	Reliability Pr(No failure occurs in time $(0, t)$ )	$R_X(t) = 1 - F_X(t)$ Typically $R(0) = 1$ and $\lim_{t\to 0} R(t) = 0$
h(t)	Instantaneous failure rate $\frac{1}{R(t)} \lim_{x \to 0} \frac{R(t+x) - R(t)}{x}$ Pr(Failure in $(t0, t1)$  System has survived till $t0$ ) $\neq \int_{t0}^{t1} h(x) dx$ It is $= 1 - \frac{R(t1)}{R(t0)}$	$h(t) = \frac{f(t)}{R(t)}$ $R(t) = \exp[-\int_0^t h(x)dx]$ If $h(t)$ as $t$ , this represents an Increasing Failure Rate (IFR). If $h(t)$ as $t$ , this represents a Decreasing Failure Rate (DFR). If $h(t)$ stays constant as $t$ , this represents a Constant Failure Rate (CFR).

Distribution	Characteristic Measures	Interpretation
Exponential distribution	$F(t) = 1 - e^{-\lambda t}$ $h(t) = \lambda$	If failures arrive with a constant rate $\lambda$ , then the time to failure follows the Exponential distribution.
		Exponential distribution represents the CFR part of a system's lifetime.
Hypoexponential distribution	parameters $\lambda_1$ and $\lambda_2$ ( $\lambda_1 \neq \lambda_2$ ) $F(t) = 1 - (\lambda_2/(\lambda_2 - \lambda_1))e^{-\lambda_1 t} + $ $(\lambda_1/(\lambda_2 - \lambda_1))e^{-\lambda_2 t}$ sequential phases, and to each is independent and exponentially distributed. It represents the IFR parameters are supported by the sequence of the sequenc	If a process passes through several sequential phases, and time spent in each is independent and exponentially distributed.
		It represents the IFR part of a systems lifetime, going from 0 to

		$\min\{\lambda_I, \lambda_2, 1\}.$
Erlang distribution	$F(t) = 1 - \sum_{k=0}^{r-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!}$ Parameters: $r, \lambda$	Process passes through $r$ sequential phases, each of which has an identical exponential distribution. Exponential is a special case of Erlang, with $r = 1$
		If a system can survive up to <i>r</i> -1 shocks and fails upon the arrival of the <i>r</i> -th shock, and the shocks arrive with a Poisson distribution, then time to failure follows Erlang.
Gamma distribution	$f(t) = \frac{\lambda^{\alpha} t^{\alpha-1} e^{-\lambda t}}{\Gamma(\alpha)}, \ \alpha > 0$ $\Gamma \text{ is the gamma function, defined as}$	If <i>r</i> in the Erlang distribution can take non integer values, it gives the Gamma distribution.
	$\Gamma(\alpha) = \int_0^\infty x^{\alpha - 1} e^{-x} dx$ Parameters: $\lambda$ , $\alpha$	For $0 < \alpha < 1$ , Gamma distribution is DFR; for $\alpha > 1$ , it is IFR; for $\alpha = 1$ , it is CFR.
Hyperexponential distribution	$F(t) = \sum_{i=1}^{k} \alpha_i (1 - e^{-\lambda it})$ where $\sum_{i=1}^{k} \alpha_i = 1$ Parameters: $\lambda_i$ , $\alpha_i$ , $k$	If a process faces <i>k</i> parallel phases and it will go through only one of these, then the time follows a Hyperexponential distribution.
		It is a DFR from $\sum_{i=1}^{k} \alpha_i \lambda_i$ to min $\{\lambda_1, \lambda_2,\}$ .
Weibull distribution	$F(t) = 1 - e^{-\lambda t} \alpha$ $h(t) = \lambda \alpha t^{\alpha - 1}$	Most widely used parametric family of distributions.
	Parameters: $\alpha$ (shape parameter), $\lambda$ (scale parameter)	$\alpha = 1 \Rightarrow CFR$ $\alpha < 1 \Rightarrow DFR$ $\alpha > 1 \Rightarrow IFR$