

Convolution

Dr. Mitch Pryor

Solving linear systems

Recall, the solution to $\dot{x} = ax$ is ce^{at}

A Taylor series of the exponential function of x is

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \frac{x^4}{4!} + \dots$$

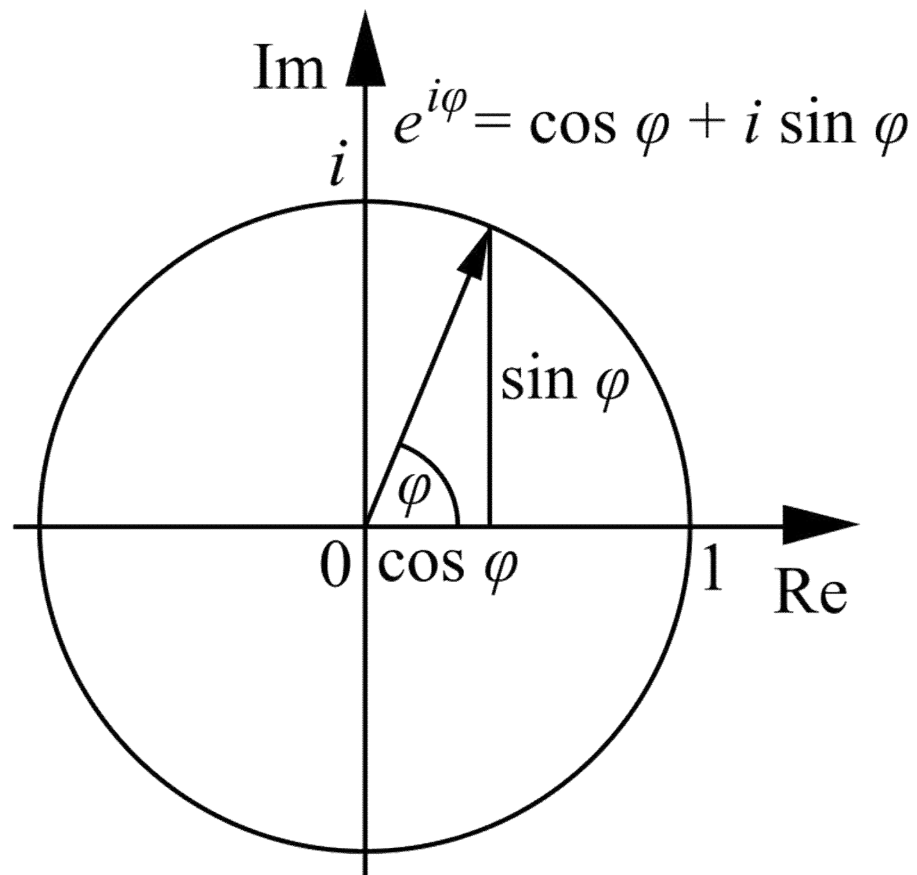
By analogy, we defined the Matrix Exponential

$$e^{\mathbf{A}} = 1 + \mathbf{A} + \frac{\mathbf{A}^2}{2!} + \frac{\mathbf{A}^3}{3!} + \frac{\mathbf{A}^4}{4!} + \dots$$

In MATLAB, the command `expm(A)` computes $e^{\mathbf{A}}$

So the solution to $\frac{d\mathbf{z}}{dt} = \mathbf{A}\mathbf{z} + \mathbf{B}u \Rightarrow \mathbf{z} = \underbrace{\mathbf{z}_0 e^{\mathbf{A}t}}_{\text{Homogeneous solution}} + \underbrace{???}_{\text{Our goal today}}$

$e^{i\varphi}$ (complex #) \Rightarrow Oscillation



Euler's Formula

Recall for linear systems...

- Today (and for most of the class) we will focus on input / single output linear systems
 - Why can we do this?
- A **linear** system $f(x)$ must obey these rules

$$f(x + y) = f(x) + f(y)$$

$$f(ax) = af(x)$$

$$f(ax + by) = af(x) + bf(y)$$

“Superposition”

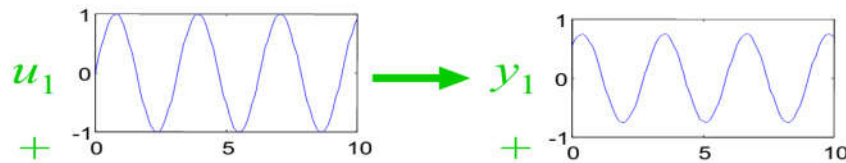
- So why is this important for controls?

Linear control systems

If we have two known responses from two different inputs to our system...

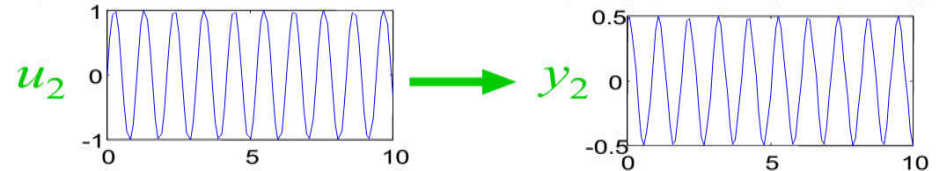
$$\frac{d}{dt}\mathbf{z} = \mathbf{A}\mathbf{z} + \mathbf{B}u_1$$

$$y_1 = \mathbf{C}\mathbf{z} + \mathbf{D}u_1$$



$$\frac{d}{dt}\mathbf{z} = \mathbf{A}\mathbf{z} + \mathbf{B}u_2$$

$$y_2 = \mathbf{C}\mathbf{z} + \mathbf{D}u_2$$

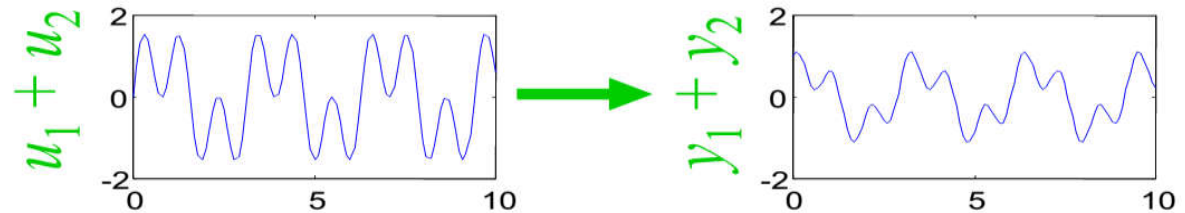


then we also can find the response to the sum of the input:

$$u_3 = \alpha u_1 + \beta u_2$$

$$\frac{d}{dt}\mathbf{z} = \mathbf{A}\mathbf{z} + \mathbf{B}u_3$$

$$y_3 = \mathbf{C}\mathbf{z} + \mathbf{D}u_3 = \alpha y_1 + \beta y_2$$



Common inputs: Impulse and Step

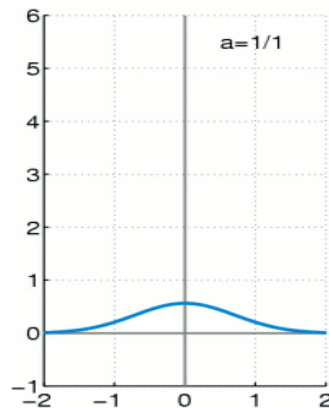
The Impulse Function

$$u(t) = u_0 \begin{cases} \infty, & t = t_0 \\ 0, & t \neq t_0 \end{cases}$$

Or more accurately...

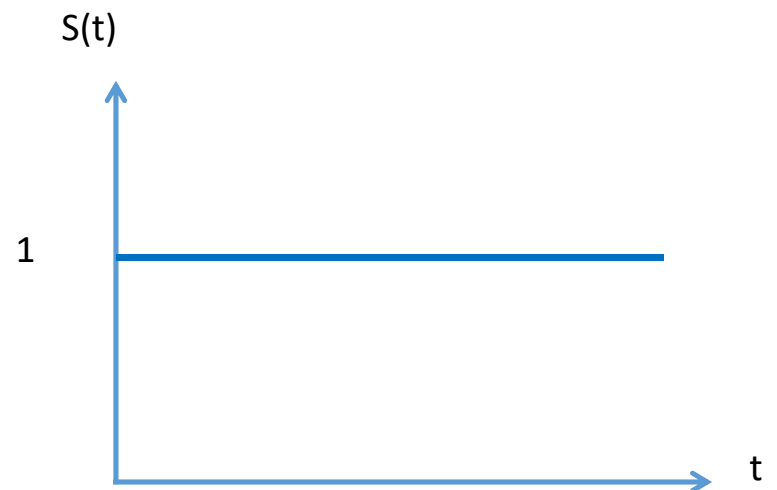
$$p_\varepsilon(t) = u_0 p_\varepsilon(t) = u_0 \begin{cases} 0 & t < 0 \\ \frac{1}{\varepsilon} & 0 \leq t < \varepsilon \\ 0 & t \geq \varepsilon \end{cases}$$

$$\delta(t) = \lim_{\varepsilon \rightarrow 0} p_\varepsilon(t)$$



The unit step function

$$s(t) = \begin{cases} 0 & t < t_0 \\ 1 & t \geq t_0 \end{cases}$$



Consider multiple step inputs

$$y(t, y_o, \delta u_1 + \cdots + \delta u_n) = y(t, y_o, \delta u_1) + \cdots + y(t, y_o, \delta u_n)$$

Principle of superposition

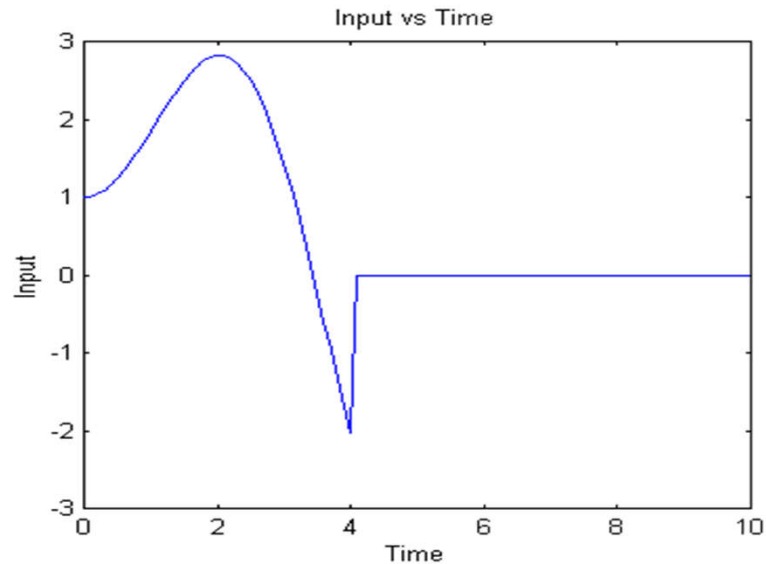
Thus, this is the solution to

$$\begin{aligned} \frac{dx}{dt} &= \mathbf{A}x + \mathbf{B}u \\ y &= \mathbf{C}x + \mathbf{D}u \end{aligned} \quad \text{for any set of inputs}$$

Let's assume that $y_o=0$ (i.e. we are only interested in the particular (not homogeneous) solution).

Any input can be viewed as a combination of other inputs!

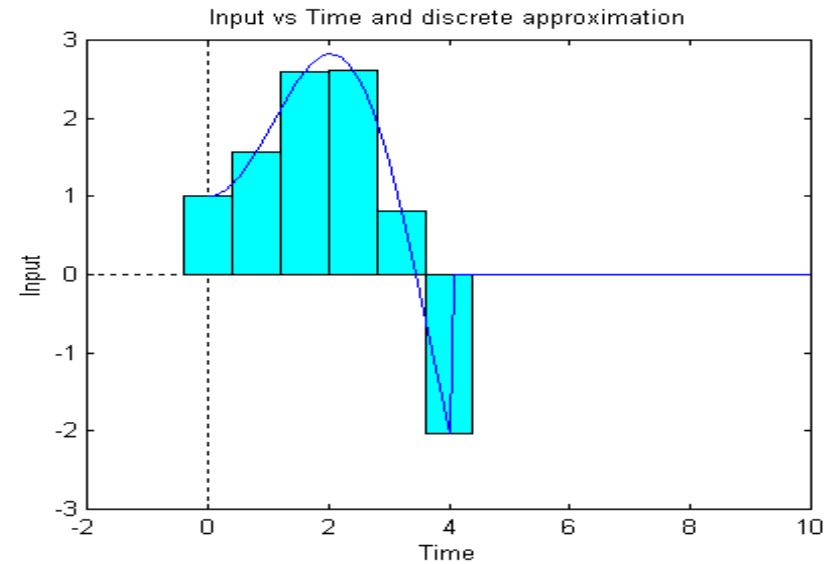
Example: “Complex” Input



Consider this rather complicated input $u(t)$ to a linear system.

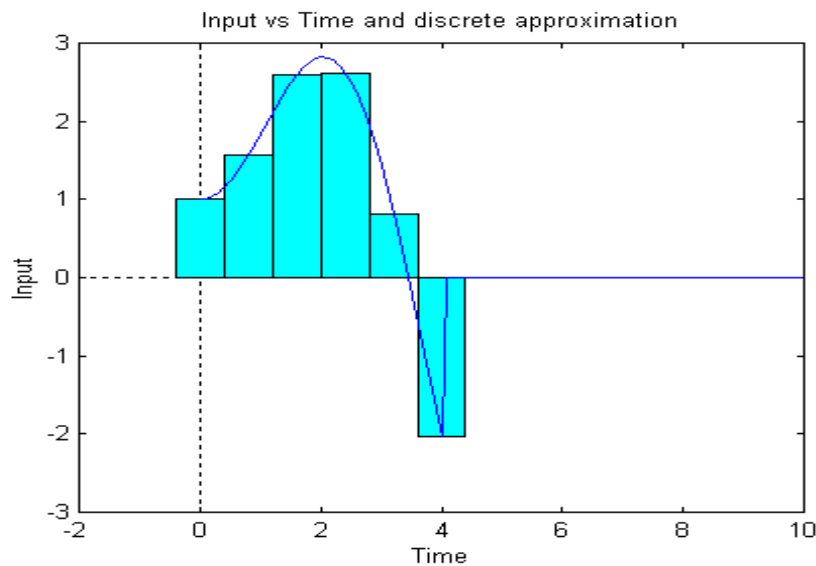
$$\ddot{z} + \dot{z} + z = u(t)$$

Objective: Find $z(t)$ if the initial conditions are zero.

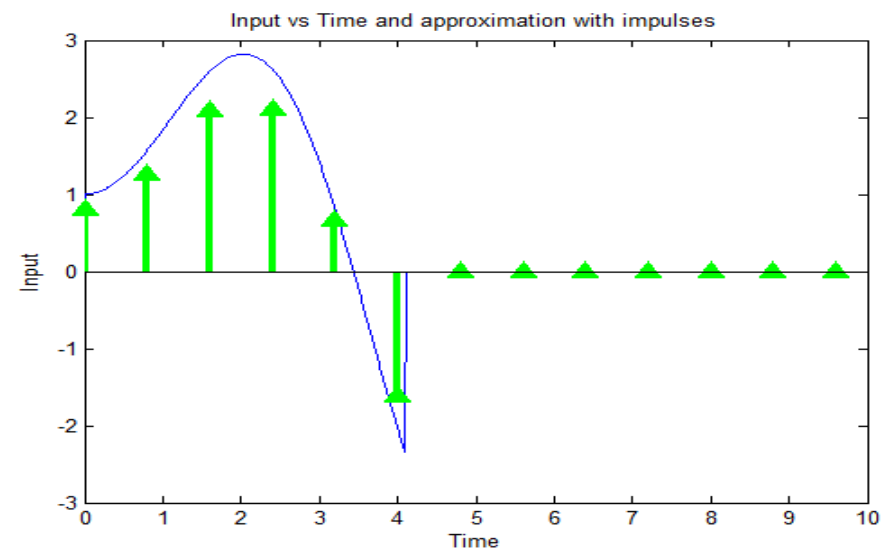


We can estimate this function as a set of discrete functions (here each is 0.8 seconds)

Example: “Complex” Input

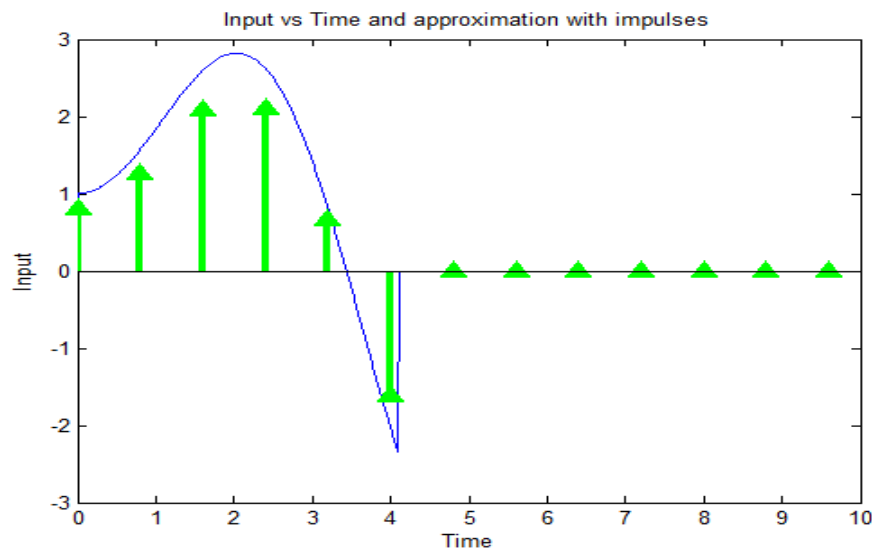


We can estimate this function as a set of discrete functions (here each is 0.8 seconds)

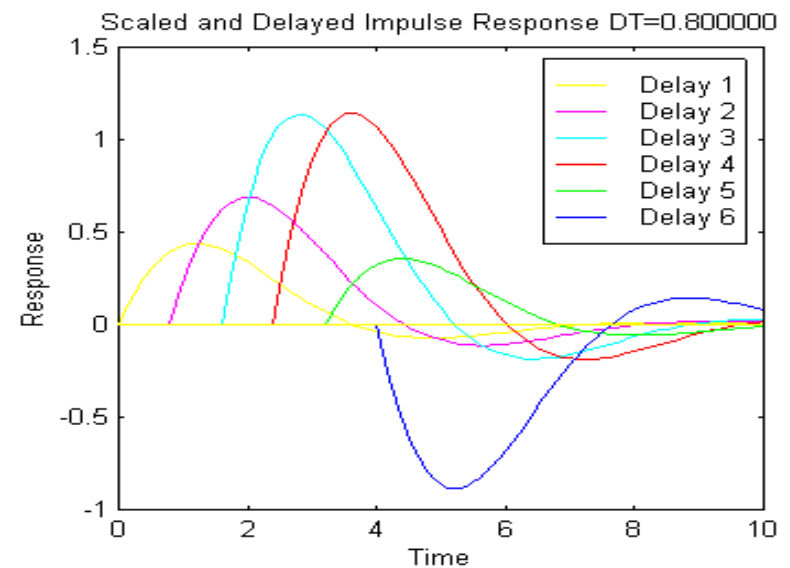


We take this one step further and replace each step with an equivalent impulse (they are shorter than the function since $\Delta t = 0.8s$)

Example: “Complex” Input

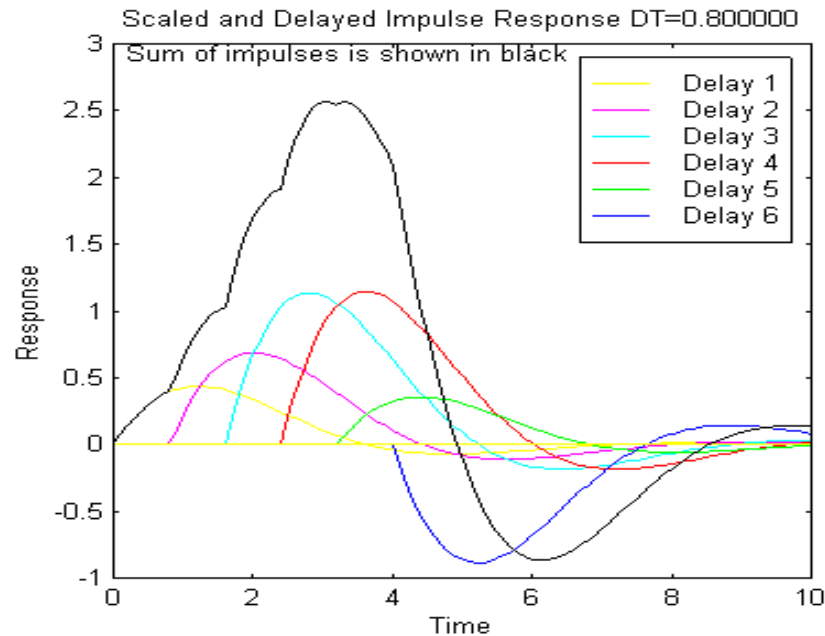


We take this one step further and replace each step with an equivalent impulse (they are shorter than the function since $\Delta t = 0.8s$)

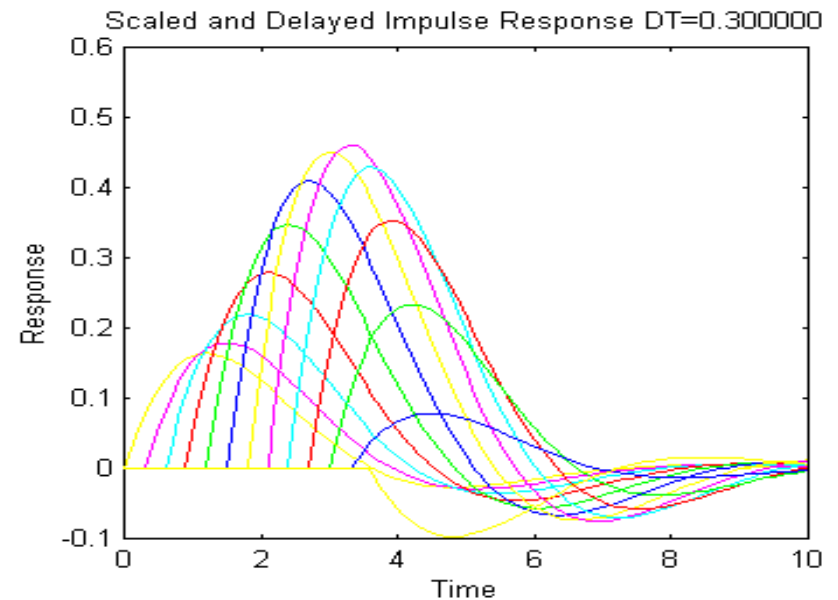


Here is the system response to each of the impulse. The final response will be sum of the impulse responses.

Example: “Complex” Input

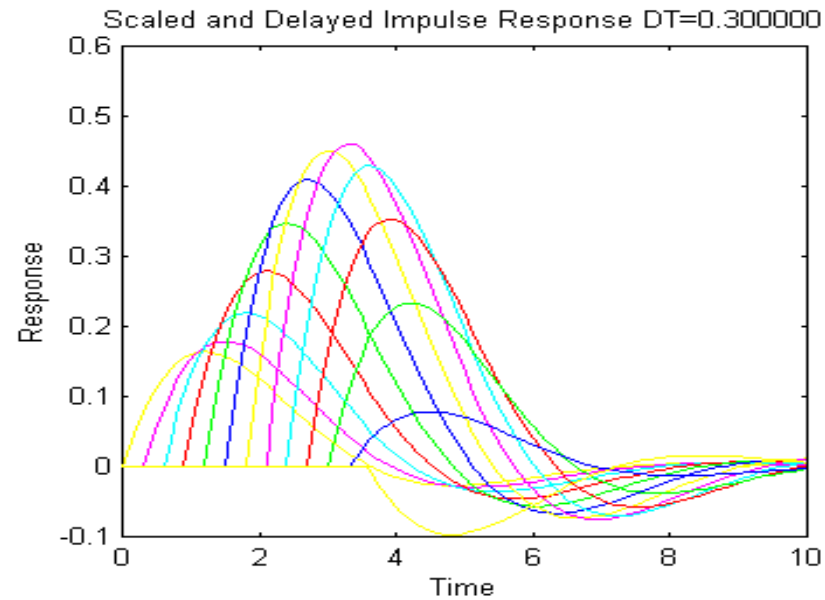


*Here is the system response to each of the impulse.
The final response will be sum of the impulse
responses. The accuracy will improve by reducing Δt .*

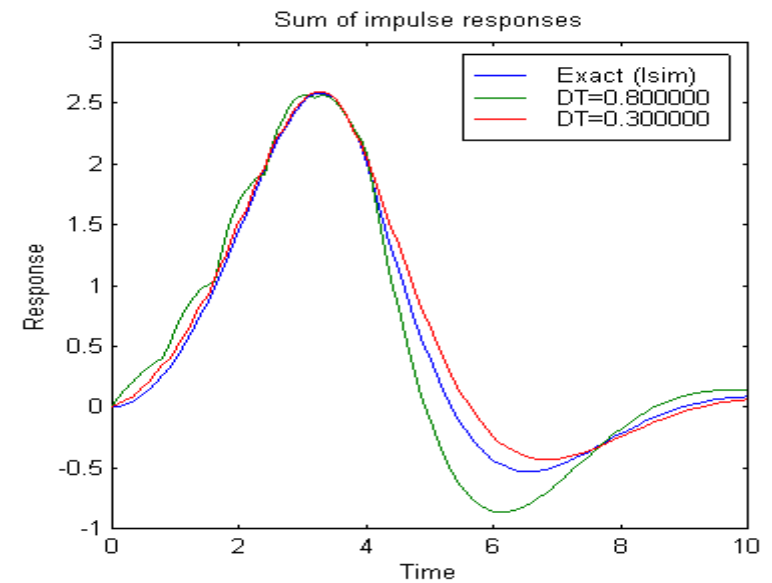


For example, here is the response if $\Delta t=0.3$.

Example: “Complex” Input



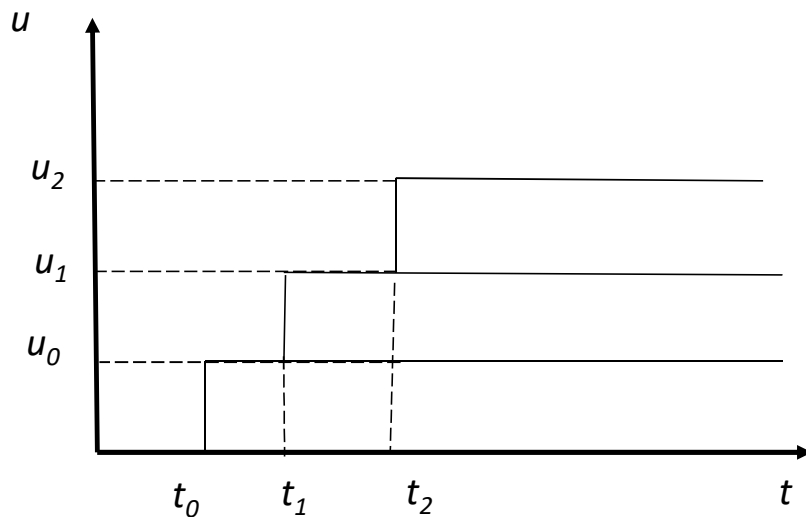
For example, here is the response if $\Delta t=0.3$.



If we sum up these impulse responses and compare them to the exact solution....

And what if we allow Δt to go to 0?

Output from a set of step inputs



Let $H(t)$ be the response to a step input at t_j .

$$\Rightarrow y(t) = H(t - t_j)u(t_j)$$

Can steps be negative? Sure.

Sum all of the inputs...

$$y(t) = H(t - t_0)u(t_0) + H(t - t_1)(u(t_1) - u(t_0)) + H(t - t_2)(u(t_2) - u(t_1)) + \dots$$

$$= [H(t - t_0) - H(t - t_1)]u(t_0) + [H(t - t_1) - H(t - t_2)]u(t_1) + \dots$$

Organize by common inputs...

$$= \sum_{i=1}^{t_n > t} [H(t - t_i) - H(t - t_{i+1})]u(t_i)$$

Rewrite as a summation of inputs...

$$= \sum_{i=1}^{t_n > t} \frac{[H(t - t_i) - H(t - t_{i+1})]}{t_{i+1} - t_i} u(t_i) (t_{i+1} - t_i)$$

Multiply numerator and denominator by delta T .

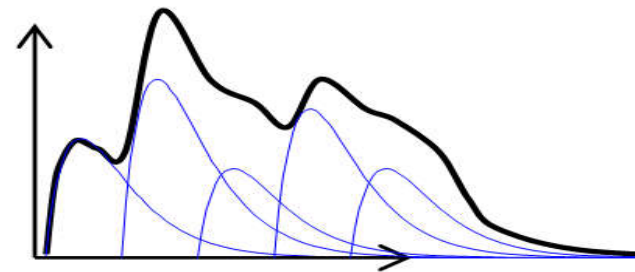
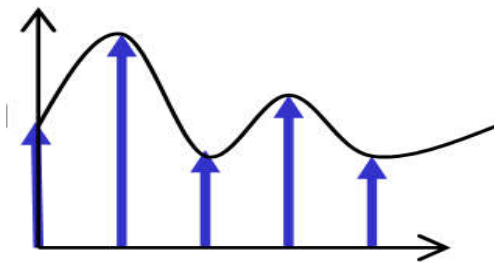
Output from a set of inputs

From previous page...

$$y(t) = \sum_{i=1}^{t_n > t} \frac{[H(t - t_i) - H(t - t_{i+1})]}{t_{i+1} - t_i} u(t_i) (t_{i+1} - t_i)$$

Let $(t_{i+1} - t_i) \rightarrow 0$

$$y(t) = \int_0^t H'(t - \tau) u(\tau) d\tau$$



We can use this to find the complete solution to:

$$\begin{aligned} \frac{dx}{dt} &= \mathbf{A}x + \mathbf{B}u \\ y &= \mathbf{C}x + \mathbf{D}u \end{aligned}$$

Complete Solution

Given: $\frac{d\mathbf{z}}{dt} = \mathbf{A} \mathbf{z} + \mathbf{B} u$
 $\mathbf{y} = \mathbf{C} \mathbf{z} + \mathbf{D} u$

We know:

$$\mathbf{z}(t) = e^{\mathbf{A}t} \mathbf{z}(0) +$$

Homogeneous solⁿ found with matrix exponential.

Response to system input

But what we are really interested in is the output

$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) +$$

Response to system input

$$\mathbf{y} = \mathbf{C} \mathbf{z} + \mathbf{D} u$$

Which must include....

Complete Solution

$$\begin{aligned} \frac{d\mathbf{z}}{dt} &= \mathbf{A}\mathbf{z} + \mathbf{B}u & \text{For a series of impulse inputs} & y(t) = \int_0^t H'(t-\tau)\delta(\tau)d\tau \\ y &= \mathbf{C}\mathbf{z} + \mathbf{D}u & & = \int_0^t \mathbf{C}e^{\mathbf{A}(t-\tau)}\mathbf{B}u(\tau)d\tau \end{aligned}$$

Put it all together, and we get...

$$\mathbf{y}(t) = \underbrace{\mathbf{C}e^{\mathbf{A}t}\mathbf{z}(0)}_{\text{Homogeneous response}} + \underbrace{\int_0^t \mathbf{C}e^{\mathbf{A}(t-\tau)}\mathbf{B}u(\tau)d\tau}_{\text{System response to input(s)}} + \underbrace{\mathbf{D}u}_{\text{Inclusion of input in measured signal}}$$

Init. condition response Transient response AND steady state response

Mismatch between ICs and steady state Only periodic if input is periodic

Does this look familiar?

It just might....

$$y(t) = c e^{at} z(0) + \int_0^t c e^{a(t-\tau)} u(\tau) d\tau + d u(t)$$

is known as the **convolution equation**. We have just derived it using the Matrix Exponential for a system of 1st order differential equations.

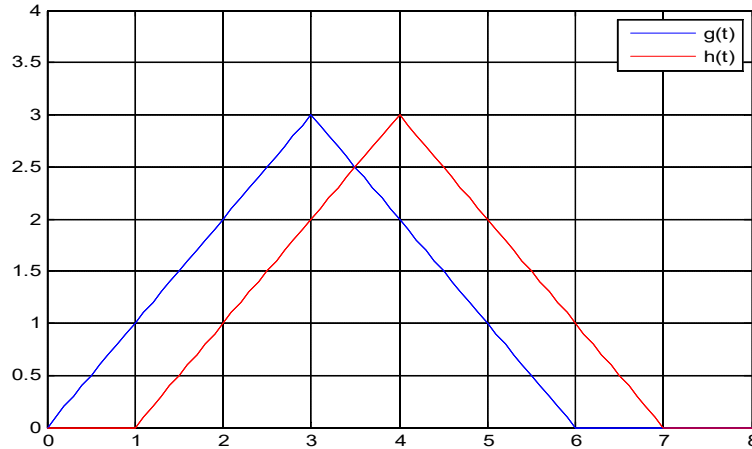
$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

- What is convolution?
 - a twisting or folding together of two things
 - A summing (using integral instead of a addition)
- Many examples:
 - Digital signal processing
 - A sound that bounces off of a wall and interacts with the source sound is a convolution
 - A shadow is a convolution between the light source and the object producing the shadow
 - In statistics, a moving average is a convolution
 - Review convolution and its state-space examples in next lesson.

Convolution example

$$f(t) = g(t) \otimes h(t) = \int g(t) * h(t) dt$$

$$g(t) = \begin{cases} t & t < 3 \\ 6 - t & t \geq 3 \\ 0 & t > 6 \end{cases} \quad h(t) = \begin{cases} 0 & t < 1 \\ t - 1 & 1 \leq t \leq 4 \\ 7 - t & t > 4 \\ 0 & t > 7 \end{cases}$$



```
t = [0:.1:10];

for i=1:length(t)
    if t(i)<3
        g(i)=t(i);
    elseif t(i) <= 6
        g(i) = 6-t(i);
    else
        g(i)=0;
    end
end

for i=1:length(t)
    if t(i)<1
        h(i)=0;
    elseif t(i)<=4;
        h(i)=t(i)-1;
    elseif t(i) <= 7
        h(i) = 7-t(i);
    else
        h(i)=0;
    end
end

plot(t,g, 'b')
hold on; grid on;
plot(t,h, 'r')
legend('g(t)', 'h(t)');
axis([0 8 0 4]);
```

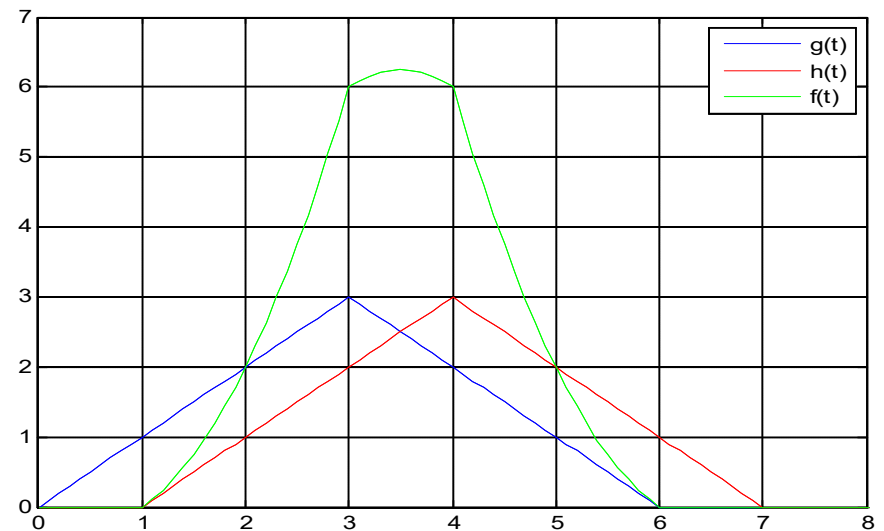
Example Result

$$f(t) = g(t) \otimes h(t) = \int_0^t g(t)h(t)dt$$

$$g(t) = \begin{cases} t & t < 3 \\ 6 - t & t \geq 3 \\ 0 & t > 6 \end{cases} \quad h(t) = \begin{cases} 0 & t < 1 \\ t - 1 & 1 \leq t \leq 4 \\ 7 - t & t > 4 \\ 0 & t > 7 \end{cases}$$

$$g(t) * h(t) = \begin{cases} 0 & t < 1 \\ t(t-1) & 1 \leq t \leq 3 \\ (6-t)(t-1) & 3 < t \leq 4 \\ (6-t)(7-t) & 4 \leq t \leq 6 \\ 0 & t > 6 \end{cases}$$

Note: This is not $\int g(t)dt \times \int h(t)dt$!



$$\int_0^t g(t)h(t)dt = 0 + \left(\frac{t^3}{3} - \frac{t^2}{2} \right) \Big|_{t=1}^{t=3} + \left(3t^2 - \frac{t^3}{3} \right) \Big|_{t=3}^{t=4} + \left(42t - \frac{13t^2}{2} + \frac{t^3}{3} \right) \Big|_{t=4}^{t=6} + 0 = 15.5$$

Another derivation of convolution

Start with our model in state-space form

$$\begin{aligned}\frac{d\mathbf{z}}{dt} &= \mathbf{A} \mathbf{z}(t) + \mathbf{B} u(t) \\ \mathbf{y} &= \mathbf{C} \mathbf{z}(t) + \mathbf{D} u(t)\end{aligned}$$

The homogeneous solution is...

$$\mathbf{z}(t) = e^{\mathbf{A}(t-t_0)} \mathbf{z}(t_0)$$

To find the particular solution, we let $\mathbf{K}(t)$ be an $n \times n$ matrix such that...

$$\frac{d\mathbf{K}(t)}{dt} = -\mathbf{K}(t)\mathbf{A}$$

So the matrix must be of the form....

$$\begin{aligned}\mathbf{K}(t) &= e^{-\mathbf{A}(t-t_0)} \\ \mathbf{K}(t)\dot{\mathbf{z}} &= \mathbf{K}(t)\mathbf{A}\mathbf{z}(t) + \mathbf{K}(t)\mathbf{B}u(t) \\ \mathbf{K}(t)\dot{\mathbf{z}}(t) - \mathbf{K}(t)\mathbf{A}\mathbf{z}(t) &= \mathbf{K}(t)\mathbf{B}u(t)\end{aligned}$$

$$\begin{aligned}\frac{d}{dt}[\mathbf{K}(t)\mathbf{z}(t)] &= \dot{\mathbf{K}}\mathbf{z} + \mathbf{K}\dot{\mathbf{z}} = \mathbf{K}(t)\mathbf{B}u(t) \\ d[\mathbf{K}(t)\mathbf{z}(t)] &= \mathbf{K}(t)\mathbf{B}u(t)dt\end{aligned}$$

Integrating from t to t_0

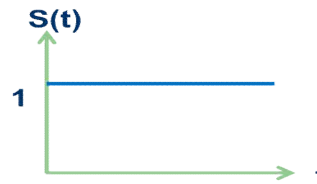
$$\begin{aligned}\mathbf{K}(t)\mathbf{z}(t) - \mathbf{K}(t_0)\mathbf{z}(t_0) &= \int_{t_0}^t \mathbf{K}(\tau)\mathbf{B}u(\tau)d\tau \\ \mathbf{x}(t) &= \mathbf{K}^{-1}(t)\mathbf{K}(t_0)\mathbf{z}(t_0) + \int_{t_0}^t \mathbf{K}^{-1}(t)\mathbf{K}(\tau)\mathbf{B}u(\tau)d\tau \\ \mathbf{x}(t) &= e^{\mathbf{A}(t-t_0)}\mathbf{z}(t_0) + \int_{t_0}^t e^{\mathbf{A}(t-\tau)}\mathbf{B}u(\tau)d\tau \\ \mathbf{y}(t) &= \mathbf{C}e^{\mathbf{A}(t-t_0)}\mathbf{z}(t_0) + \int_{t_0}^t \mathbf{C}e^{\mathbf{A}(t-\tau)}\mathbf{B}u(\tau)d\tau + \mathbf{D}u(t)\end{aligned}$$

If $t_0=0$

$$\mathbf{y}(t) = \mathbf{C}e^{\mathbf{A}t}\mathbf{z}(0) + \int_0^t \mathbf{C}e^{\mathbf{A}(t-\tau)}\mathbf{B}u(\tau)d\tau + \mathbf{D}u(t)$$

Convolution w/ one step input

Recall our step input is....



$$S(t) = \begin{cases} 0 & t \leq 0 \\ 1 & t > 0 \end{cases}$$

Plug into our convolution equation....

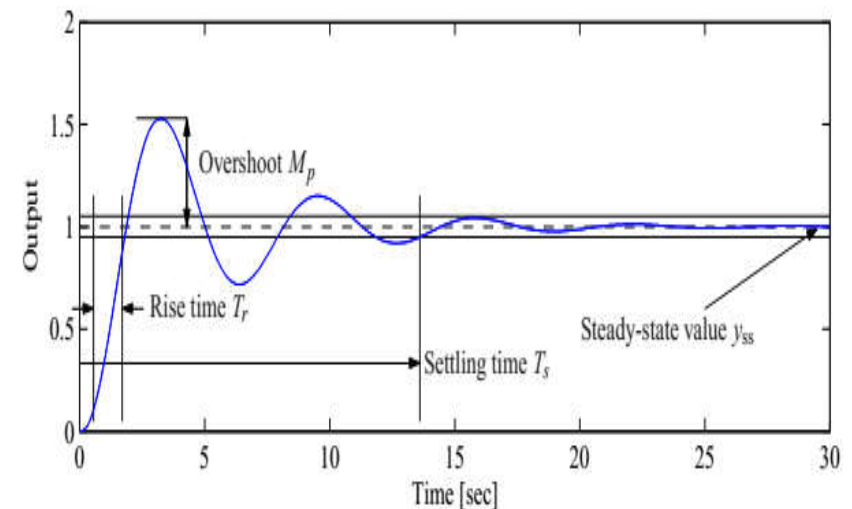
$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}(t-t_0)} \mathbf{z}(t_0) + \int_{t_0}^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

Let the initial conditions be zero...

$$\begin{aligned} \mathbf{y}(t) &= \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} S(\tau) d\tau + \mathbf{D} S(t) \\ &= \mathbf{C} \int_0^t e^{\mathbf{A}(t-\tau)} \mathbf{B} d\tau + \mathbf{D} \\ &= \mathbf{C} \int_0^t e^{\mathbf{A}\sigma} \mathbf{B} d\sigma + \mathbf{D} = \mathbf{C} \left(\mathbf{A}^{-1} e^{\mathbf{A}\sigma} \mathbf{B} \right) \Big|_{\sigma=0}^{\sigma=t} + \mathbf{D} \\ &= \underbrace{-\mathbf{C} \mathbf{A}^{-1} e^{\mathbf{A}t} \mathbf{B}}_{\text{transient response}} + \underbrace{\mathbf{C} \mathbf{A}^{-1} \mathbf{B} + \mathbf{D}}_{\text{steady state terms}} \end{aligned}$$

$\mathbf{A}^{-1} \neq \frac{1}{\mathbf{A}}$

Note: order matters for matrix multiplication



Convolution w/ sinusoidal input

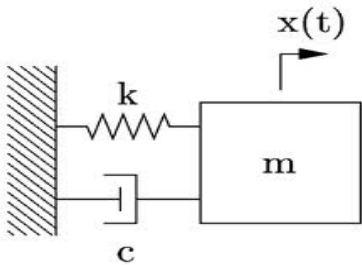
- Another common test function is a sinusoid for frequency response

$$u(t) = e^{st} \quad \text{where } s = \pm i\omega \quad u(t) = \cos \omega t = \frac{e^{i\omega t} + e^{-i\omega t}}{2}$$

- Since we have a linear system, and assuming that the eigenvalues of A do not equal s

$$\begin{aligned} y(t) &= \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} e^{s\tau} d\tau + \mathbf{D} e^{st} \\ &= \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \mathbf{C} e^{\mathbf{A}t} \int_0^t e^{(s\mathbf{I}-\mathbf{A})\tau} \mathbf{B} e^{s\tau} d\tau + \mathbf{D} e^{st} \\ &= \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \mathbf{C} e^{\mathbf{A}t} (s\mathbf{I} - \mathbf{A})^{-1} \left[e^{(s\mathbf{I}-\mathbf{A})t} - \mathbf{I} \right] \mathbf{B} + \mathbf{D} e^{st} \\ &= \underbrace{\mathbf{C} e^{\mathbf{A}t} \left(\mathbf{z}(0) - (s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} \right)}_{\text{Transient}} + \underbrace{\left(\mathbf{C} (s\mathbf{I} - \mathbf{A})^{-1} \mathbf{B} + \mathbf{D} \right) e^{st}}_{\text{Steady State}} \end{aligned}$$

Example: Mass – spring -damper



$$\begin{aligned} \frac{d\mathbf{z}}{dt} &= \mathbf{A} \mathbf{z} + \mathbf{B} u & \frac{d\mathbf{z}}{dt} &= \begin{bmatrix} 0 & 1 \\ -\frac{c}{m} & -\frac{k}{m} \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ \frac{F}{m} \end{bmatrix} u & u(t) &= \begin{cases} 0 & t < t_0 \\ 1 & t \geq t_0 \end{cases} \\ \mathbf{y} &= \mathbf{C} \mathbf{z} + \mathbf{D} u & \mathbf{y} &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u & z_1(0) &= 0 \\ & & & & z_2(0) &= 0 \end{aligned}$$

$$\begin{aligned} y(t) &= \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u \\ &= 0 + \mathbf{C} \int_0^t e^{\mathbf{A}(\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} \\ &= \mathbf{C} \left[\mathbf{A}^{-1} e^{\mathbf{A}(\tau)} \mathbf{B} \right]_{\tau=0}^{\tau=t} + \mathbf{D} \\ &= \mathbf{C} \left[\mathbf{A}^{-1} e^{\mathbf{A}(t)} \mathbf{B} - \mathbf{A}^{-1} e^{\mathbf{A}(0)} \mathbf{B} \right] + \mathbf{D} \\ &= \mathbf{C} \mathbf{A}^{-1} e^{\mathbf{A}t} \mathbf{B} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} + \mathbf{D} \end{aligned}$$

Example: Mass – spring -damper

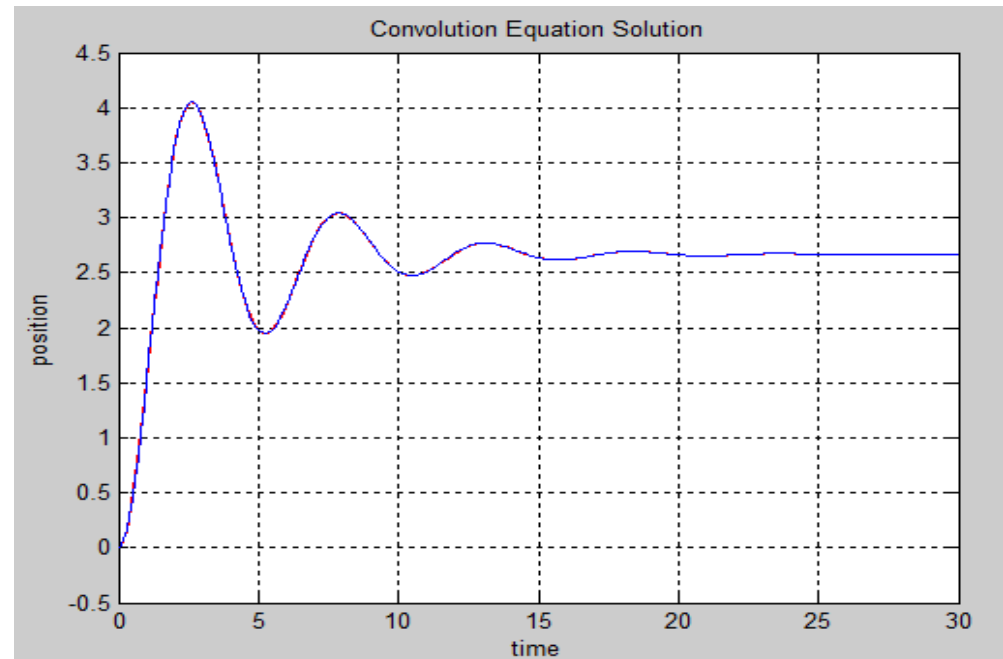
```
clear all;

m = 2; c = 1; k = 3; F = 4;

A = [ 0 1; -(k/m) -(c/m) ];
B = [ 0; F ];
C = [ 1 0; 0 1 ];
D = [ 0; 0 ];

[S, E] = eig(A);

tSpan = 0:.1:30;
for i = 1:length(tSpan)
    ME = [ exp(E(1,1)*tSpan(i)) 0; 0 exp(E(2,2)*tSpan(i)) ]; %could have used ME = expm(A)
    temp = C*inv(A)*S*ME*inv(S)*B - C*inv(A)*B + D;
    y(i,:) = real(temp'); %first check, then remove small numerical errors.
end
plot( tSpan, y(:,1), 'b' );
```



Example: Mass – spring -damper

Double Check...

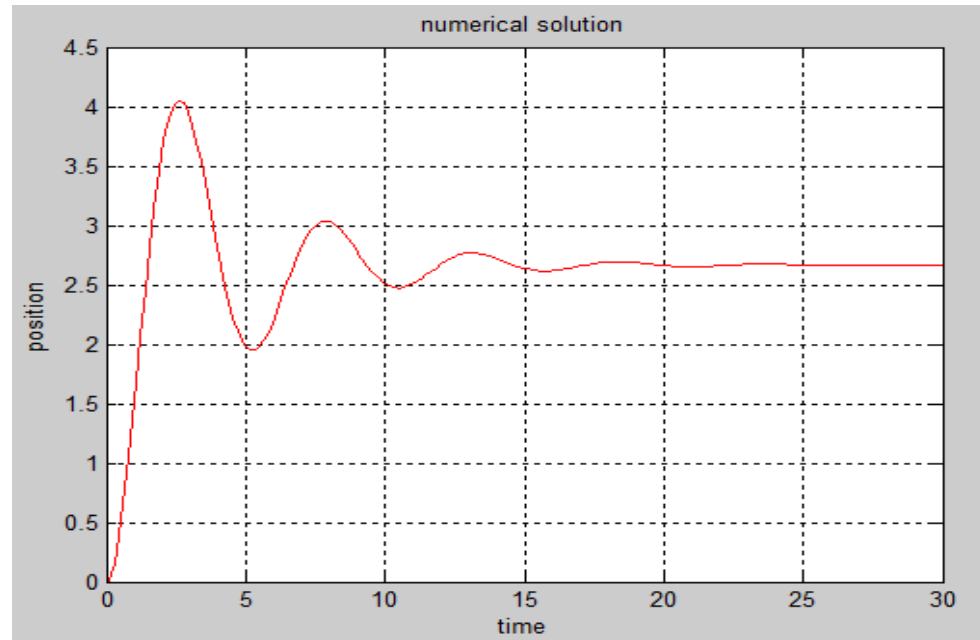
```
%numerical solution
[t, z] = ode45( @test, [0 30 ], [ 0 0 ]);

figure(2)
plot( t, z(:,1), 'r');

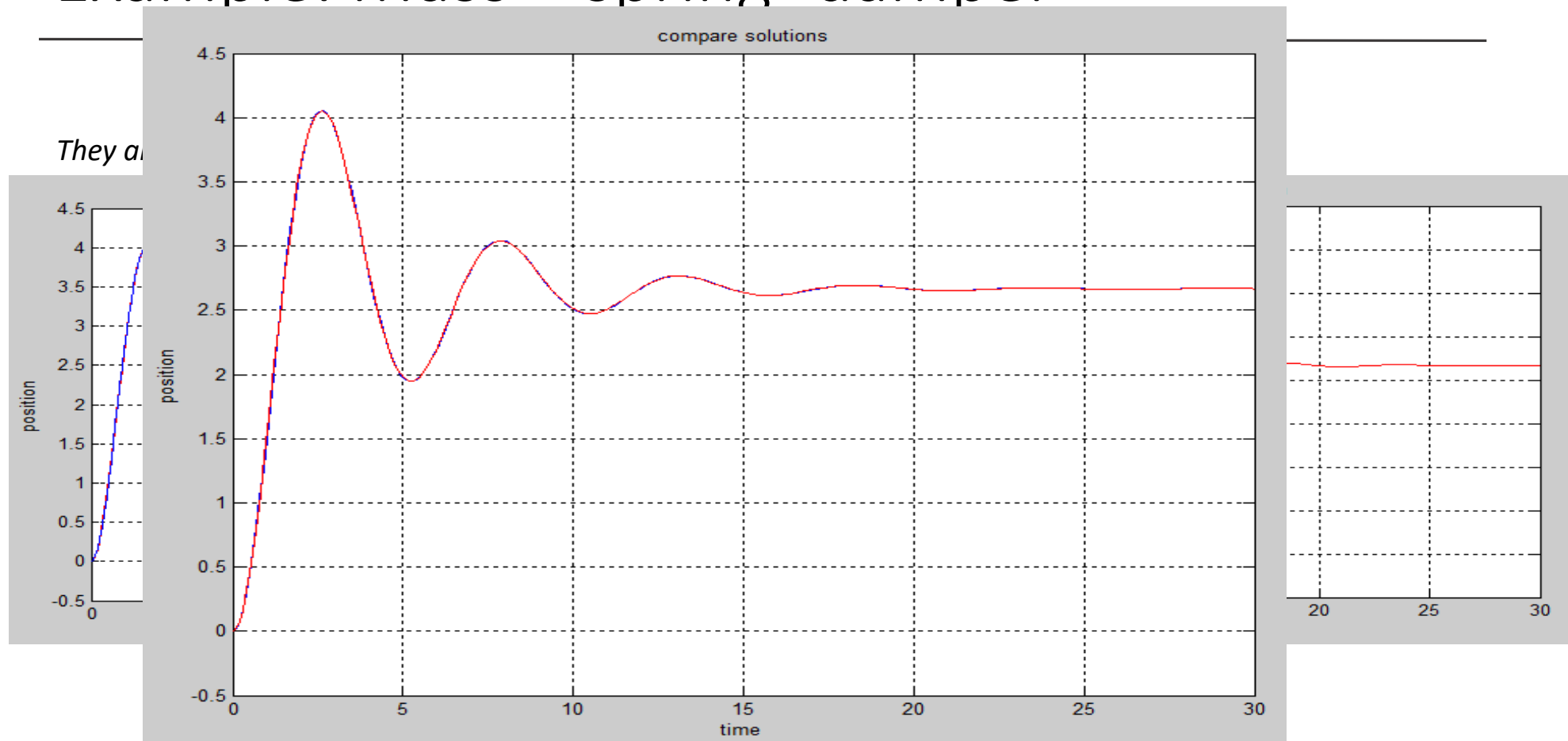
function zprime = test( t, z )

m = 2; c = 1; k = 3; F = 4;

zprime = [
    z(2);
    -(c/m)*z(2) - (k/m)*z(1) + 4;
];
```



Example: Mass – spring -damper



Convolution and Transformations

- A property of matrix exponentials is that

$$e^{\mathbf{A}} = e^{\mathbf{T} \mathbf{\Lambda} \mathbf{T}^{-1}} = \mathbf{T} e^{\mathbf{\Lambda}} \mathbf{T}^{-1}$$

- Therefore

$$y(t) = \mathbf{C} e^{\mathbf{A}(t-t_0)} \mathbf{z}(t_0) + \int_{t_0}^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

$$y(t) = \mathbf{C} \mathbf{T} e^{\mathbf{\Lambda}(t-t_0)} \mathbf{T}^{-1} \mathbf{T} \mathbf{z}(t_0) + \int_{t_0}^t \mathbf{C} \mathbf{T} e^{\mathbf{\Lambda}(t-\tau)} \mathbf{T}^{-1} \mathbf{T} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

$$y(t) = \mathbf{C} \mathbf{T} e^{\mathbf{\Lambda}(t-t_0)} \mathbf{z}(t_0) + \int_{t_0}^t \mathbf{C} \mathbf{T} e^{\mathbf{\Lambda}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

$$y(t) = \mathbf{C} \mathbf{T} e^{\mathbf{\Lambda}(t-t_0)} \mathbf{z}(t_0) + \mathbf{C} \mathbf{T} \int_{t_0}^t e^{\mathbf{\Lambda}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

We have a complete solution!

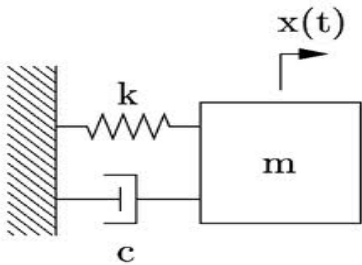
$$y(t) = c e^{at} x(0) + \int_0^t c e^{a(t-\tau)} u(\tau) d\tau + d u(t)$$

is known as the convolution equation. We have just derived it using the Matrix Exponential for a system of 1st order differential equations.

$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

- Now have complete analytical solution for sets of first order equations
 - Handle non-zero initial conditions
 - Handle multiple inputs
- MATLAB provides many tools to solve these equations for you.
- Next step: Let's look at some of these tricks.

Example: Mass – spring -damper



$$\frac{d\mathbf{z}}{dt} = \mathbf{A} \mathbf{z} + \mathbf{B} u$$

$$\mathbf{y} = \mathbf{C} \mathbf{z} + \mathbf{D} u$$

$$\frac{d\mathbf{z}}{dt} = \begin{bmatrix} 0 & 1 \\ -\frac{c}{m} & -\frac{k}{m} \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ \frac{F}{m} \end{bmatrix} u$$

$$\mathbf{y} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u$$

$$u(t) = \begin{cases} 0 & t < t_0 \\ 1 & t \geq t_0 \end{cases}$$

$$z_1(0) = 0$$

$$z_2(0) = 0$$

$$y(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u$$

$$= 0 + \mathbf{C} \int_0^t e^{\mathbf{A}(\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D}$$

$$= \mathbf{C} \left[\mathbf{A}^{-1} e^{\mathbf{A}(\tau)} \mathbf{B} \right]_{\tau=0}^{\tau=t} + \mathbf{D}$$

$$= \mathbf{C} \left[\mathbf{A}^{-1} e^{\mathbf{A}(t)} \mathbf{B} - \mathbf{A}^{-1} e^{\mathbf{A}(0)} \mathbf{B} \right] + \mathbf{D}$$

$$= \mathbf{C} \mathbf{A}^{-1} e^{\mathbf{A}t} \mathbf{B} - \mathbf{C} \mathbf{A}^{-1} \mathbf{B} + \mathbf{D}$$

Other Cool MATLAB tricks

$$\frac{d\mathbf{z}}{dt} = \begin{bmatrix} 0 & 1 \\ -\frac{c}{m} & -\frac{k}{m} \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ \frac{F}{m} \end{bmatrix} u$$

$$\mathbf{y} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \mathbf{z} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} u$$

$$y(t) = \underbrace{\mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0)} + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

```
clear all;
```

```
m = 2; c = 1; k = 3; F = 4;
```

```
A = [ 0 1; -(k/m) -(c/m) ];
```

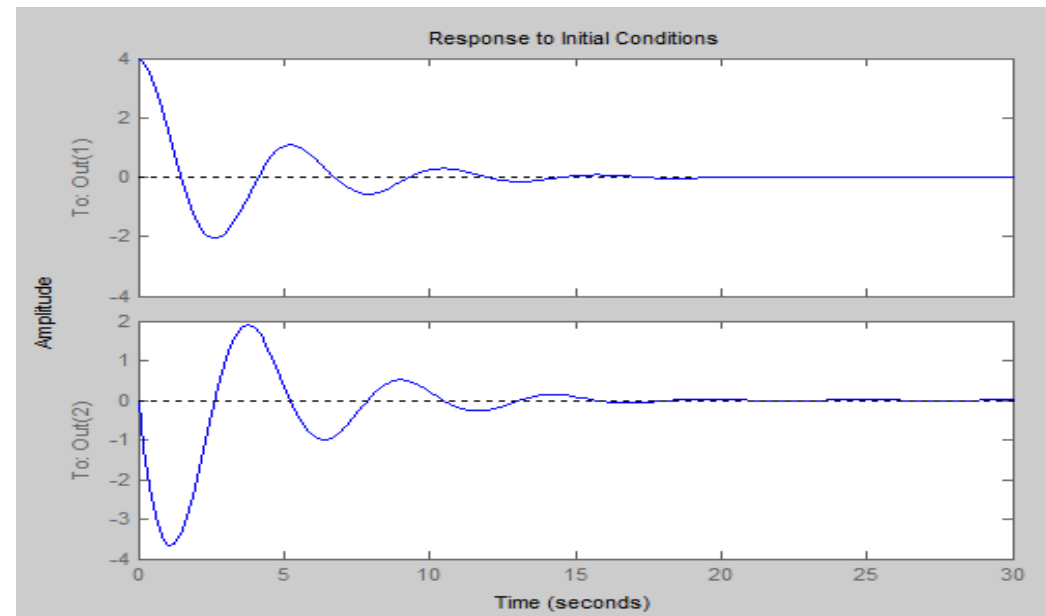
```
B = [ 0; F ];
```

```
C = [ 1 0; 0 1];
```

```
D = [ 0; 0 ];
```

```
sys = ss( A, B, C, D );
```

```
initial( sys, [ 4 0 ] );
```



Other Cool MATLAB tricks

$$y(t) = \underbrace{\mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau}_{\text{Impulse Response}} + \mathbf{D} u(t)$$

```
clear all;
```

```
m = 2; c = 1; k = 3; F = 4;
```

```
A = [ 0 1; -(k/m) -(c/m) ];
```

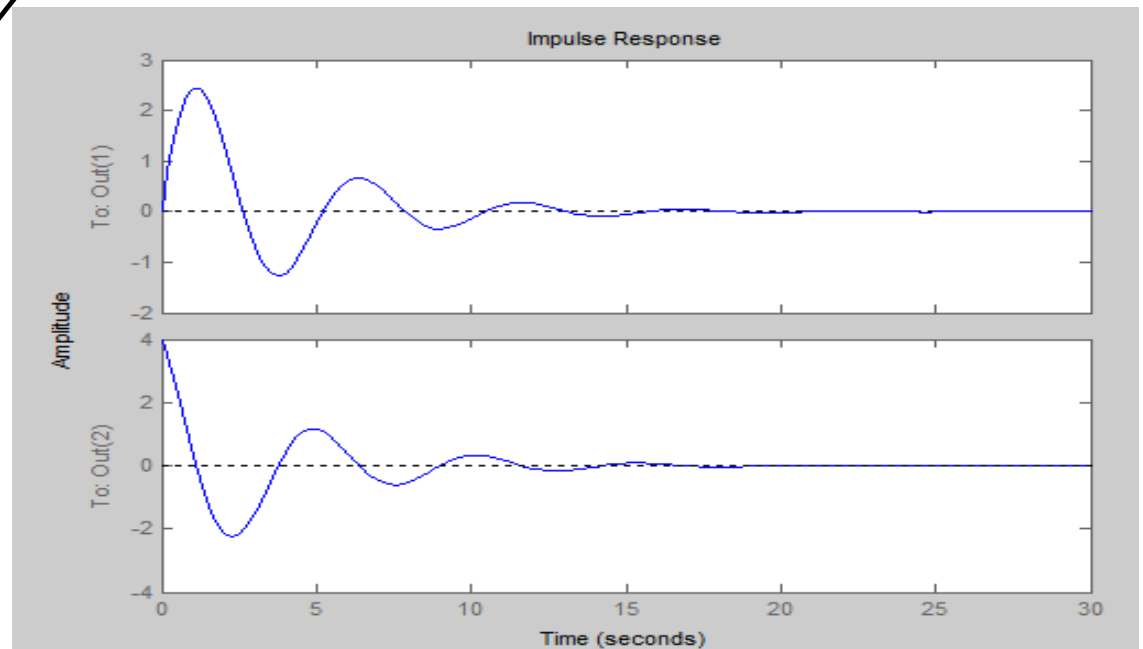
```
B = [ 0; F ];
```

```
C = [ 1 0; 0 1];
```

```
D = [ 0; 0 ];
```

```
sys = ss( A, B, C, D );
```

```
impz( sys );
```



Other Cool MATLAB tricks (lsim)

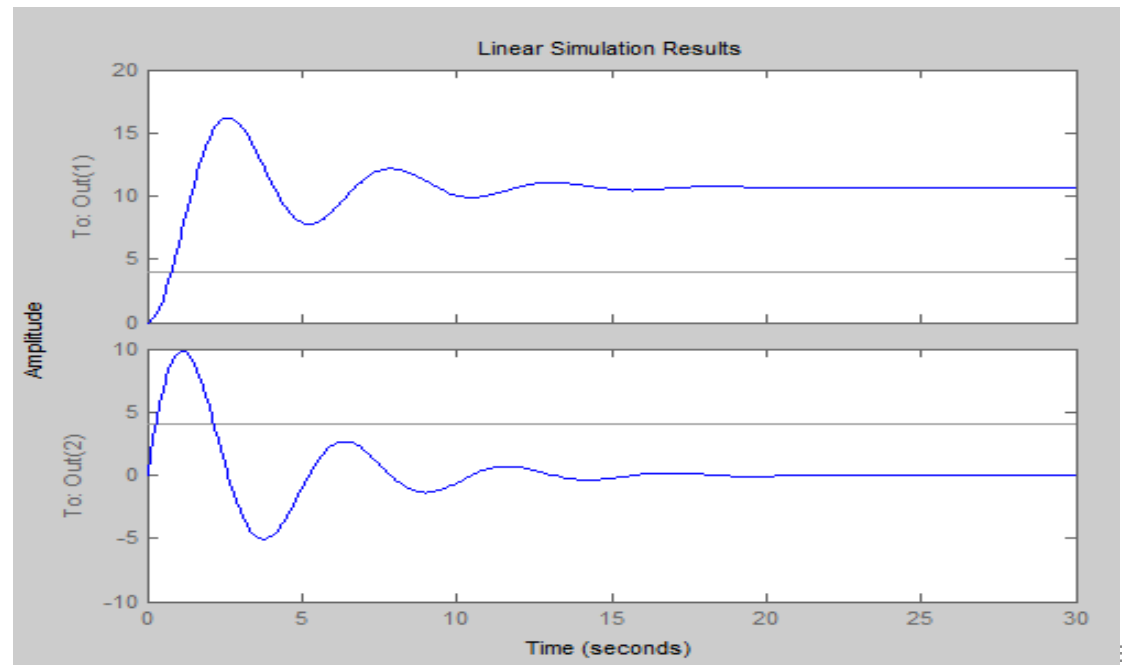
$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

```
clear all;

m = 2; c = 1; k = 3; F = 4;

A = [ 0 1; -(k/m) -(c/m) ];
B = [ 0; F ];
C = [ 1 0; 0 1 ];
D = [ 0; 0 ];

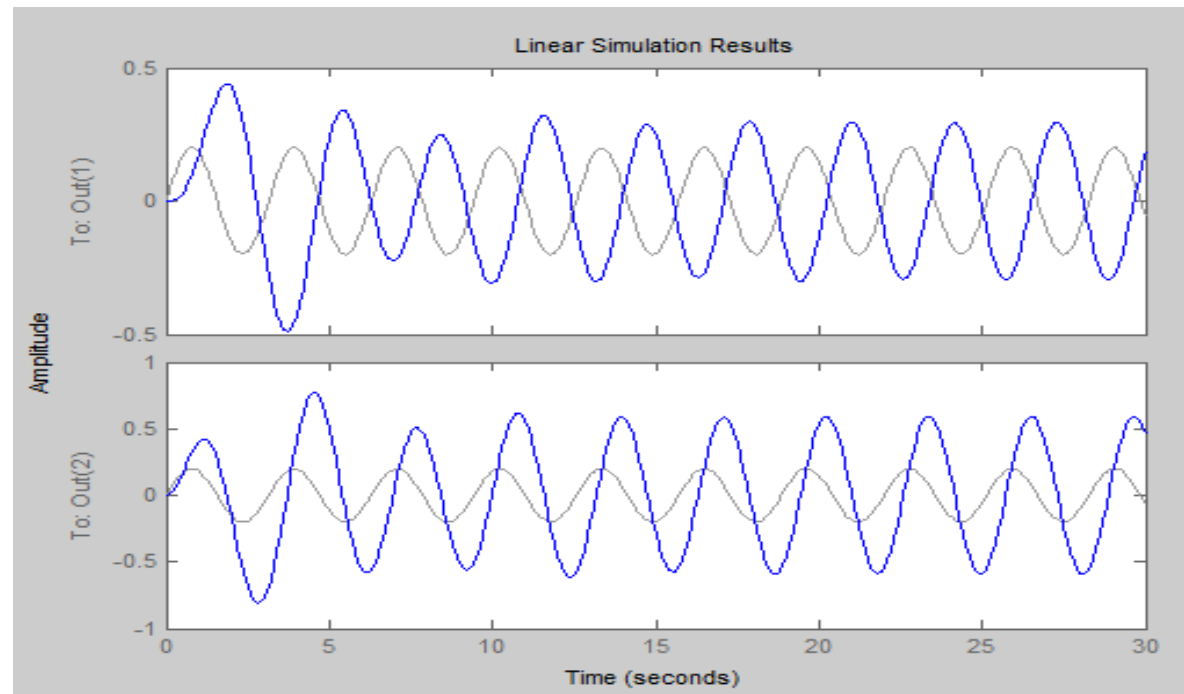
sys = ss( A, B, C, D );
t = 0:0.1:30;
for i = 1:length(t)
    u(i) = 4;
end
lsim(sys, u, t, [0 0]);
```



Other Cool MATLAB tricks (lsim)

$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

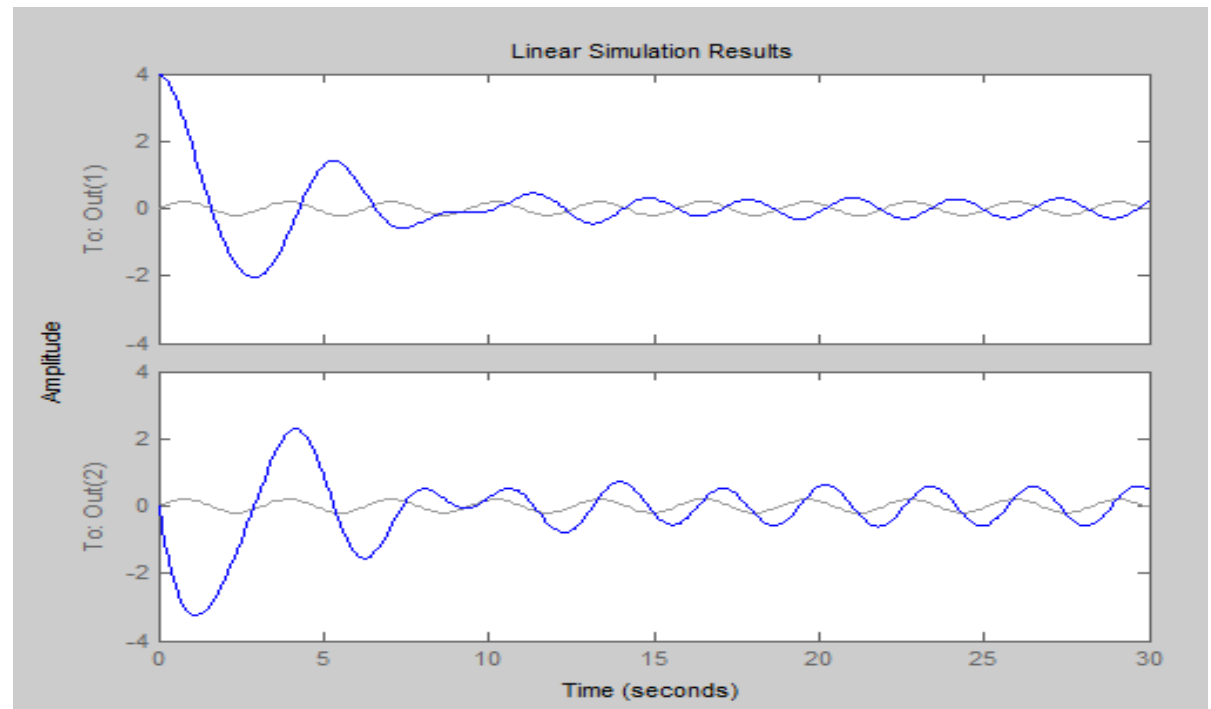
```
clear all;  
  
m = 2; c = 1; k = 3; F = 4;  
  
A = [ 0 1; -(k/m) -(c/m) ];  
B = [ 0; F ];  
C = [ 1 0; 0 1 ];  
D = [ 0; 0 ];  
  
sys = ss( A, B, C, D );  
u = 0.2*sin(2*t);  
lsim(sys, u, t, [0 0 ]);
```



Other Cool MATLAB tricks

$$\mathbf{y}(t) = \underbrace{\mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0)}_{\text{Transient response}} + \underbrace{\int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)}_{\text{Steady-state response}}$$

```
clear all;  
  
m = 2; c = 1; k = 3; F = 4;  
  
A = [ 0 1; -(k/m) -(c/m) ];  
B = [ 0; F ];  
C = [ 1 0; 0 1 ];  
D = [ 0; 0 ];  
  
sys = ss( A, B, C, D );  
u = 0.2*sin(2*t);  
lsim(sys, u, t, [4 0 ]);
```



Summary

$$y(t) = c e^{at} x(0) + \int_0^t c e^{a(t-\tau)} u(\tau) d\tau + d u(t)$$

is known as the convolution equation. We have just derived it using the Matrix Exponential for a system of 1st order differential equations.

$$\mathbf{y}(t) = \mathbf{C} e^{\mathbf{A}t} \mathbf{z}(0) + \int_0^t \mathbf{C} e^{\mathbf{A}(t-\tau)} \mathbf{B} u(\tau) d\tau + \mathbf{D} u(t)$$

- Now have complete analytical solution for sets of first order equations
 - Handle non-zero initial conditions
 - Handle the summation of scalar inputs
- MATLAB provides many tools to solve these equations for you.