

A Random Oscillator

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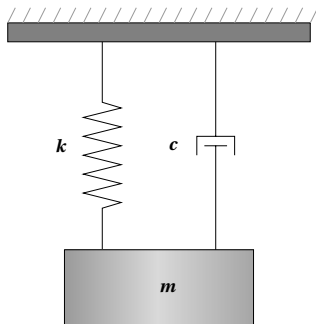


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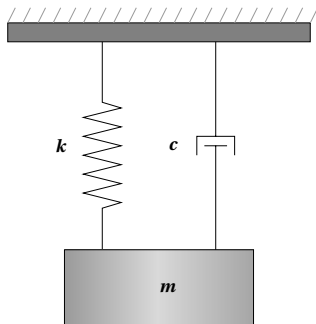
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Mass-Spring-Damper Oscillator



$$m\ddot{x} + c\dot{x} + kx = 0 \quad \dot{x}(0) = v_0 \quad x(0) = x_0$$

Mass-Spring-Damper Oscillator



$$m\ddot{x} + c\dot{x} + kx = 0 \quad \dot{x}(0) = v_0 \quad x(0) = x_0$$

What happens if the stiffness k is random ?

Parametric probabilistic approach

Probability space: $(\Omega, \Sigma, \mathcal{P})$

Stiffness k is modeled as a **random variable**

$$K : \omega \in \Omega \mapsto K(\omega) \in \mathbb{R}.$$

Displacement x has to be modeled as a **random process**

$$X_t : (\omega, t) \in \Omega \times \mathcal{T} \mapsto X(\omega, t) \in \mathbb{R},$$

which respect the stochastic equation of motion

$$m \ddot{X}(\omega, t) + c \dot{X}(\omega, t) + K(\omega) X(\omega, t) = 0,$$

$$\dot{X}(\omega, 0) = v_0, \quad X(\omega, 0) = x_0.$$

Contraction of the probabilistic model

Known theoretical information:

- positive support – $\text{Supp } p_K \subset (0, +\infty) \implies K > 0 \text{ a.s.}$
- finite variance – $\mathbb{E} \{ K^2 \} < +\infty$
- known mean – $\mathbb{E} \{ K \} = \mu_K$
- inverse finite variance – $\mathbb{E} \{ K^{-2} \} < +\infty$

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$$p_K(k) = \mathbb{1}_{(0,+\infty)}(k) \frac{1}{\mu_K} \frac{\delta_K^{-2\delta_K^{-2}}}{\Gamma(\delta_K^{-2})} \left(\frac{k}{\mu_K} \right)^{\delta_K^{-2}-1} \exp \left\{ -\frac{k/\mu_K}{\delta_K^2} \right\}$$

Gamma distribution is obtained via **MaxEnt Principle**.

rand_oscillator.m (1/7)

```
1  clc
2  clear all
3  close all
4
5  t0 = 0.0;    % initial time of analysis (s)
6  t1 = 30.0;  % final time of analysis (s)
7  Ndt = 300;  % number of time steps
8
9  m = 1.0;    % system mass (kg)
10 c = 0.1;    % damper constant (N.s/m)
11 k = 5.0;    % stiffness constant (N/m)
12 x0 = 1.0;   % initial position (m)
13 v0 = 1.0;   % initial velocity (m/s)
```

rand_oscillator.m (2/7)

```
1 % define stochastic parameters
2 rng_stream = RandStream('mt19937ar','Seed',30081984);
3 RandStream.setGlobalStream(rng_stream);
4
5 % number of samples
6 Ns = 256;
7 % preallocate memory for displacement and velocity
8 Qd = zeros(Ndt,Ns);
9 Qv = zeros(Ndt,Ns);
10 % stiffness mean (N/m)
11 mean_k = k;
12 % stiffness coef. var
13 coefvar_k = 0.15;
14 % generate stiffness with Gamma distribution (N/m)
15 k = gamrnd(1/coefvar_k^2,mean_k*coefvar_k^2,[Ns,1]);
```


rand_oscillator.m (3/7)

```
1 % init. cond. and interval of analysis
2 IC = [x0 v0]; tspam = linspace(t0,t1,Ndt);
3
4 % Monte Carlo method
5 for n=1:Ns
6
7     % system of equations
8     dydt=@(t,y)[0 1; -k(n) -c]*y;
9
10    % ODE solver Runge-Kutta45
11    [t,y] = ode45(dydt,tspam,IC);
12
13    % time series of system displacement (m)
14    Qd(:,n) = y(:,1);
15
16    % time series of system velocity (m/s)
17    Qv(:,n) = y(:,2);
18 end
```

rand_oscillator.m (4/7)

```
1 % sample mean
2 Qd_smean = mean(Qd');
3 Qv_smean = mean(Qv');
4
5 % temporal mean
6 Qd_tmean = mean(Qd);
7 Qv_tmean = mean(Qv);
8
9 % std. dev.
10 Qd_std = std(Qd');
11 Qv_std = std(Qv');
12
13 % confidence band
14 Pc = 95;
15 r_plus = 0.5*(100 + Pc); r_minus = 0.5*(100 - Pc);
16 Qd_upp = prctile(Qd',r_plus);
17 Qv_upp = prctile(Qv',r_plus);
18 Qd_low = prctile(Qd',r_minus);
19 Qv_low = prctile(Qv',r_minus);
```

rand_oscillator.m (5/7)

```
1 % histogram of temporal mean
2 Nbins = round(sqrt(Ns));
3 [Qd_bins,Qd_freq] = randvar_pdf(Qd_tmean,Nbins);
4 [Qv_bins,Qv_freq] = randvar_pdf(Qv_tmean,Nbins);
5
6 % kernel density estimator for temporal mean
7 [Qd_ksd,Qd_supp] = ksdensity(Qd_tmean);
8 [Qv_ksd,Qv_supp] = ksdensity(Qv_tmean);
```

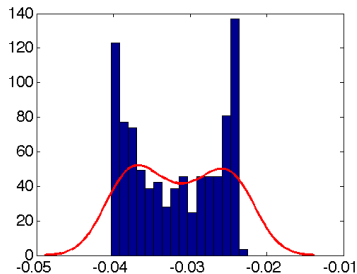
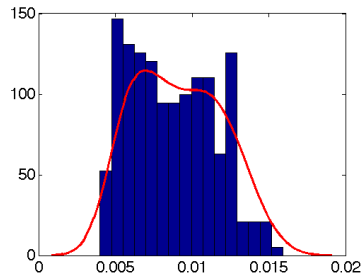
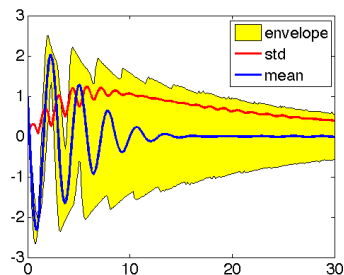
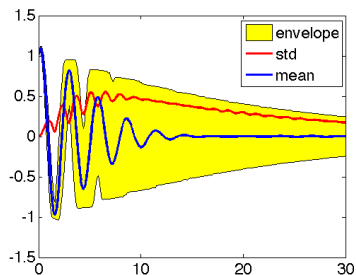
rand_oscillator.m (6/7)

```
1 figure(1)
2 fh1 = plot(t,Qd_smean,'b','linewidth',3); hold on
3 fh2 = plot(t,Qd_std,'r','linewidth',3);
4 fh3 = fill([t' fliplr(t')],[Qd_upp fliplr(Qd_low)],'y');
5 uistack(fh3,'top');
6 uistack(fh2,'top');
7 uistack(fh1,'top');
8 legend('envelope','std','mean')
9 hold off
10
11 figure(2)
12 fh1 = plot(t,Qv_smean,'b','linewidth',3); hold on
13 fh2 = plot(t,Qv_std,'r','linewidth',3);
14 fh3 = fill([t' fliplr(t')],[Qv_upp fliplr(Qv_low)],'y');
15 uistack(fh3,'top');
16 uistack(fh2,'top');
17 uistack(fh1,'top');
18 legend('envelope','std','mean')
19 hold off
```

rand_oscillator.m (7/7)

```
1 figure(3)
2 bar(Qd_bins,Qd_freq,1.0);
3 hold on
4 plot(Qd_supp,Qd_ksd,'r','linewidth',3)
5 hold off
6
7 figure(4)
8 bar(Qv_bins,Qv_freq,1.0);
9 hold on
10 plot(Qv_supp,Qv_ksd,'r','linewidth',3)
11 hold off
```

Random oscillator response



References



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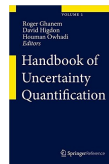
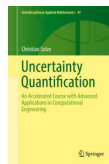
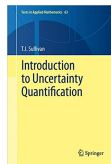
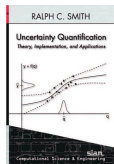
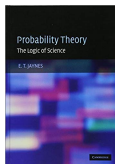
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