

# ECE 510 Lecture 3

## Functions

Reliability Functions, T&T 2.1-6, 9  
Distributions, T&T 3.1-4, 4.1-4, 5.1-3

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# Reliability Functions

# Reliability Functions

- Functions of time
  - CDF(x)  $\rightarrow$  F(t)
- Survival function  $S(t) = 1 - F(t)$
- PDF(x)  $\rightarrow$  f(t)

$$f(t) = \frac{\text{fraction of ORIGINAL population that fails in } dt}{dt}$$

$$= \frac{dF(t)}{dt} = -\frac{dS(t)}{dt}$$

- Hazard function  $h(t)$

$$h(t) = \frac{\text{fraction of CURRENT population that fails in } dt}{dt}$$

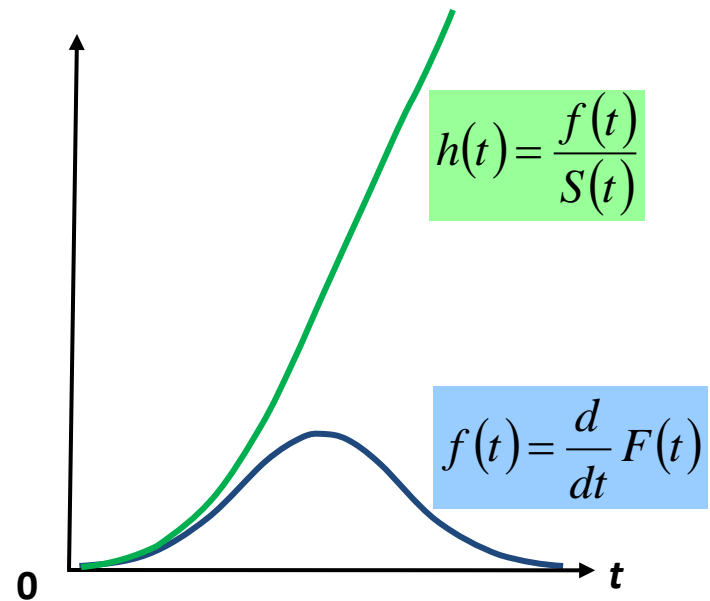
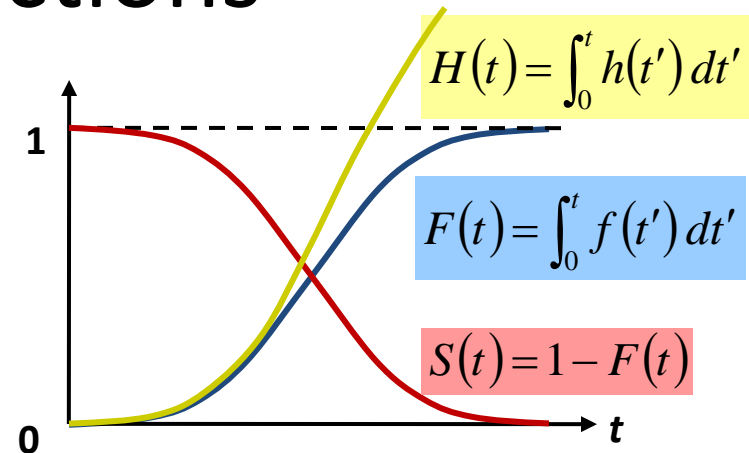
$$= \frac{f(t)}{S(t)} = -\frac{dS(t)}{dt} \frac{1}{S(t)} = -\frac{d \ln S(t)}{dt}$$

- Cum hazard function  $H(t)$

$$H(t) = \int_0^t h(t) dt$$

$$S(t) = \exp[-H(t)]$$

$$F(t) = 1 - \exp[-H(t)]$$



# Exercise 3.1a

- Calculate  $H(t)$ ,  $S(t)$ , and  $F(t)$  for the given human mortality data, and plot  $h(t)$ ,  $S(t)$ , and  $F(t)$ . The data is given as  $h(t)$  for each age, that is, the probability of a living person dying at the given age. Use a sum to approximate the integral for  $H(t)$ .

# Exercise 3.1a Solution, Part 1

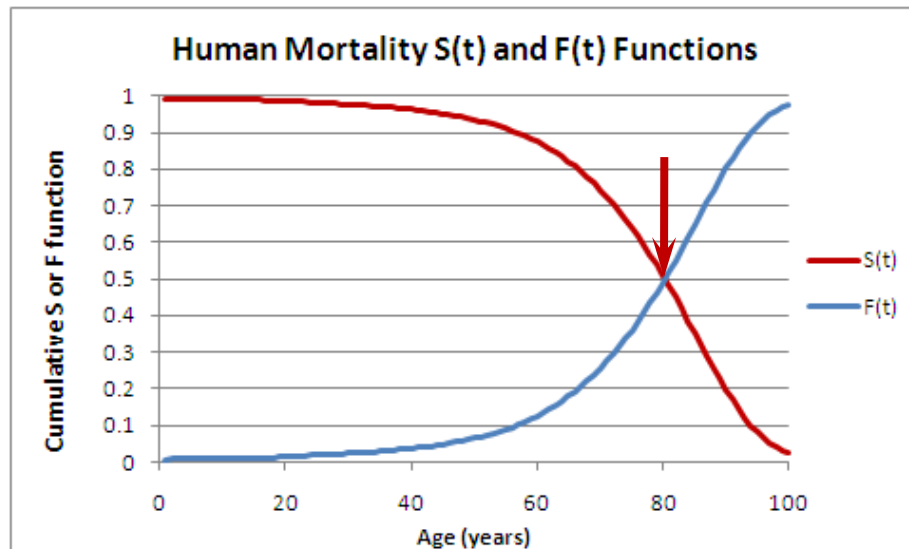
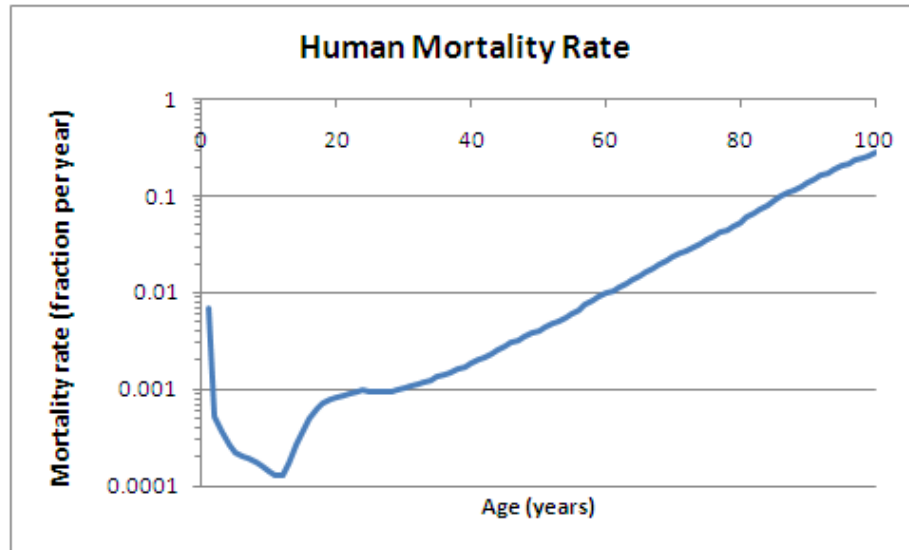
OFFSET					
=SUM(C\$6:C10)					
A	B	C	D	E	F
1	<b>Exercise 3 – Hazard Function for Human Mor</b>				
2	Calculate H, S, and F. (For H, use a sum to approximate t				
3					
		Mortality rate (hazard function)	Cumulative hazard function	Cumulative survival function	Cumulative fail function
4	Age	h(t)	H(t)	S(t)	F(t)
5	1	0.00706	0.00706	0.9929649	0.0070351
6	2	0.00053	0.00759	0.9924387	0.0075613
7	3	0.00036	0.00795	0.9920815	0.0079185
8	4	0.00027	0.00822	0.9918137	0.0081863
9	5	0.00022	C\$6:C10)	0.9915955	0.0084045
10	6	0.00018	0.0084	0.9914175	0.0085825

$$H(t) = \int_0^t h(t) dt$$

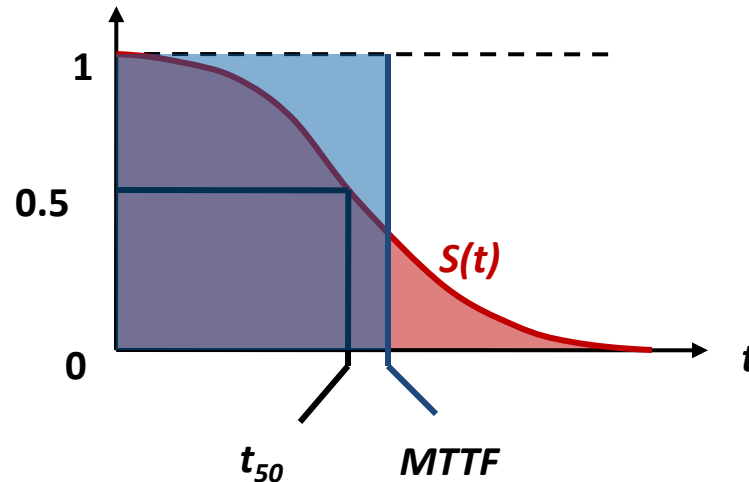
$$S(t) = \exp[-H(t)]$$

$$F(t) = 1 - \exp[-H(t)]$$

# Human Mortality Graphs



# Reliability Indicators



- Mean time to failure (MTTF)

$$MTTF = \int_0^{\infty} t f(t) dt = \frac{1}{N} \sum_{j=1}^N t_N = \int_0^{\infty} S(t) dt$$

- Median time to failure ( $t_{50}$ ) is the solution of

$$S(t_{50}) = 0.5$$

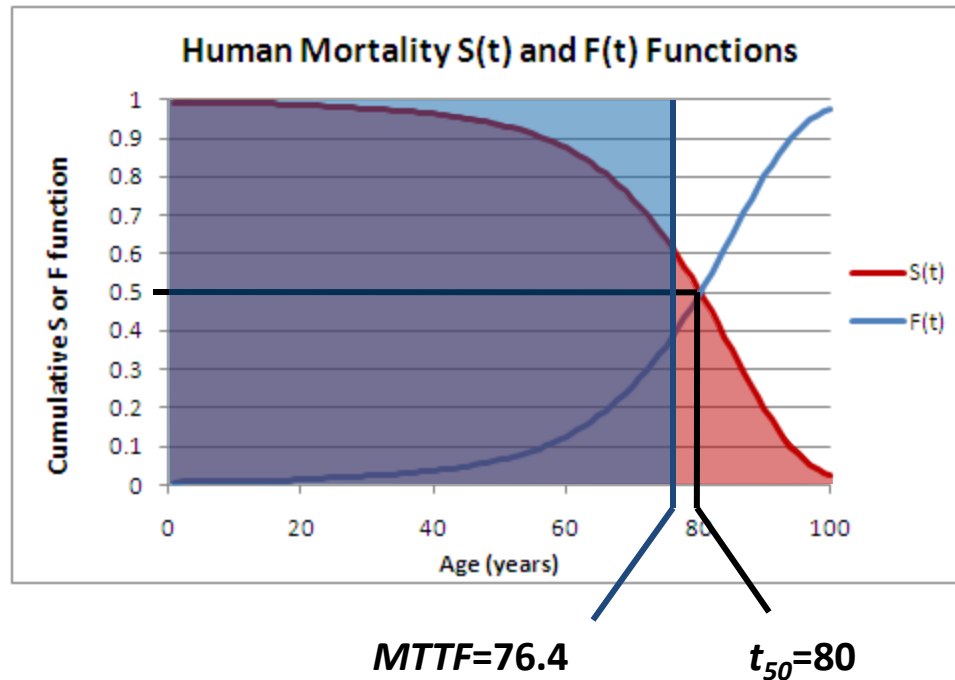
- Time at which half of the initial population fails

# Exercise 3.1b

- Find the mean and median times to failure for the human mortality data set from the last exercise



# Exercise 3.1b Solution



- Sum  $S(t)$  to get MTTF

# Reliability Measures: DPM

- Metric designed for low fail rates
- DPM = Defects Per Million

% pass	% fail	DPM
99	1	10,000
99.9	0.1	1000
99.95	0.05	500
99.99	0.01	100
99.999	0.001	10

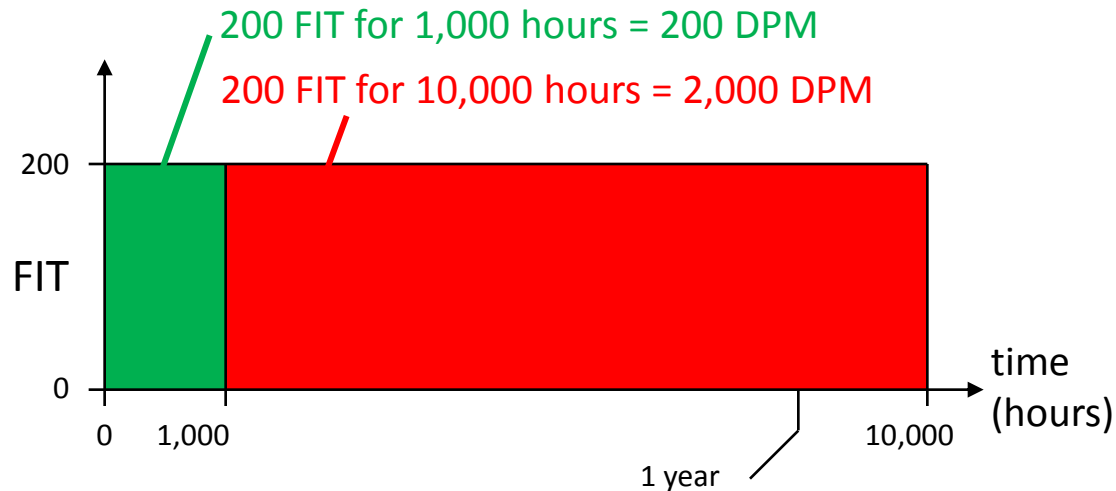
Typical target at end of life

Typical target at t=0

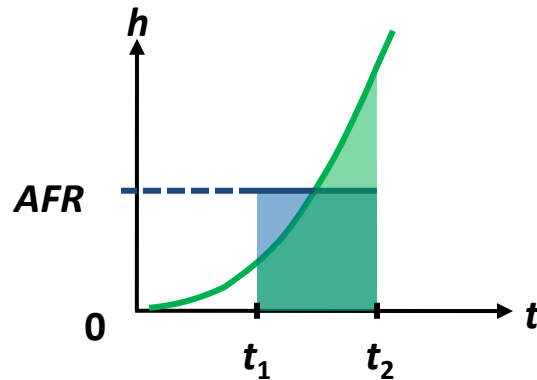
Typical range for semiconductor reliability

# Reliability Measures: FIT

- FIT = Failures In Time
- FIT is a fail *rate*, fails per billion device hours
  - FIT = DPM per 1,000 hours
- DPM is a fail total, fails per million total devices
  - DPM = FIT \* hours / 1,000



# Reliability Indicators: AFR



- AFR, Average Fail Rate

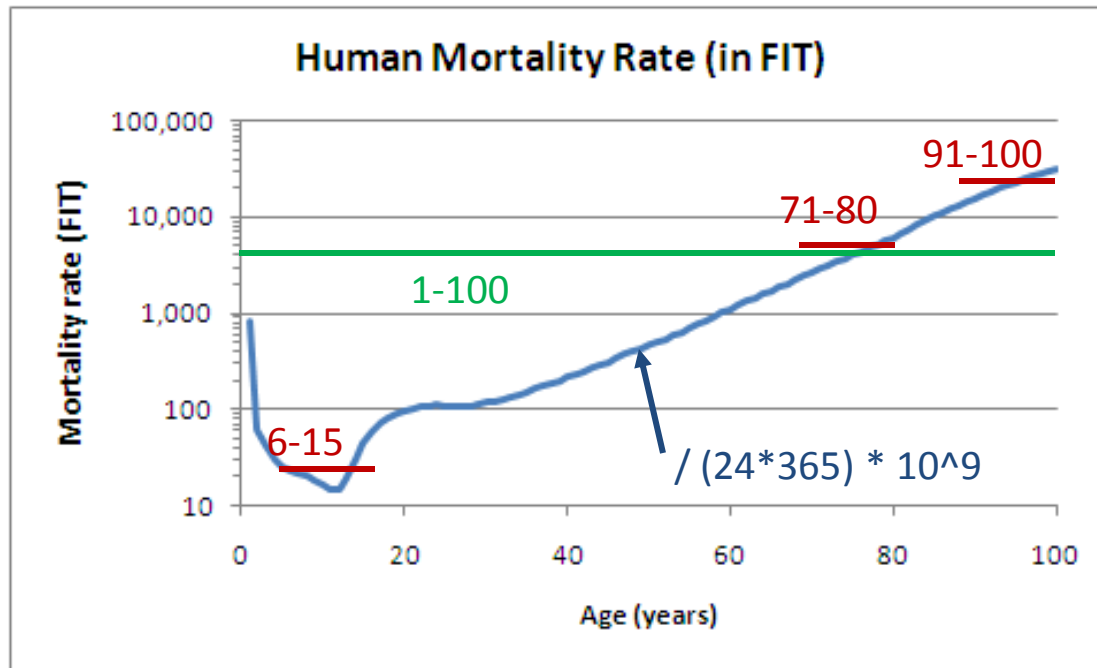
$$AFR(t_1, t_2) = \frac{\int_{t_1}^{t_2} h(t) dt}{t_2 - t_1} = \frac{H(t_2) - H(t_1)}{t_2 - t_1} = \frac{\ln S(t_1) - \ln S(t_2)}{t_2 - t_1}$$

- If  $t$  in hours, units are fail fraction per hour
- Multiply by  $10^9$  for units of FIT

# Exercise 3.1c

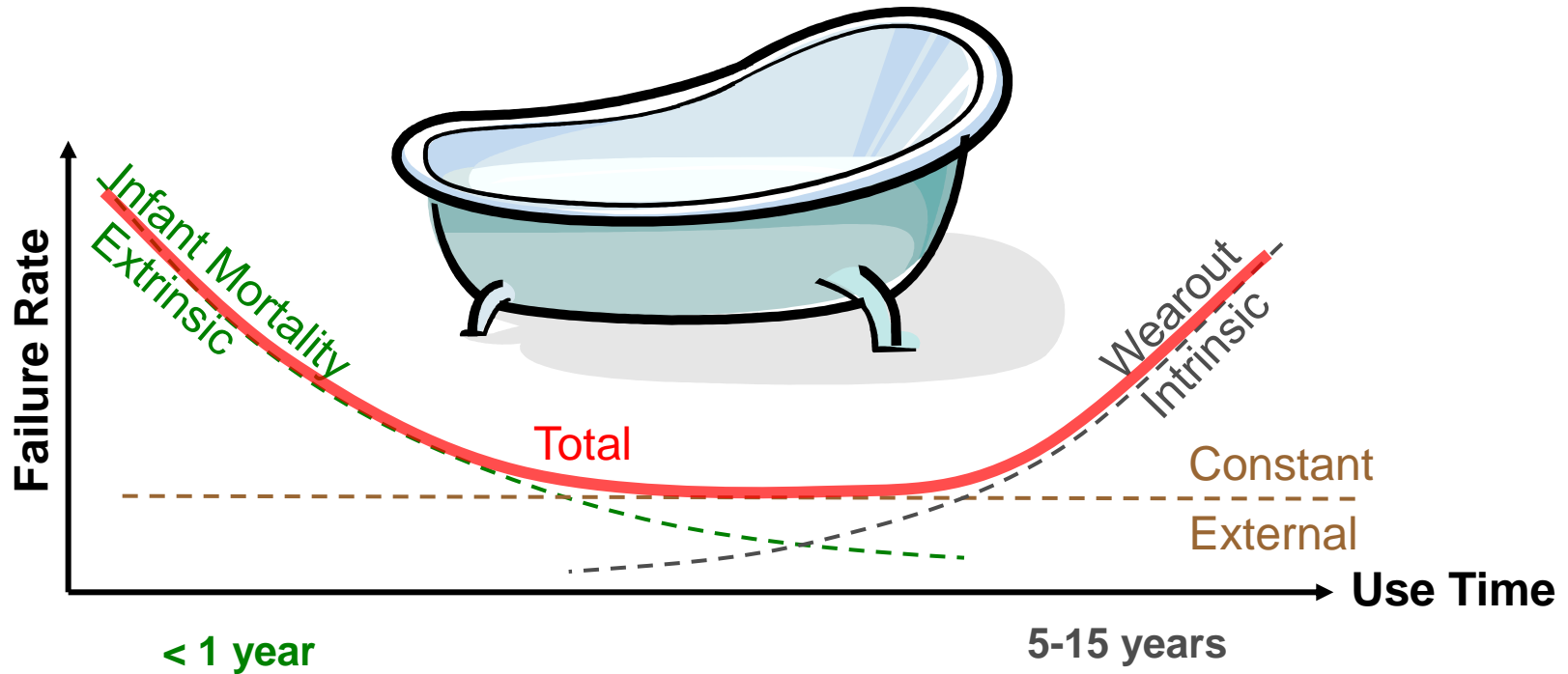
1. Plot the hazard function in FIT
2. Find the AFR (in FIT) for:
  - The 10-year range from ages 6 to 15
  - The 10-year range from ages 71 to 80
  - The 10-year range from ages 91 to 100
  - The entire 100-year range from ages 1 to 100

# Exercise 3.1c Solution



Age Range	AFR (FIT)
6-15	22
71-80	4,311
91-100	24,116
1-100	4,270

# The Bathtub Curve

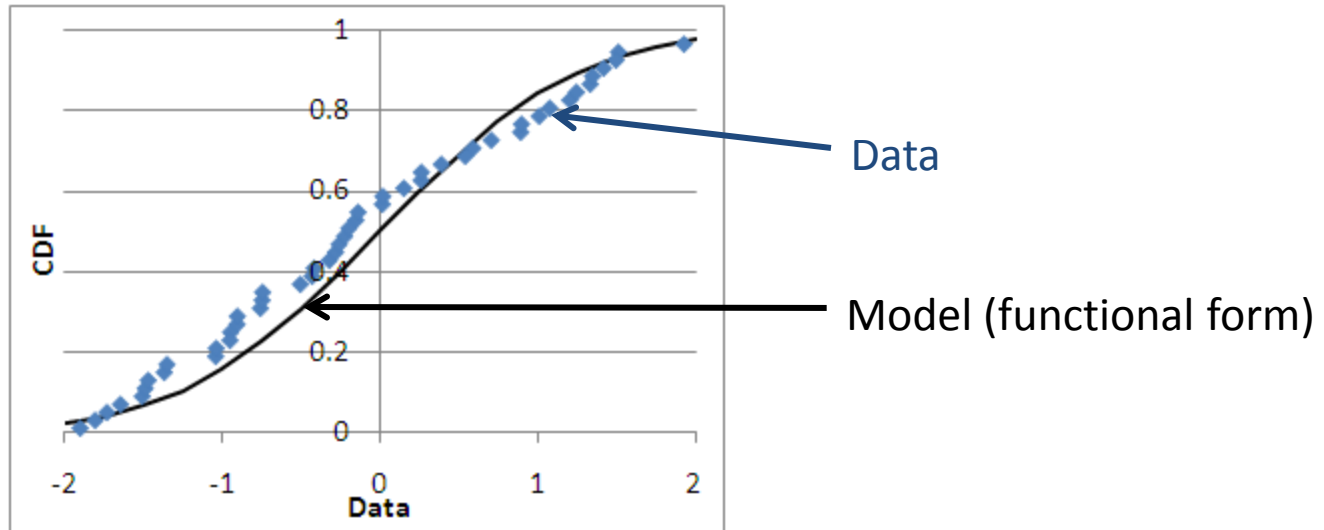


- Infant Mortality (IM) from latent reliability defects
- Wearout from reliability mechs like oxide wearout
- Constant from external effects like radiation
- Many versions of this graph – it is a very important concept

# Functional Forms



# Reliability Functional Forms

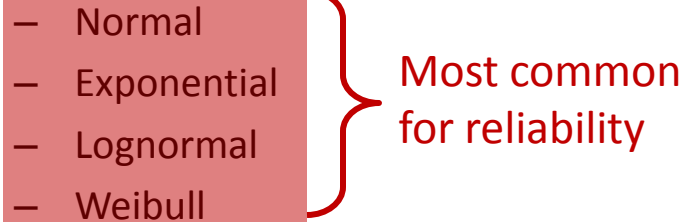


- Choose functional form for model to fit data

# A Function Bestiary

– *Bestiary: A medieval collection of stories providing physical and allegorical descriptions of real or imaginary animals*

- Continuous distributions

- Normal
  - Exponential
  - Lognormal
  - Weibull
- 
- Most common  
for reliability

- Gamma

- Beta

- Discrete distributions

- Hypergeometric

- Binomial

- Poisson

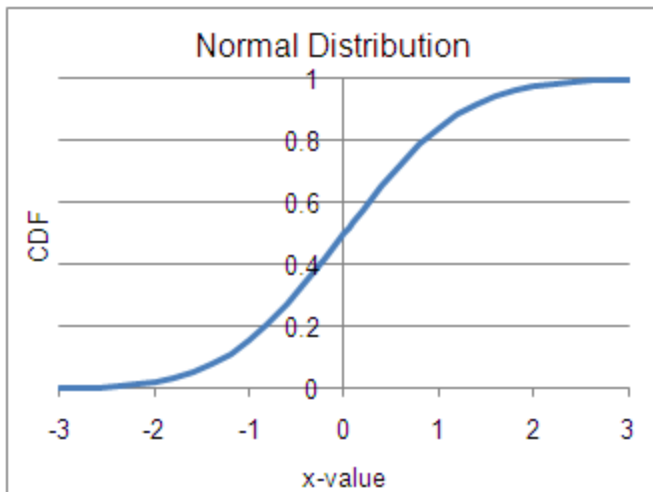
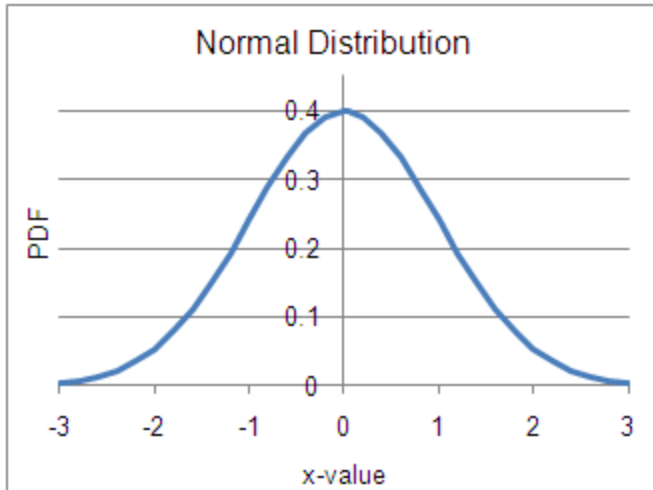
- Statistical distributions

- Chi-square

- Student's t

- F

# Normal Distribution



$\mu$  = mean  
 $\sigma$  = standard deviation

$\sigma^2$  = variance

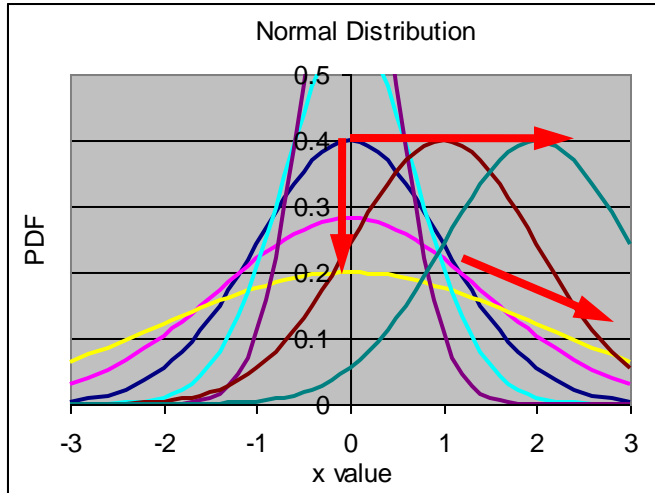
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left[\frac{x-\mu}{\sigma}\right]^2} \leftarrow e^{-x^2}$$

$$F(x) = \int_{-\infty}^x dx' \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left[\frac{x'-\mu}{\sigma}\right]^2}$$

rand normal =  $NORMSINV(CDF)$   
 where CDF is rand uniform

- Using Excel:
  - PDF =  $NORMDIST(x, \mu, \sigma, FALSE)$
  - CDF =  $NORMDIST(x, \mu, \sigma, TRUE)$
- Plot using:
  - y-axis = probit =  $NORMSINV(CDF)$
  - x-axis = x
  - $\sigma = 1/\text{slope}$
  - $\mu = \text{x-intercept} = -(\text{y-intercept}) / \text{slope}$

# Normal Distribution



mean	0	0	0	0	0	1	2
std	1	1.41	2	0.71	0.5	1	1

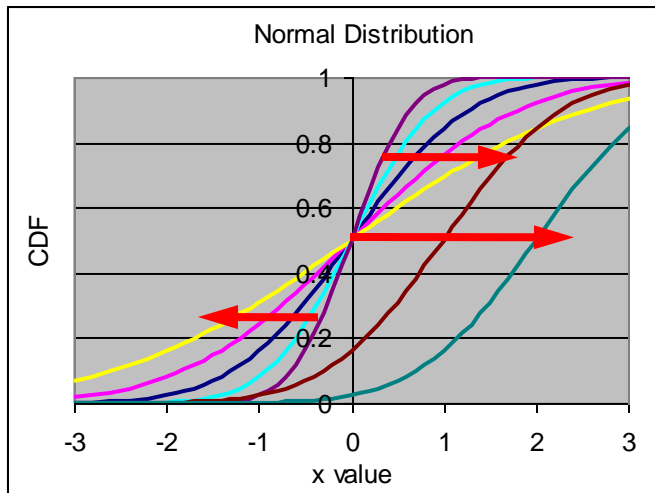
$\mu$  = mean  
 $\sigma$  = standard deviation

$\sigma^2$  = variance

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left[\frac{x-\mu}{\sigma}\right]^2}$$

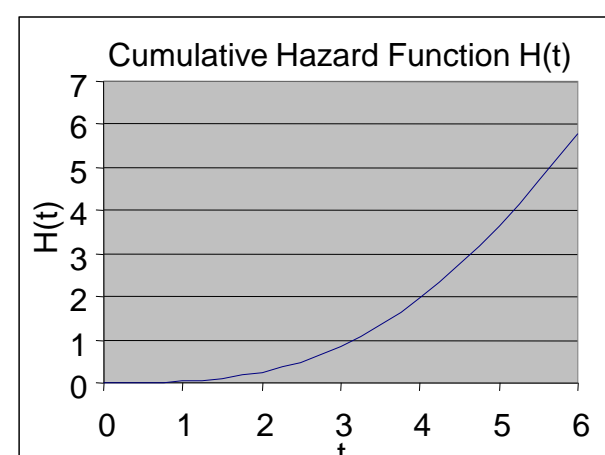
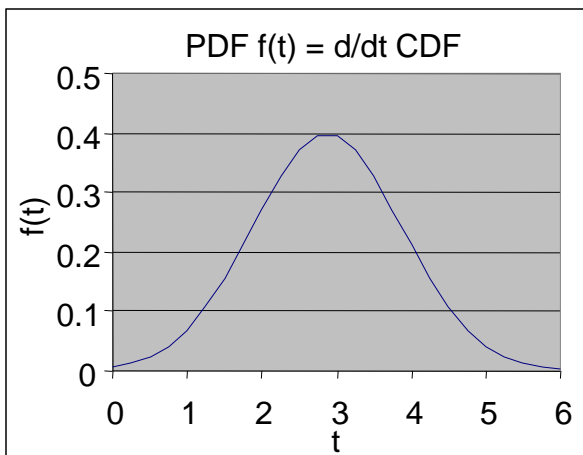
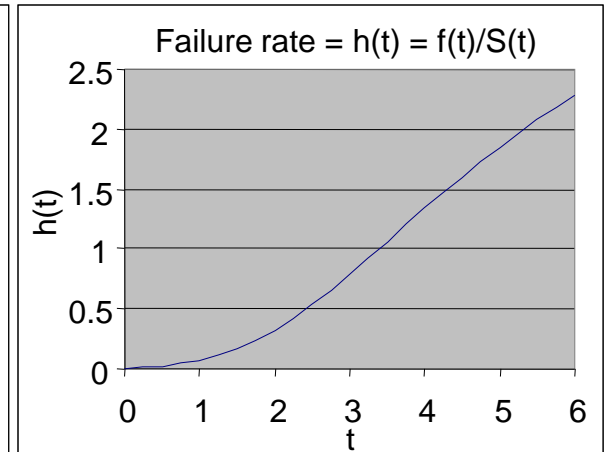
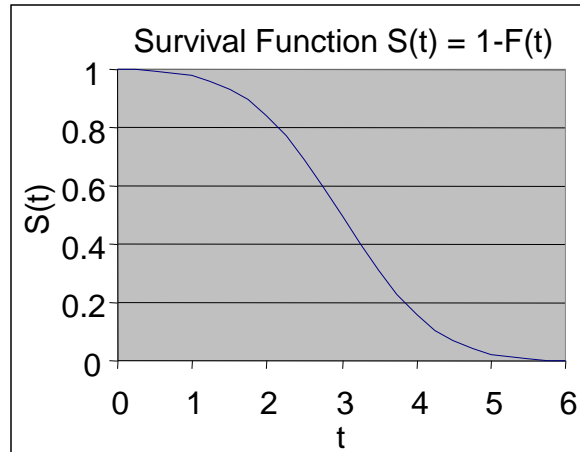
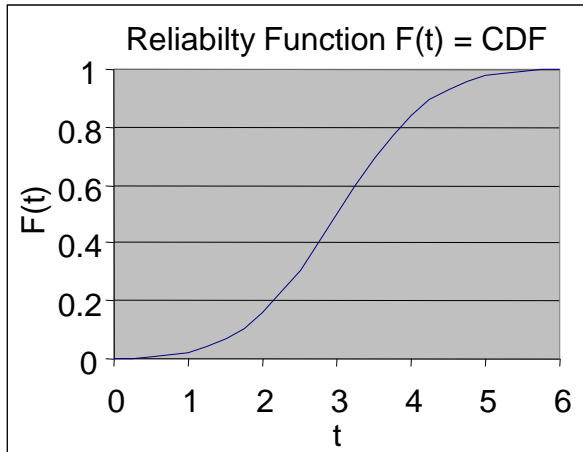
$$F(x) = \int_{-\infty}^x dx' \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left[\frac{x'-\mu}{\sigma}\right]^2}$$

rand normal =  $NORMSINV(CDF)$   
 where CDF is rand uniform

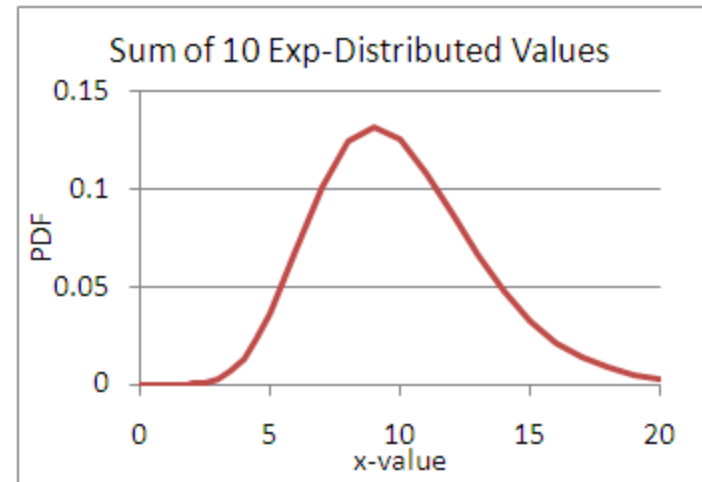
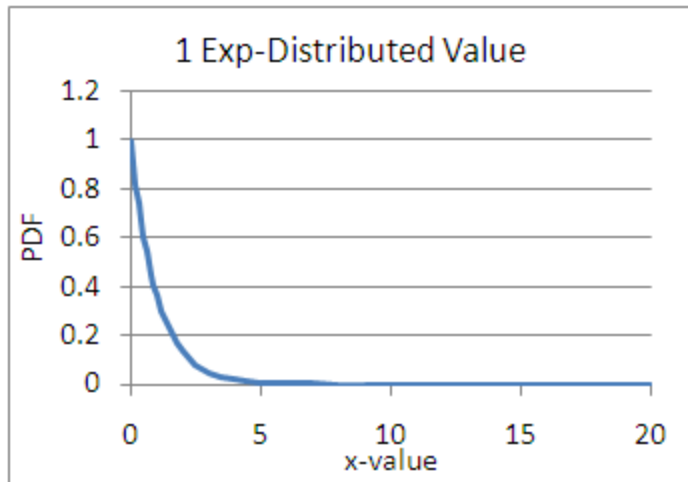


- Using Excel:
  - PDF =  $NORMDIST(x, \mu, \sigma, FALSE)$
  - CDF =  $NORMDIST(x, \mu, \sigma, TRUE)$
- Plot using:
  - y-axis = probit =  $NORMSINV(CDF)$
  - x-axis = x
  - $\sigma = 1/\text{slope}$
  - $\mu = \text{x-intercept}$

# Normal Distribution Reliability Plots

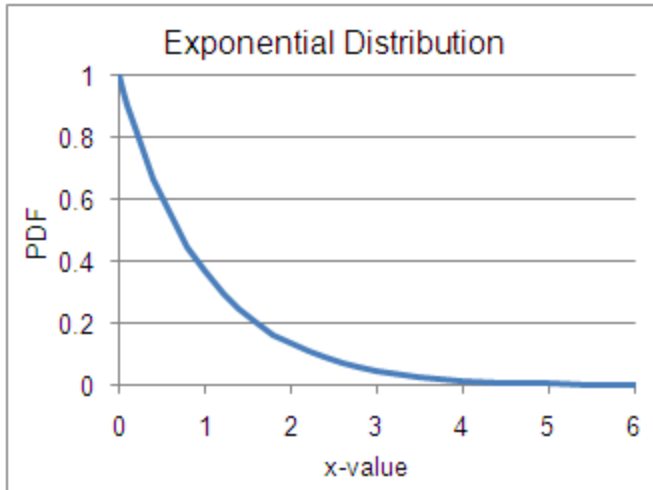


# Use of Normal Distributions



- Most measurement error
- Sum of random things is normal

# Exponential Distribution



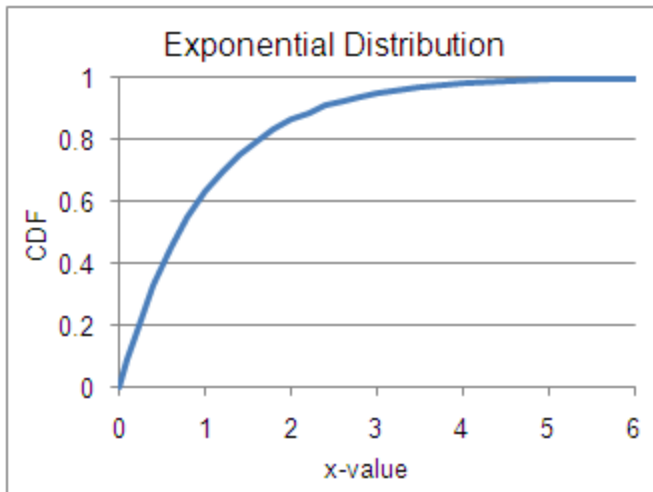
$\lambda$  = scale factor

$$f(x) = \lambda e^{-\lambda x}$$

$$F(x) = 1 - e^{-\lambda x}$$

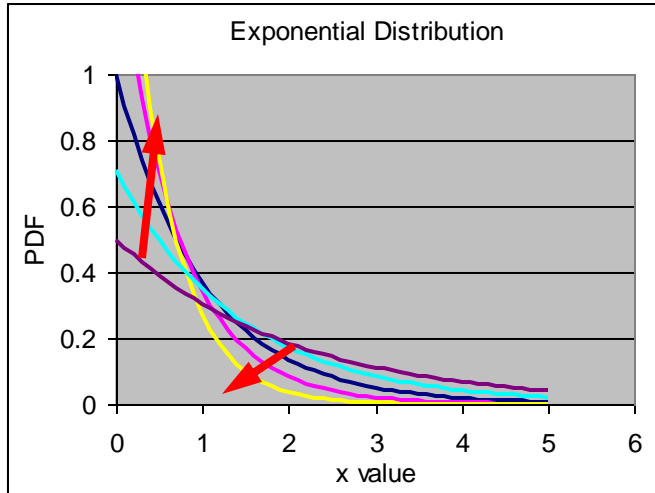
$$\text{rand exponential} = -\frac{\ln(1 - \text{CDF})}{\lambda}$$

where CDF is rand uniform



- Using Excel:
  - PDF =  $\lambda * \text{EXP}(-\lambda x)$
  - CDF =  $1 - \text{EXP}(-\lambda x)$
- Plot using:
  - y-axis = “exbit” =  $-\text{LN}(1 - \text{CDF})$
  - x-axis =  $x$
  - $\lambda$  = slope

# Exponential Distribution



lambda	1	1.41	2	0.71	0.5
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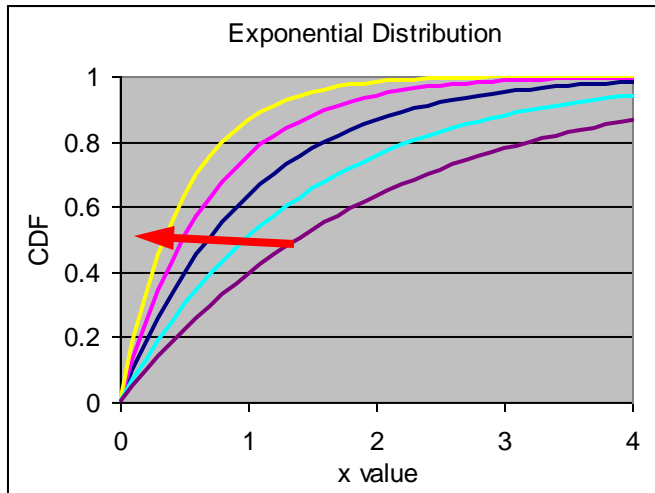
$\lambda$  = scale factor

$$f(x) = \lambda e^{-\lambda x}$$

$$F(x) = 1 - e^{-\lambda x}$$

$$\text{rand exponential} = -\frac{\ln(1 - \text{CDF})}{\lambda}$$

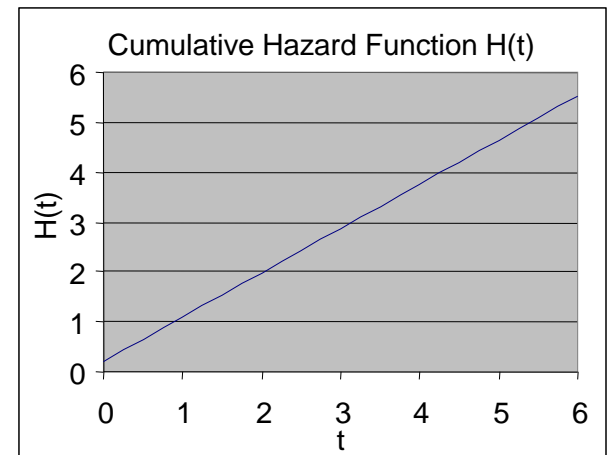
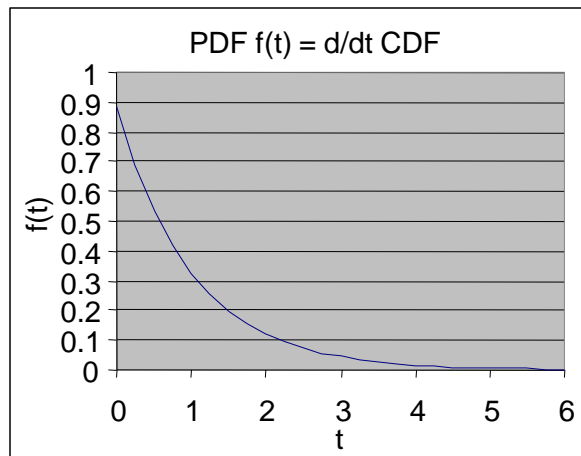
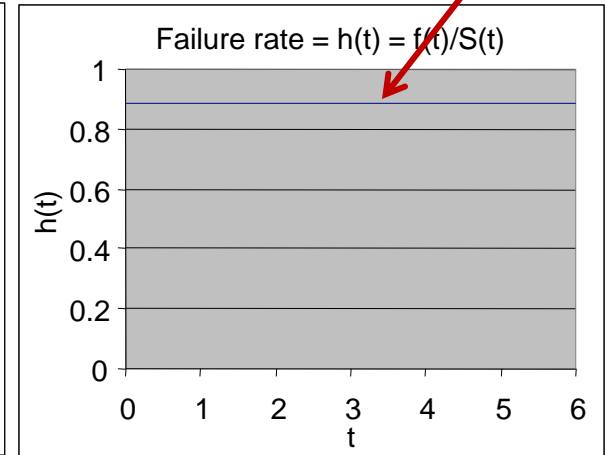
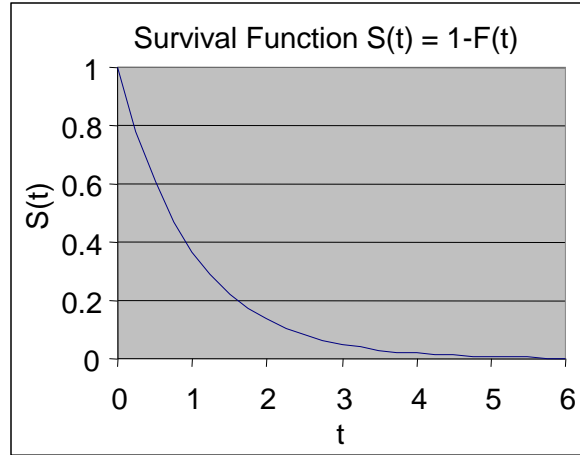
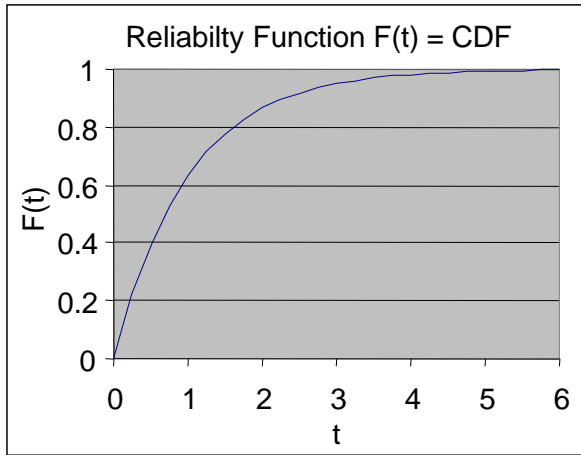
where CDF is rand uniform



- Using Excel:
  - PDF =  $\lambda * \text{EXP}(-\lambda x)$
  - CDF =  $1 - \text{EXP}(-\lambda x)$
- Plot using:
  - y-axis = “exbit” =  $-\text{LN}(1 - \text{CDF})$
  - x-axis = x
  - $\lambda$  = slope



# Exponential Reliability Plots



# Use of Exponential Distributions

- Constant fail rate
  - No “memory” of the past; no age
  - Radioactive decay
  - Soft errors, external environment

- Easy to calculate

- MTTF =  $1/\lambda$

- Median time to fail from

$$F(t_{50}) = 1 - e^{-\lambda t_{50}} = 0.5 \quad \text{so} \quad t_{50} = \frac{\ln 2}{\lambda}$$

# Exercise 3.2

- Given an exponential fail distribution with

$$\lambda = \frac{0.04\%}{\text{khr}}$$

what is the probability of failure within 15,000 hours of use?  
What is the MTTF?

# Solution 3.2

- Convert to “pure” units

$$\lambda = \frac{0.04\%}{\text{khr}} = 0.000\,000\,4 \frac{\text{fails}}{\text{hour}}$$

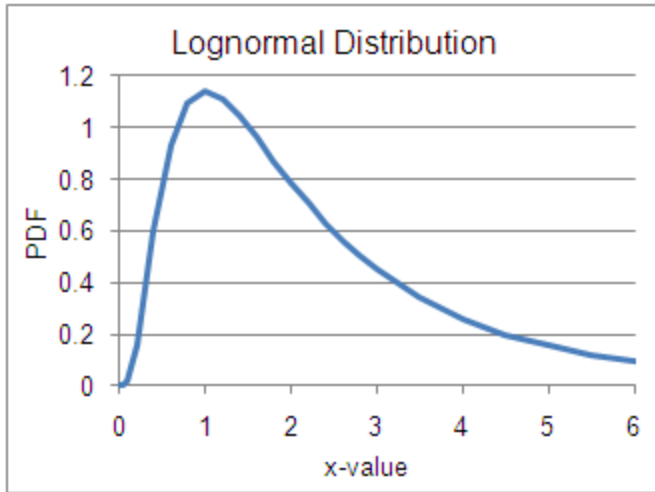
then evaluate the fail function at 15,000 hours

$$F(t) = 1 - e^{-\lambda t} = 1 - e^{-0.000\,000\,4 \times 15,000} = 0.006 = 0.6\%$$

The MTTF is even easier

$$MTTF = \frac{1}{\lambda} = 2,500,000 \text{ hours}$$

# LogNormal Distribution

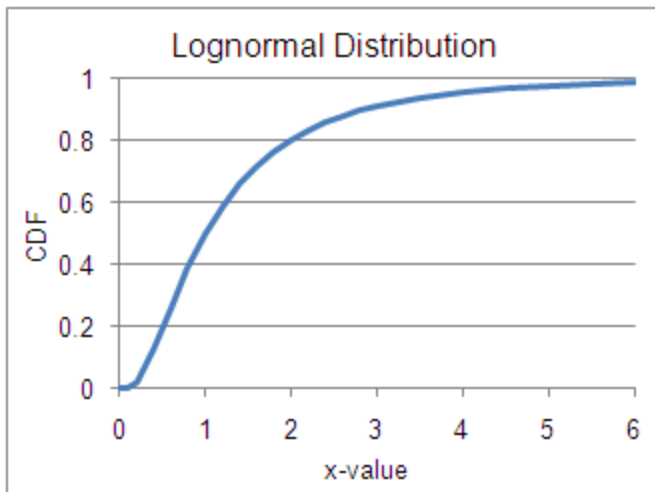


$t_{50}$  = median time to fail  
 $\sigma$  = standard deviation

$$f(t) = \frac{1}{\sigma t \sqrt{2\pi}} e^{-\frac{1}{2} \left[ \frac{\ln(t) - \ln(t_{50})}{\sigma} \right]^2}$$

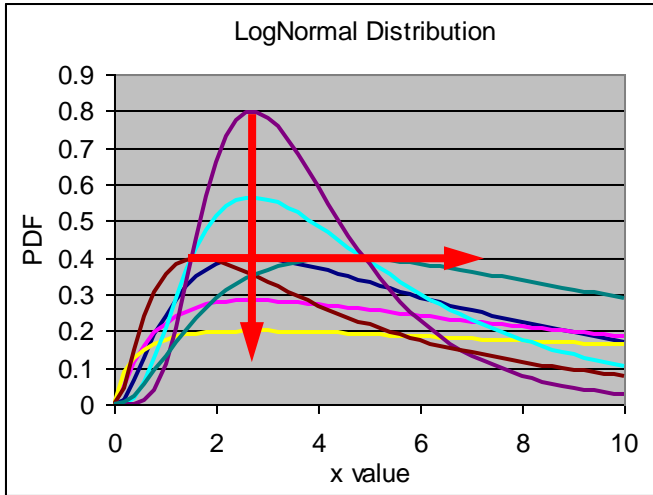
$$F(t) = \int_{-\infty}^t dt' \frac{1}{\sigma t' \sqrt{2\pi}} e^{-\frac{1}{2} \left[ \frac{\ln(t') - \ln(t_{50})}{\sigma} \right]^2}$$

rand normal =  $\exp(\text{NORMSINV}(\text{CDF}))$   
 where CDF is rand uniform



- Using Excel:
  - PDF = NORMDIST(ln(t),ln(t50),σ,FALSE)/t
  - CDF = NORMDIST(ln(t),ln(t50),σ,TRUE)
- Plot using:
  - y-axis = probit = NORMSINV(CDF)
  - x-axis = ln(t)
  - $\sigma = 1/\text{slope}$
  - $\ln(t_{50}) = \text{x-intercept}$

# LogNormal Distribution



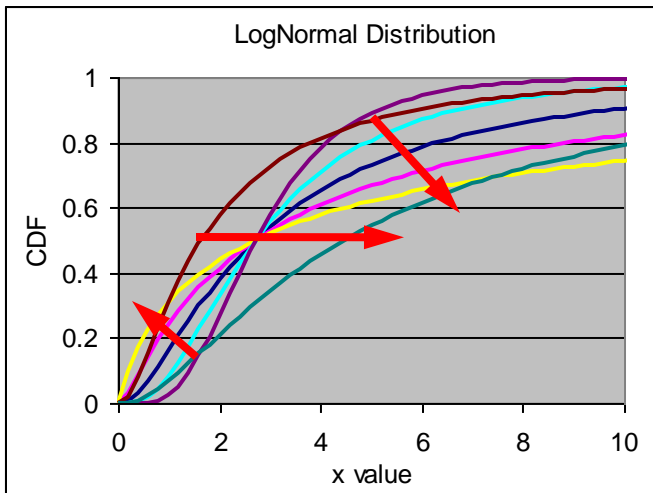
t50	1	1	1	1	1	0.5	1.5
std	1	1.41	2	0.71	0.5	1	1

t50 = median time to fail  
 σ = standard deviation

$$f(t) = \frac{1}{\sigma t \sqrt{2\pi}} e^{-\frac{1}{2} \left[ \frac{\ln(t) - \ln(t50)}{\sigma} \right]^2}$$

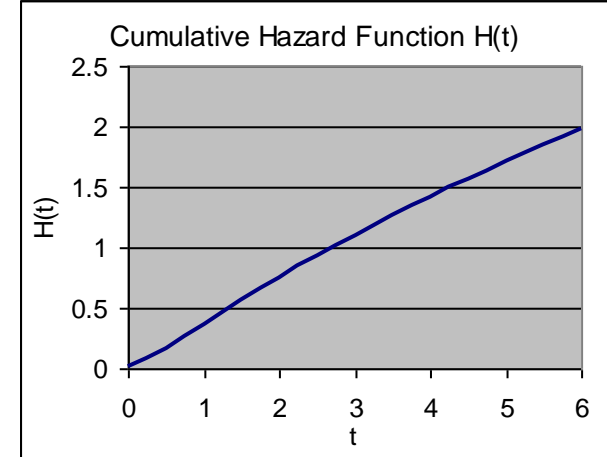
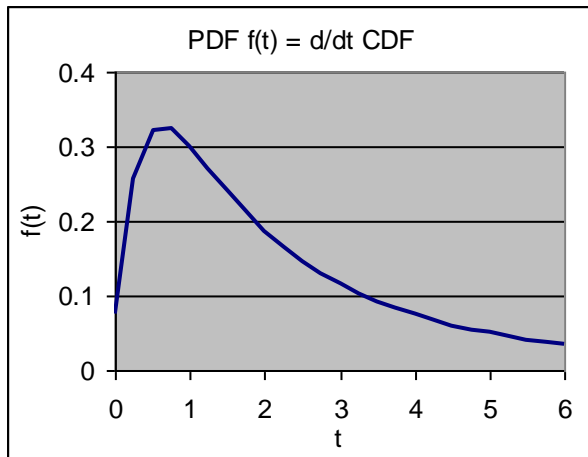
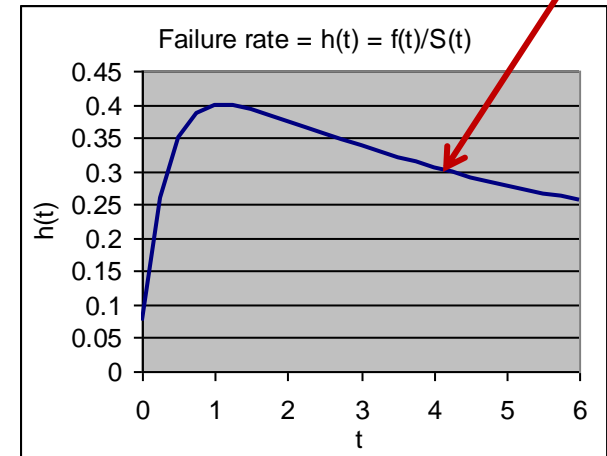
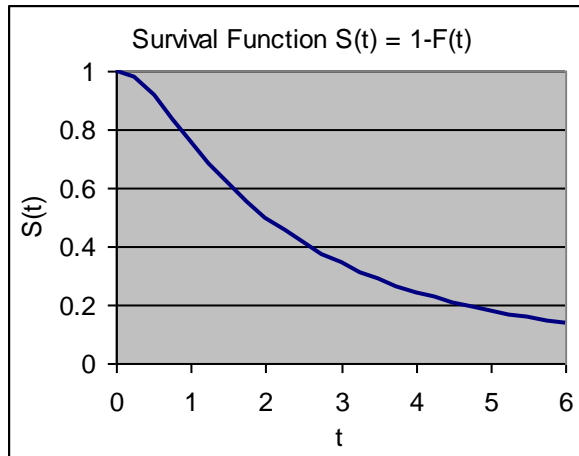
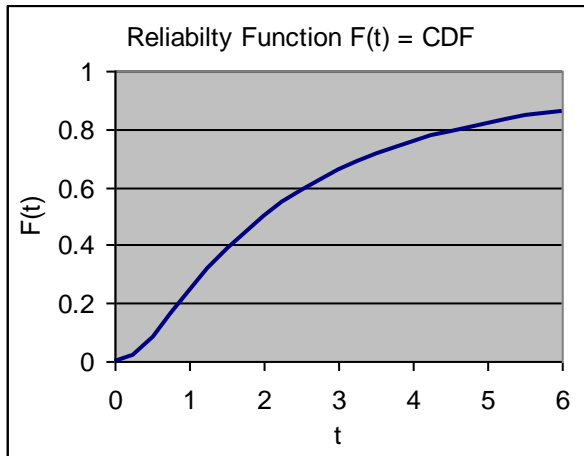
$$F(t) = \int_{-\infty}^t dt' \frac{1}{\sigma t' \sqrt{2\pi}} e^{-\frac{1}{2} \left[ \frac{\ln(t') - \ln(t50)}{\sigma} \right]^2}$$

rand normal =  $\exp(NORMSINV(CDF))$   
 where CDF is rand uniform



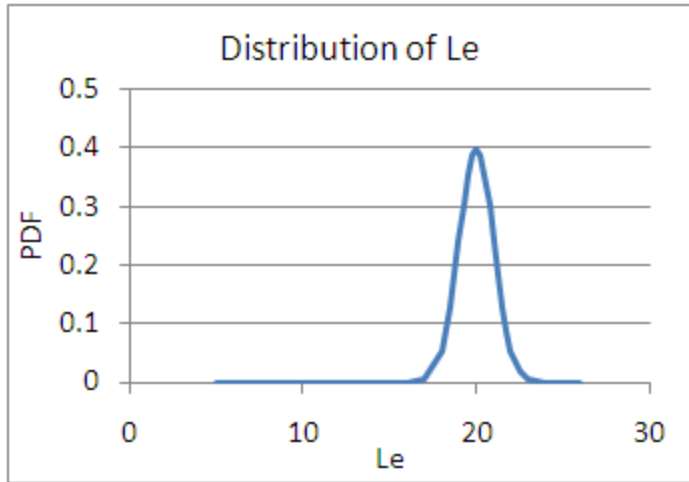
- Using Excel:
  - PDF = NORMDIST(ln(t),ln(t50),σ,TRUE)/t
  - CDF = NORMDIST(ln(t),ln(t50),σ,FALSE)
- Plot using:
  - y-axis = probit = NORMSINV(CDF)
  - x-axis = ln(t)
  - σ = 1/slope
  - ln(t50) = x-intercept

# Lognormal Reliability Plots

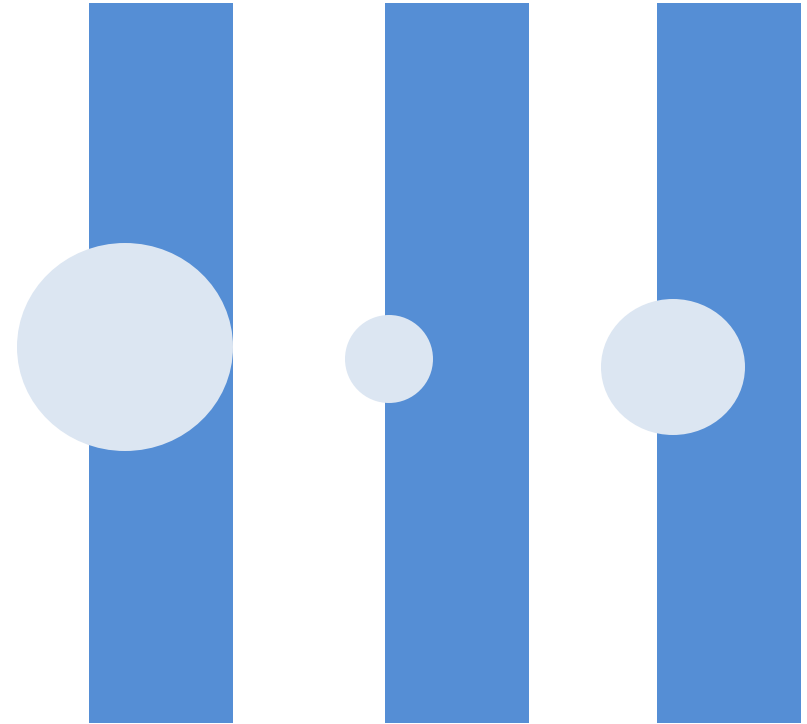
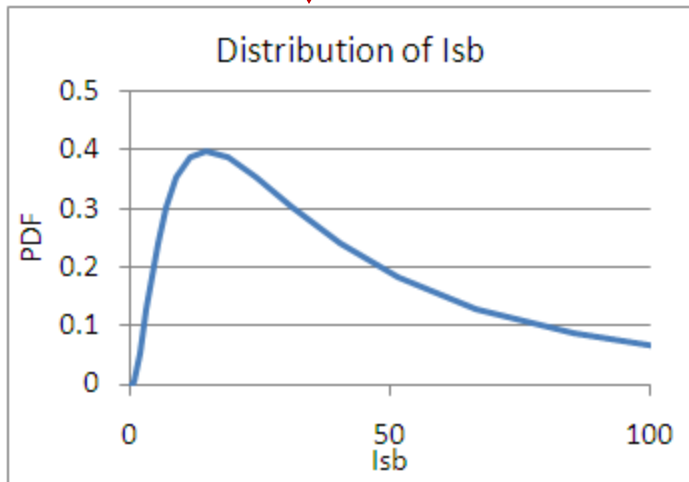


Mostly decreasing failure rate:  
IM-type mechanism

# Use of Lognormal Distributions



↓  $I_{SB} \sim e^{Le}$

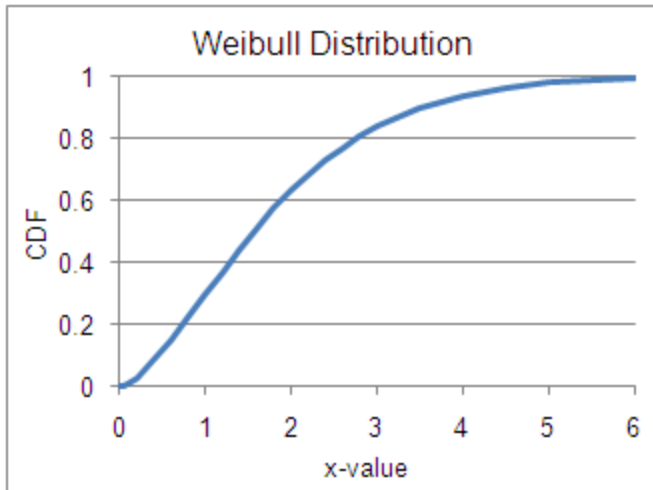
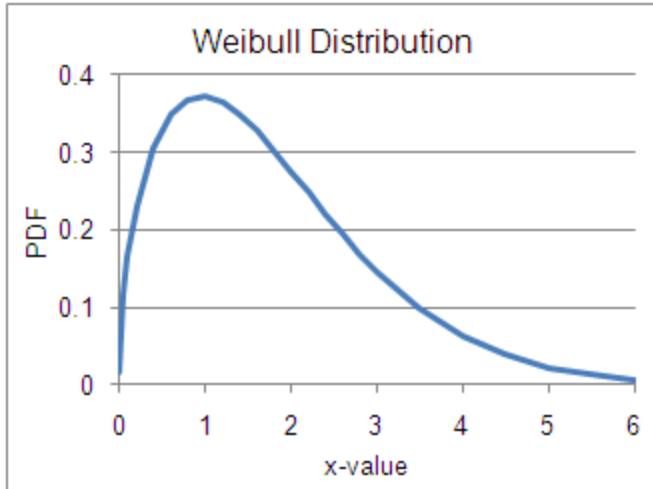


$$R_t = (1 + \delta) \times R_{t-1}$$



# Weibull Distribution

$$e^{-\left(\frac{t}{\alpha}\right)^\beta}$$



$$f(x) = \frac{\beta}{\alpha} \left(\frac{x-\gamma}{\alpha}\right)^{\beta-1} \exp\left[-\left(\frac{x-\gamma}{\alpha}\right)^\beta\right]$$

$$F(x) = 1 - \exp\left[-\left(\frac{x-\gamma}{\alpha}\right)^\beta\right]$$

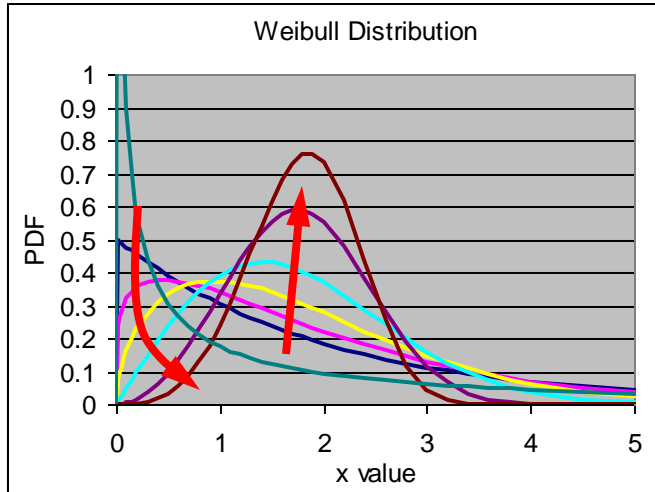
$\beta$  = shape parameter  
 $\alpha$  = scale parameter  
 $\gamma$  = location parameter

Note:  $\alpha$  and  $\beta$  are often swapped in meaning!  
 Excel swaps them (below).  
 T&T use  $\beta \rightarrow m$  and  $\alpha \rightarrow c$ .

rand Weibull =  $\alpha[-\ln(1 - CDF)]^{1/\beta}$   
 where CDF is rand uniform

- Using Excel:
  - PDF = WEIBULL(x,β,α,FALSE)
  - CDF = WEIBULL(x,β,α,TRUE) = 1-EXP(-((x/α)^β))
  - Note  $\gamma=0$  in Excel
- Plot using:
  - y-axis = weibit =  $\ln(-\ln(1-CDF))$
  - x-axis =  $\ln(x)$
  - $\beta$  = slope
  - $\alpha = \exp(-\text{intercept/slope})$

# Weibull Distribution



	1	1.2	1.5	2	3	4	0.5
beta	2	2	2	2	2	2	2
alpha	2	2	2	2	2	2	2

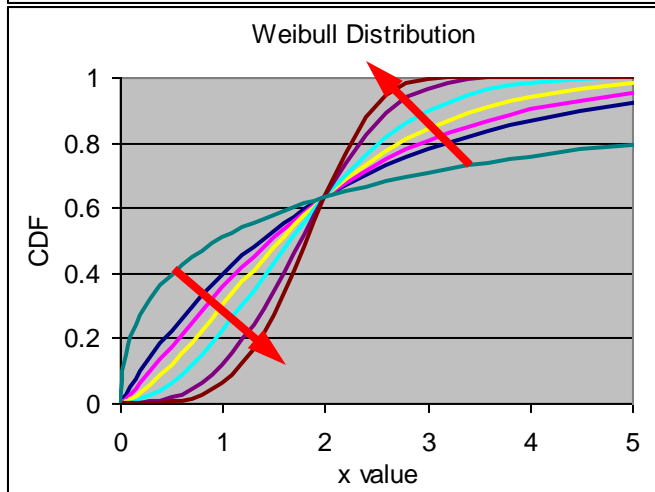
$$f(x) = \frac{\beta}{\alpha} \left( \frac{x-\gamma}{\alpha} \right)^{\beta-1} \exp \left[ - \left( \frac{x-\gamma}{\alpha} \right)^{\beta} \right]$$

$\beta$  = shape parameter  
 $\alpha$  = scale parameter  
 $\gamma$  = location parameter

$$F(x) = 1 - \exp \left[ - \left( \frac{x-\gamma}{\alpha} \right)^{\beta} \right]$$

Note:  $\alpha$  and  $\beta$  are often swapped in meaning!  
 Excel swaps them (below).  
 T&T use  $\beta \rightarrow m$  and  $\alpha \rightarrow c$ .

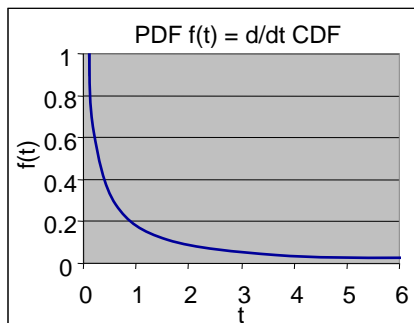
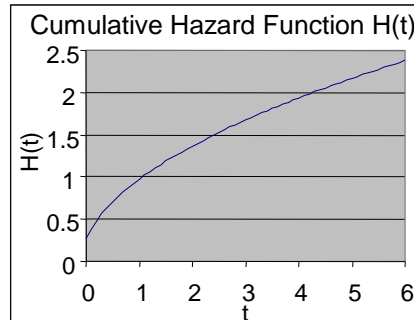
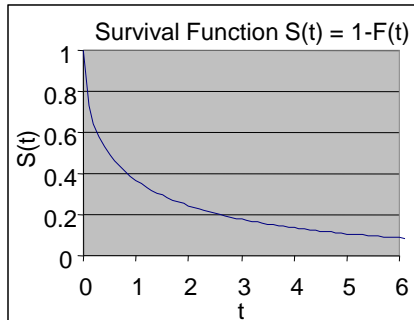
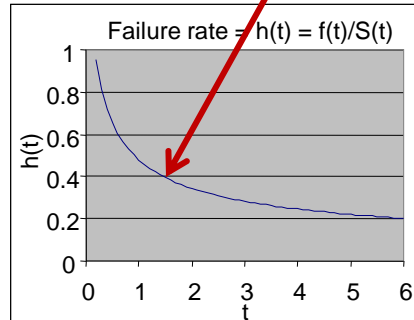
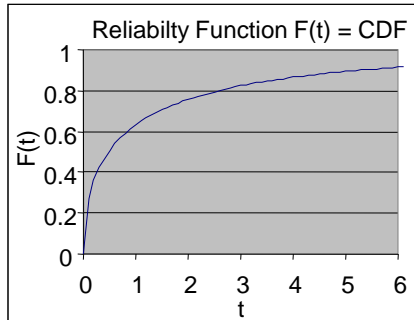
rand Weibull =  $\alpha [-\ln(1 - CDF)]^{1/\beta}$   
 where CDF is rand uniform



- Using Excel:
  - PDF = WEIBULL(x,β,α,FALSE)
  - CDF = WEIBULL(x,β,α,TRUE) = 1-EXP(-(x/α)^β)
  - Note γ=0 in Excel
- Plot using:
  - y-axis = weibit = ln(-ln(1-CDF))
  - x-axis = ln(x)
  - β = slope
  - α = exp(-intercept/slope)

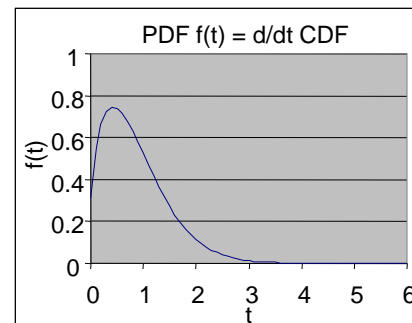
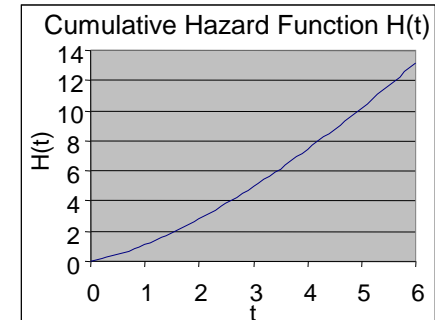
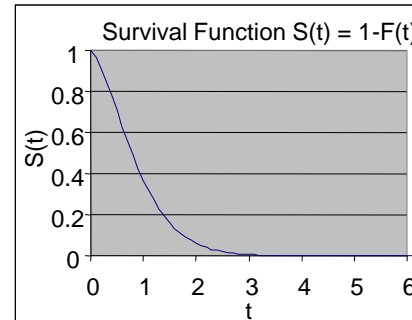
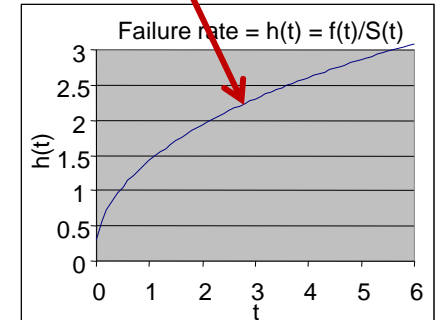
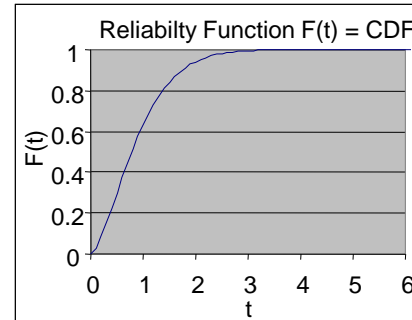
# Weibull Reliability Plots

Weibull,  $\beta=0.5 (<1)$



Decreasing failure rate:  
Infant Mortality (IM)  
type mechanism

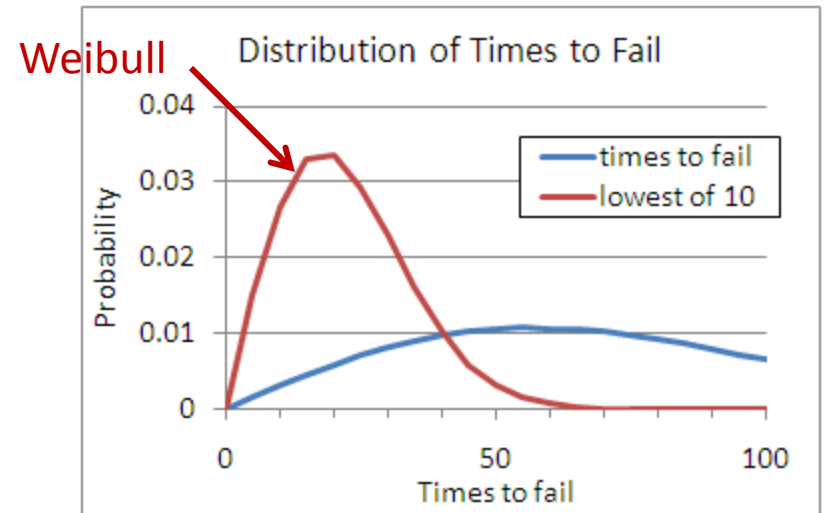
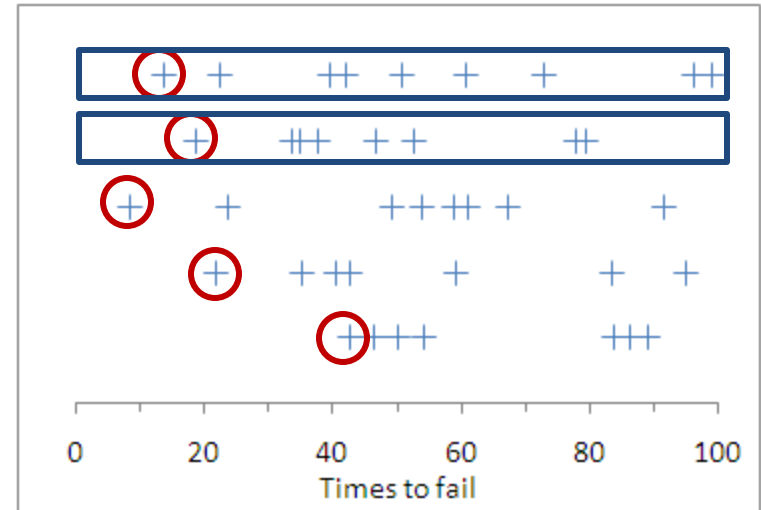
Weibull,  $\beta=1.5 (>1)$



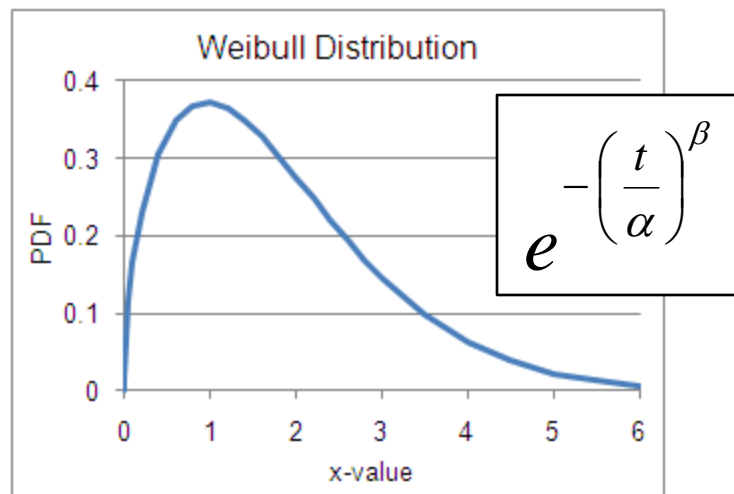
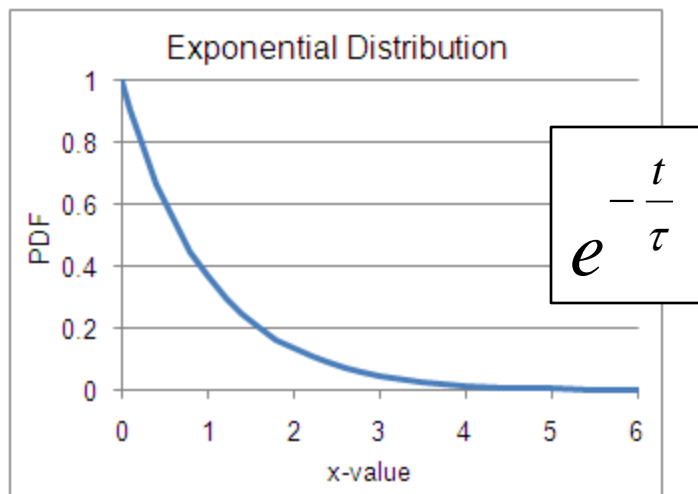
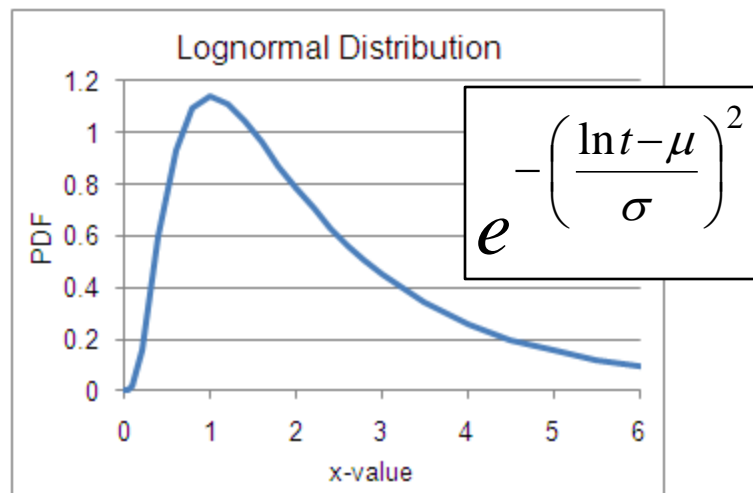
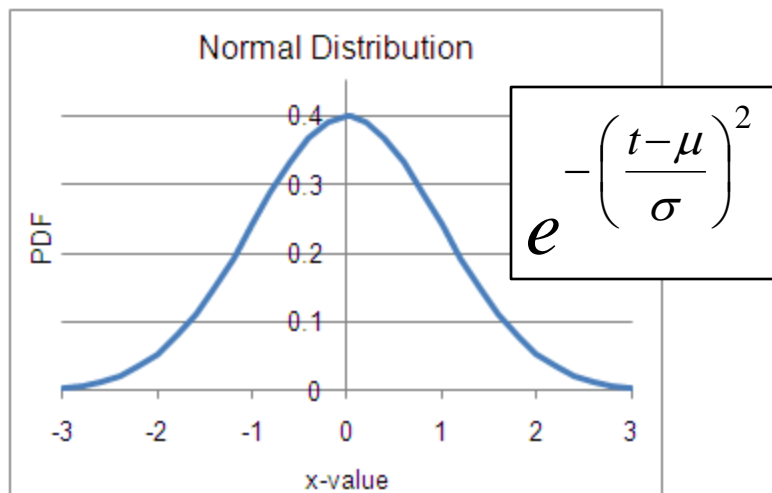
Increasing failure rate:  
Wearout (WO)  
type mechanism

# Use of Weibull Distributions

- When fail is caused by the worst of many items
- When it fits the data well

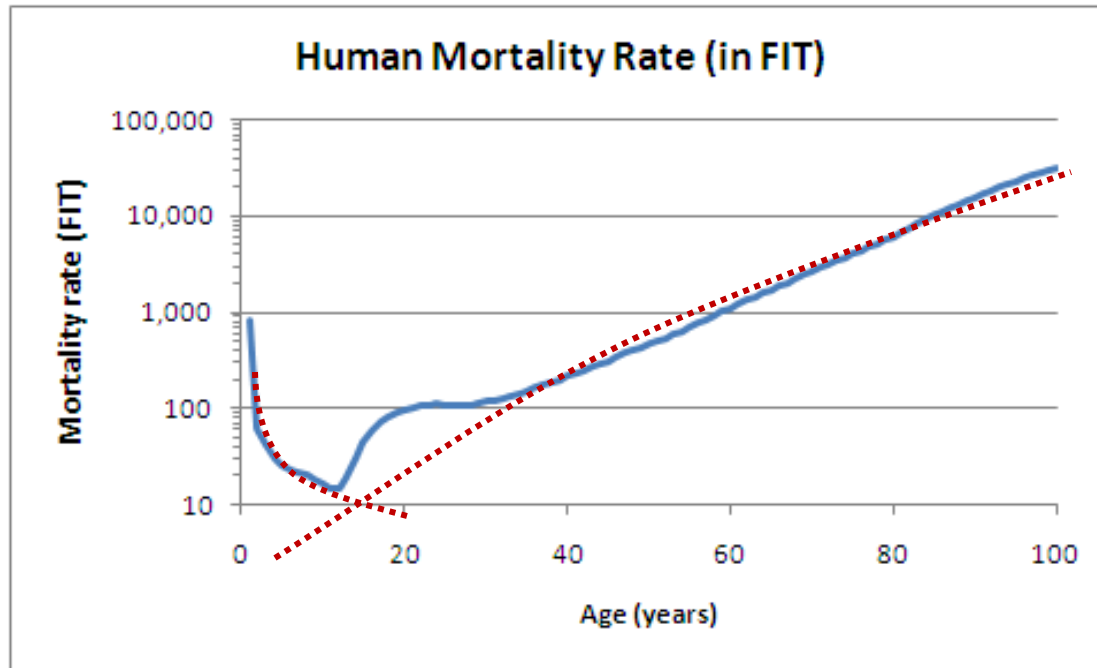


# Main Reliability Functions



# Multiple Mechanisms

# Multiple Mechanisms



Survivals multiply, hazard rates add:

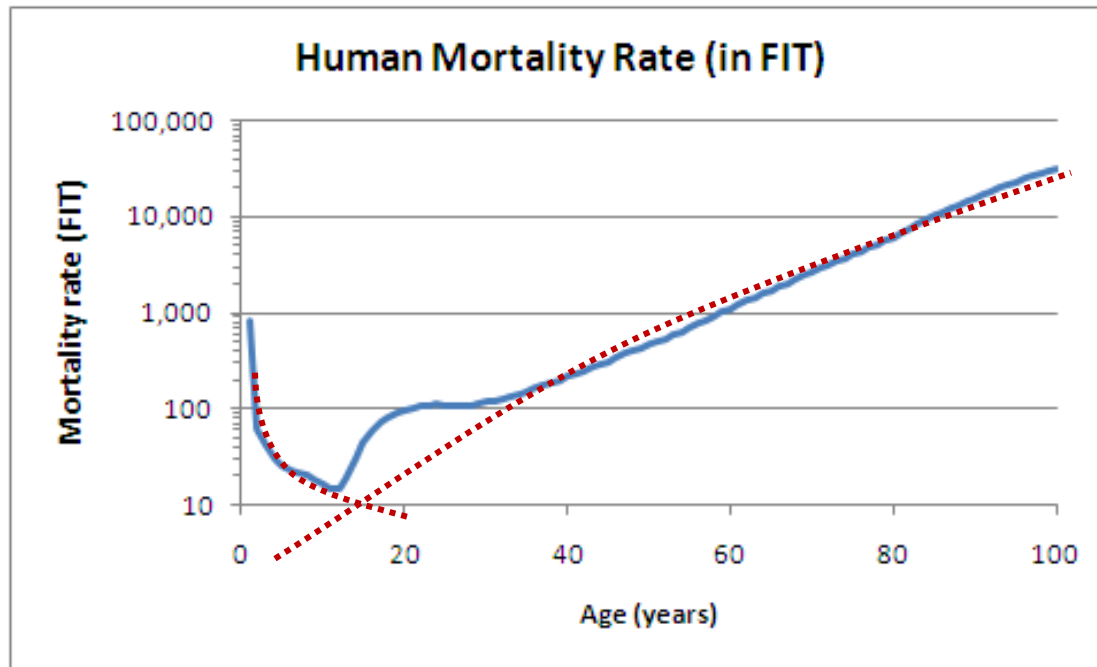
$$S_{tot}(t) = S_1(t) S_2(t)$$

$$F_{tot}(t) = 1 - S_1(t) S_2(t) \approx F_1(t) + F_2(t)$$

$$h_{tot}(t) \approx h_1(t) + h_2(t)$$

# Exercise 3.3

Hand fit 2 Weibull distributions to the human mortality data like this:



Plot both the hazard rate  $h(t)$  (like above) and the fail function  $F(t)$ .

Useful: for the Weibull, from T&T table 4.3:

$$h(t) = \frac{\beta}{\alpha} \left( \frac{t}{\alpha} \right)^{\beta-1}$$



# Solution 3.3

$$h(t) = \frac{\beta}{\alpha} \left( \frac{t}{\alpha} \right)^{\beta}$$

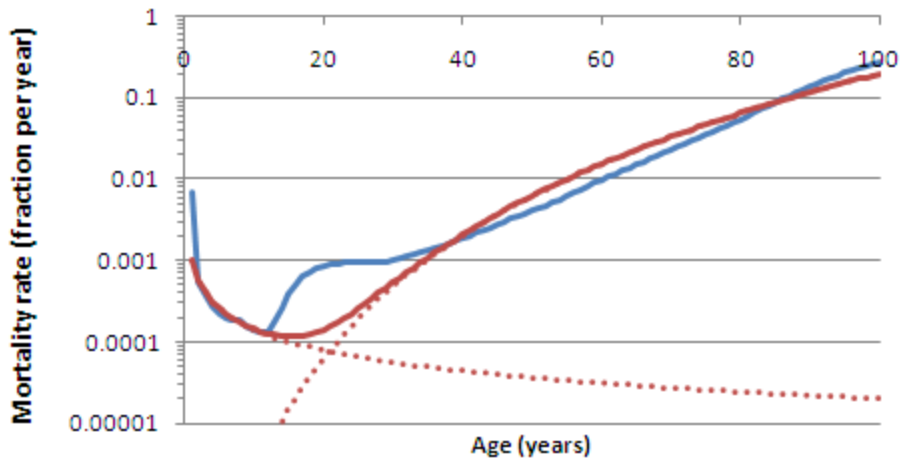
$$h_1(t) + h_2(t)$$

$$F(t) = 1 - e^{-\left(\frac{t}{\alpha_1}\right)^{\beta_1}} e^{-\left(\frac{t}{\alpha_2}\right)^{\beta_2}}$$

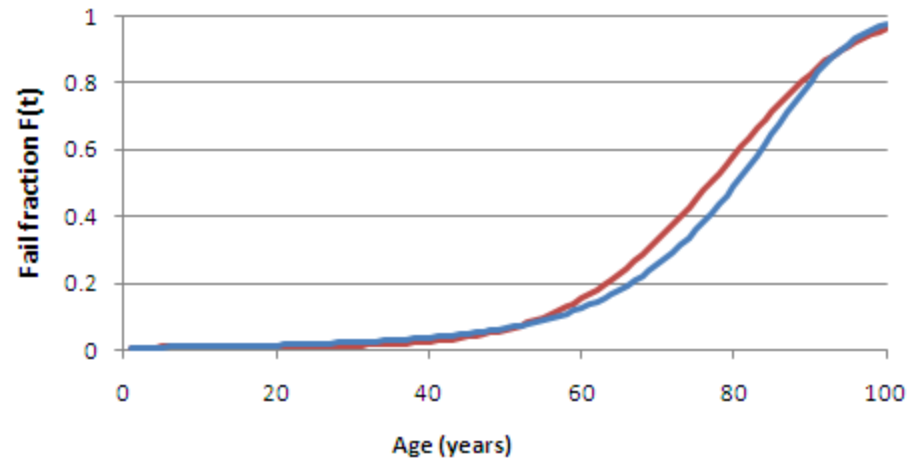
Age	data h(t)	data H(t)	data F(t)	Weib1 h(t)	Weib2 h(t)	Weib h(t)	Weib F(t)
1	0.00706	0.00706	0.007035	0.0010105	1.974E-11	0.00101	0.006714
2	0.00053	0.00759	0.007561	0.0005606	6.316E-10	0.000561	0.007447
3	0.00036	0.00795	0.007918	0.0003972	4.796E-09	0.000397	0.007912
4	0.00027	0.00822	0.008186	0.000311	2.021E-08	0.000311	0.008259

alpha	3E+14	82
beta	0.15	6

Human Mortality Rate (fraction per year)



Fail Fraction F(t), Data and Dual Weibull



The End