

## WNeiburll

## Distribution

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Dept of Mechanical \& Industrial Engineering
The University of Iowa

MLE: Weibull

$$
L(t ; k, u)=\frac{k^{n}}{u^{n k}}\left[\prod_{i=1}^{n} t_{i}^{k-1}\right] \exp \left\{-u^{-k} \sum_{i=1}^{n} t_{i}^{k}\right\}
$$

We wish to choose values of $\mathrm{u} \& \mathrm{k}$ which maximize L (or equivalently, the logarithm of L), i.e., which make the observed values of $t$ as large as possible!

The log-likelihood function is

$$
\ln L(t ; k, u)=n \ln k-n k \ln u+(k-1) \sum_{i=1}^{n} \ln t_{i}-u^{-k} \sum_{i=1}^{n} t_{i}^{k}
$$

## Weibull Distribution:

pdf: $\quad f(t)=\frac{k}{u}\left(\frac{t}{u}\right)^{k-1} \exp \left\{-\left(\frac{t}{u}\right)^{k}\right\}$
Suppose $t_{1}, t_{2}, \ldots t_{n}$ are times to failure of a group of $n$ mechanisms.

The likelihood function is

$$
\begin{aligned}
L(t ; k, u) & =\prod_{i=1}^{n} \frac{k}{u}\left(\frac{t_{i}}{u}\right)^{k-1} \exp \left\{-\left(\frac{t_{i}}{u}\right)^{k}\right\} \\
& =\frac{k^{n}}{u^{n k}}\left[\prod_{i=1}^{n} t_{i}^{k-1}\right] \exp \left\{-u^{-k} \sum_{i=1}^{n} t_{i}^{k}\right\}
\end{aligned}
$$

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The optimality conditions for the maximum of the loglikelihood function are

$$
\left\{\begin{array}{l}
\frac{\partial}{\partial u} \ln L(t ; u, k)=0 \\
\frac{\partial}{\partial k} \ln L(t ; u, k)=0
\end{array}\right.
$$

This gives us a pair of nonlinear equations in two unknowns ( $u \& k$ ):

$$
\left\{\begin{array}{l}
-\frac{n \hat{k}}{\hat{u}}+\hat{k} \hat{u}^{-\hat{k}-1} \sum_{i=1}^{n} t_{i}^{\hat{k}}=0 \\
\frac{n}{\hat{k}}-n \ln \hat{u}+\sum\left(\frac{t_{i}}{\hat{u}}\right)^{\hat{k}} \ln \left(\frac{t_{i}}{\hat{u}}\right)=0
\end{array}\right.
$$

But the left side of the first equation can be factored:

$$
-\frac{n \hat{k}}{\hat{u}}+\hat{k} \hat{u}^{-\hat{k}-1} \sum_{i=1}^{n} t_{i}^{\hat{k}}=0 \Rightarrow \hat{k} \hat{u}^{-1}\left[-n+\hat{u}^{-\hat{k}} \sum_{i=1}^{n} t_{i}^{\hat{k}}\right]=0
$$

Since the first factor cannot be zero, we set the second factor equal to zero and solve for $\hat{u}$ in terms of $\hat{k}$ :

$$
\hat{u}=\left(\frac{1}{n} \sum_{i=1}^{n} t_{i}^{\hat{k}}\right)^{1 / 1}
$$

Eliminating $\hat{u}$ in the second equation by substituting the first, we get the following nonlinear equation in $\hat{k}$ alone:

$$
\frac{1}{\hat{k}}-\frac{\sum_{i=1}^{n} t_{i}^{\hat{k}} \ln t_{i}}{\sum_{i=1}^{n} t_{i}^{\hat{k}}}+\frac{1}{n} \sum_{i=1}^{n} \ln t_{i}=0
$$

This can now be solved by, for example, the secant method.

The CDF of the Weibull distribution is

$$
\left.F(t ; k, u)=1-\exp \left\{-\left(\frac{t}{u}\right)^{k}\right)\right\}
$$

and so the likelihood function is

$$
\begin{aligned}
L(t ; k, u) & =\left[\exp \left\{-\left(\frac{\tau}{u}\right)^{k}\right\}\right]^{n-r} \times \prod_{i=1}^{r} \frac{k}{u}\left(\frac{t_{i}}{u}\right)^{k-1} \exp \left\{-\left(\frac{t_{i}}{u}\right)^{k}\right\} \\
& =\frac{k^{r}}{u^{n k}}\left[\prod_{i=1}^{r} t_{i}^{k-1}\right] \exp \left\{-u^{-k}\left[\sum_{i=1}^{r} t_{i}^{k}+(n-r) \tau^{k}\right]\right\}
\end{aligned}
$$

## Maximum Likelihood Estimation

 with "censored" dataSuppose that an experiment was terminated at time $\tau$ after only $r$ of the $n$ units in a lifetest had failed. This is accounted for by defining the likelihood as

$$
L(t, \theta)=[1-F(\tau ; \theta)]^{n-r} \times \prod_{i=1}^{r} f\left(t_{i} ; \theta\right)
$$

The log-likelihood function is therefore

$$
\ln L(t ; \theta)=(n-r) \ln [1-F(t ; \theta)]+\sum_{i=1}^{r} \ln f\left(t_{i} ; \theta\right)
$$

The log-likelihood function is

$$
\ln L(t ; k, u)=r \ln k-n k \ln u+(k-1) \sum_{i=1}^{r} \ln t_{i}-u^{-k}\left[\sum_{i=1}^{r} t_{i}^{k}+(n-r) \tau^{k}\right]
$$

The optimality conditions for a maximum of the log-likelihood at $(\hat{k}, \hat{u})$ are

$$
\left\{\begin{array}{l}
\frac{\partial}{\partial u} \ln L(t ; \hat{u}, \hat{k})=0 \\
\frac{\partial}{\partial k} \ln L(t ; \hat{u}, \hat{k})=0
\end{array}\right.
$$

A result similar to the uncensored case can be derived:

$$
\hat{u}=\left(\frac{\left.\sum_{i=1}^{r} t_{i}^{\hat{k}}+(n-r) \tau^{\hat{k}}\right)^{1 / \hat{k}}}{n}\right)
$$

and

$$
\frac{1}{\hat{k}}-\frac{\sum_{i=1}^{r} t_{i}^{\hat{k}} \ln t_{i}+(n-r) \tau^{\hat{k}} \ln \tau}{\sum_{i=1}^{r} t_{i}^{\hat{k}}+(n-r) \tau^{\hat{k}}}+\frac{1}{r} \sum_{i=1}^{r} \ln t_{i}=0
$$

This second equation can be solved for $\hat{k}$ by the secant method, and
then $\hat{k}$ can be used to calculate $\hat{u}$ by the first equation.

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A plot of Y vs X , obtained by the transformations:
$Y=\log \log \frac{1}{R(t)}$ where $R(t)$ is the observed fraction of the
devices which have survived until time $t$, and
$X=\log t$
should be a line if the Weibull model were to fit the data perfectly.

EXAMPLE: Twenty devices are tested simultaneously until 500 days have passed, at which time the following failure times (in days) have been recorded:

$$
\begin{array}{r}
31.5 \\
74.0 \\
87.5 \\
100.1 \\
103.3 \\
181.9 \\
279.9 \\
297.1 \\
462.5 \\
465.4
\end{array}
$$

Estimate the lifetime for which the device is $90 \%$ reliable.

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## LEAST SQUARES REGRESSION RESULTS:


$u$ (scale parameter) $=653.504$
$k$ (shape parameter) $=0.908313$
so that
mean $=\quad 630.396$
standard deviation $=754.336$

Note: this is determined by minimizing the sum of the squared errors in the linearized version of $F(t)=1-e^{-(t / u)^{k}}$, namely $y=k x-k \ln u$ where $x=\ln t \& y=\ln \ln \frac{1}{R(t)}$,
rather than in the original equation!

If we use these parameters found by linear regression, the reliability function would have the values:

| $\frac{t}{}$ | $\frac{F(t)}{0.01}$ | $\frac{1-F(t)}{0.99}$ |
| :--- | :--- | ---: |
| 4.12824 | 0.02 | 0.98 |
| 8.90435 | 0.03 | 0.97 |
| 13.993 | 0.04 | 0.96 |
| 19.3163 | 0.05 | 0.95 |
| 24.837 | 0.06 | 0.94 |
| 30.5337 | 0.07 | 0.93 |
| 36.3925 | 0.08 | 0.92 |
| 42.4042 | 0.09 | 0.91 |
| 48.5623 | 0.10 | 0.90 |
| 54.8622 | 0 |  |

Hence, according to this model, $90 \%$ of the devices should be operating at 54.8 (approximately 55) days.

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## SECANT METHOD



If our first two "guesses" at the value of $k$ are 0.5 and 2.0, then we determine that

$$
g(0.5)=1.13739 \& \text { and } g(2.0)=-0.618085 .
$$

## Maximum Likelihood result:

Solving the nonlinear equation for $k$ :

$$
g(k)=\frac{1}{\hat{k}}-\frac{\sum_{i=1}^{r} t_{i}^{\hat{k}} \ln t_{i}+(n-r) \tau^{\hat{k}} \ln \tau}{\sum_{i=1}^{r} t t_{i}^{\hat{k}}+(n-r) \tau^{\hat{k}}}+\frac{1}{r} \sum_{i=1}^{r} \ln t_{i}=0
$$



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The secant joining the two points on the graph of g cross the $k$ axis at 1.47187 .

We then repeat, with the 2 improved "guesses" $k=0.5$ and $\mathrm{k}=1.47187$.

| SECANT METHOD RESULTS: | k | error | Once we determine the value of $\hat{k}$ which maximizes the likelihood function, then the corresponding value of the parameter $\hat{u}$ is found by |  |
| :---: | :---: | :---: | :---: | :---: |
|  | 0.5 | 1.13739 |  |  |
|  | 2.0 | -0.618085 |  |  |
|  | 1.47187 | -0.397478 |  |  |
|  | 0.5203 | 1.05148 | $\hat{u}=\left(\frac{\left.\sum_{i=1}^{r} t_{i}^{\hat{k}}+(n-r) \tau^{\hat{k}}\right)^{1 / \hat{k}}}{n}\right)$ |  |
|  | 1.21083 | -0.217582 |  |  |
|  | 1.09244 | -0.108608 |  |  |
|  | 0.974445 0.996528 | 0.025006 -0.00227829 |  |  |
|  | 0.994684 | -0.0000438242 |  |  |
|  | 0.994648 | $7.83302 \mathrm{E}^{-8}$ |  |  |
|  | 0.994648 | ${ }^{-2} 2.68896 \mathrm{E}^{-12}$ |  |  |

## Maximum Likelihood result:

$u$ (scale parameter) $=710.339$,
$k($ shape parameter $)=0.994648$

| t | $\mathrm{F}(\mathrm{t})$ | $1-\mathrm{F}(\mathrm{t})$ |
| ---: | ---: | ---: |
| 6.9646 | 0.01 | 0.99 |
| 14.0526 | 0.02 | 0.98 |
| 21.2337 | 0.03 | 0.97 |
| 28.5026 | 0.04 | 0.96 |
| 35.8579 | 0.05 | 0.95 |
| 43.2993 | 0.06 | 0.94 |
| 50.8272 | 0.07 | 0.93 |
| 58.4427 | 0.08 | 0.92 |
| 66.1467 | 0.09 | 0.91 |
| 73.9408 | 0.10 | 0.90 |

According to this model, then, $90 \%$ of the devices should be operating at 73.94 (approximately 74) days.

