



STUDENT VERSION

PARAMETER ESTIMATION THROUGH STEADY STATE

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STATEMENT

We attempt to estimate parameters for a spring-mass-dashpot model differential equation (1) in which the mass is known ($m = 4$ kg) but the spring constant, k N/m, and the resistance coefficient, c N/(m/s), are unknown. We will subject the device/system to a driver force we control and use the output steady state data to estimate parameters c and k . This is a form of input output analysis in which we input a driver, analyze the output, and tell something about the device or model.

$$4 \cdot y''(t) + c \cdot y'(t) + k \cdot y(t) = f(t), \quad y(0) = 0 \quad \text{and} \quad y'(0) = 0 \quad (1)$$

You will do this with the following process. Apply a driver, say $f(t) = \sin(2t)$ and collect data on the steady state output of our system. See Table 1. From these observations in the steady state time period recover c and k .

We render plots of data taken from a system with unknown c and k and known $m = 4$ that is driven by $f(t) = \sin(2t)$ with initial conditions $y'(0) = 0$ and $y(0) = 0$ in Figure 1 and Figure 2. In Figure 2 we see the oscillating motion of the steady state more clearly than in Figure 1.

In Table 1 we offer the data. In whatever software you use you should plot this data set to see that you get a reasonable plot of the steady state phenomena.

- a) Consider the steady state or nonhomogeneous solution, which must be of the form, $ys(t) = A \sin(2t) + B \cos(2t)$ for our driver is $\sin(2t)$.
- b) Put $ys(t)$ into (1) and collect like terms, i.e. all the $\sin(2t)$ and $\cos(2t)$ terms on the left hand side of the resulting equation.

t s	20.0	20.5	21.0	21.5	22.0	22.5	23.0	23.5	24.0	24.5	25.0
$y(t)$ m	0.006	0.038	0.036	0.001	-0.033	-0.037	-0.006	0.031	0.040	0.013	-0.026
t s	25.5	26.0	26.5	27.0	27.5	28.0	28.5	29.0	29.5	30.0	
$y(t)$ m	-0.040	-0.017	0.022	0.041	0.023	-0.016	-0.040	-0.027	0.011	0.039	

Table 1. Steady state data from a a system with unknown c and k and known $m = 4$ that is driven by $f(t) = \sin(2t)$ with initial conditions $y'(0) = 0$ and $y(0) = 0$.

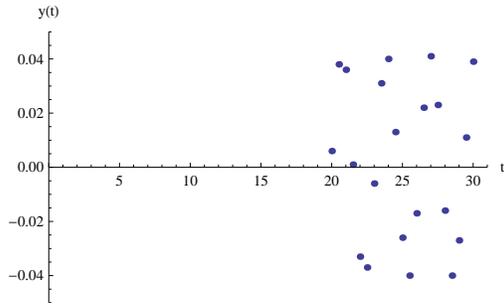


Figure 1. Plot of steady state data over entire time domain.

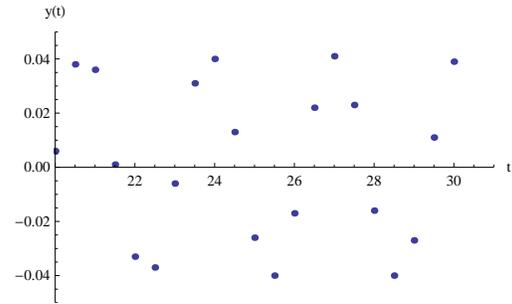


Figure 2. Plot of steady state data well into the steady state time frame.

- c) Equate the coefficients from $\sin(2t)$ and $\cos(2t)$ on both sides of our equation. You should get the following:

$$\begin{aligned} 2Ac + Bk - 16B &= 0 \\ -2Bc + Ak - 16A &= 1 \end{aligned} \quad (2)$$

- d) Solve (2) for A and B . Call your solutions for A and B , A_s and B_s respectively. These will be fractions in terms of c and k .
- e) Replace A and B in $ys(t) = A \sin(2t) + B \cos(2t)$ with A_s and B_s into respectively and obtain a steady state solution in terms of c and k . Call this solution $ys_s(t)$. It will have t , c , and k , as variables.
- f) Form the sum of square errors function, $SS(c, k)$, as offered in (3), between the model prediction at each data point and the observed data,

$$SS(c, k) = \sum_{i=1}^n (ys(t_i) - \hat{y}_i)^2, \quad (3)$$

where t_i is the time of the i^{th} observation, $ys(t_i)$ is the value of our steady state solution at time t_i , and \hat{y}_i is the observed dependent variable y for the i^{th} observation. Figure 3 shows a contour plot of the sum of square errors function in which you can see the region to best hunt for candidates for c and k which will produce a minimum $SS(c, k)$.

