A Random Oscillator

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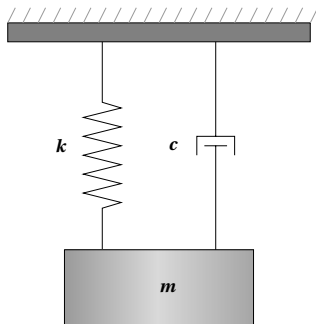
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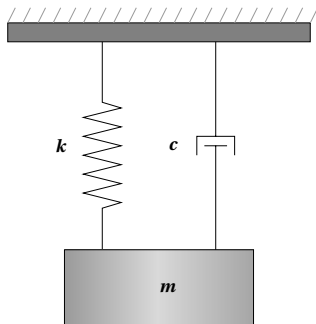
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Mass-Spring-Damper Oscillator

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

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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

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

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

Contruction of the probabilistic model

Known theoretical information:

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

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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_oscilator.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_oscilator.m (2/7)

```matlab
% define stochastic parameters
rng_stream = RandStream('mt19937ar','Seed',30081984);
RandStream.setGlobalStream(rng_stream);

% number of samples
Ns = 256;
% preallocate memory for displacement and velocity
Qd = zeros(Ndt,Ns);
Qv = zeros(Ndt,Ns);
% stiffness mean (N/m)
mean_k = k;
% stiffness coef. var
coefvar_k = 0.15;
% generate stiffness with Gamma distribution (N/m)
k = gamrnd(1/coefvar_k^2,mean_k*coefvar_k^2,[Ns,1]);
```

rand_oscilator.m (3/7)

```matlab
% init. cond. and interval of analysis
IC = [x0 v0]; tspam = linspace(t0,t1,Ndt);

% Monte Carlo method
for n=1:Ns

    % system of equations
    dydt=@(t,y)[0 1; -k(n) -c]*y;

    % ODE solver Runge-Kutta45
    [t,y] = ode45(dydt,tspam,IC);

    % time series of system displacement (m)
    Qd(:,n) = y(:,1);

    % time series of system velocity (m/s)
    Qv(:,n) = y(:,2);
end
```

rand_oscilator.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_oscilator.m (5/7)

```matlab
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);
```

```
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
```
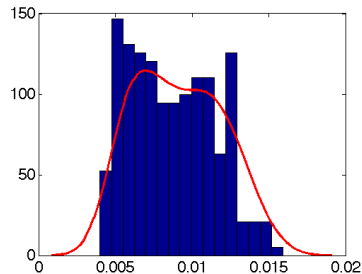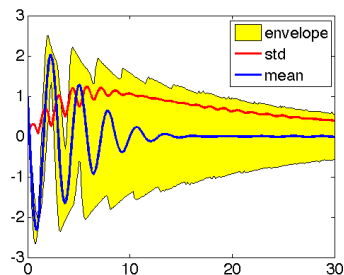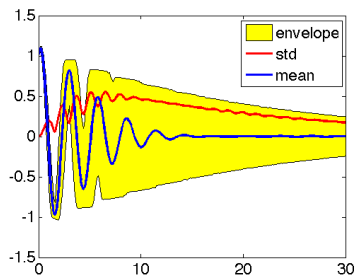
rand_oscilator.m (7/7)

```
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
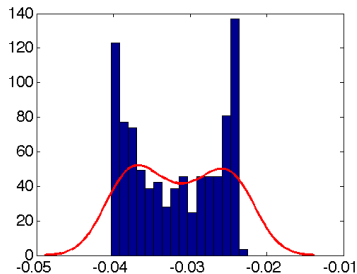
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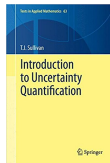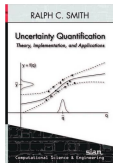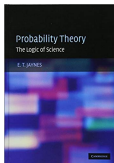
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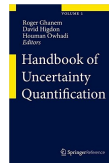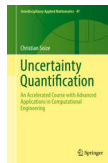
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