Using Logarithms to model Curves (4 pages; 7/8/17)

Experimental data are often collected with a view to demonstrating some form of mathematical relation between the variables involved.

For example, we might measure the temperature of a cup of tea at various times after it has been made; or the volume of a given mass of gas may be recorded at different pressures.

In simple cases there may be a straight line relation of the form y = mx + c. The data points (x, y) can be plotted and a line of 'best fit' can be drawn, enabling the gradient m and y-intercept c to be established.

[Note that, even if two variables are connected by a precise scientific law, there is bound to be some experimental error, so that the original data points will not lie exactly on a straight line.]

However, the mathematical relation will often take a more complicated form. In this section, we consider two possibilities. In both cases, we can convert the given relation into a straight line form by taking logarithms.

Relation of the form $y = kx^n$

Taking logarithms (for example, to base 10) of both sides gives $logy = logk + log(x^n) = logk + nlogx$

[Note: *log* to the base 10 is commonly denoted by *log* on its own, without any subscript. As an alternative to using logs to base 10, we could equally well use natural logs; ie to the base *e*, denoted by *ln*.]

Ignoring the question of experimental error, if we now plot the points (logx, logy), we would expect them to lie on a straight line with gradient n and 'logy-intercept' of logk.

Often we won't be sure that the relation $y = kx^n$ holds. It is only when the points (*logx*, *logy*) are plotted that it becomes clear whether a straight line relation exists between logx and logy. If it does then n and logk can be deduced from the graph.

Example: The corresponding values of two variables *x* and *y* found by experiment are:

x	1	2	3	4	5
у	0.49	2.29	5.79	11.96	18.97

If we wish to investigate a relation of the form $y = kx^n$, then we would construct a table of values of *logx* and *logy* (or alternatively *lnx* and *lny*) to give:

logx	0	0.30	0.48	0.60	0.70
logy	-0.31	0.36	0.76	1.08	1.28

Plotting *logy* against *logx* gives the following line of best fit:





The gradient can be calculated approximately as:

 $\frac{1.30 - (-0.36)}{0.7 - 0} = 2.37$ Thus $n \approx 2.4$ And the 'logy-intercept' is -0.36, so that logk = -0.36, and hence $k = 10^{-0.36} = 0.437$ and $k \approx 0.4$

[Note that we have to be careful not to claim any greater accuracy than can be justified from our readings from the graph.]

So the proposed relation is $y = 0.4x^{2.4}$

Relation of the form $y = ab^x$

In this case, taking logarithms of both sides gives

logy = loga + xlogb

So this time we need to plot *logy* against *x*, and the gradient of the line of best fit will then give us an estimate for *logb*, whilst the '*logy*-intercept' will give us an estimate for *loga*.

Example: The corresponding values of two variables *x* and *y* found by experiment are:

x	1	2	3	4	5
у	4.34	4.75	6.53	6.34	8.88

To investigate a relation of the form $y = ab^x$, we plot *logy* against x to give the following table and graph:

X	1	2	3	4	5
logy	0.64	0.68	0.81	0.80	0.95



Figure 2

The gradient can be calculated approximately as:

 $\frac{0.92-0.55}{5-0} = 0.074$ Thus logb = 0.074, and hence $b = 10^{0.074} = 1.186 \approx 1.2$ And the 'logy-intercept' is 0.55, so that logk = 0.55, and hence $k = 10^{0.55} = 3.548 \approx 3.5$ So the proposed relation is $y = 3.5(1.2)^x$

Note: In some situations, the scales may not be convenient: for example, the *logx* values may all lie between 9 and 10. However, we mustn't introduce a break in either of the axes, as this would distort the

'*logy*-intercept'.