

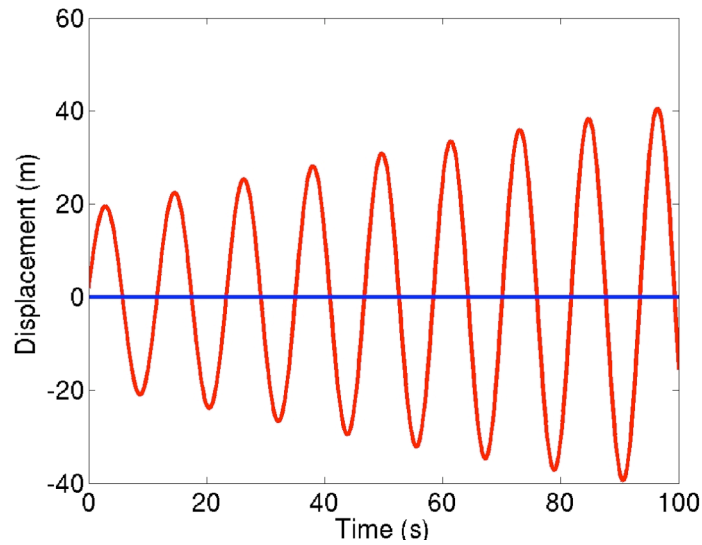
# Hypothesis Testing for Model Comparison

*“It is the mark of an educated mind to rest satisfied with the degree of precision which the nature of the subject admits and not to seek exactness where only an approximation is possible.” Aristotle*

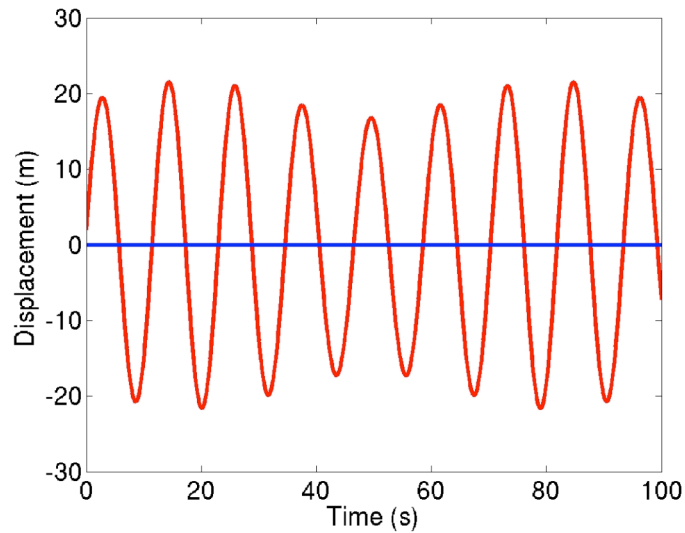
# Motivation

**Question:** Do the following forced spring responses exhibit damping?

**Example 1:**

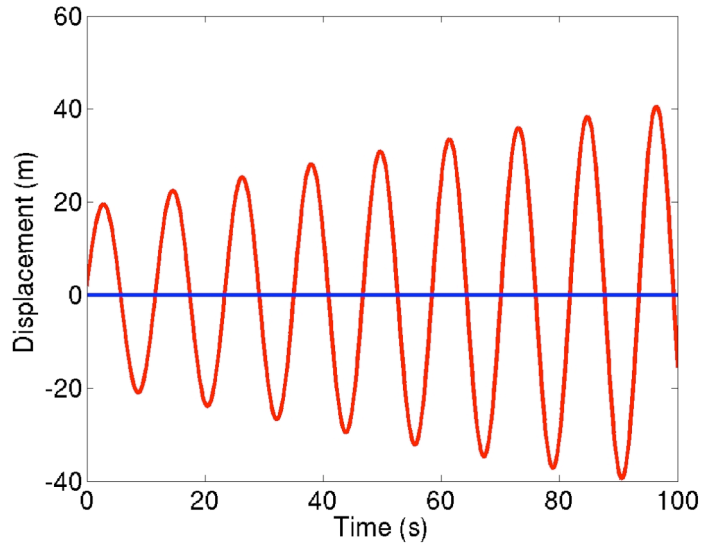


**Example 2:**



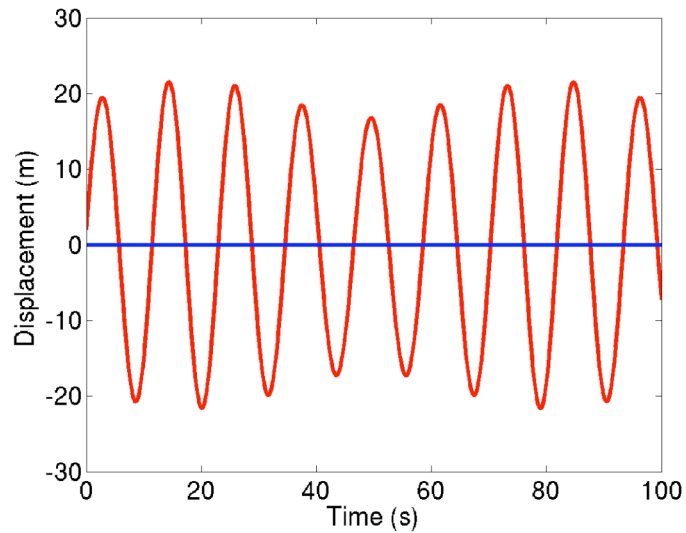
# Motivation

Example 1:



$$m = 7, k = 2, c = 0.01$$
$$\omega = 0.5445, \omega_0 = 0.5345$$

Example 2:



$$m = 7, k = 2, c = 0$$
$$\omega = 0.6345, \omega_0 = 0.5345$$

# Model Comparison

**Model:** Consider the model

$$m\ddot{y} + c\dot{y} + ky = F_0 \cos(\omega t)$$

$$y(0) = y_0, \dot{y}(0) = y_1$$

which has the solution

$$y(t) = e^{(-c/2m)t} [A \cos(\nu t) + B \sin(\nu t)] + \frac{F_0}{\Delta} \cos(\omega t - \delta)$$

where  $\omega_0 = \sqrt{k/m}$  and

$$\nu = \frac{\sqrt{4km - c^2}}{2m}, \Delta = \sqrt{m^2(\omega_0^2 - \omega^2)^2 + c^2\omega^2}, \delta = \arccos(m(\omega_0^2 - \omega^2)/\Delta)$$

$$A = y_0 - \frac{F_0}{\Delta} \cos \delta, B = \frac{1}{\nu} \left( y_1 + \frac{cA}{2m} - \frac{F_0\omega}{\Delta} \sin \delta \right)$$

**Parameter Spaces:**

$$Q_0 = \{q = (m, c, k) \mid m > 0, c = 0, k > 0\}$$

$$Q_1 = \{q = (m, c, k) \mid m > 0, c > 0, k > 0\}$$

# Hypothesis Testing

Hypotheses:

$$H_0 : c = 0$$

$$H_1 : c \neq 0$$

**Test Statistic:** Let  $\bar{q}_0, \bar{q}_1$  denote minimizers over  $Q_0, Q_1$ . Take

$$J(q) = \sum_{j=1}^n [y_j - y(t_j; q)]^2$$

and consider

$$U_n = n \frac{J(\bar{q}_0) - J(\bar{q}_1)}{J(\bar{q}_1)}$$

**Note:**  $U_n \rightarrow U$  where  $U$  is a random variable having a  $\chi^2(r)$  distribution

**Hypothesis Test:**

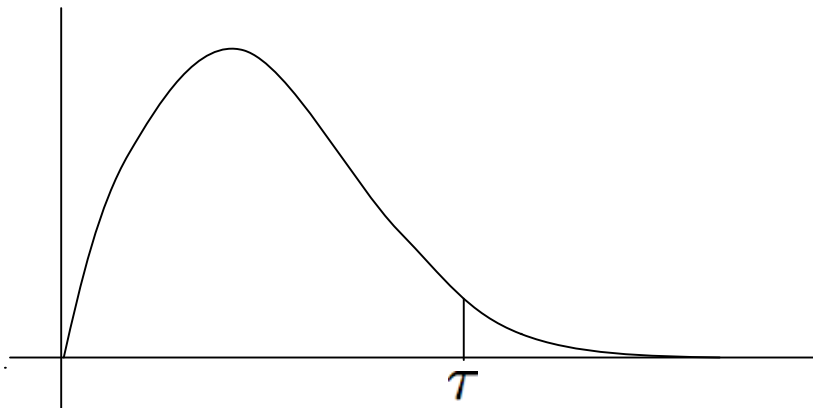
- Choose significance level  $\alpha$  and use  $\chi^2(r)$  tables to compute threshold  $\tau$  so that  $P(\chi^2(r) > \tau) = \alpha$

# Hypothesis Testing

## Hypothesis Test:

- Compute  $U_n$  and compare to  $\tau$ . If  $U_n > \tau$ , reject  $H_0$  as false. Otherwise, accept.
- $r$  is determined by the dimensionality of the parameter space;  $r = 1$  if linear constraint. In general,  $Hq = c$  where  $H$  is  $r \times p$ .

e.g.,  $H = [0, 1, 0]$  so  $Hq = c = 0$



# Hypothesis Testing

Example: [Banks and Fitzpatrick, 1990]

Data Set 1:

$$J(\bar{q}_0) = 180.17$$

$$J(\bar{q}_1) = 106.15$$

$$\Rightarrow U_8 = 8 \cdot \frac{180.17 - 106.15}{106.15} = 5.579$$

Table:

	$\chi^2(1)$
$\alpha = 0.25$	$\tau = 1.32$
$\alpha = 0.10$	$\tau = 2.71$
$\alpha = 0.05$	$\tau = 3.84$
$\alpha = 0.01$	$\tau = 6.63$
$\alpha = 0.005$	$\tau = 10.83$

Conclusion: Reject  $H_0$  at  $\alpha$  level of 0.05 or larger