

**Structural Reliability**  
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**Lecture –144**  
**Component Reliability - Time Defined (Part - 23)**

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### TTF statistics - estimations from test data

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**Example : censored reliability tests**

The TTF on an item is known to follow the exponential distribution. A test of 9 samples was designed to run until 3 items failed. The 3 failure instances were recorded as 144 hr, 182 hr and 243 hr, respectively. Estimate the mean time to failure of the item.

There are  $n$  samples with TTF  $X_i, i = 1, \dots, n$   
 $X_i$  are IID exponential RVs with parameter  $\lambda$ .  
 All  $n$  are put to test at time  $t = 0$   
 $r$  items fail at times  $\tau_1, \tau_2, \dots, \tau_r$ .

Time between successive failures:  
 $\Delta T_j$  - exponential with parameter  $(n - j + 1)\lambda$   
 mutually independent

If  $T \sim$  exponential ( $a\lambda$ )  
 then  $aT \sim$  exponential ( $\lambda$ )

Hence,  
 $(n - j + 1)\Delta T_j \sim$  exponential with parameter  $\lambda$

Thus we have  $r$  samples from an exponential distribution with parameter  $\lambda = 1/\mu$ :

$n\tau_1, (n-1)(\tau_2 - \tau_1), (n-2)(\tau_3 - \tau_2), (n-3)(\tau_4 - \tau_3), \dots, (n-r+1)(\tau_r - \tau_{r-1})$

Their sum  $\approx r \times \text{mean}$


Or,  $\tau_1 + \tau_2 + \tau_3 + \dots + \tau_r + (n-r)\tau_r = r \times \hat{\mu}$

$3 \times \hat{\mu} = 144 + 182 + 243 + (9-3)243 = 2027$   
 $\hat{\mu} = 676 \text{ hr}$

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We will be able to sort this out if we can answer this question which is if  $y$  is exponentially distributed with parameter  $\lambda$  then which random variable is distributed exponentially with  $\lambda$ . So, if we can answer that question we can make all the previous that those  $\Delta T_j$ s IID also by maybe doing some little bit of scaling. So, let us start with the basics. So, if  $y$  is exponential with parameter  $\lambda$  then the CDF is  $1 - \exp(-\lambda y)$ .

So, then if we were to divide  $y$  by  $a$  then the right hand side becomes  $1 - \exp(-\lambda y/a)$ . So, clearly then what is exponential with parameter  $\lambda$  it is a  $y/a$  times capital  $Y$  is the random variable that we are looking for. So, with that in mind that if  $y$  is exponentially  $\lambda$  then  $a y$  is exponential  $\lambda$  we can fix that that non-identical problem that we just mentioned.

So, all the  $T_j$ s as I said are exponential with parameter  $n - j + 1 \lambda$  they are mutually independent and we already know if  $y$  is exponentially a  $\lambda$  then  $a y$  is exponential  $\lambda$  hence  $n - j + 1 \Delta T_j$  has to be exponential with parameter  $\lambda$ . So, now we have scaled all these exponentials to have the same parameters. So, that gives us a way forward. So, then we can have our samples from an exponential distribution with the same parameter  $\lambda$  our  $\lambda$  is  $1$  by mean.

So, what are these our samples the  $r$  samples are  $n$  times  $\tau_1$   $n - 1$  times  $\tau_2 - \tau_1$ . So, all these differences are basically the time between failures. So,  $\Delta T$  is the time between successive failures. So, that is how we have entire one which is the time to the first failure  $n$   $1$   $n - 1$  times  $\tau_2 - \tau_1$  which is the time between the first and second failures  $n - 2$  times  $\tau_3 - \tau_2$  which is the time between second and third phase and so on until  $n + r + 1$   $\tau_r - \tau_{r - 1}$ .

So, each of these terms is a realization of an exponential random variable with the same parameter  $\lambda$ . So, their sum will be  $r$  times mean and here we are going towards an estimation problem which is a bit too simple we are not going for method of maximum likelihood or interval estimation as we have said but we want to get a good answer quickly. So, this simplifies to all that all the  $\tau_1$  through  $r$  the sum of all  $\tau_1$  all the up to  $\tau_r$  and then  $n - r$  times  $\tau_r$ .

So, the last recorded failure time that is equal to  $r$  times the estimated mean and if now we plug in the three values  $r$  is equal to three in this example we get a mean an estimated mean of about 676 hours. So, this is this is quite good uh. So, what we did was we just had three failures and three recorded instances we started with nine samples we did not wait for all nine of them to fail. And now because we knew that the time to failure is exponential for each of them we were able to go through the mathematics and come up with a good estimate of the mean time to failure.

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## TTF statistics - estimations from test data

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Functional relation:  $R = \exp(-\lambda t)$   
 linearizing:  $\ln R = -\lambda t$


What if we solved the problem with regression:  
 $\min \sum_{i=1}^r [\ln R_i + \lambda t_i]^2 \text{ wrt } \lambda$   
 $\Rightarrow 2 \sum_{i=1}^r [\ln R_i + \lambda t_i] t_i = 0$

We have  $r$  point estimates of reliability  $R_i$  at times  $t_i$   
 of an item with exponential TTF with parameter  $\lambda = 1/\mu$

$R_i = [8/9 \quad 7/9 \quad 6/9]$   
 $t_i = [144 \quad 182 \quad 243]$

$\Rightarrow \hat{\lambda} = \frac{\sum_{i=1}^r t_i \ln R_i}{\sum_{i=1}^r t_i^2}$   
 $= \frac{-0.9811433}{108} \text{ hr}^{-1}$   
 $\hat{\mu} = 700 \text{ hr}$

Obtain the best fit line



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Now we could wonder that with this problem statement that we started with what if we instead of going with this independent exponentials which by the way would lead to a gamma distributed or an Erlang distributed random variable for the sum and we could show later on that this estimate is biased because the way it is been computed but we will not go into those details of more advanced statistical concepts.

But let us come back to this and what if we did not go through all that trouble and we just solve this problem with regression what do I mean by that I mean is that we have basically what we have is our point estimates our number of point estimates of the reliability  $r_i$  at times  $t_i$ . So, three times have been recorded we have three reliability estimates corresponding to that of an item with exponential TTF that is important that is very helpful.

So, can we not estimate lambda from just this information can we not fit a curve can we not do regression analysis. So, we have  $r = 8$  by  $9$ ,  $7$  by  $9$  and  $6$  by  $9$  and the corresponding T's are 144, 182 and 243. So, we need to obtain the best fit line between these three points uh. So, what is the step? So, we will just go step by step as we do in a linear regression analysis. So, we have a function relationship which is  $r$  equals exponential minus lambda T we need to linearize and that is  $\log r$  is minus lambda T and then we just minimize the squared error between the theoretical value and the observed value.

So, that would be minimize sum of the square deviations there is a plus sign there because of the negative  $\lambda T$  at the previous line. So, we need to minimize this sum with regards to  $\lambda$ . So, now we are in familiar territory we can get the expression which leads to this minimization and it turns out that the estimated  $\lambda$  would be the ratio of the sum of  $T \log r$  over the sum of  $T$  squared.

So, this would be another way of estimating the mean time to failure. And we can solve this with the numbers that we have and it is not too bad the mean is 700, so, very much within the estimate that we did from the independent time between failures approach.