

Reliability-HW3

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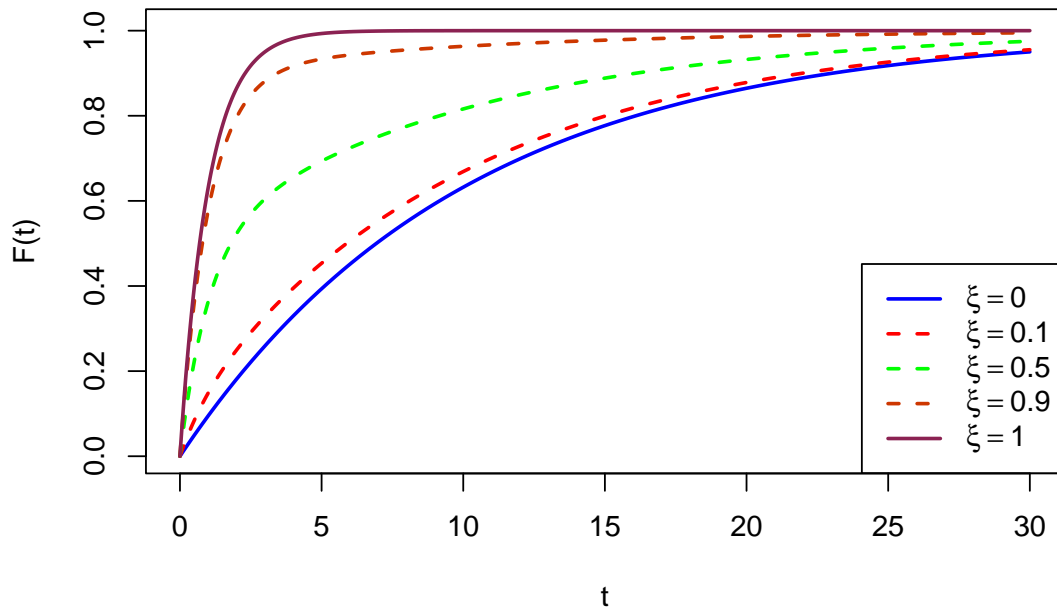
1.

(a)

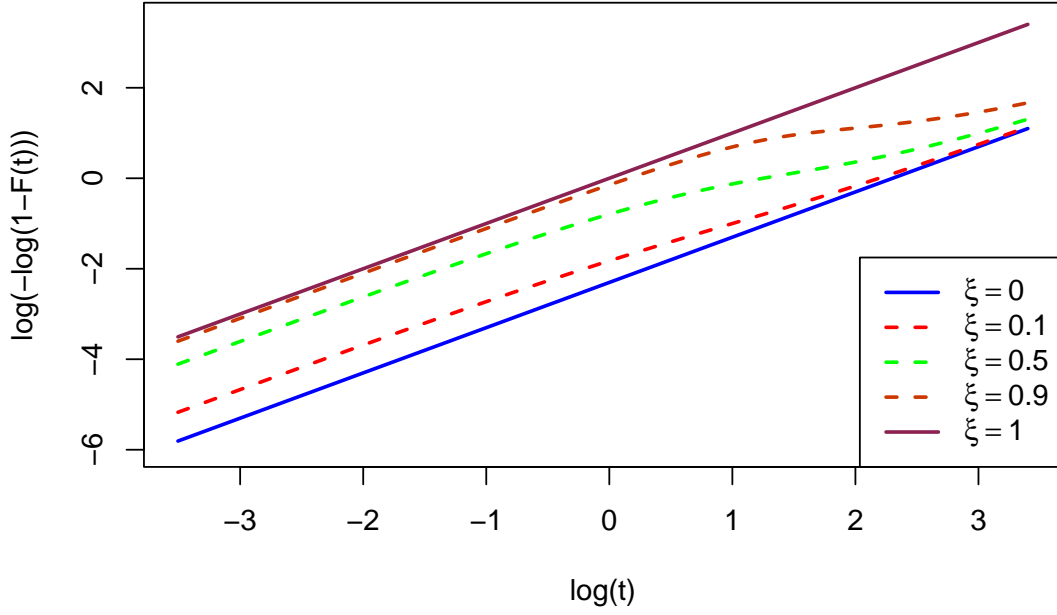
$$F(t) = \xi \int_0^t e^{-x} dx + (1 - \xi) \int_0^t \frac{1}{10} e^{-x/10} dx = \xi(1 - e^{-t}) + (1 - \xi)(1 - e^{-t/10})$$

, where $x > 0, \xi \in [0, 1]$.

(b)



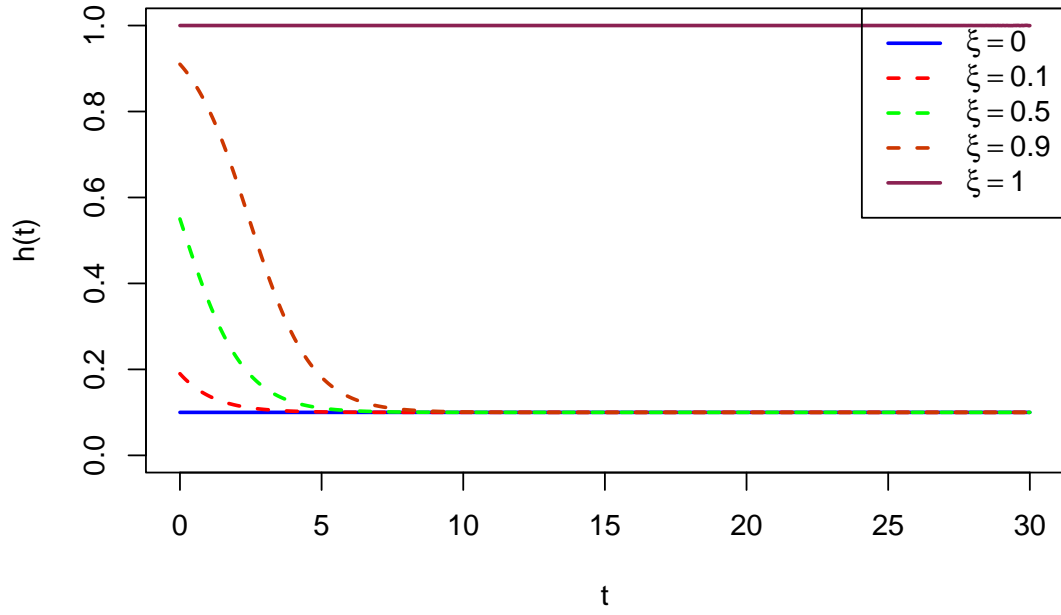
(c)



The shape of a mixture of two exponential distributions depends on the value of the mixing parameters ξ and the parameter $\theta = (\theta_1, \theta_2)$. See part (b), as ξ increases, the mixture distribution reflects which distribution the sources of variation is weighted on and relative to the pure exponential distribution, the shape of the mixture distribution is similar no matter what ξ is. See above Weibull plot, as ξ increases except for 0 and 1, we can find the shape of mixture distribution may be skewed toward the distribution with the higher mean. In contrast to the plot from part (b), Weibull distribution has a more flexible shape, which can capture more situations of the mixture distribution.

(d)

$$h(t) = \frac{f(t)}{1 - F(t)} = \frac{\xi e^{-t} + (1 - \xi) \frac{1}{10} e^{-t/10}}{1 - (\xi(1 - e^{-t}) + (1 - \xi)(1 - e^{-t/10}))}$$



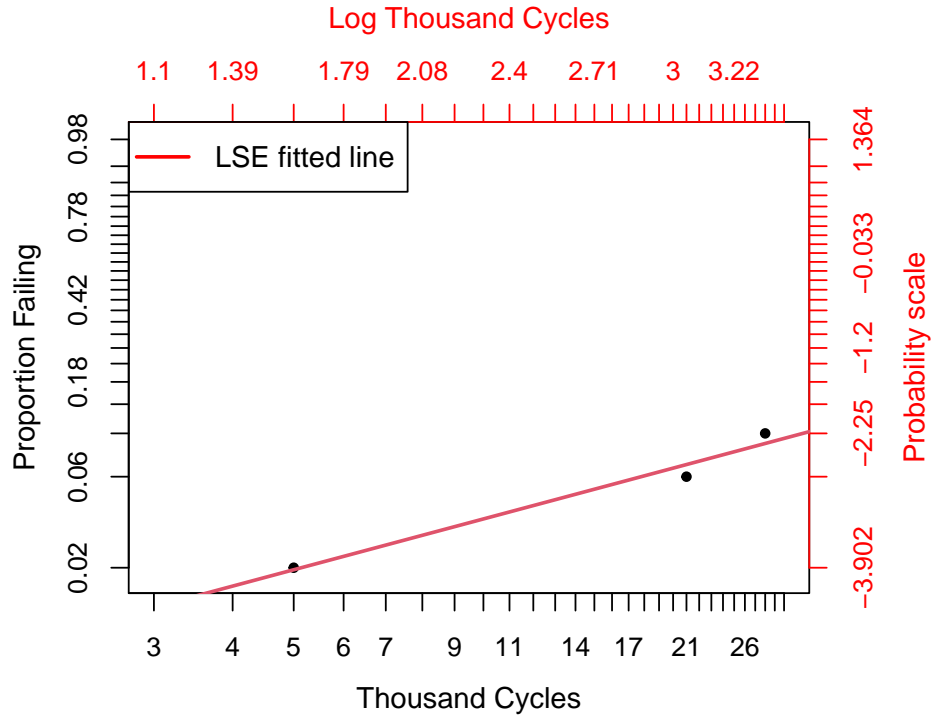
(e)

From part (c), we know that when $\xi=0$ or 1 , an exact straight line is presented, and when $\xi \neq 0$ or 1 , the line is not exact straight line. See the plot of part (d), the pure exponential presents the constant hazard function, and the impure exponential distribution presents the non-constant hazard function. It suggests to observe the Weibull plot to confirm whether it is a mixed exponential distribution.

2.

(a)

Here I attach LSE fitted line to observe whether there is a linear relationship:



(b)

With the results of the LSE, the probability scale before 10 thousand cycle can be estimated:

$$\log(-\log(1 - \hat{F}(t))) = \hat{\beta}_0 + \hat{\beta}_1 \log(t)$$

$$\hat{F}(t) = 1 - \exp(-\exp(\hat{\beta}_0 + \hat{\beta}_1 \log(t)))$$

The proportion: 0.03613221

(c)

According to Weibull distribution is log-location-scale family, its plotted point in probability plot (by (a)) should be close to the straight line. Since it looks very close, we subjectively think that Weibull distribution fits the data well.

(d)

The slope and intercept are estimated by LSE method from (a):

The estimated intercept coefficient: -5.377057

The estimated slope coefficient: 0.9010872

So,

$$\log(t_{(i)}) = \log(\theta) + \frac{1}{\beta} \log(-\log(1 - \hat{F}(t_{(i)}))),$$

where θ : scale parameter; β : shape parameter.

$$\hat{\beta} = 0.9010872$$

$$\hat{\theta} = \exp(-(-5.377057)/\hat{\beta}) = 390.4498$$

(e)

With the results of the LSE, the proportion before 100 thousand cycle can be estimated:

The proportion: 0.2540186

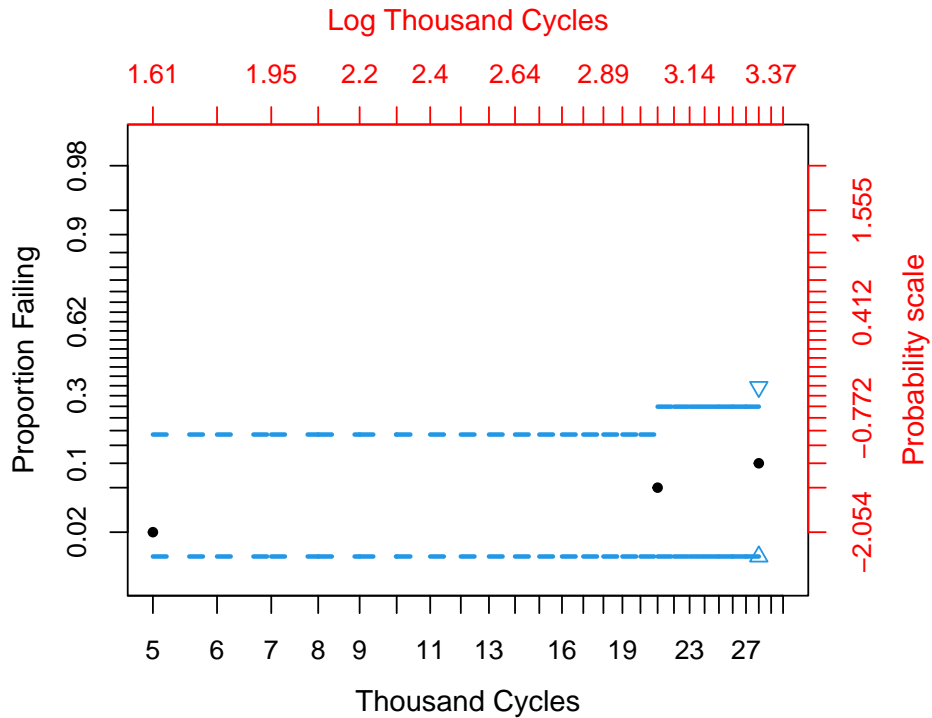
Estimating the proportion that is beyond the test time 30 thousand cycle can be challenging and subject to higher uncertainty! The estimated proportion is based on the linear structure, that is the same failure mechanism will continue to occur beyond the observed failure times. This extrapolation can be risky. Therefore, the estimate may not be reliable.

(f)

By KM estimation, the $100(1-\alpha)\%$ confidence band of $\Phi_{norm}^{-1}(F)$ is given by

$$[F_L(t), F_U(t)] = \hat{F}(t) \mp e_{(0.01, 0.95, 1-\alpha/2)} \text{se}_{\hat{F}}(t)$$

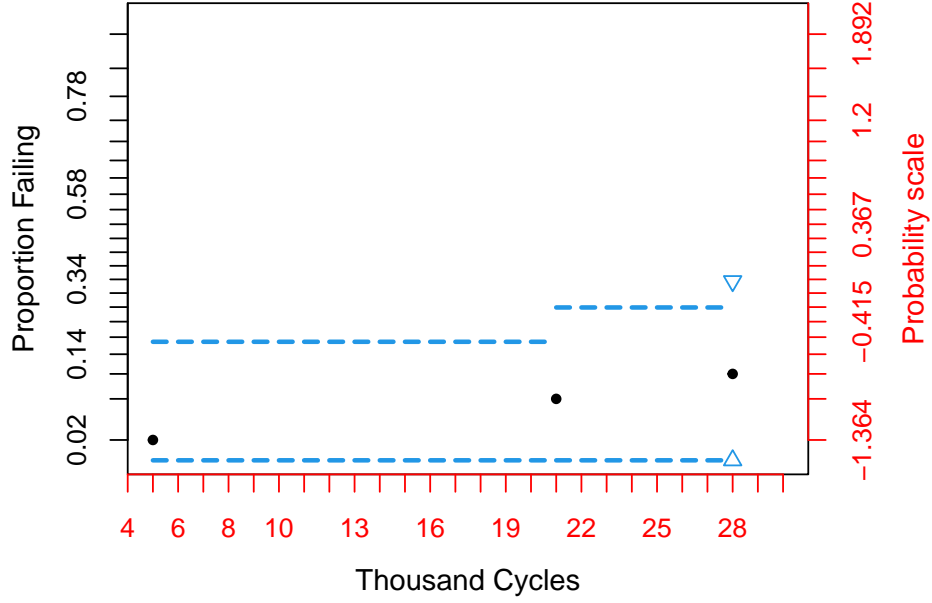
$\text{qnorm}[F_L(t), F_U(t)]$ is the confidence band.



(g)

Similar to part (f):

$-\log(-\log[F_L(t), F_U(t)])$ is the confidence band.



3.

(a)(b)

$$F(t) = 1 - [1 + (\frac{\log(t) - \mu}{\sigma})]^{-1}$$

$$[1 + (\frac{\log(t) - \mu}{\sigma})]^{-1} = 1 - F(t)$$

$$\frac{\log(t) - \mu}{\sigma} = \frac{1}{1 - F(t)} - 1 = \frac{F(t)}{1 - F(t)}$$

$$\text{Thus, } \log(t_p) = \mu + \frac{F(t_p)}{1 - F(t_p)}\sigma : \text{linear form}$$

For generating t_p , we can use inverse CDF method:

$$t_p = \exp(\mu) \times \exp(\sigma \frac{U_p}{1 - U_p}), \text{ where } U_p \sim U(0, 1)$$

(c)

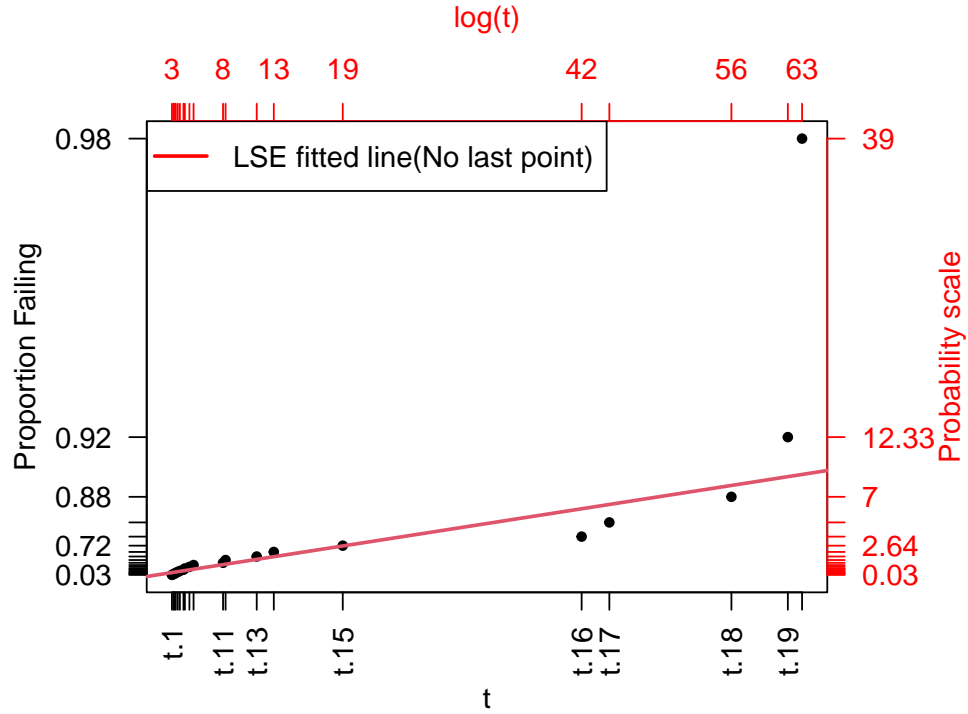
The generated t :

t.1	t.2	t.3	t.4	t.5	t.6	t.7	t.8	t.9	t.10
26.74	30.69	34.72	38.55	46.55	56.69	79.47	90.75	139.25	205.38

t.11	t.12	t.13	t.14	t.15
3311.34	4197.5	79008.82	395559.4	262937410

t.16	t.17	t.18	t.19	t.20
1.573509e+18	2.143892e+19	2.12417e+24	4.325007e+26	1.653688e+27

Next,draw the probability plot:



Since the probability scale of the last point is too large, I don't consider it to construct LSE. Thus, most of the points will be closer to the straight line.

(d)

$$\log(t_p) = \mu + \frac{F(t_p)}{1 - F(t_p)} \sigma$$

$$\frac{F(t_p)}{1 - F(t_p)} = -\frac{\mu}{\sigma} + \frac{1}{\sigma} \log(t_p)$$

From the above, in the probability plot,

- $-\frac{\mu}{\sigma}$: intercept
- $\frac{1}{\sigma}$: slope