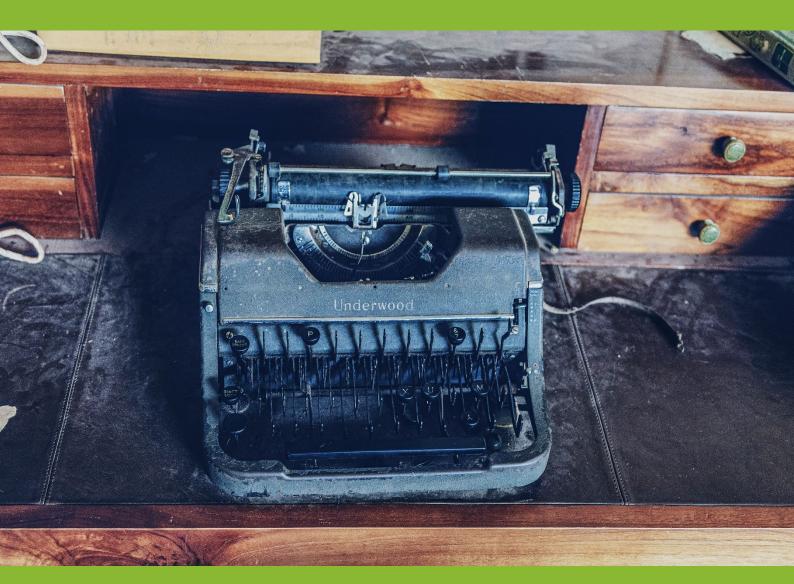
# Reliability, Availability and Maintainability (RAM) Practitioners Course

If reliability engineering isn't making you money, then you are doing it wrong



# Workbook

Presented by

# **Dr Chris Jackson**

Reliability engineer



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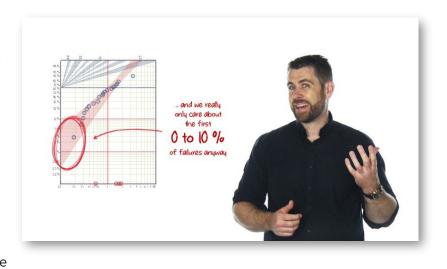
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# Who is your teacher? ... Dr Chris Jackson

Dr Jackson holds a Ph.D. and MSc in Reliability Engineering from the University of Maryland's Center of Risk and Reliability. He also holds a BE in Mechanical Engineering, a Masters of Military Science from the Australian National University (ANU) and has substantial experience in the field of leadership, management, risk, reliability and maintainability. He is a Certified Reliability Engineer (CRE) through the American Society of Quality (ASQ) and a Chartered Professional Engineer (CPEng) through Engineers Australia.

Dr Jackson is the cofounder of www.is4.com online learning, was the inaugural director of the University of California, Los Angeles (UCLA's) Center of Reliability and Resilience Engineering and the founder of its Center for the Safety and Reliability of **Autonomous** Systems (SARAS). He has had an extensive career as a Senior Reliability Engineer in



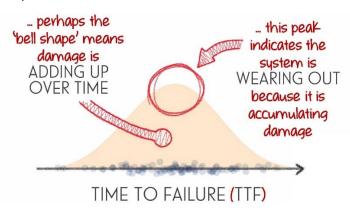
Australian Defence Force, and is the Director of Acuitas Reliability Pty Ltd.

He has supported many complex and material systems develop their reliability performance and assisted in providing reliability management frameworks that support business outcomes (that is, make more money). He has been a reliability management consultant to many companies across industries ranging from medical devices to small satellites. He has also led initiatives in complementary fields such as systems engineering and performance-based management frameworks (PBMFs). He is the author of two reliability engineering books, co-author of another, and has authored several journal articles and conference papers.

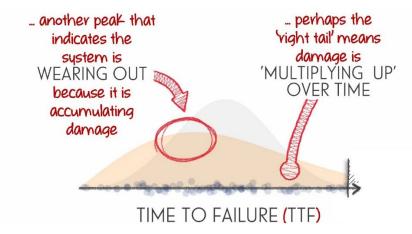
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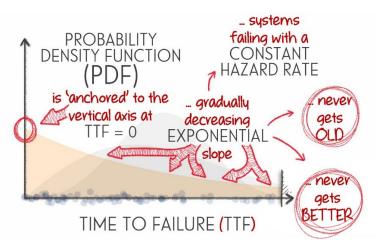
The **Weibull distribution** is talked about a *lot* in **reliability engineering**. And there is a really good reason why. But first, some revision. We know that the **probability density function (PDF)** defines a curve the defined the relatively likelihood of a **random variable** having a certain value. That means, random variable values are more likely where the **PDF curve** is the highest (as in the illustration below). And the shape of that curve can tell us a lot about how failure is occurring ... if the random variable in question is **time to failure (TTF)**.



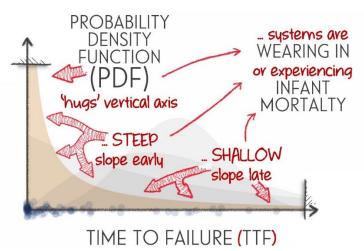
If we become really proficient, at this we can not only tell if our system is wearing out, but HOW it is wearing out. Another PDF curve with a 'wear-out' peak is illustrated below.



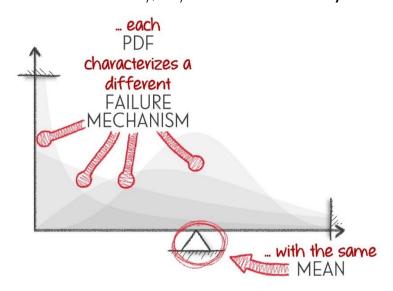
Then there are **PDF** curves with this shape ...



#### And PDF curves with this shape ...



All the **PDF curves** in the previous illustrations (each one modelling a different **failure mechanism**) were all specific examples of the **Weibull distribution**. And ... they all had the same **mean** or **mean time between failure (MTBF)**. This further illustrates that just understanding the mean of the **random process** that is **failure** reveals very, very little about the **reliability characteristics**.



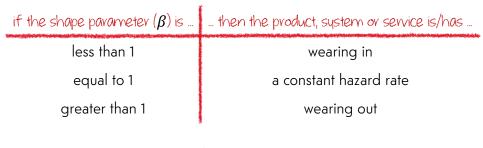
The ability to model lots of different, well known failure mechanisms is primarily why the **Weibull distribution** is so popular. If you don't know what your data 'says' about the nature of **failure**, conducting what we call a '**Weibull analysis**' can not only find (perhaps) a really good **TTF** model for your product, system or service, but it can tell you where you need to look in order to improve **reliability**.

#### ... Weibull distribution PARAMETERS

The parameters of the **Weibull distribution** are talked about ... a lot. Beyond helping us define the shape of the **PDF** and **reliability curves**, they tell us a lot about the nature of **failure**.

shape parameter - 
$$\beta$$

The **shape parameter** is represented by the lowercase Greek letter beta ( $\beta$ ). The **shape parameter** tells us if our product, system or service is wearing in, wearing out, or has a constant **hazard rate** 

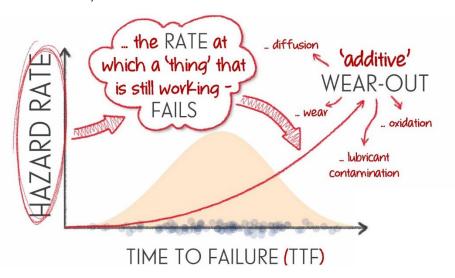


scale parameter –  $\eta$ 

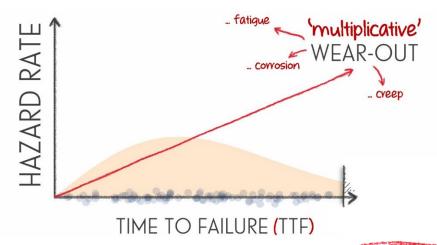
The scale parameter is represented by the lowercase Greek letter eta ( $\eta$ ). The scale parameter tells represents an indicative size of the **Weibull random variable** in question. The scale parameter is the 63<sup>rd</sup> percentile (which is just a value that makes it easy for us to express the **probability distribution functions**).

#### ... some examples

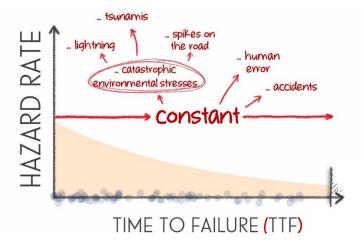
Because the **Weibull distribution** can model lots of different **failure mechanisms**, it (very understandably) it can look very similar to lots of other **probability distribution**. For example, the **PDF** and **hazard rate curves** for a **Weibull distribution** with a **shape parameter of 3.5** are illustrated below. This closely mimics the **normal distribution**.



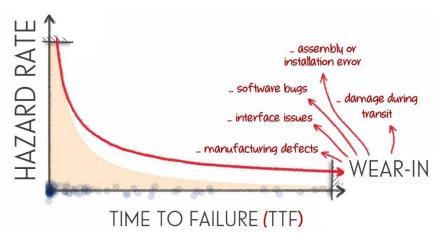
The PDF and hazard rate curves for a Weibull distribution with a shape parameter of 2 are illustrated below. This closely mimics some lognormal distributions.



The **PDF** and **hazard rate curves** for a **Weibull distribution** with a **shape parameter of 1** are illustrated below. This is the **exponential distribution**.

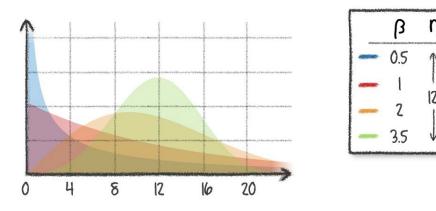


And finally, the **PDF** and **hazard rate curves** for a **Weibull distribution** with a **shape parameter of 0.4** are illustrated below. This does not mimic a specific distribution, but has been found to accurately model wear-in failure mechanisms.



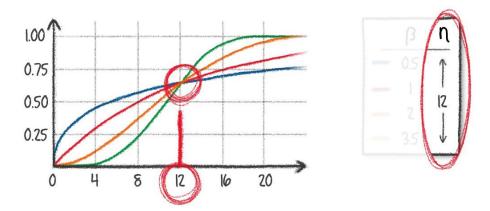
#### PROBABILITY DENSITY FUNCTION (PDF)

The equations and formulae for the **PDF** and other functions and statistics based on the **Weibull distribution** (including the **mean**) are at the end of this chapter. The most important thing to understand when dealing with **PDFs** is knowing what they represent in terms of failure. Four example **Weibull distribution PDF curves** are illustrated below. They all have different **shape parameters**, but identical **scale parameters** (12).



#### CUMULATIVE DISTRIBUTION FUNCTION (CDF)

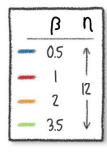
Four example **Weibull distribution CDF curves** are illustrated below. Each **CDF curve** corresponds to one of the **PDF curves** above. You can see here that the **Weibull distribution CDF** (like all other **CDFs**) starts at 0 and approaches 1 (or 100 %) as **TTF** increases.

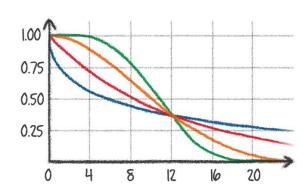


You can also see that regardless of the **shape parameter** ( $\beta$ ), all **CDF curves** above pass through the same point. This point happens to occur when the **TTF** is equal to the **scale parameter** ( $\eta$ ). This is a very useful characteristic that can be exploited when we are trying to identify **parameters** for a given failure process. We will explore that later.

#### RELIABILITY FUNCTION

Four example **Weibull distribution reliability curves** are illustrated below. Each **reliability curve** corresponds to one of the **PDF and CDF curves** above. You can see here that the **Weibull distribution reliability curve** (like all other **reliability curves**) starts at 1 (or 100 %) and approaches 0 as **TTF** increases.





You can see that again, all these curves pass through the same point when the TTF is equal to the scale parameter ( $\eta$ ). There is a reason for this. The Weibull distribution reliability function is ...

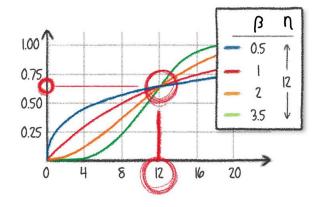
$$R(t) = e^{-\left(\frac{t}{\eta}\right)^{\beta}}$$

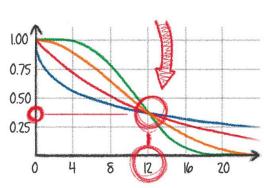
Now we can use this equation to understand 'what is happening' when our system has reached its characteristic life ...

$$R(\eta) = e^{-\frac{\eta}{\eta}} = e^{-(1)^{\frac{3}{2}}} = 36.788 ... \%$$

... when the system has reached its CHARACTERISTIC LIFE ( $t = \eta$ ) ...

Which is why we always see this ...



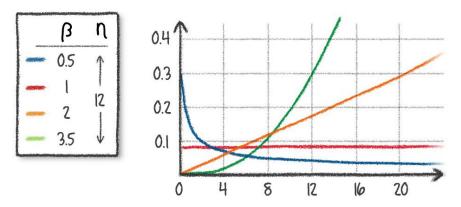


#### HAZARD RATE

One way of looking at the **hazard rate** is by seeing it as the rate at which working (non-failed) systems are **failing**.



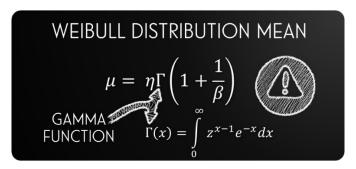
Four example **Weibull distribution hazard rate curves** are illustrated below. Each **hazard rate curve** corresponds to one of the **PDF**, **CDF** and reliability curves above.



This is where you can see the way in which the **Weibull distribution** can model all sorts of different types of **failures**, with it possible to have decreasing, constant and increasing **hazard rates**. But ... the **Weibull distribution** cannot model products, systems or services that have a little bit of each. You need to create a **system reliability model** for that.

#### **MEAN**

The Weibull distribution mean is a little complicated to calculate.



This is obviously difficult (impossible) to calculate by hand. Microsoft Excel formulae that can help you find the **Weibull mean** are ...

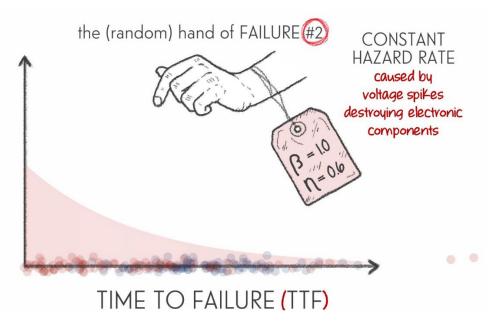


#### ... a quick look at 'WEIBULL ANALYSIS'

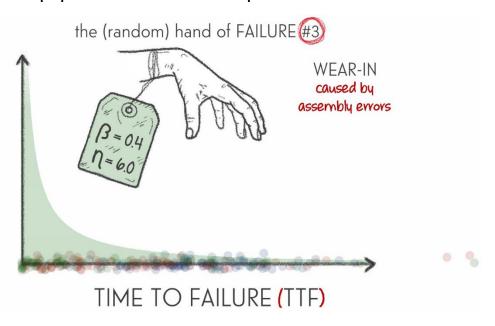
Let's go back to our 'random hand of failure.' It can represent a **reliability test**, actual use, historical data or anything else where your product, system or service is allowed to operate until **failure**. Let's say we conduct a lifetime test on a **failure process** described by the **Weibull distribution** with a **shape parameter** of 3.5 and a **scale parameter** of 1.0.



Let's say we conduct another lifetime test on a **failure process** described by the **Weibull distribution** with a **shape parameter** of 1.0 and a **scale parameter** of 0.6.

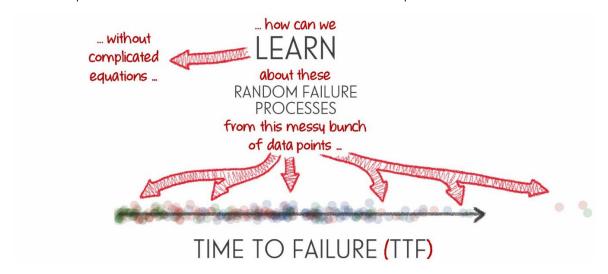


Let's now say we conduct another lifetime test on a **failure process** described by the **Weibull distribution** with a **shape parameter** of 0.4 and a **scale parameter** of 6.0.

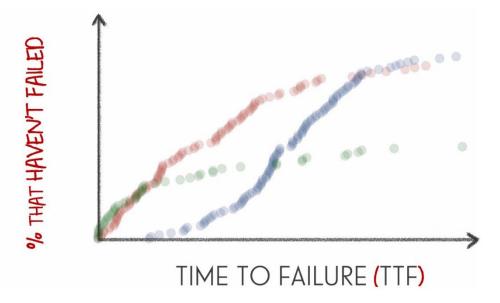


We can see that each process will result in test data which looks different to every other piece of test data.

But ... in the real world we don't know what those 'random hands of failure' represent. We don't know that the shape parameter of the first process is 3.5. We don't know that the scale parameter of the third process is 6.0. All we have is a bunch of 'dots' that represent the TTFs we observed.

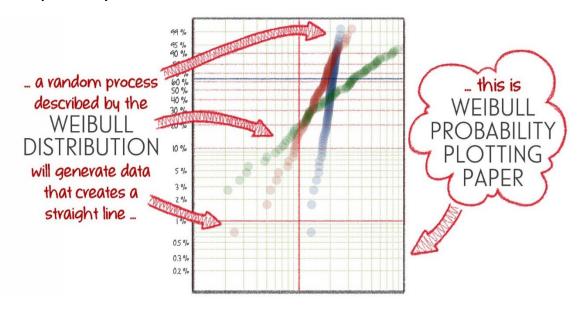


We can get a better idea of what is happening by plotted these data points on a chart where the vertical axis represented the percentage of things that have **failed**. This makes a 'smooth' line that helps us better visualize the 'nature' of **failure**. And we can better 'see' the three separate failure processes below.

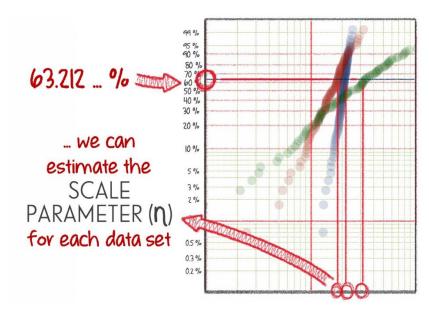


While these lines are 'smooth' ... which usually means it is easier to spot patterns ... it is simply not possible to use it to work out what the hypothetical **Weibull distribution** parameters might be. Or if a Weibull distribution is the right distribution to use.

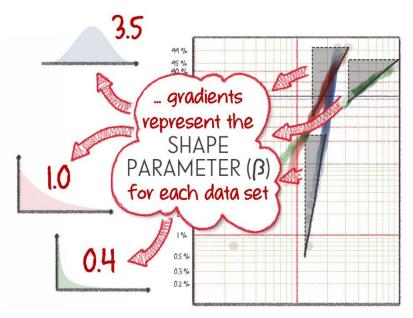
The key to a lot of data analysis activities is to make a 'straight line.' We can manipulate the **Weibull CDF** a little bit to create a 'special' chart that is often referred to as **Weibull probability plotting paper**. The reason we do this is that any data that is described by a **Weibull distribution** will create a straight line on **Weibull probability plotting paper**. The data from the three **failure processes** are illustrated on **Weibull plotting paper below**. Note that the vertical axis represents the **CDF** or **failure probability**.



And that straight line can be used to estimate our **Weibull distribution parameters** directly. We know that because the **CDF** of any **Weibull** will be 63.212 % at the **characteristic life**. So we can find the **scale parameter** from the **Weibull plot** where the **CDF** is 63.212 %.



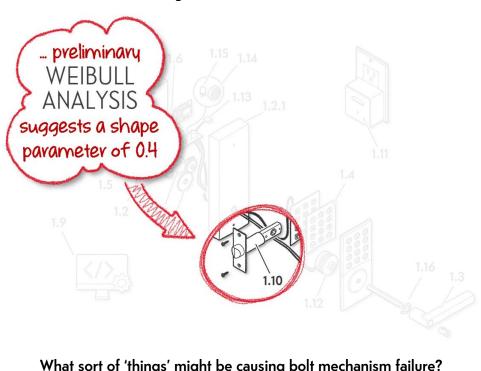
We can use the slopes of the lines of best fit to estimate the **shape parameter** of each process.



Weibull probability plotting is a lot easier than you might expect. You can do it on specially printed paper or using software to automatically generate Weibull plots. Neither approach requires you to use any equations ... but you do need to really understand what every Weibull **plot** represents.

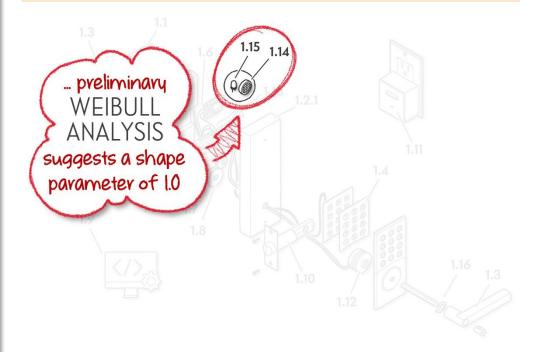


#### Let's go back to the smart lock.



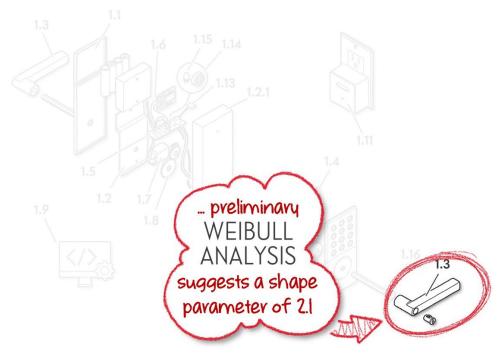
What sort of 'things' might be causing bolt mechanism failure?

What sorts of things MIGHT we be able to do to improve the reliability of our bolt mechanism?



What sort of 'things' might be causing LED or speaker failure?

What sorts of things MIGHT we be able to do to improve the reliability of our LED and speaker?



What sort of 'things' might be causing handle failure?

What sorts of things MIGHT we be able to do to improve the reliability of our handle?

#### WEIBULL DISTRIBUTION

#### **Equations and Formulae**

#### **Parameters**

 $\beta$  - shape parameter

... defines the nature of the random failure process we are modelling (wear-in, constant hazard rate, wear-out, and different types therein)

 $\eta$  – scale parameter

... defines the magnitude of the random variable (TTF)

#### **Probability Density Function (PDF)**

$$f(t) = \frac{\beta t^{\beta-1}}{n^{\beta}} e^{-\left(\frac{t}{\eta}\right)^{\beta}} \dots$$
 = WEIBULL.DIST([t],[ $\beta$ ],[ $\eta$ ],FALSE)

#### **Cumulative Distribution Function (CDF)**

$$F(t) = 1 - e^{-\left(\frac{t}{\eta}\right)^{\beta}}$$
...  $\mathbf{X} = \text{WEIBULL.DIST}([t],[\beta],[\eta],\text{TRUE})$ 

#### **Reliability Function**

$$R(t) = e^{-\left(\frac{t}{\eta}\right)^{\beta}}$$
... =1 - WEIBULL.DIST([t],[\beta],[\eta],TRUE)

#### Hazard rate

$$\lambda(t) = \frac{\beta t^{\beta-1}}{\eta^{\beta}} \dots = [\beta]^*([t]^{([\beta]-1))/([\eta]^{\alpha} [\beta])$$

#### Mean

$$\mu = \eta \Gamma \left( 1 + \frac{1}{\beta} \right) \dots = [\eta]^* \text{GAMMA} (1 + (1/[\beta]))$$

$$\dots = [\eta]^* \text{EXP} (\text{GAMMALN} (1 + (1/[\beta])))$$

... where is the gamma function, which is defined as  $\Gamma(x) = \int_0^8 z^{x-1} e^{-x} dx$ 

One of the approaches in volved using a **probability distribution** to model **TTF**. And the **Weibull distribution** is really AWESOME as it can model wear-in, wear-out and constant **hazard rate failure processes**.

That means you can gather data and use the **Weibull distribution** to identify how your product, system or service is failing. Which can help you make better business decisions, and (perhaps even better) now what you need to do to make it better. And **probability plotting** can really help.

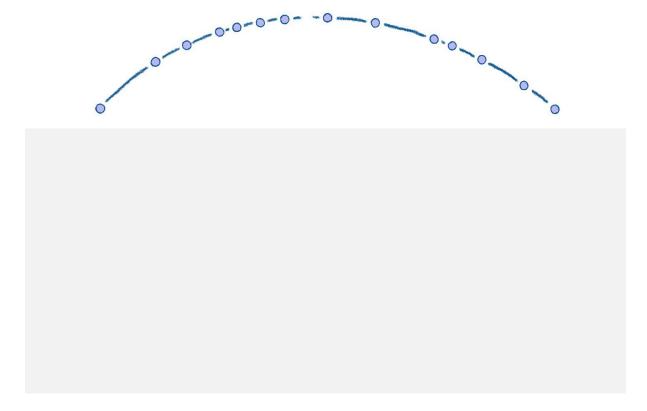
#### ... human beings are great at seeing straight lines

And we use this characteristic when it comes to **probability plotting**. For example, we can really easily see that the following points are in a straight line.

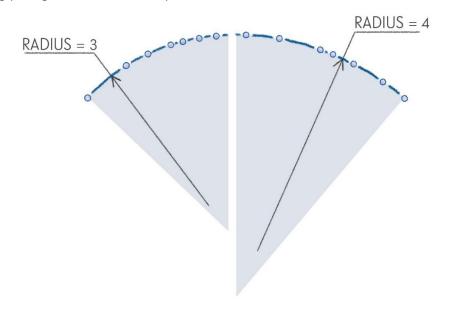
We can also see that he following points are not in a straight line.

In fact, we can easily that there are two straight line segments. And if we get practiced, some of us might be able to estimate that the angle deviation of the two line segments above is 12 degrees.

But, can you identify the two different curves below?



This seemingly 'single' curve is made up of two, distinct and different curves.



It is very, very difficult for us to characterize lines that aren't straight just by looking at them. If we can make lines straight, then we can make all sorts of conclusions. And that is what **probability plotting** is all about.

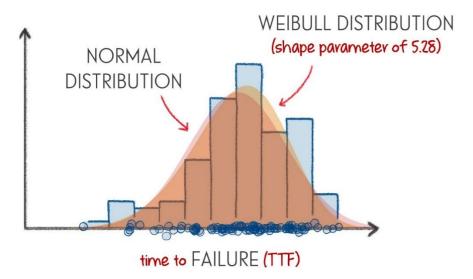
#### ...so this is 'WEIBULL ANALYSIS'

**Weibull analysis** is all about using the **Weibull distribution** to try and learn as much about a data set as possible. So Let's go back to our 'random hand of failure.' It can represent a **reliability test**, actual use, historical data or anything else where your product, system or service is allowed to operate until **failure**.

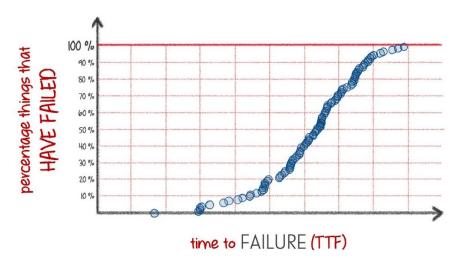


One of the key elements of **Weibull analysis** (or any other statistical analysis) is to try and identify what (if any) **probability distributions** explain the data we see. And one of the reasons we like

going straight to the **Weibull distribution** is that it can mimic many other probability distributions ... like the **normal distribution**.

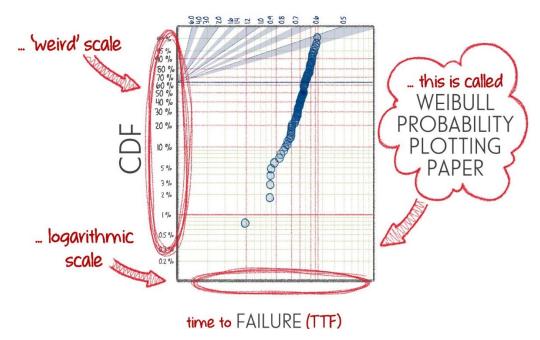


We can plot these data points on a chart where the vertical axis represented the percentage of things that have failed. This makes a 'smooth' line that helps us better visualize the 'nature' of **failure**.

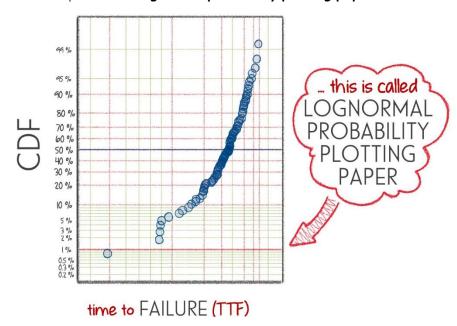


While this line is 'smooth' ... which usually means it is easier to spot patterns ... it is simply not possible to use it to work out what the parameters of any hypothetical **probability distribution** might be.

As we said earlier, we want to make that straight line. We can use the **probability distribution** functions to create 'special' charts that we call '**probability plotting papers**.' The reason we do this is down to the fact that any data that is described by the underlying **probability distribution** will create a straight line. The **TTF** data points on the previous page are shown plotted below on **Weibull probability plotting paper** ... and it produces (what looks like) a straight line.

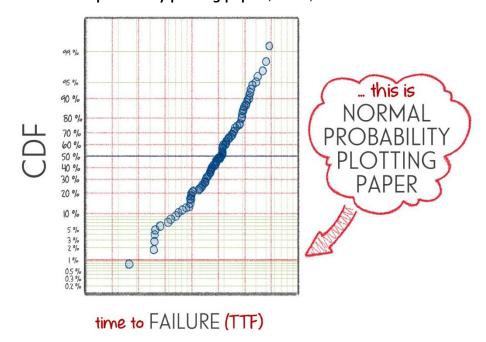


Compare this to the same data plotted on lognormal probability plotting paper below.



You can see that there is a distinct curve in the data line. This suggests that **the lognormal distribution** is not the **probability distribution** that best describes the **random (failure) process**.

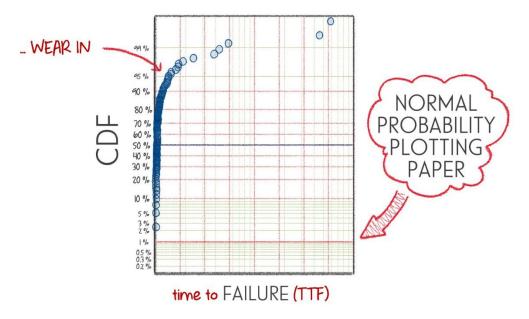
Now let's plot our data on normal probability plotting paper (below).



You could argue that this also creates a straight line ... as it does in **Weibull probability plotting paper** on the previous page. And this would make sense as we have already seen how the **normal distribution PDF (bell) curve** looks like it could match the density of our observed random data.

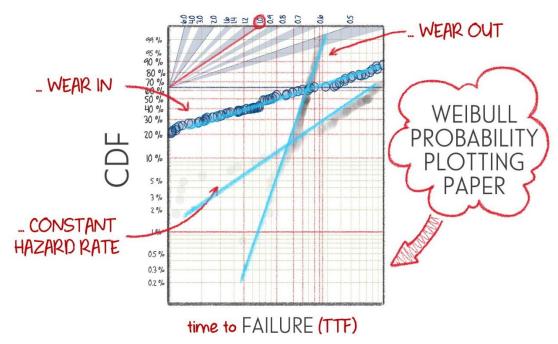
### ... we can tell SO MUCH MORE than this though

Let's look at how OTHER **probability plotting papers** are somewhat limited. For example, we have gathered some **TTF** data for a product that exhibits WEAR-IN failure. That is, it has a decreasing **hazard rate**. This is what that data looks like on **normal probability plotting paper**.

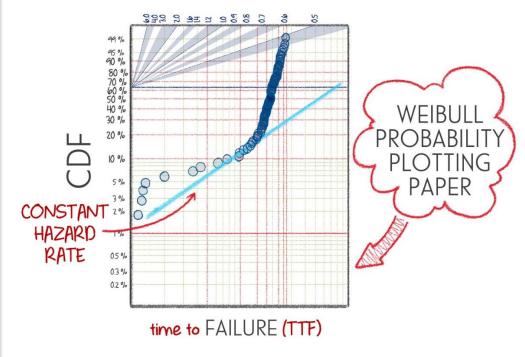


As the **normal distribution** can only model wear-out, there is no chance of understanding other types of failure, like wear-in **failure**.

The **Weibull probability plotting paper** below has three separate data sets plotted on it. One data set (which creates the steepest straight line) is based on a wear-out **failure mechanism**. Another data set (which creates the least steep straight line) is based on a wear-in **failure mechanism**. And the third data set (which creates the straight line parallel to the red line in the top left corner that terminates at '1.0') is based on a constant **hazard rate failure mechanism**.



TTF data from a relatively complex packaging machine is plotted on Weibull probability plotting paper below. The pale blue line represents a Weibull distribution that has a constant hazard rate.



What can we tell about the 'nature' of how our system fails?



What can you say about the failures that occur first? ... why?

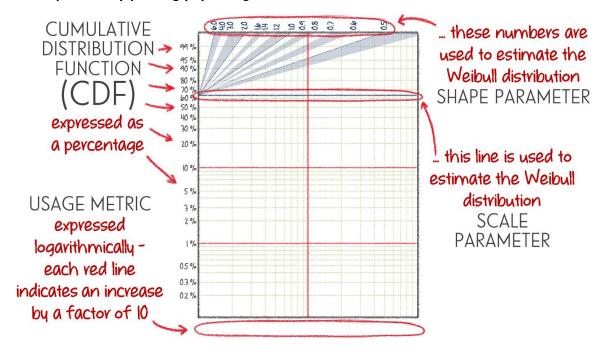
What can you say about the failures that occur last? ... why?

What can you say about the failures that occur in the middle? ... why?

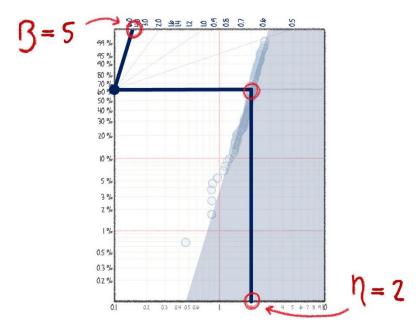
#### ...so how do we do WEIBULL PROBABILITY PLOTTING

PROBABILITY PLOTTING is a graphical technique for comparing TWO DATA SETS.

In the examples we have looked at on the previous pages, the first 'theoretical' data set is embedded in the **plotting paper** (with those weird scales on the axes) and the second 'actual' data set is the raw or empirical data sets that were observed in a test or operation. So let's look at **Weibull probability plotting paper** in greater detail.



When we follow the eight steps of **Weibull probability plotting**, we can identify key distribution parameters directly from the **plotting paper**. The following illustration summarizes **Weibull analysis** that we go through in the next section. You can see that we are able to estimate the **scale**  $(\eta = 2)$  and **shape**  $(\beta = 5)$  **parameters**.



And this is the Weibull distribution we get from these parameters! Not a bad fit ...

