

# Pragmatic Weibull Analysis

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# Prerequisites

- Calculus
- Linear Algebra
- Probability Theory
- Mathematical, Critical and Analytical Thinking

$$\frac{e^{-x^2/2}}{\sqrt{2\pi}}$$

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

## Take aways

- Practical knowledge of Weibull analysis
- Self-reliant confidence in fitting curves

$$e^{-x}$$

$$\lambda e^{-\lambda x}$$

$$\frac{\beta}{\eta} \left(\frac{x}{\eta}\right)^{\beta-1} e^{-(x/\eta)^\beta}$$

# History

- Fréchet (fray-SHAY) 1927 ( $F = \frac{1}{W}$ ; *recipricol of the Weibull*)
  - While investigating the limiting behavior of the maximum of independent and identically distributed (i.i.d.) random variables, Fréchet showed that under certain conditions, the distribution of the maximum converges to:

$$f(x) = \frac{\alpha}{s} \left( \frac{x-m}{s} \right)^{-1-\alpha} e^{-\left( \frac{x-m}{s} \right)^{-\alpha}}$$

- Rammler (RAHM-ler) 1933 (*Equivalent to the Weibull*)
  - The Rammler distribution assumes that an exponential relationship exists between the particle diameter (d) and the mass fraction of particles with diameter greater than d.

$$Y_d \stackrel{\text{def}}{=} e^{-(d/\bar{d})^n}$$

- Weibull (VY-bull) 1939 ( $W = \frac{1}{F}$ ; *recipricol of the Fréchet*)

# Probability Distributions ( $\Omega = \mathbb{R}_{\geq 0}$ case)

A **Probability Distribution** is a mathematical model used to construct probability measures. **Probability Density Functions** are one such characterization for continuous random variables. They are analogous to **Histograms**, and they have useful geometric interpretations.

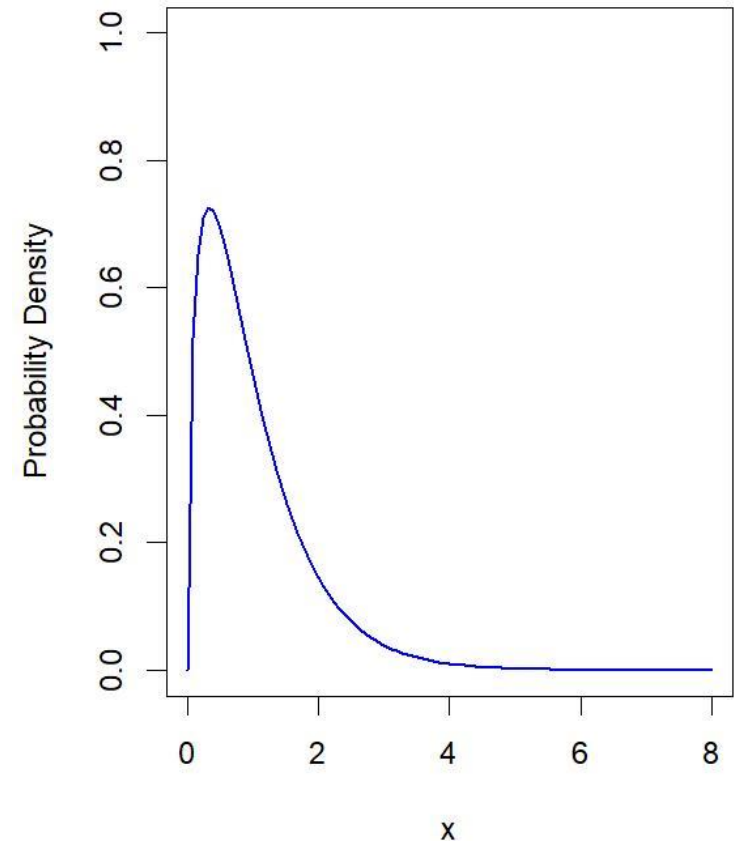
A **Probability Density Function** is a density function  $f(x)$  defined on  $\mathbb{R}_{\geq 0}$  such that,

$$P(\mathbb{R}_{\geq 0}) = \int_0^{\infty} f(x)dx = 1$$

And,

$$P([a, b]) = \int_a^b f(x)dx$$

PDF



# Definitions ( $\Omega = \mathbb{R}_{\geq 0}$ case)

The **Cumulative Distribution Function**  $F(x)$  is defined as

$$F(x) \stackrel{\text{def}}{=} \int_0^x f(z) dz = P(X \leq x)$$

The **Inverse Cumulative Distribution Function**  $F^{-1}(p)$  is defined at  $p$  to be the value  $x$  such that  $F(x) = p$

The **Reliability Function** or **Survival Function** is defined as the complement of the Cumulative Distribution Function:

$$S(x) \stackrel{\text{def}}{=} 1 - F(x) = P(X > x)$$

# Weibull Distribution Overview

$$f(x) = \frac{\beta}{\eta} \left(\frac{x}{\eta}\right)^{\beta-1} e^{-(x/\eta)^\beta}$$

$$F(x) \stackrel{\text{def}}{=} \int_0^x f(t)dt = 1 - e^{-(x/\eta)^\beta}$$

Example Weibull plots for beta = 4, eta = 3

- beta represents the acceleration of failure rate
- eta represents characteristic life independent of beta

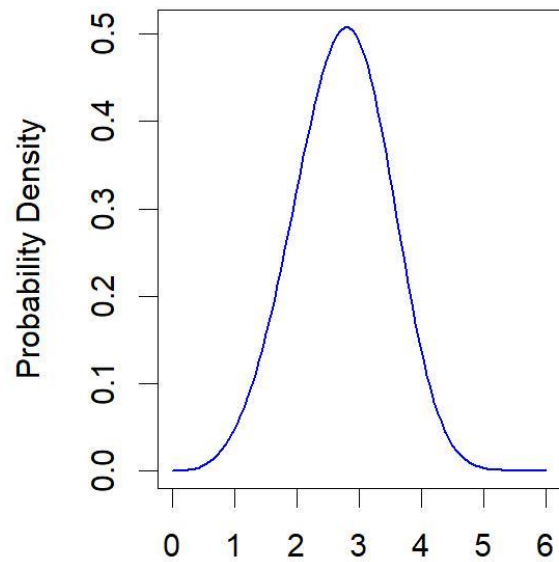
Note1:  $F(\eta) \cong 0.63$  and  $S(\eta) \cong 0.37$

Note2: For  $\beta > 1$ ,  $h(x)$  increases over "time"

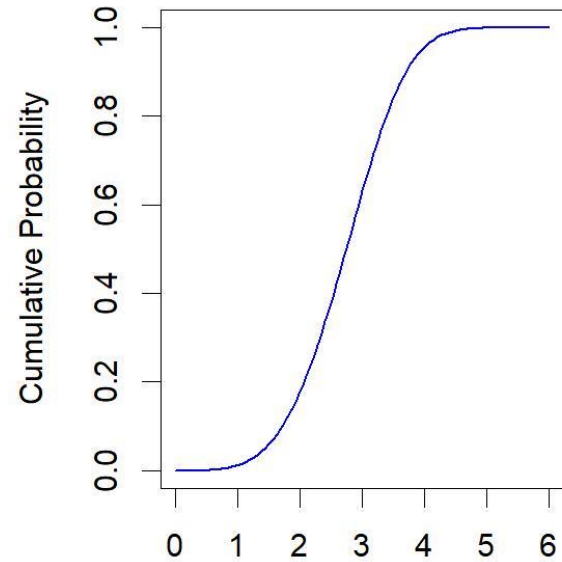
$f(x)$  ← “Failure Space” →  $F(x)$

$S(x) = 1 - F(x)$

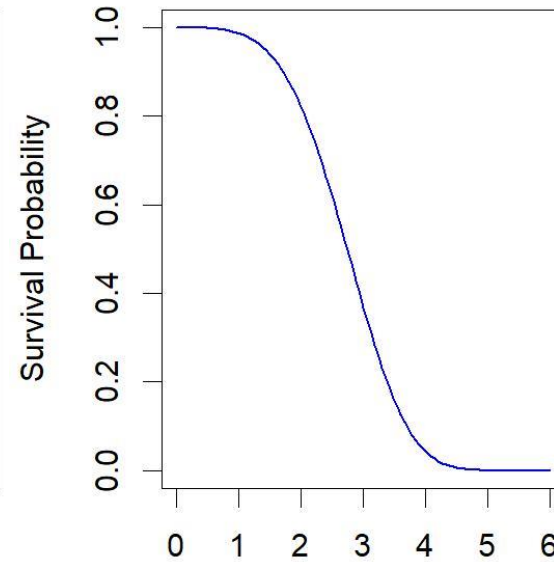
$h(x) = f(x) / S(x)$



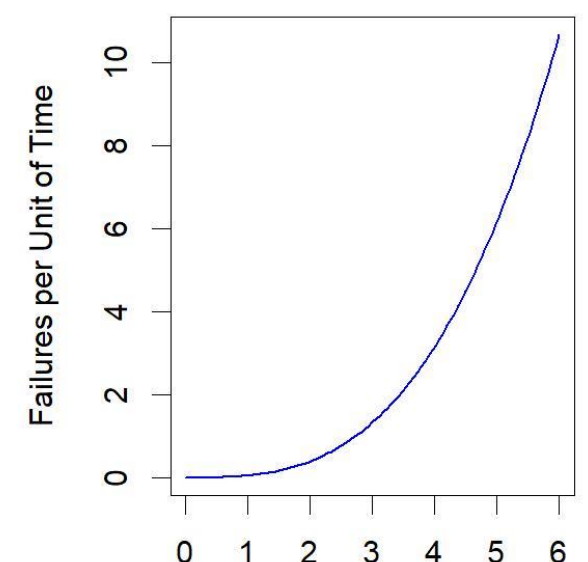
Design Lives



Design Lives



Design Lives

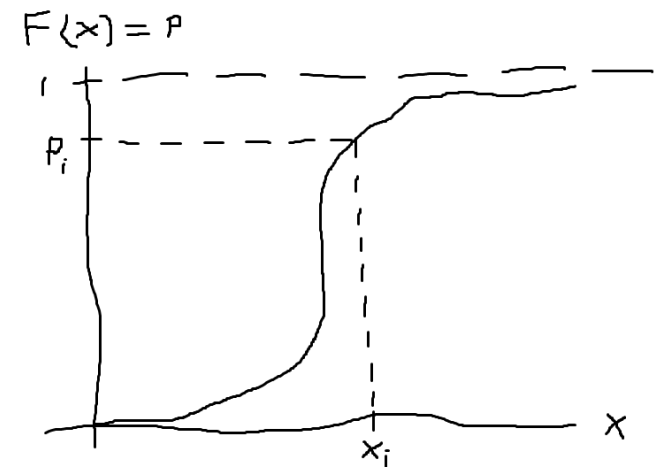


Design Lives

# Notional Data

## Generate evolving data sets from an arbitrary Weibull Distribution:

1. Select *representative* parameters
2. Define the inverse cumulative distribution
3. Sample random numbers  $p_i$  uniformly from  $[0,1]$
4. Compute  $F^{-1}(p_i)$



beta =	3
eta =	6

$$f(x) = \frac{\beta}{\eta} \left(\frac{x}{\eta}\right)^{\beta-1} e^{-(x/\eta)^\beta}$$

$$F(x) = p$$

$$F(x) = 1 - e^{-(x/\eta)^\beta}$$

$$F^{-1}(p) = x$$

$$F^{-1}(p) = \eta(-\ln(1 - p))^{\frac{1}{\beta}}$$

x
4.03
7.91
5.35

x
4.03
7.91
5.35
3.64
5.48
4.07

x	
$x_1$	4.03
$x_2$	7.91
$x_3$	5.35
$x_4$	3.64
$x_5$	5.48
$x_6$	4.07
$x_7$	4
$x_8$	7.8
$x_9$	1.92

## Apply curve fitting strategies:

1. Fit curves (illustrated on subsequent slides) to each of the three evolving data sets
2. Compare estimated parameters to the selected *representative* parameters
3. Visually examine the distribution of the samples against the “fitted” curves
4. Analyze what each strategy is doing at key steps in their respective procedures

# Summary of Two Strategies

- Maximum Likelihood Estimate (MLE)
  - Solves for the density curve that **maximizes** the likelihood of the data given the curve
    1. Begins with defining the Likelihood Function
    2. Follows with taking the log of the Likelihood and expressing eta as a function of beta
    3. Ends with numerically “maximizing” the log-likelihood function
  - Pros: automatically accommodates censored data sets
  - Cons: solutions do not always exist
- Median Rank Regression (MRR)
  - Solves for the cumulative curve that **minimizes** the “errors” between the data and curve
    1. Begins with linearizing the Weibull curve
    2. Follows with finding “Median Ranks”
    3. Ends with finding the parameters of the regression line then solving for eta and beta
  - Pros: credible solutions for small data sets
  - Cons: requires additional assumptions to address censored data

# Maximum Likelihood Estimate

Newton's Method

$$x_{n+1} = x_n - \frac{f'(x_n)}{f''(x_n)}$$

Density Function

$$f(x) = \frac{\beta}{\eta} \left(\frac{x}{\eta}\right)^{\beta-1} e^{-(x/\eta)^\beta}$$

Likelihood Function

$$L(\eta, \beta) = \prod_{i=1}^n f(x_i)$$

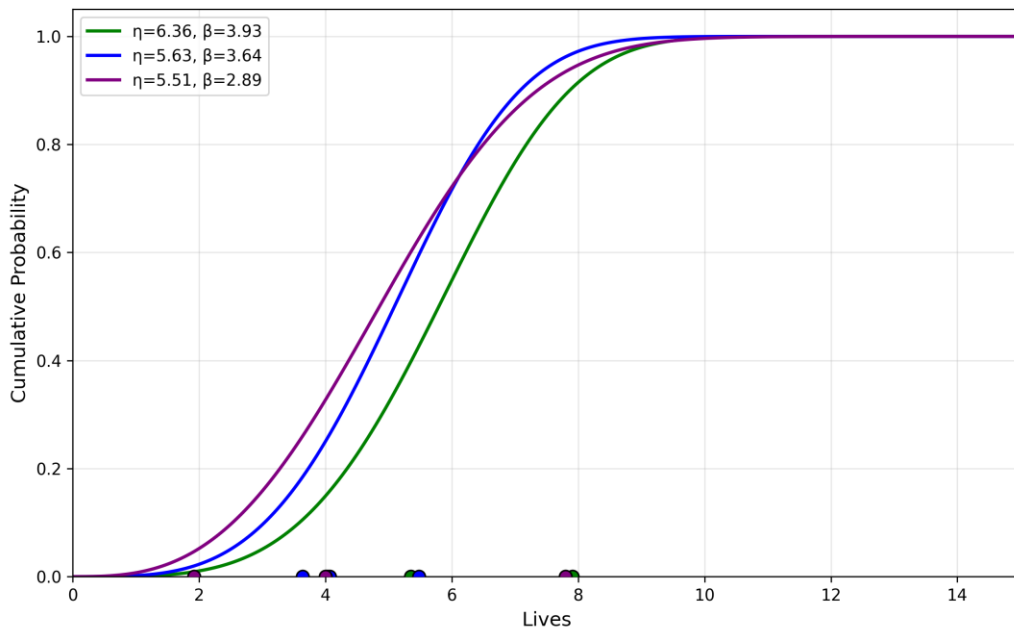
Log-Likelihood Function

$$LL(\eta, \beta) = \sum_{i=1}^n \ln(f(x_i))$$

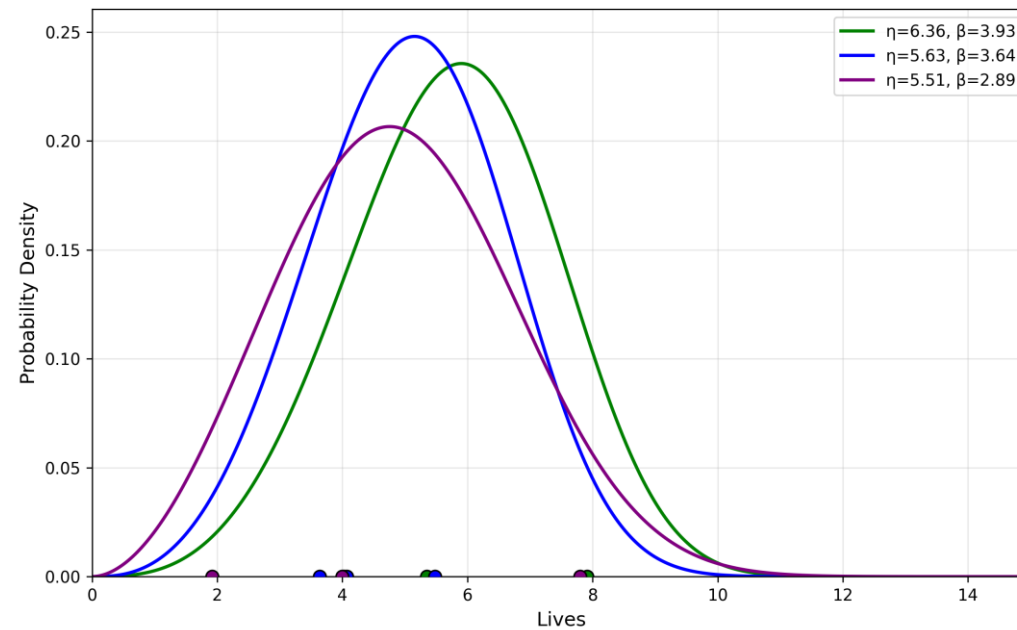
Eta as a function of beta and data

$$\hat{\eta}(\hat{\beta}) = \frac{\frac{1}{n} \sum_{i=1}^n x_i}{\Gamma\left(1 + \frac{1}{\hat{\beta}}\right)}$$

MLE Weibull CDF Evolution



MLE Weibull PDF Evolution



x	
$x_1$	4.03
$x_2$	7.91
$x_3$	5.35
$x_4$	3.64
$x_5$	5.48
$x_6$	4.07
$x_7$	4
$x_8$	7.8
$x_9$	1.92

# Median Rank Regression

$$i - th \text{ Median Rank} = F_{beta}(0.5, i, n + 1 - i)$$

Cumulative Probability

$$F(x) = 1 - e^{-(x/\eta)^\beta}$$

Linearized Cumulative Probability

$$\ln(-\ln(1 - F(x))) = \beta \cdot \ln(x) - \beta \cdot \ln(\eta)$$

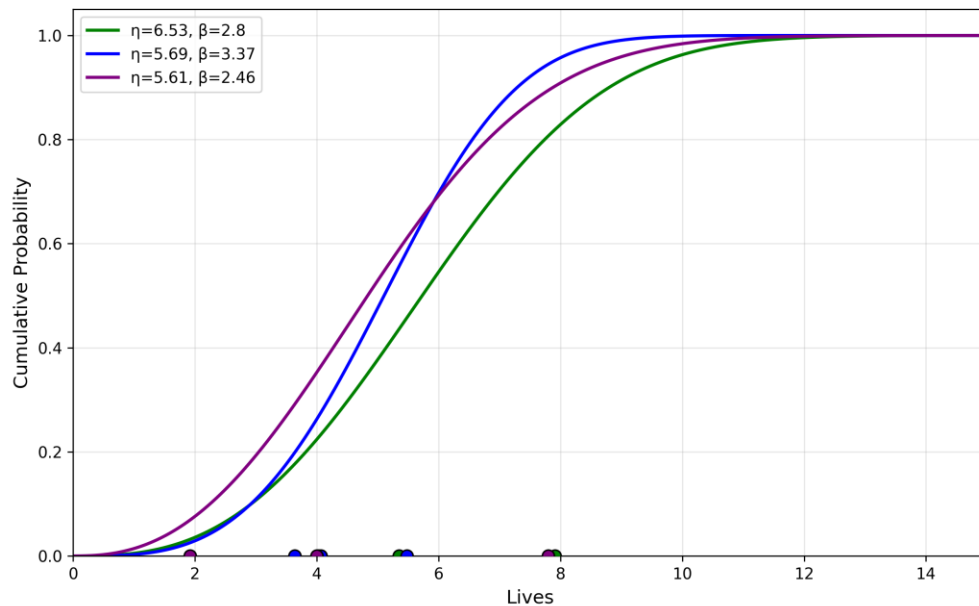
Regression Coefficients

$$m = \frac{\rho_{xy} \cdot \sigma_y}{\sigma_x}$$

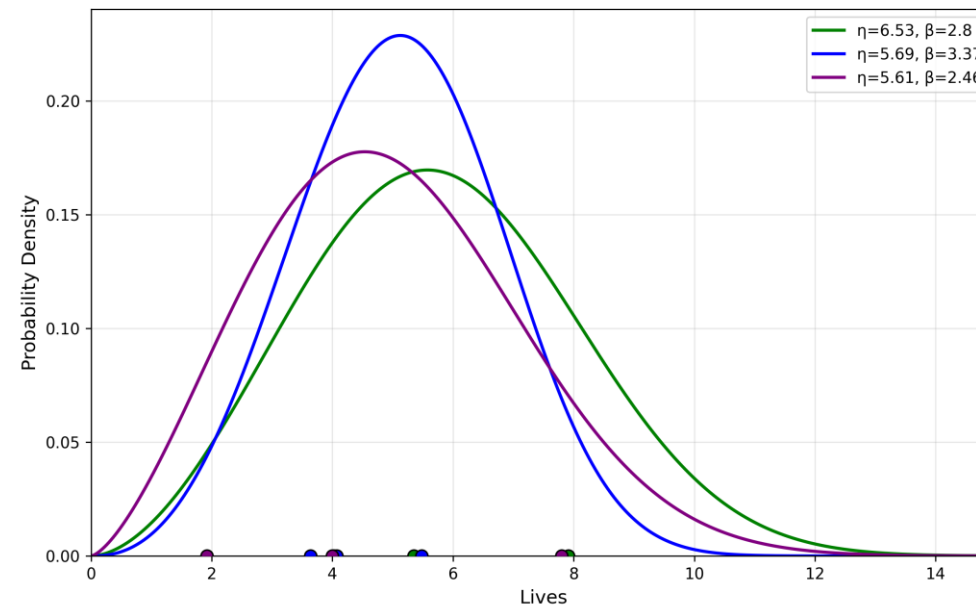
$$Y = m \cdot X + c$$

$$c = \mu_y - m \cdot \mu_x$$

MRR Weibull CDF Evolution



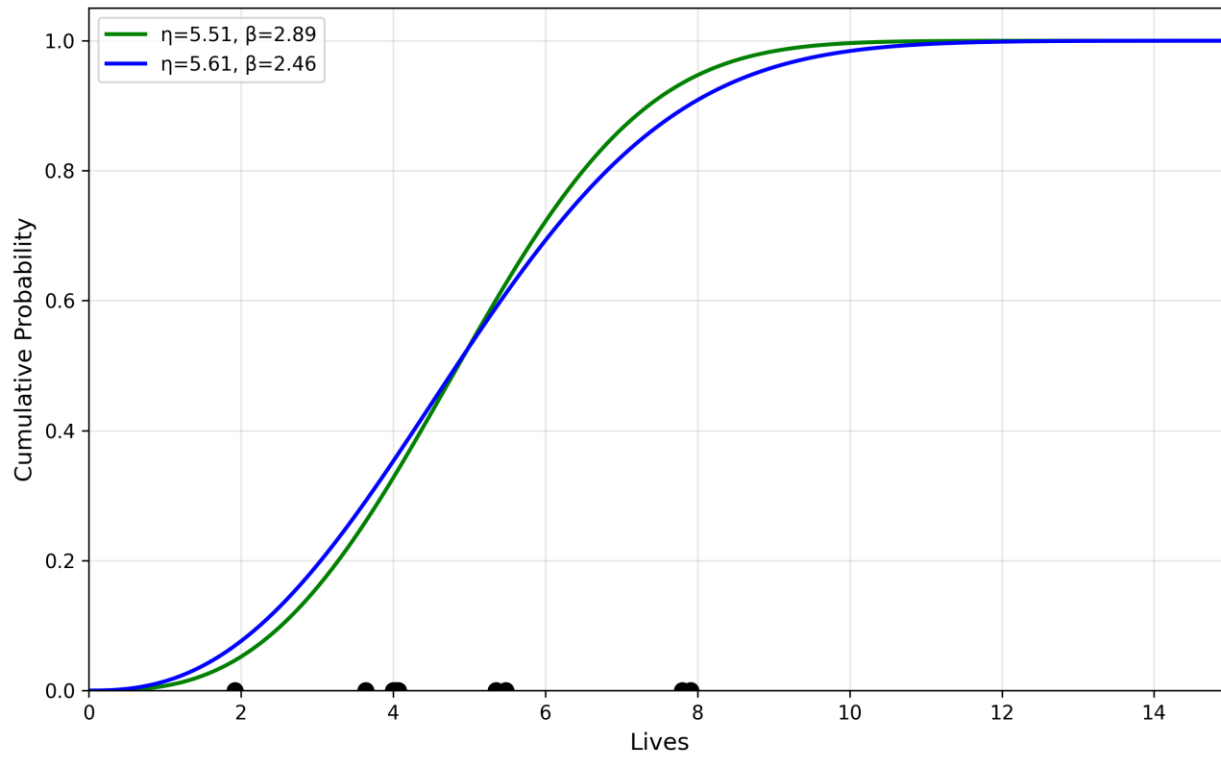
MRR Weibull PDF Evolution



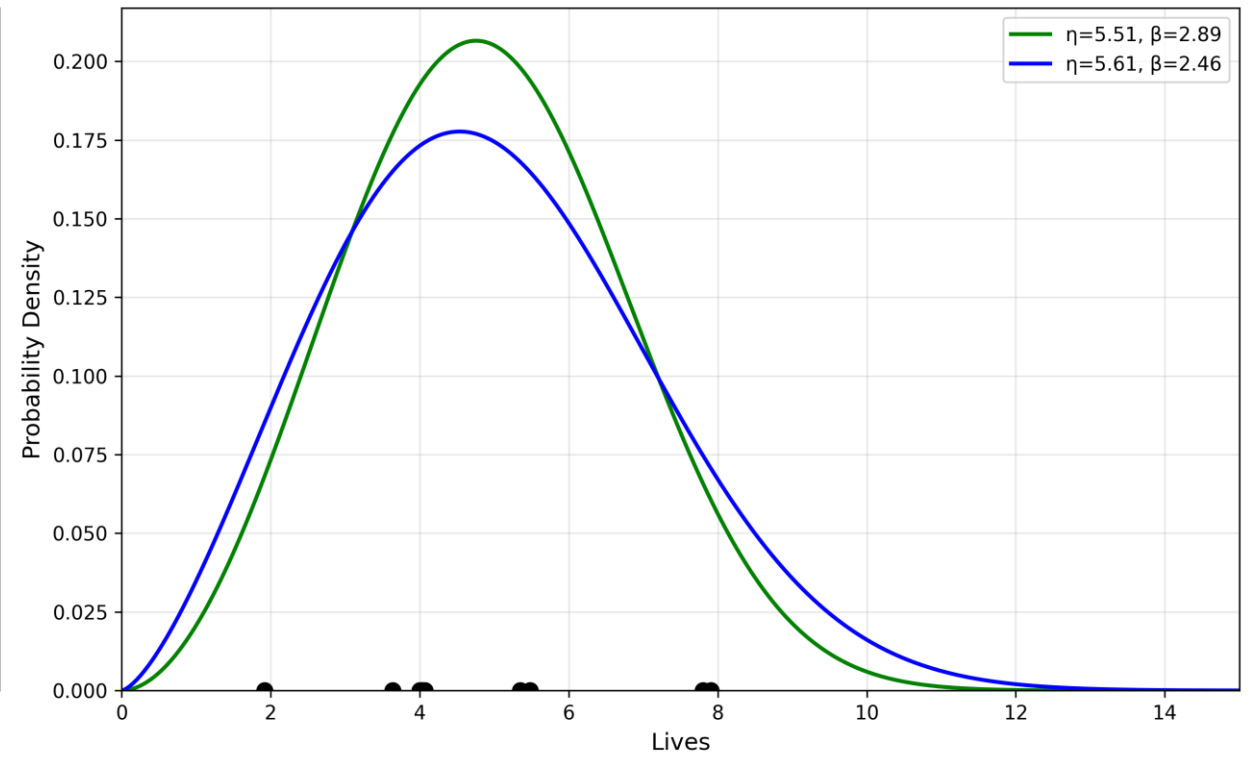
	x	Rank
$x_1$	4.03	4
$x_2$	7.91	9
$x_3$	5.35	6
$x_4$	3.64	2
$x_5$	5.48	7
$x_6$	4.07	5
$x_7$	4	3
$x_8$	7.8	8
$x_9$	1.92	1

# MLE vs MRR

MLE MRR Weibull CDF Comparison



MLE MRR Weibull PDF Comparison



Back up

# Median Rank Table

	$i - th \text{ Median Rank} = F_{beta}(0.5, i, n + 1 - i)$							
i	n = 2	n = 3	n = 4	n = 5	n = 6	n = 7	n = 8	n = 9
1	0.293	0.206	0.159	0.129	0.109	0.094	0.083	0.074
2	0.707	0.500	0.386	0.314	0.264	0.228	0.201	0.180
3		0.794	0.614	0.500	0.421	0.364	0.321	0.286
4			0.841	0.686	0.579	0.500	0.440	0.393
5				0.871	0.736	0.636	0.560	0.500
6					0.891	0.772	0.679	0.607
7						0.906	0.799	0.714
8							0.917	0.820
9								0.926