



Exploring Reliability with JMP: Competing Risks

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Topics

- Reliability Concepts and Terminology
- Life Distribution Fitting
- Competing Cause Analysis
- Accelerated Life Testing
- Recurrence Analysis
- Degradation Studies

Jmp.



Multiple Modes: Series Model

Consider a system made up with *n* components in series. If the *i*th component has reliability **R**_{*i*}(**t**), the system reliability is the product of the individual reliabilities, that is,

$$R_{s}(t)$$
 R_{1} t R_{2} t \ldots R_{n} t

The system reliability cannot be greater than the lowest component reliability (weakest link).

The **system failure rate** is the <u>sum</u> of the individual **component failure rates**. The system failure rate is higher than the highest individual failure rate.



Reliability Block Diagram for Components in Series

For two components in series:



The first component to fail causes system failure.

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The **bathtub curve** is a popular representation for the reliability of products with competing risks failures.



Example of Bathtub Curve



Product Reliability Issue

Company ABC is becoming aware of an increase in field failures from a mode of failure not previously seen in testing to 15,000 cycles, considered equivalent to three years in the field with normal use.

Production stress testing through 15,000 cycles showed the hazard rate was steadily decreasing. Management was actually considering extending the warranty period to 25,000 cycles before this issue arose.

Some customers were heavy users who were cycling the product at two to three types the expected rate and began experiencing failures after only one year.



Product Reliability Stress Test

A **sample** of 40 devices will be stressed for 30,000 cycles to see if the new failure mode can be reproduced.

- The exact time at which each system fails is recorded. All failure are then sent to failure analysis for identification of the mode of failure.
- At the end of the 30,000 cycle test, 28 units had failed and 12 survived.

MD.



Objectives of Reliability Test

- Learn whether the new failure mode can be reproduced.
- Model the distribution of both failure modes.
- Predict the failure fraction after 25,000 cycles in the field.
- Estimate how long it will take to reach 10% failures in the field for the second mode.
- Learn if burn-in can improve product reliability for the existing known mode.

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Analysis of Multiple Failure Modes

- Nonparametric analysis can reveal unusual patterns in the EDF.
- Failures associated with one failure modes will be censoring times for other failure modes.
- With information on modes associated with each failure, it is possible to do separate analysis on each mode.
- Independence of failure modes is an important assumption.

MD



JMP Data Table

We will analyze using JMP's Life Distribution Platform.

Two Modes Stress				
 Source 		TF	Mode Type	
	1	20	A	
	2	173	A	
	3	404	A	
	4	777	A	
	5	2755	A	
	6	2848	A	
	7	2954	A	
	8	4025	A	
	9	4043	A	
	10	5808	A	
	11	7472	A	
	12	7598	Α	
Columns (2/0)	13	8948	Α	
	14	9156	A	
Mode Type	15	11574	A	
	16	12438	A	
	17	13289	A	
	18	17802	В	
	19	17976	A	
	20	18257	A	
	21	19949	В	
	22	20118	В	
	23	20165	В	
	24	24117	В	
	25	26389	В	
	26	26754	В	
	27	27006	В	
Rows	28	27190	В	
All rows 4	0 29	30000	Censored	
Selected	0 30	30000	Censored	
Excluded	0 31	30000	Censored	
Hidden	32	30000	Censored	
Lapelled	33	30000	Censored	
	34	30000	Censored	
	35	30000	Censored	
	36	30000	Censored	
	37	30000	Censored	
	38	30000	Censored	
	39	30000	Censored	
	40	30000	Censored	



Nonparametric Plot All Points



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Lognormal Probability Plots



Full Horizontal Scale

Adjusted Horizontal Scale

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Lognormal Probability Plot with Modes





Statistics and Profiler



Fraction failing at 25000 cycles

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Nonparametric Plot Mode A



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_ognormal Probability Plot Mode A





Statistics and Profiler Mode A

Statistics									
Model Co	omparisons								
Distribution	AICc	-2Loglikeliho	od B	IC					
Lognormal	432.94702	428.622	436.0004	45				Rurn-In	Possibility
Summary	of Data							Dum-in	1 033101111
Nonpara	metric Estim	ate							
🕈 🛡 Parame	etric Estimate	e - Lognorn	nal						
Parameter	Estimate	Std Error	Lower 95%	Upper 95%	Criterion		/		
location	10.415732	0.56091492	9.3163587	11.515105	-2 LogLike	ihood 428.62269			
scale	2.797599	0.50600249	1.8058525	3.789346	AICc	432.94702			
		_			BIC	436.00045			
Covaria	ance Matrix								
🕈 💌 Distri	bution Profil	ler	🕈 💌 Quantil	e Profiler		🕽 🕈 🔍 Hazard Pro	filer	💙 🖻 Density Pr	ofiler
Probability 0.387556 [0.26035, 0.52825]	0.5 0.4 0.3 0.2 0.1 0		300 200 100 11118/100 1100 1	000 - 000 - 000 - 000 - 0 -		2000.0 9 14836-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856-2 14856		0.0002 	
	-0000 9 15	12000 300000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 300000 300000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 30000 3000000		E O	9.5 0.5 bability	ť			



Nonparametric Plot Mode B



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Lognormal Probability Plot Mode B







Statistics and Profiler Mode B

Statistics									
Model Co	mparisons								
Distribution Lognormal	AICc 206.03633	-2Loglikeliho 201.712	od B 01 209.089	IC 77					
Summary	of Data								
Nonparam	etric Estim	ate							
🛡 Paramet	tric Estimat	e - Lognorn	nal						
Parameter ocation scale	Estimate 10.354463 0.320804	Std Error 0.09481169 0.08395323	Lower 95% 10.168636 0.156259	Upper 95% 10.540291 0.485350	Criterion -2 LogLikelih AICc BIC	ood 201.71201 206.03633 209.08977			
Covaria	nce Matrix								
🕈 🛡 Distrib	oution Profi	ler	🕈 💌 Quantil	e Profiler		🕈 🖻 Hazard Profi	ler	🗢 Density Pr	ofiler
Probability 0.010712 [0.00052, 0.09358]	0.4 0.3 0.2 0 0 0 0 0 0 0 0 0 0 0 0 0	20000- 255000- 30000- 110- 255000- 100- 100- 100- 100- 100- 100- 100	100 102 101 101 101 101 101 101 101 101			E 0.00015 9 9 9 9 9 9 0.0001 9 0.00005 0 0.00005 0 0.00005 0 0 0.00005		0.00004 4 9 0.00002 0.00001 0 0 0 0	5000- 10000- 15000- 120000- 1220000- 125000-
				Ti	me to I	Reach 10	% Fractior	n Failing 1	for Mode



Likelihood Contours

💌 Parame	tric Estimate	e - Lognorn	nal			
Parameter location scale	Estimate 10.415732 2.797599	Std Error 0.56091492 0.50600249	Lower 95% 9.3163587 1.8058525	Upper 95% 11.515105 3.789346	Criterion -2 LogLikelihood AICc BIC	428.62269 432.94702 436.00045
			Mode	4		
		T50	= 33,3 σ = 2	81 cycl .80	es	
(—						









Lessons Learned

- Although the early failure mode showed a decreasing hazard rate, the wearout mode began appearing late in product life and had to be remediated.
- Analysis of multiple failure modes, called competing risk analysis, can provide valuable information for improving product reliability.
- Engineers can see the effect of eliminating different failure modes on product reliability.

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