



# Forced Response

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Industrial AI Lab.**

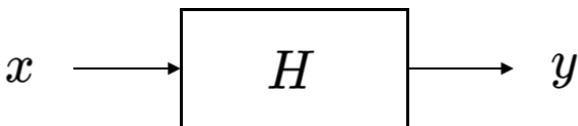
# Outline

- LTI Systems
- Time Response to Constant Input
- Time Response to Singularity Function Inputs
- Response to General Inputs (in Time)
- Response to Sinusoidal Input (in Frequency)
- Response to Periodic Input (in Frequency)
- Response to General Input (in Frequency)
- Fourier Transform

# Linear Time-Invariant (LTI) Systems

# Systems

- $H$  is a transformation (a rule or formula) that maps an input signal  $x(t)$  into a time output signal  $y(t)$



- System examples

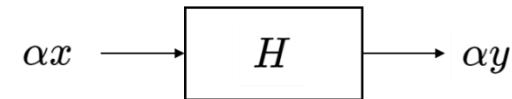
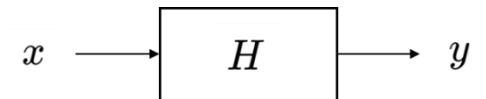
Identity	$y(t) = x(t)$	$\forall t$
Scaling	$y(t) = 2x(t)$	$\forall t$
Offset	$y(t) = x(t) + 2$	$\forall t$
Square signal	$y(t) = (x(t))^2$	$\forall t$
Shift	$y(t) = x(t + 2)$	$\forall t$
Decimate	$y(t) = x(2t)$	$\forall t$
Square time	$y(t) = x(t^2)$	$\forall t$

# Linear Systems

- A system  $H$  is linear if it satisfies the following two properties:

- Scaling

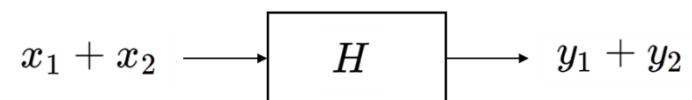
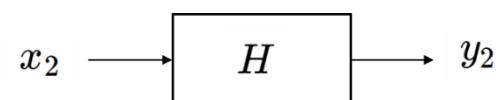
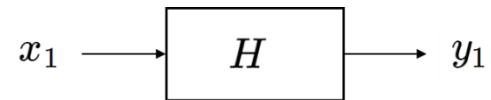
$$H\{\alpha x\} = \alpha H\{x\} \quad \forall \alpha \in \mathbb{C}$$



- Additivity

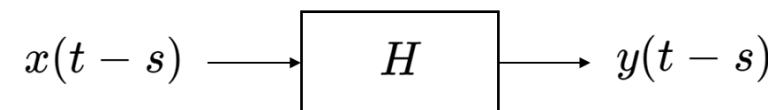
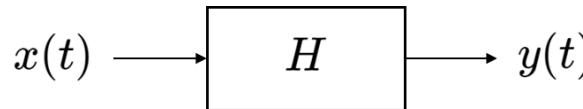
If  $y_1 = H\{x_1\}$  and  $y_2 = H\{x_2\}$

then  $H\{x_1 + x_2\} = y_1 + y_2$



# Time-Invariant Systems

- A system  $H$  processing infinite-length signals is time-invariant (shift-invariant) if a time shift of the input signal creates a corresponding time shift in the output signal



# Linear Time-Invariant (LTI) Systems

- We will only consider Linear Time-Invariant (LTI) systems
- Examples

Identity	$y(t) = x(t)$	$\forall t$	Linear	Time Invariant
Scaling	$y(t) = 2x(t)$	$\forall t$	Linear	Time Invariant
Offset	$y(t) = x(t) + 2$	$\forall t$	Non Linear	Time Invariant
Square signal	$y(t) = (x(t))^2$	$\forall t$	Non Linear	Time Invariant
Shift	$y(t) = x(t + 2)$	$\forall t$	Linear	Time Invariant
Decimate	$y(t) = x(2t)$	$\forall t$	Linear	Time Variant
Square time	$y(t) = x(t^2)$	$\forall t$	Linear	Time Variant

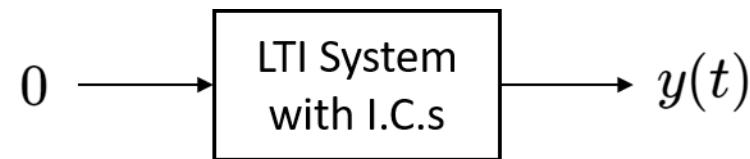
# Time Response to Constant Input

# Natural Response

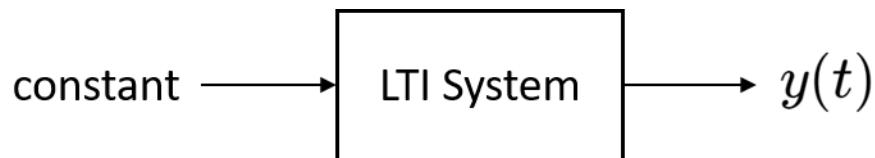
- So far, natural response of zero input with non-zero initial conditions are examined

$$\dot{y} + \frac{1}{\tau}y = 0$$

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2y = 0$$



# Response to Non-Zero Constant Input



- Assume all the systems are stable
- Inhomogeneous ODE

$$\dot{y} + \frac{1}{\tau}y = 0$$



$$\dot{y} + \frac{1}{\tau}y = q'$$

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2y = 0$$

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2y = p'$$

- Same dynamics, but it reaches different steady state
- Good enough to sketch

$$(\dot{y} - \dot{q}) + \frac{1}{\tau}(y - q) = 0$$

$$(\ddot{y} - \ddot{p}) + 2\zeta\omega_n(\dot{y} - \dot{p}) + \omega_n^2(y - p) = 0$$

# Response to Non-Zero Constant Input

- Dynamic system response = transient + steady state
- Transient response is present in the short period of time immediately after the system is turned on
  - It will die out if the system is stable
- The system response in the long run is determined by its steady state component only
- In steady state, all the transient responses go to zero

$$\frac{dy}{dt} \rightarrow 0$$

# Example

- Example

$$\dot{y} + \frac{1}{\tau}y = q, \quad y(0) = 0$$

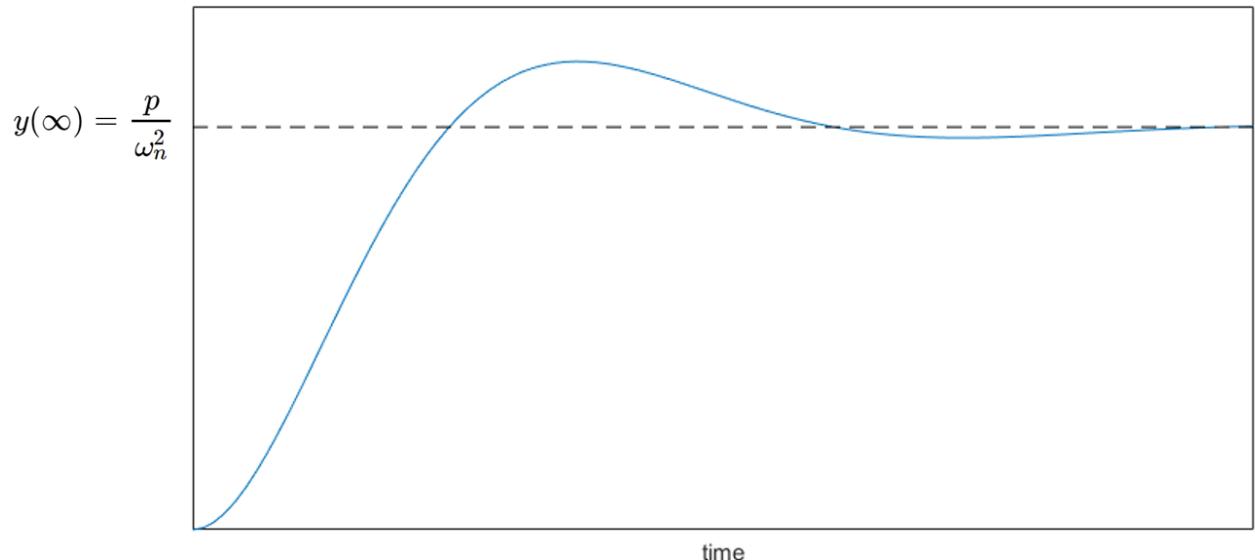
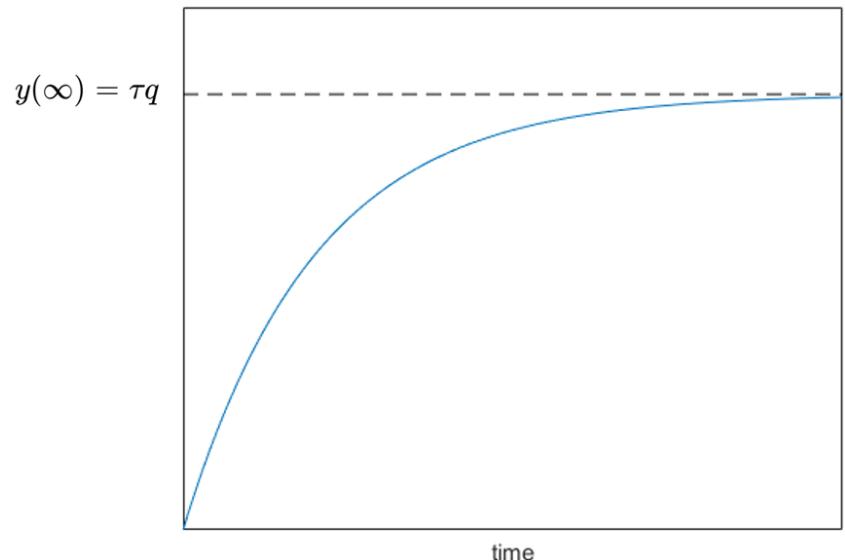
$$\dot{y}(\infty) = 0$$

$$\implies y(\infty) = \tau q$$

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2y = p, \quad y(0) = 0, \dot{y}(0) = 0$$

$$\ddot{y}(\infty) = \dot{y}(\infty) = 0$$

$$\implies y(\infty) = \frac{p}{\omega_n^2}$$

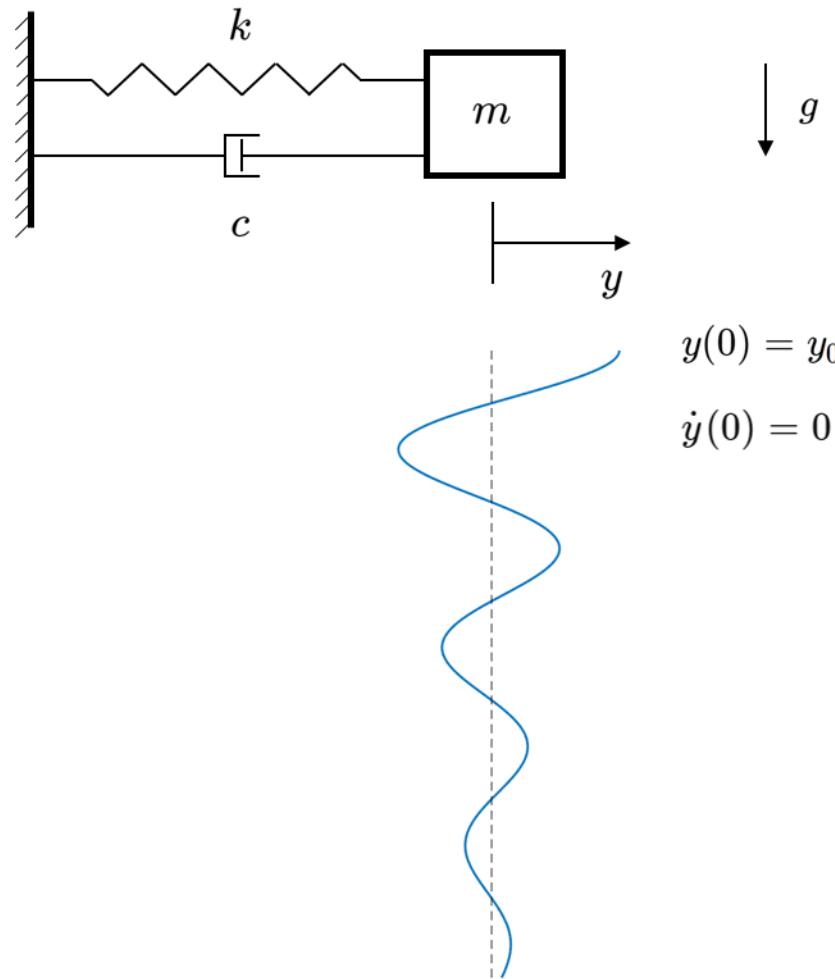


# Response to Non-Zero Constant Input

- Think about mass-spring-damper system in horizontal setting

$$\ddot{y} + \frac{c}{m}\dot{y} + \frac{k}{m}y = 0,$$

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



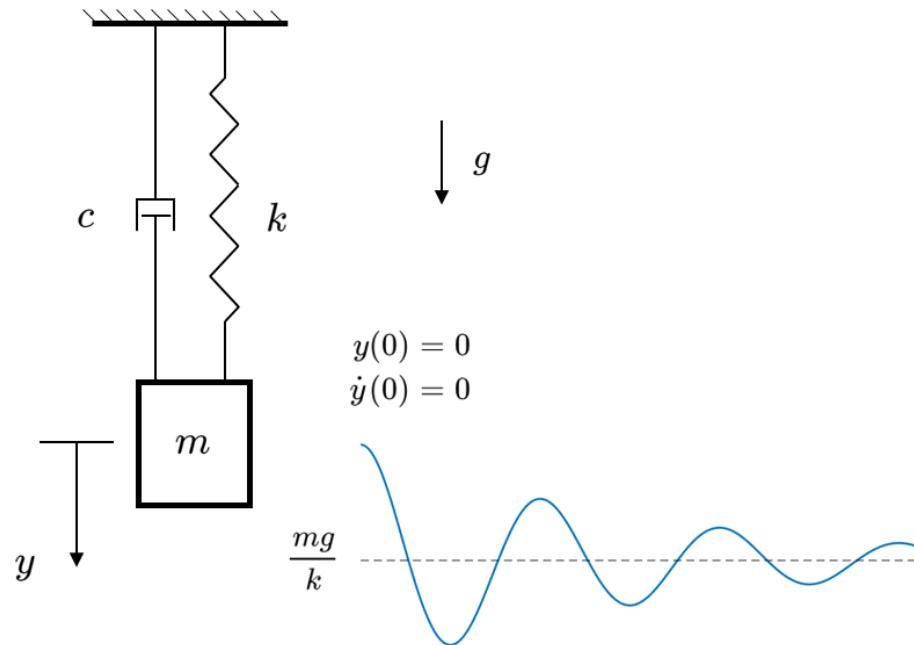
$$y(0) = y_0, \\ \dot{y}(0) = 0$$

# Response to Non-Zero Constant Input

- Mass-spring-damper system in vertical setting
- $y(0) = 0$  no initial displacement
- $\dot{y}(0) = 0$  initially at rest

$$m\ddot{y} + c\dot{y} + ky = mg$$
$$\ddot{y} + \frac{c}{m}\dot{y} + \frac{k}{m}y = g,$$

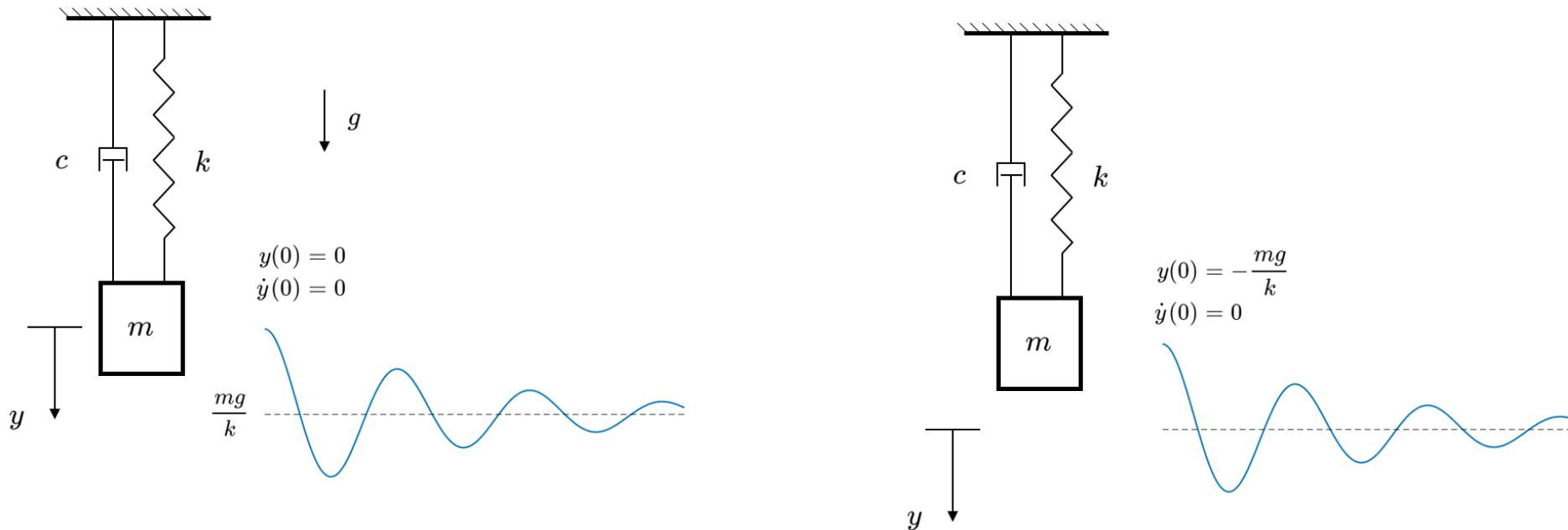
where  $y(0) = 0$ ,  $\dot{y}(0) = 0$



# Response to Non-Zero Constant Input

- Shift the origin of  $y$  axis to the static equilibrium point, then act like a natural response with
- $y(0) = -\frac{mg}{k}$  and  $\dot{y}(0) = 0$  as initial conditions

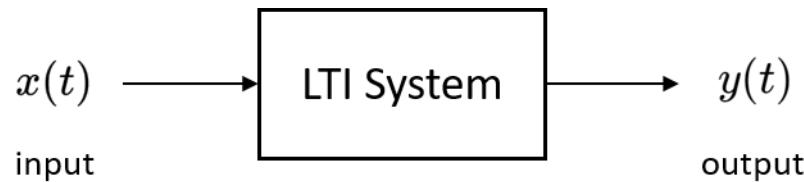
$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2 y = \omega_n^2 p \quad \longrightarrow \quad (\ddot{y} - \ddot{p}) + 2\zeta\omega_n(\dot{y} - \dot{p}) + \omega_n^2(y - p) = 0$$



# Time Response to Singularity Function Inputs

# Time Response to General Inputs

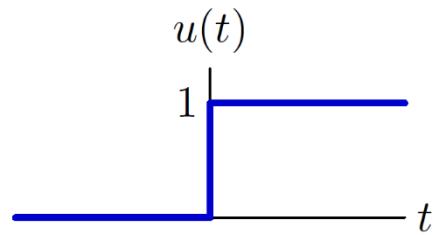
- We studied output response  $y(t)$  when input  $x(t)$  is constant
- Ultimate Goal: output response of  $y(t)$  to general input  $x(t)$



- Consider singularity function inputs first
  - Step function
  - Impulse function (Delta Dirac function)

# Step Function

- Step function



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

# Step Response

- Start with a step response example

$$\dot{x} + 5x = 1 \quad \text{for } t \geq 0, \quad x(0) = 0$$

- Or

$$\dot{x} + 5x = u(t), \quad x(0) = 0 \quad \text{where } u(t) = \begin{cases} 1 & t > 0 \\ 0 & \text{otherwise} \end{cases}$$

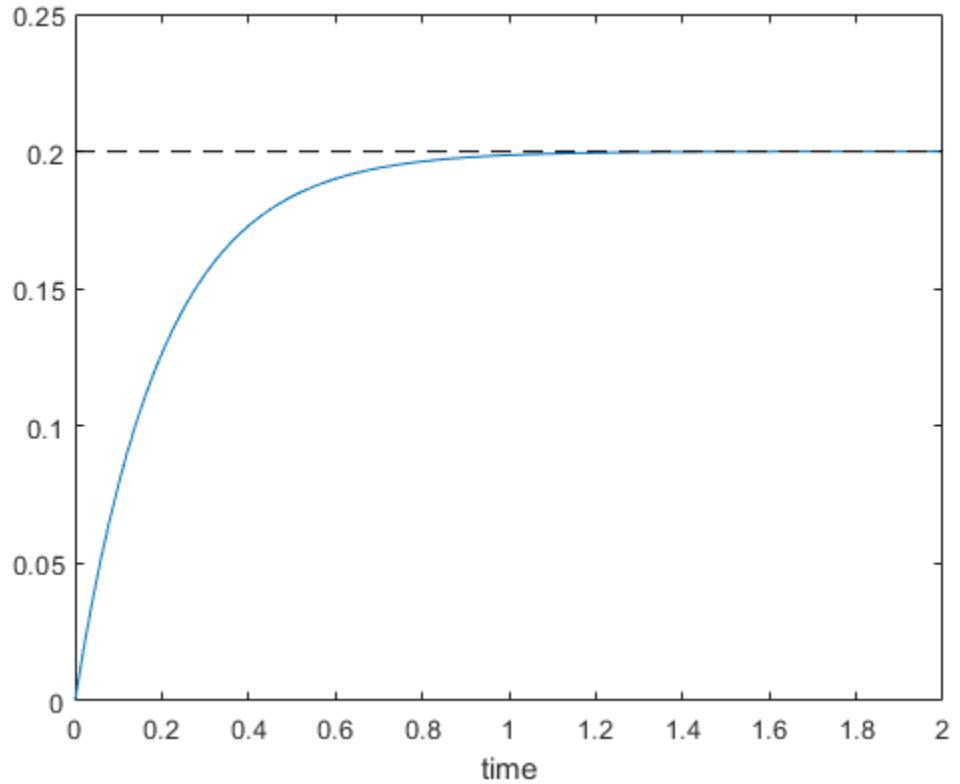
- The solution is given:

$$x(t) = \frac{1}{5}(1 - e^{-5t})$$

# Step Response

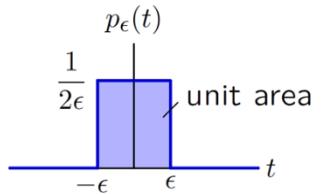
```
t = linspace(0,2,100);
x = 1/5*(1-exp(-5*t));

plot(t,x,t,0.2*ones(size(t)), 'k--')
ylim([0,0.25])
xlabel('time')
```

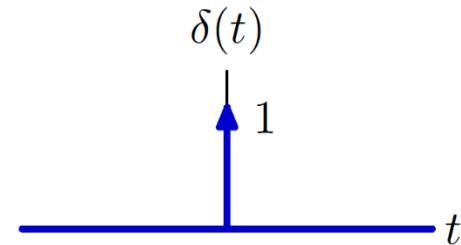
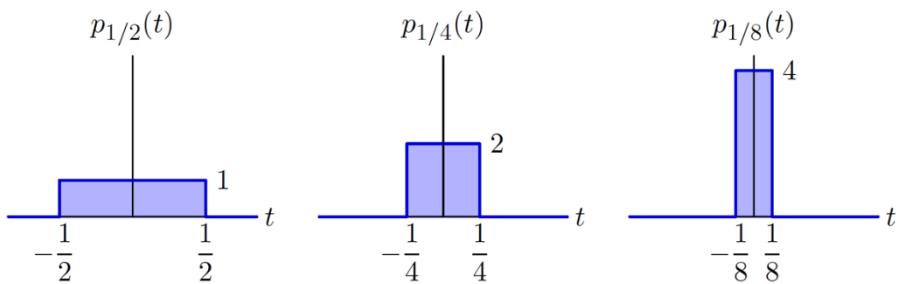


# Impulse

- Impulse: difficult to image
- The unit-impulse signal acts as a pulse with unit area but zero width



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



- The unit-impulse function is represented by an arrow with the number 1, which represents its area
- It has two seemingly contradictory properties :
  - It is nonzero only at  $t = 0$  and
  - Its definite integral  $(-\infty, \infty)$  is 1

# Properties of Dirac Delta Function

- The Dirac delta can be loosely thought of as a function on the real line which is zero everywhere except at the origin, where it is infinite,

$$\delta(t) = \begin{cases} \infty & x = 0 \\ 0 & x \neq 0 \end{cases}$$

and which is also constrained to satisfy the identity

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

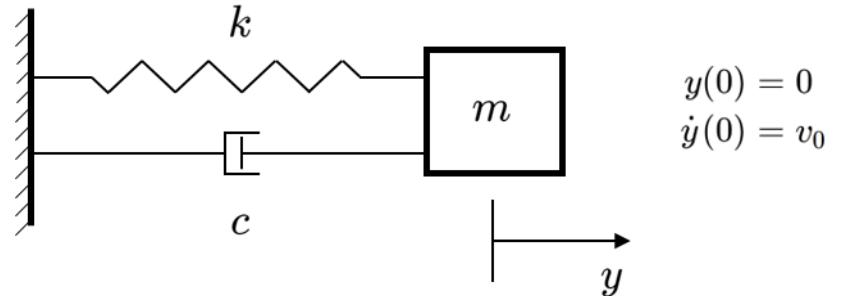
- Sifting property

$$\begin{aligned} \int_{-\infty}^{\infty} x(t) \delta(t) dt &= \int_{-\infty}^{0^-} x(t) \delta(t) dt + \int_{0^-}^{0^+} x(t) \delta(t) dt + \int_{0^+}^{\infty} x(t) \delta(t) dt \\ &= 0 + x(0) \int_{0^-}^{0^+} \delta(t) dt + 0 \\ &= x(0) \end{aligned}$$

$$\int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau = x(t)$$

# Impulse Response

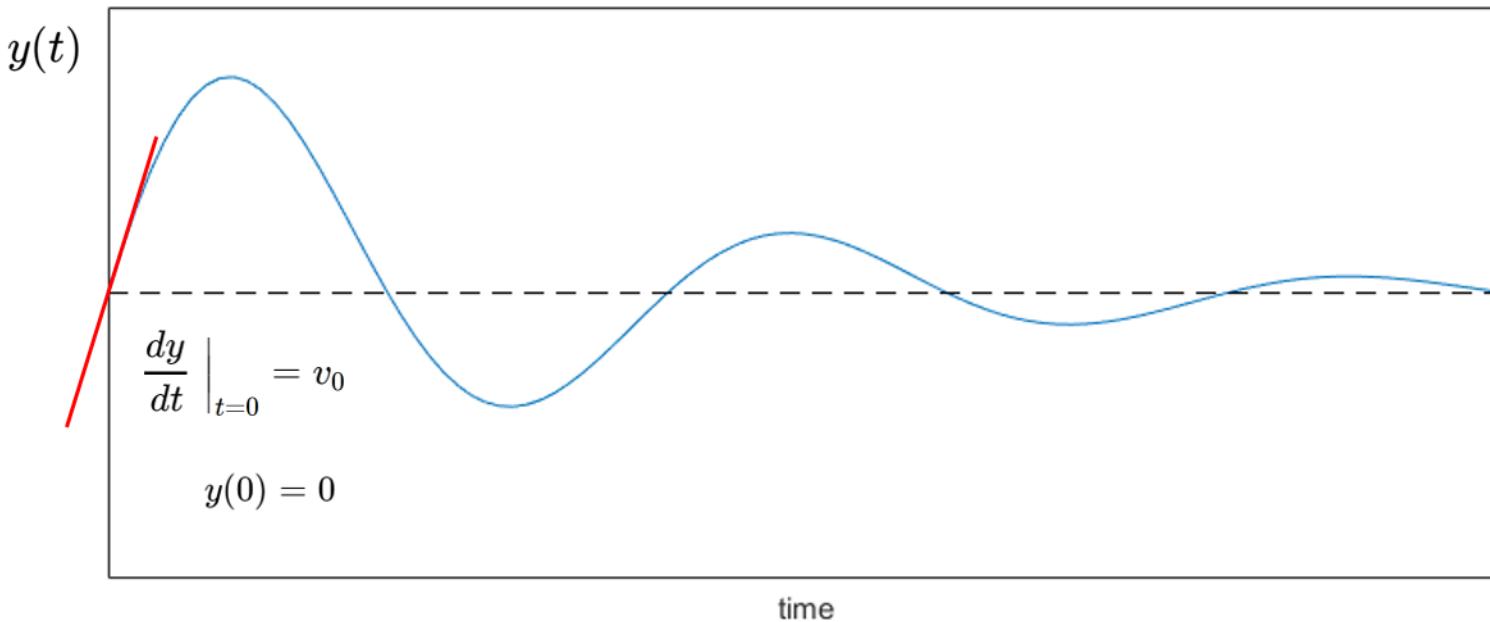
- Impulse response: difficult to image
- Question: how to realize initial velocity of  $v_0 \neq 0$



- Momentum and impulse in physics I
  - Consider an "impulse" which is a sudden increase in momentum  $0 \rightarrow mv$  of an object applied at time 0
  - To model this,
$$mv - 0 = \int_{0_-}^{0^+} f(t)dt$$
where force  $f(t)$  is strongly peaked at time 0
  - Actually the details of the shape of the peak are not important, what is important is the area under the curve
$$f(t) = mv\delta(t)$$
  - This is the motivation that mathematician and physicist invented the delta Dirac function

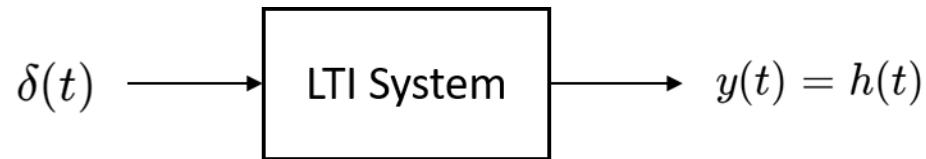
$$m\ddot{y} + c\dot{y} + ky = mv_0\delta(t) \quad \ddot{y} + \frac{c}{m}\dot{y} + \frac{k}{m}y = v_0\delta(t)$$

# Impulse Response



$$v_0\delta(t) \quad \longrightarrow \quad \begin{aligned} y(0) &= 0 \\ \dot{y}(0) &= v_0 \end{aligned} \quad \longrightarrow \quad \text{natural response}$$

# Impulse Response to LTI system



$$\ddot{y} + 2\zeta\omega_n \dot{y} + \omega_n^2 y = \delta(t), \quad \begin{aligned} \delta(t) &: \text{input} \\ y &: \text{output} \end{aligned}$$

- Later, we will discuss why the impulse response is so important to understand an LTI system

# Impulse Response to LTI system

- Example: now think about the impulse response

$$\dot{y} + 5y = \delta(t), \quad y(0) = 0$$

- The solution is given: (why?)

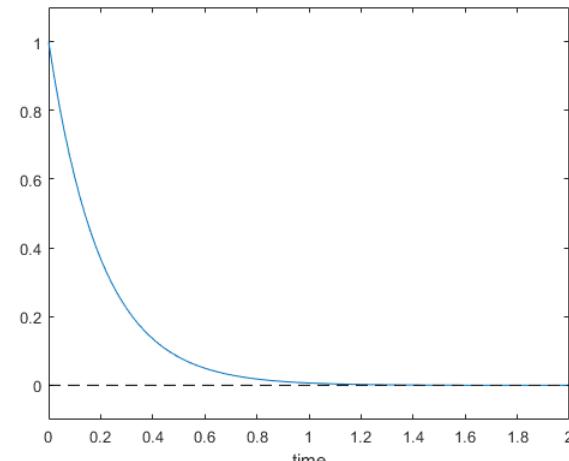
$$y(t) = h(t) = e^{-5t}, \quad t \geq 0$$

- Impulse input can be equivalently changed to zero input with non-zero initial condition

$$\int_{0^-}^{0^+} \delta(t) dt = u(0^+) - u(0^-) = 1$$
$$\dot{y} + 5y = 0, \quad y(0) = 1$$

```
t = linspace(0,2,100);
h = exp(-5*t);

plot(t,h,t,zeros(size(t)), 'k--'),
ylim([-0.1,1.1])
xlabel('time')
```

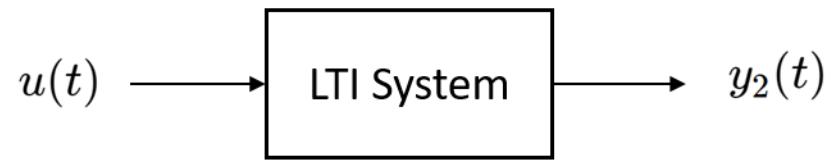
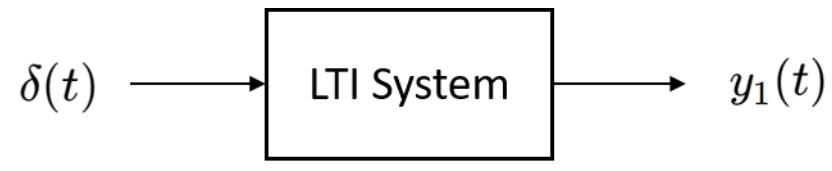


# Step Response Again

- Relationship between impulse response and unit-step response

$$\ddot{x} + 2\zeta\omega_n \dot{x} + \omega_n^2 x = u(t) \quad \text{or}$$

$$\ddot{x} + 2\zeta\omega_n \dot{x} + \omega_n^2 x = 1, \quad t > 0$$



$$\frac{du(t)}{dt} = \delta(t)$$

$$\frac{dy_2(t)}{dt} = y_1(t)$$

- Impulse response is the derivative of the step response

# **Response to General Inputs (in Time)**

# Response to a General Input (in Time)

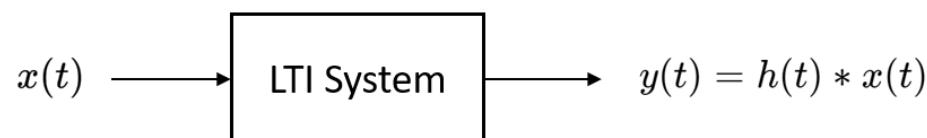
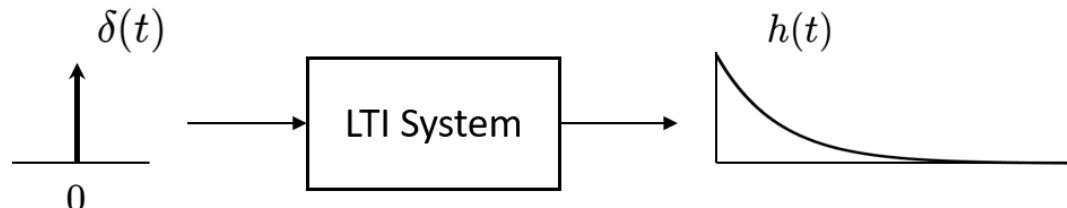
- Finally, think about response to a "general input" in time

$$\dot{y} + 5y = x(t), \quad y(0) = 0$$

- The solution is given

$$y(t) = h(t) * x(t) \quad \text{where } h(t) \text{ is impulse response}$$

- If this is true, we can compute output response to any general input if an impulse response is given
  - Impulse response = LTI system

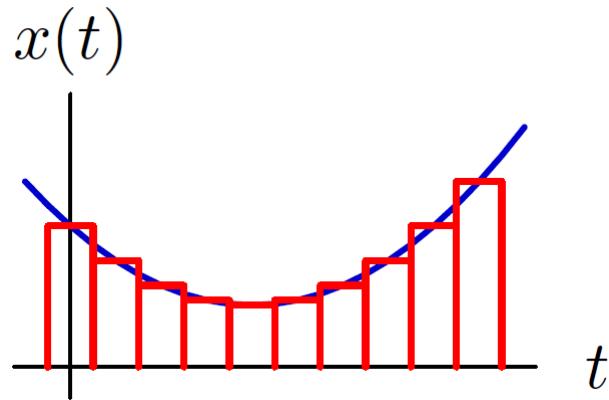


# Convolution: Definition

- $y(t)$  is the integral of the product of two functions after one is reversed and shifted by  $t$

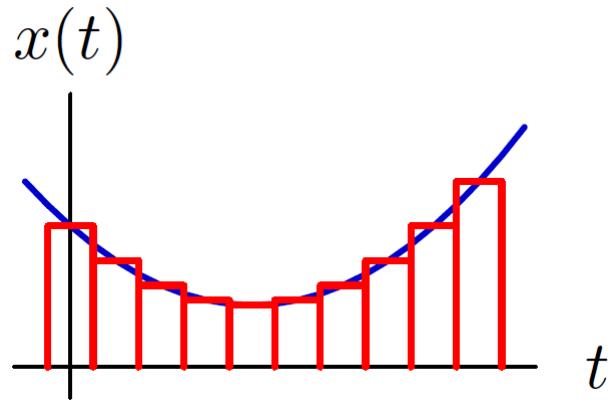
$$\begin{aligned}y(t) &= h(t) * x(t) = \int_{-\infty}^{\infty} h(\tau)x(t - \tau)d\tau \\&= \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau\end{aligned}$$

# Easier Way to Understand Continuous Time Signal

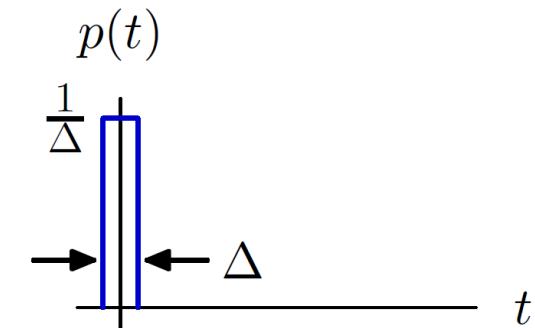


$$x(t) = \lim_{\Delta \rightarrow 0} \sum_k x(k\Delta)p(t - k\Delta)\Delta$$

# Easier Way to Understand Continuous Time Signal



$$x(t) = \lim_{\Delta \rightarrow 0} \sum_k x(k\Delta)p(t - k\Delta)\Delta$$

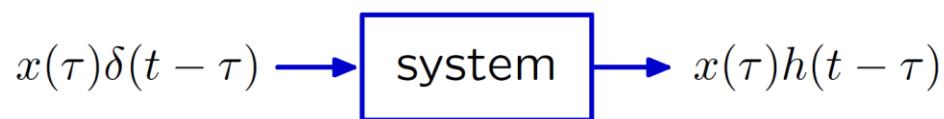
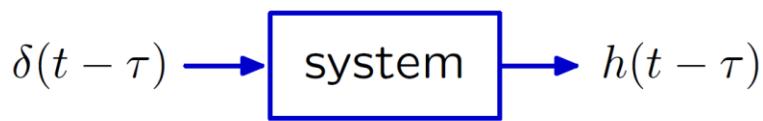
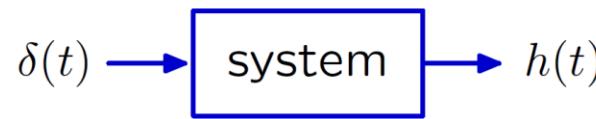
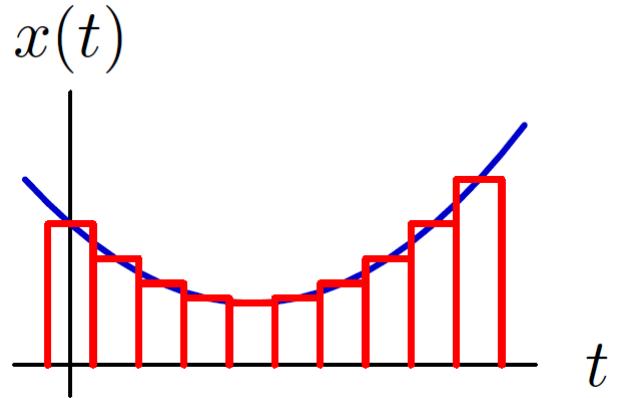


As  $\Delta \rightarrow 0$ ,  $k\Delta \rightarrow \tau$ ,  $\Delta \rightarrow d\tau$ , and  $p(t) \rightarrow \delta(t)$ :

$$x(t) \rightarrow \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau$$

# Structure of Superposition

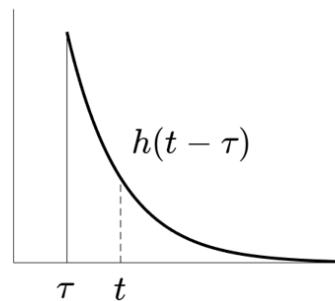
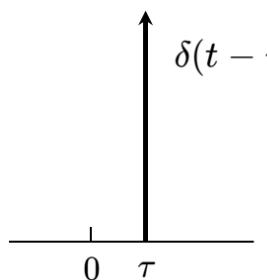
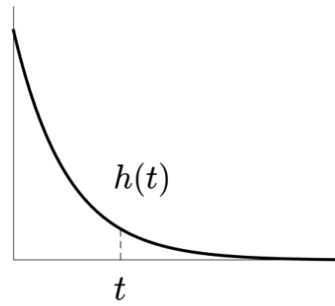
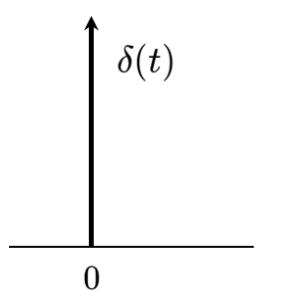
- If a system is linear and time-invariant (LTI) then its output is the integral of weighted and shifted unit-impulse responses.



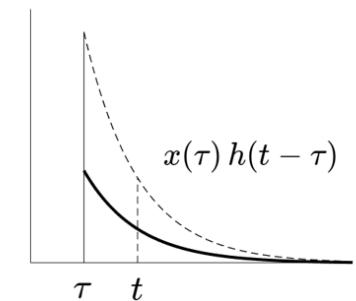
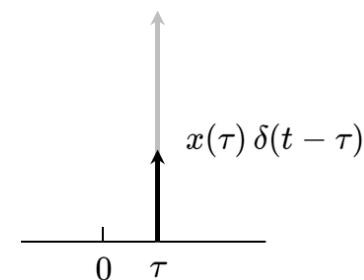
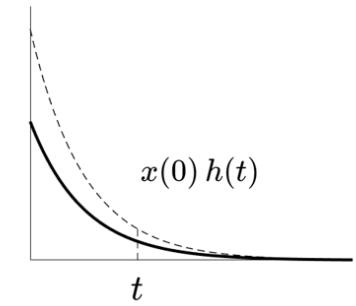
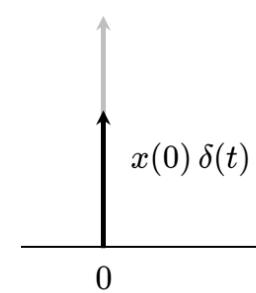
$$x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau \rightarrow \text{system} \rightarrow y(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau$$

# Impulse Response to LTI System

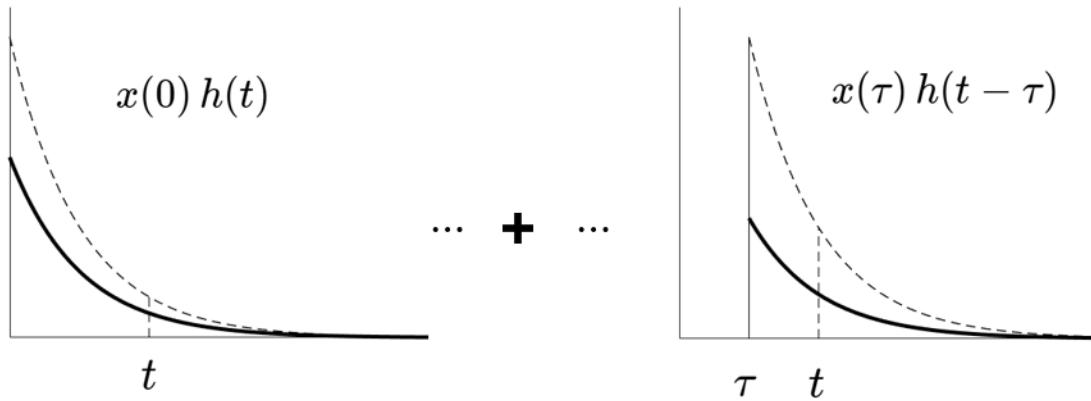
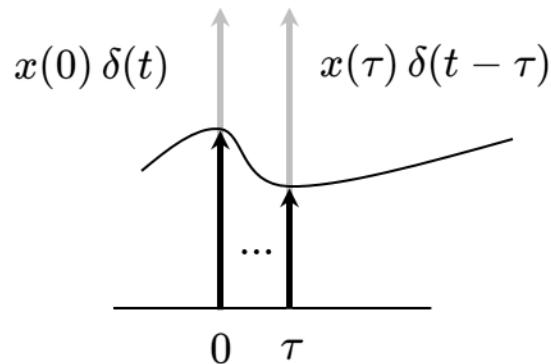
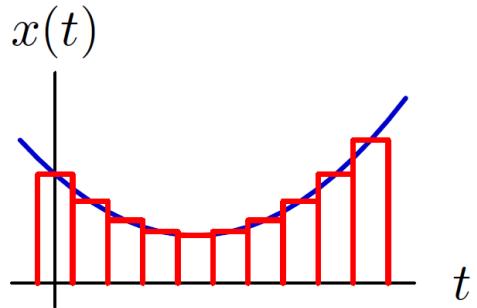
Time-invariant



Linear (scaling)



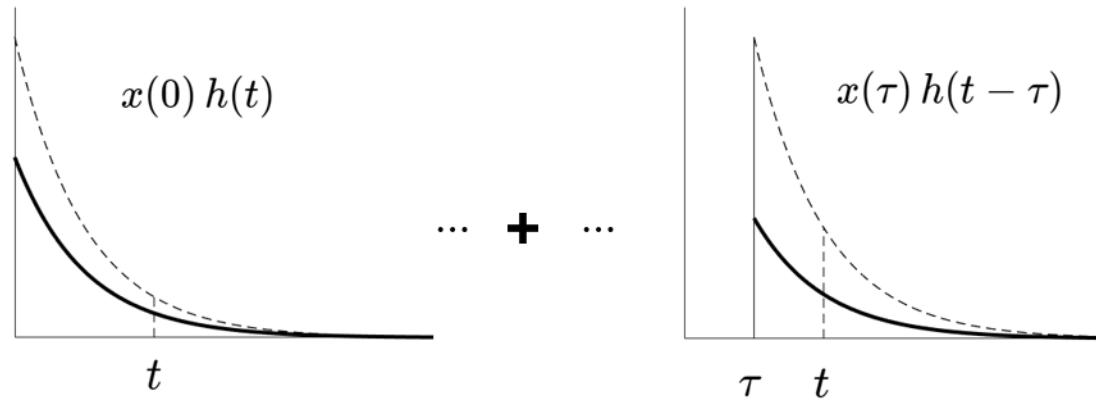
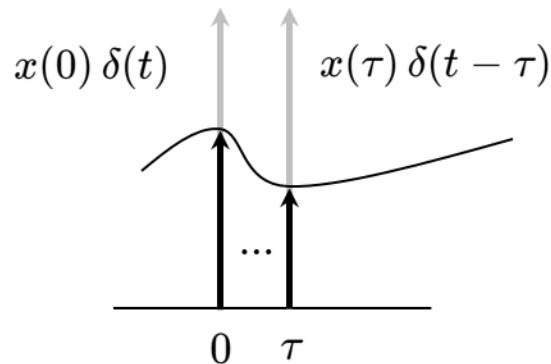
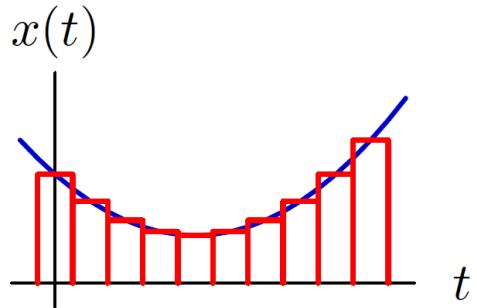
# Response to Arbitrary Input $x(t)$ (1/2)



$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau \quad \rightarrow$$

$$y(t) = \int_{-\infty}^{\infty} x(\tau) h(t - \tau) d\tau = x(t) * h(t)$$

## Response to Arbitrary Input $x(t)$ (2/2)



$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

$$y(t) = \int_{-\infty}^{\infty} x(\tau) h(t - \tau) d\tau = x(t) * h(t)$$

# Response to Arbitrary Input: MATLAB (1/2)

- Example

$$\dot{y} + 3y = 3x(t)$$

- The solution is given:

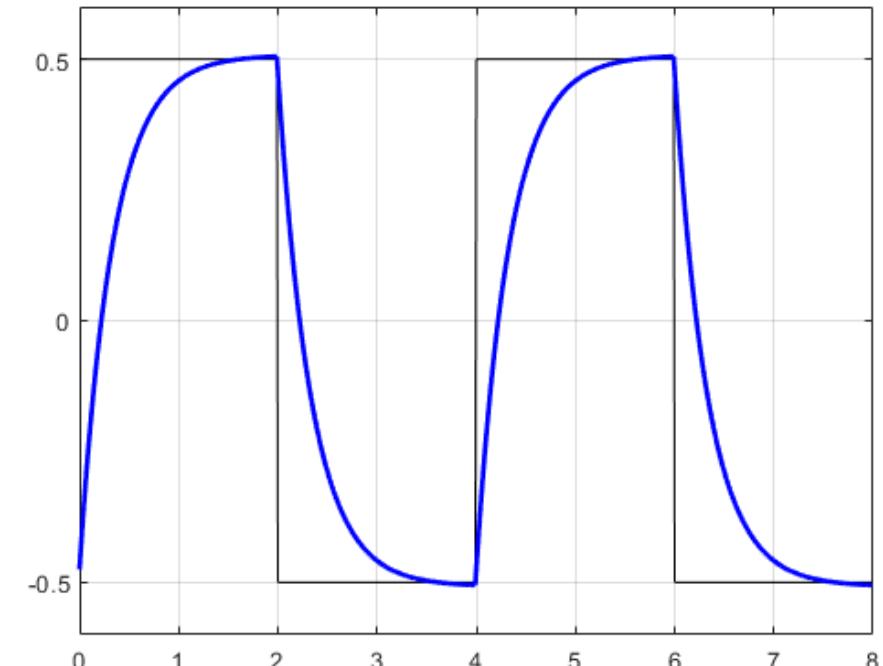
$$y(t) = h(t) * x(t), \quad \text{where } h(t) = 3e^{-3t}$$

```
w0 = 2*pi/T;
x = 1/2*square(w0*t);
```

% convolution (circular, will be discussed later)

```
→ h = 3*exp(-3*t);
y = cconv(x,h,length(t))*Ts;

plot(t,x, 'k'), hold on
plot(t,y, 'b', 'linewidth',2), hold off
grid on
```



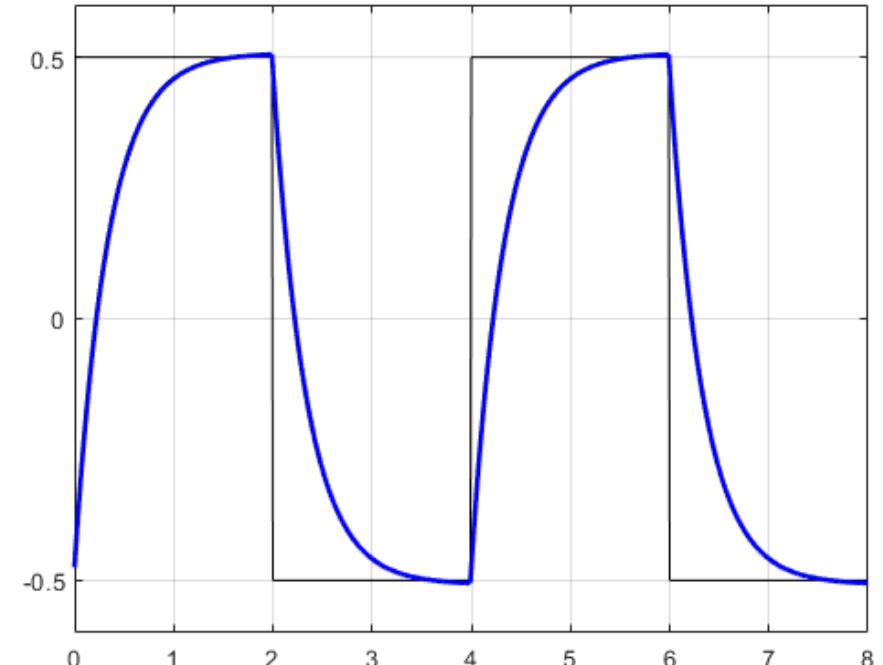
## Response to Arbitrary Input: MATLAB (2/2)

- Example

$$\dot{y} + 3y = 3x(t)$$

- The solution is given:

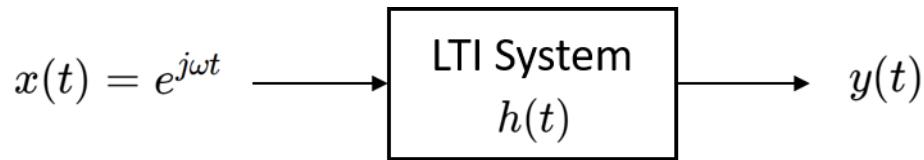
$$y(t) = h(t) * x(t), \quad \text{where } h(t) = 3e^{-3t}$$



# **Response to Sinusoidal Input (in Frequency)**

# Response to a Sinusoidal Input

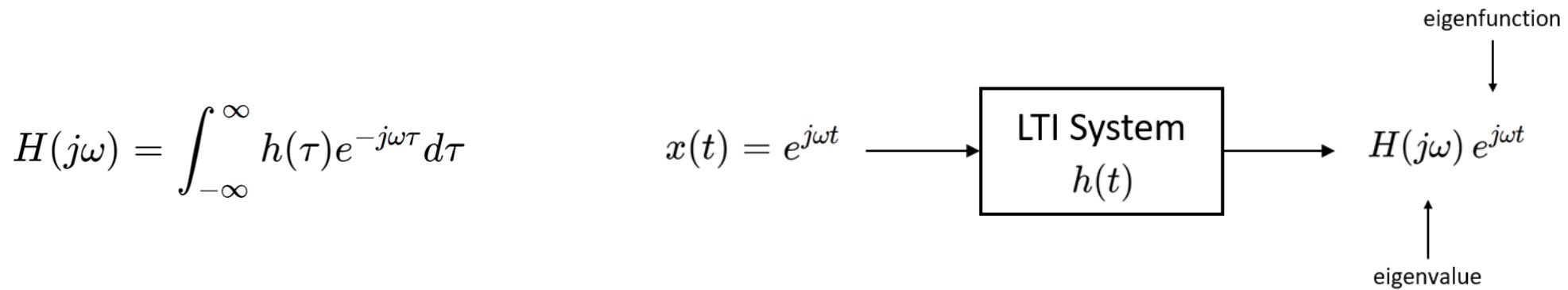
- When the input  $x(t) = e^{j\omega t}$  to an LTI system



$$\begin{aligned}y(t) &= h(t) * x(t) = \int_{-\infty}^{\infty} h(\tau) e^{j\omega(t-\tau)} d\tau = e^{j\omega t} \underbrace{\int_{-\infty}^{\infty} h(\tau) e^{-j\omega\tau} d\tau}_{\text{complex function of } \omega} \\&= e^{j\omega t} H(j\omega)\end{aligned}$$

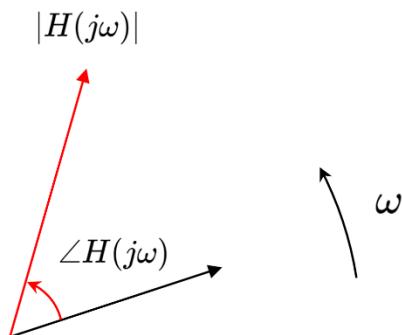
# Fourier Transform

- Definition: Fourier transform



- $H(j\omega)e^{j\omega t}$  rotates the same angular velocity  $\omega$

$$H(j\omega) = |H(j\omega)| e^{j\angle H(j\omega)}$$



# Response to a Sinusoidal Input: MATLAB

$$\dot{y} + 5y = x(t)$$

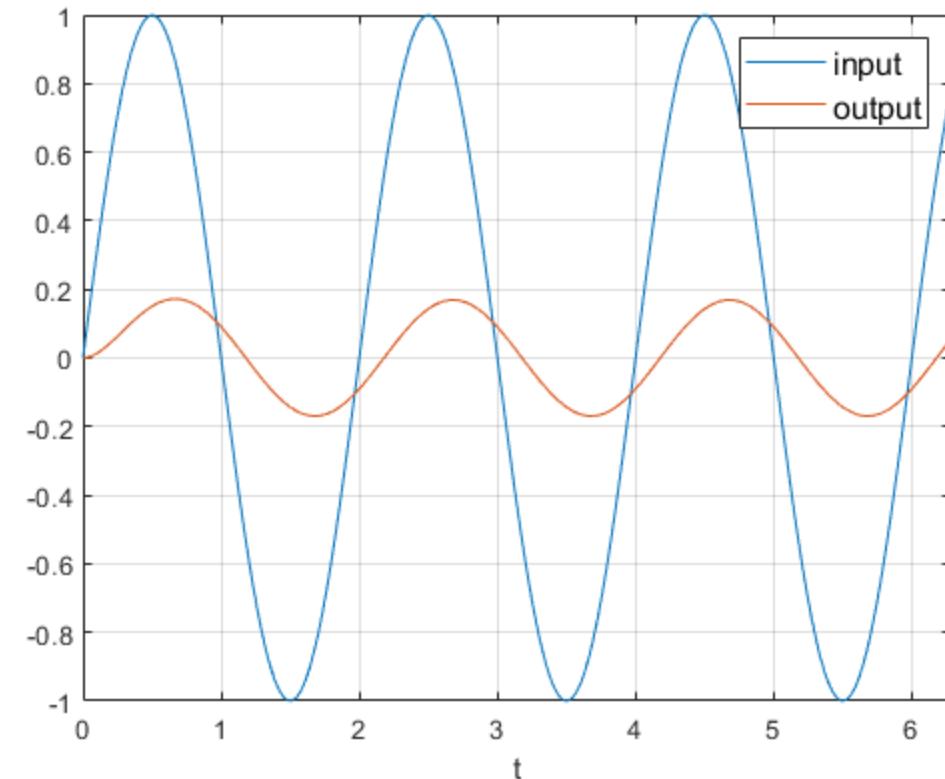
```
A = -5;
B = 1;
C = 1;
D = 0;
G = ss(A,B,C,D);

w = pi;
t = linspace(0,2*pi,200);

x0 = 0;

x = sin(w*t);

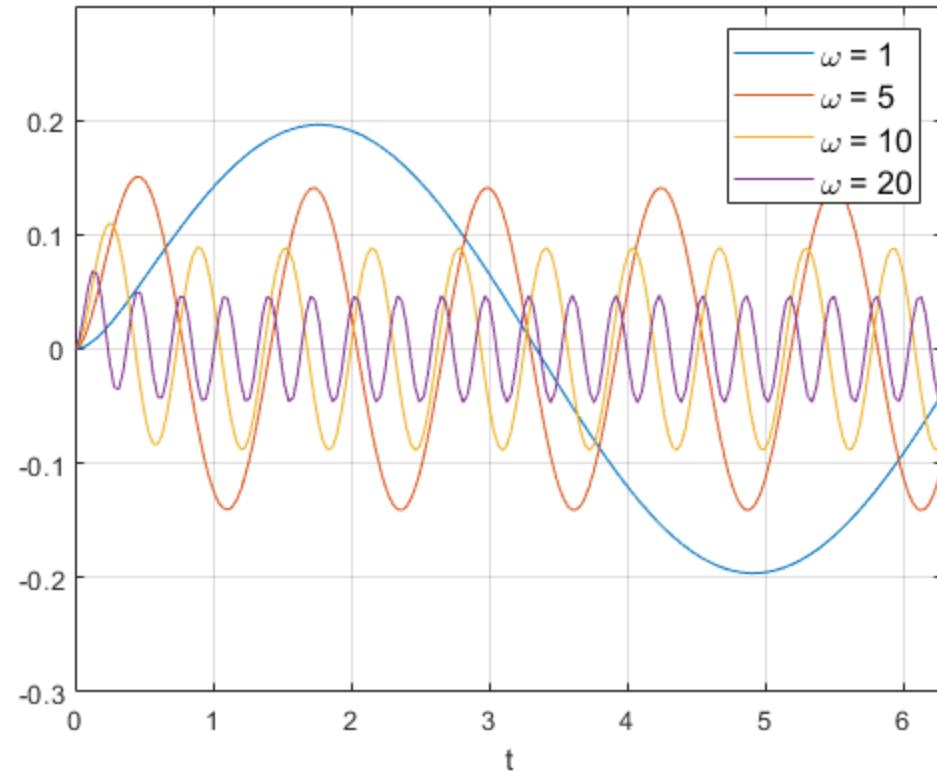
[y,tout] = lsim(G,x,t,x0);
```



# Response to a Sinusoidal Input: MATLAB

$$\dot{y} + 5y = x(t)$$

```
w = [1,5,10,20];
for w = W
    x = sin(w*t);
    [y,tout] = lsim(G,x,t,x0);
    plot(tout,y), hold on
end
```



# Response to Periodic Input (in Frequency)

# Response to a Periodic Input (in Frequency Domain)

- Periodic signal: Definition

$$x(t) = x(t + T)$$

Fundamental frequency  $\omega_0 = 2\pi f$

$$\text{Fundamental period } T = \frac{2\pi}{\omega_0}$$

- Fourier series represent periodic signals in terms of sinusoids (or complex exponential of  $e^{j\omega t}$ )
- Fourier series represent periodic signals by their harmonic components

$$x(t) = a_0 + a_1 e^{j\omega_0 t} + a_2 e^{j2\omega_0 t} + a_3 e^{j3\omega_0 t} + \dots$$

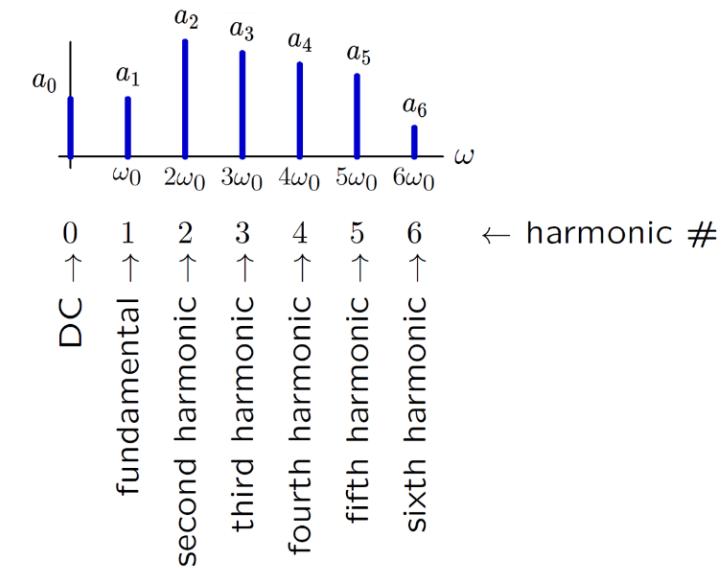
$$= \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$



# Response to a Periodic Input (in Frequency)

- Fourier series represent periodic signals by their harmonic components

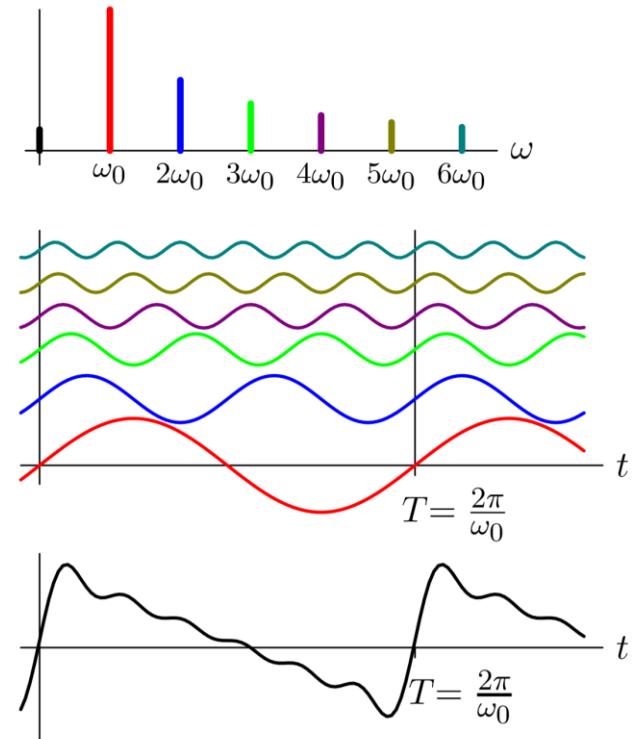
$$\begin{aligned}x(t) &= a_0 + a_1 e^{j\omega_0 t} + a_2 e^{j2\omega_0 t} + a_3 e^{j3\omega_0 t} + \dots \\&= \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}\end{aligned}$$



# Response to a Periodic Input (in Frequency)

- What signals can be represented by sums of harmonic components?

$$\begin{aligned}x(t) &= a_0 + a_1 e^{j\omega_0 t} + a_2 e^{j2\omega_0 t} + a_3 e^{j3\omega_0 t} + \dots \\&= \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}\end{aligned}$$



# Harmonic Representations

- It is possible to represent all periodic signals with harmonics
- Question: how to separate harmonic components given a periodic signal
- Underlying properties

$$e^{jk_1\omega_0 t} \cdot e^{jk_2\omega_0 t} = e^{j(k_1+k_2)\omega_0 t}$$

$$\int_t^{t+T} e^{jk\omega_0 \tau} d\tau = \begin{cases} 0 & k \neq 0 \\ T & k = 0 \end{cases}$$
$$= T\delta[k] \quad \text{where } \delta[k] \text{ is Kronecker delta}$$

# Harmonic Representations

- Assume that  $x(t)$  is periodic in  $T$  and is composed of a weighted sum of harmonics of  $\omega_0 = \frac{2\pi}{T}$

$$x(t) = x(t + T) = \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t}$$

- Then

$$\int_T x(t) e^{-jn\omega_0 t} dt = \int_T \sum_{k=-\infty}^{\infty} a_k e^{jk\omega_0 t} \cdot e^{-jn\omega_0 t} dt$$

$$= \sum_{k=-\infty}^{\infty} a_k \int_T e^{j(k-n)\omega_0 t} dt$$

$$= a_n T$$

$$\therefore a_k = \frac{1}{T} \int_T x(t) e^{-jk\omega_0 t} dt = \frac{1}{T} \int_T x(t) e^{-jk\frac{2\pi}{T} t} dt$$

# Fourier Series

- Fourier series: determine harmonic components of a periodic signal

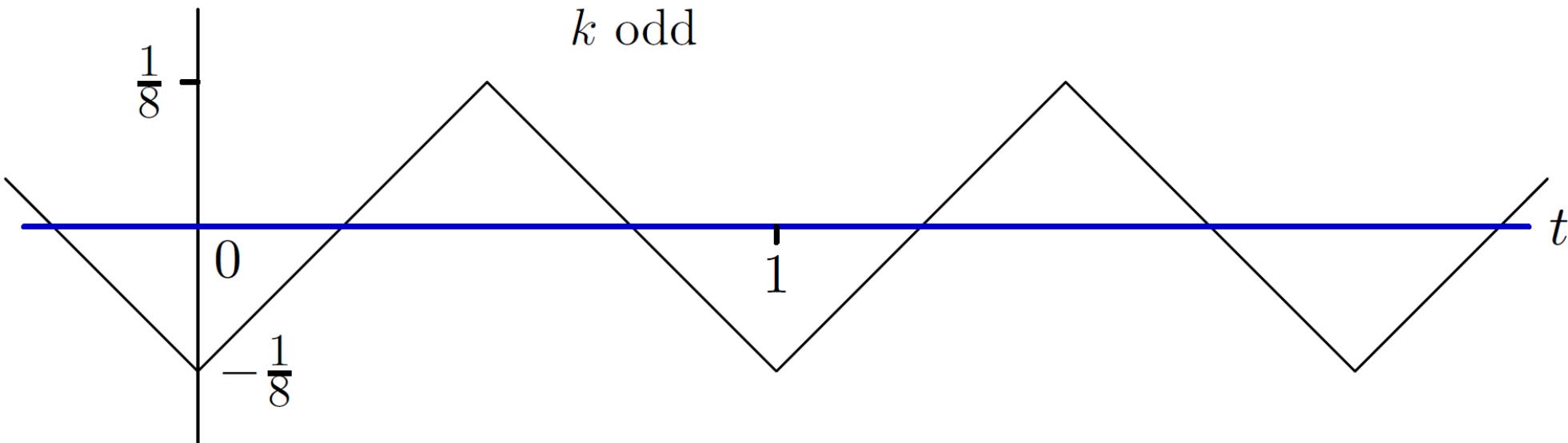
$$a_k = \frac{1}{T} \int_T x(t) e^{-j\omega_0 kt} dt, \quad \text{analysis}$$

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{j\omega_0 kt}, \quad \text{synthesis}$$

## Example: Triangle Waveform

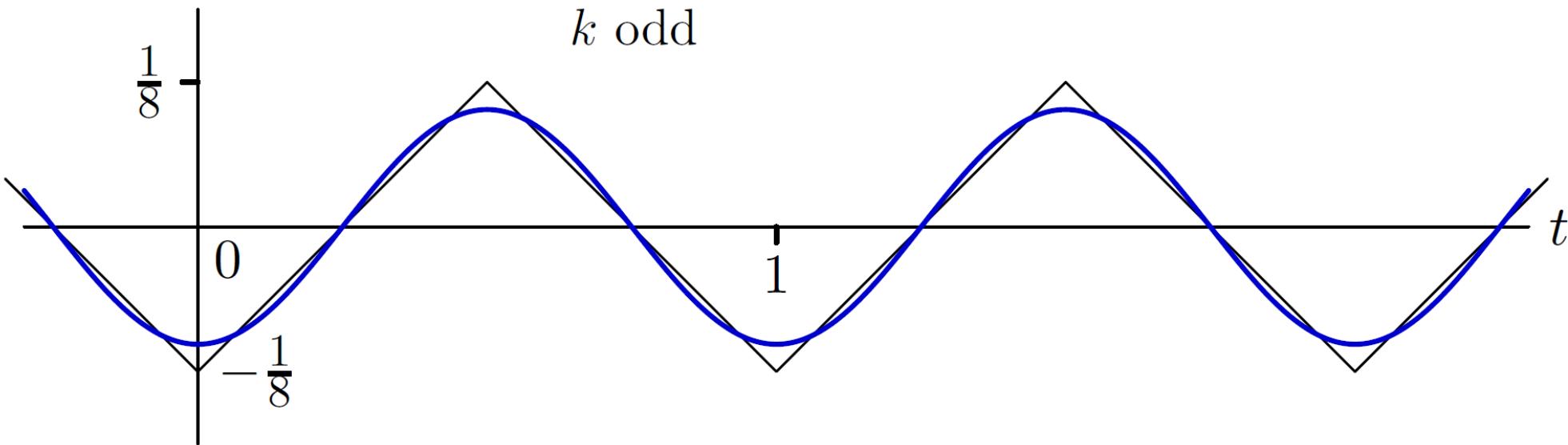
- One can visualize convergence of the Fourier Series by incrementally adding terms.

$$\sum_{\substack{k = -0 \\ k \text{ odd}}}^0 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



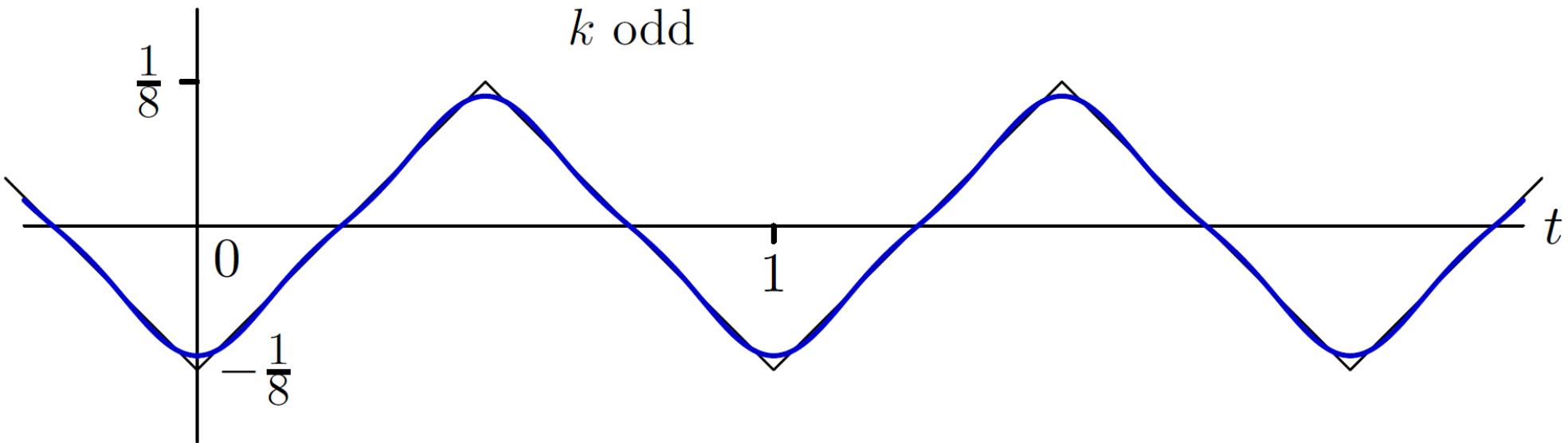
## Example: Triangle Waveform

$$\sum_{\substack{k = -1 \\ k \text{ odd}}}^1 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



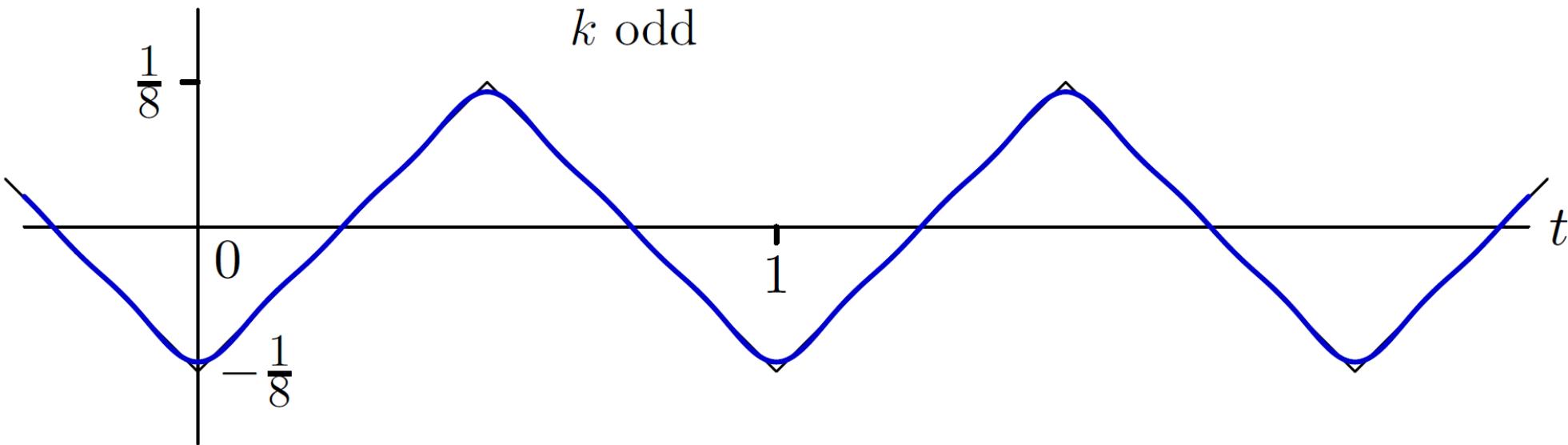
## Example: Triangle Waveform

$$\sum_{\substack{k = -3 \\ k \text{ odd}}}^3 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



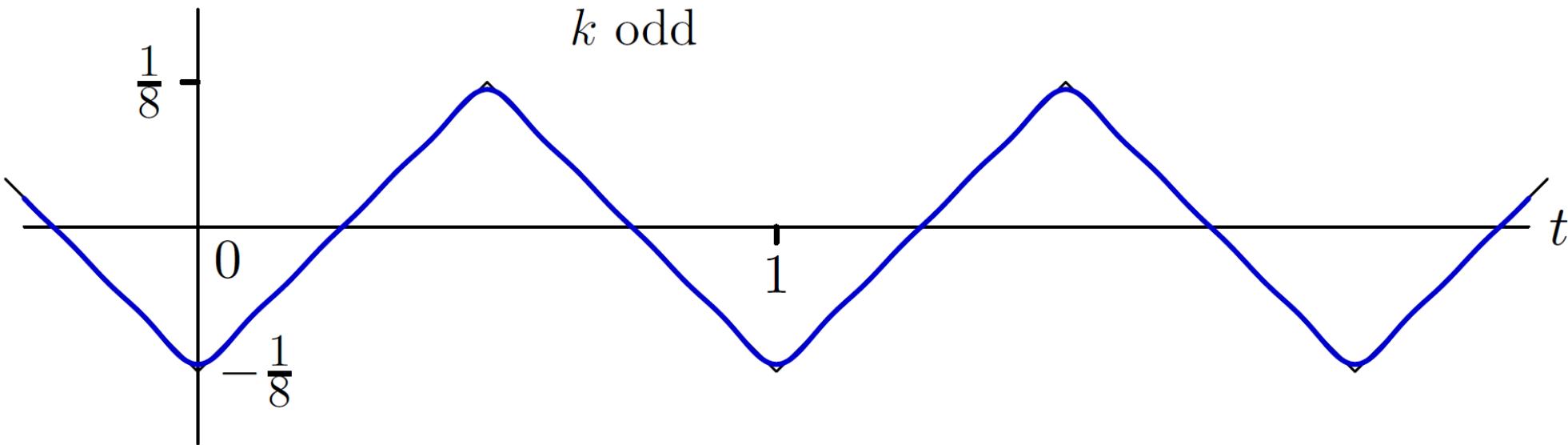
## Example: Triangle Waveform

$$\sum_{\substack{k = -5 \\ k \text{ odd}}}^5 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



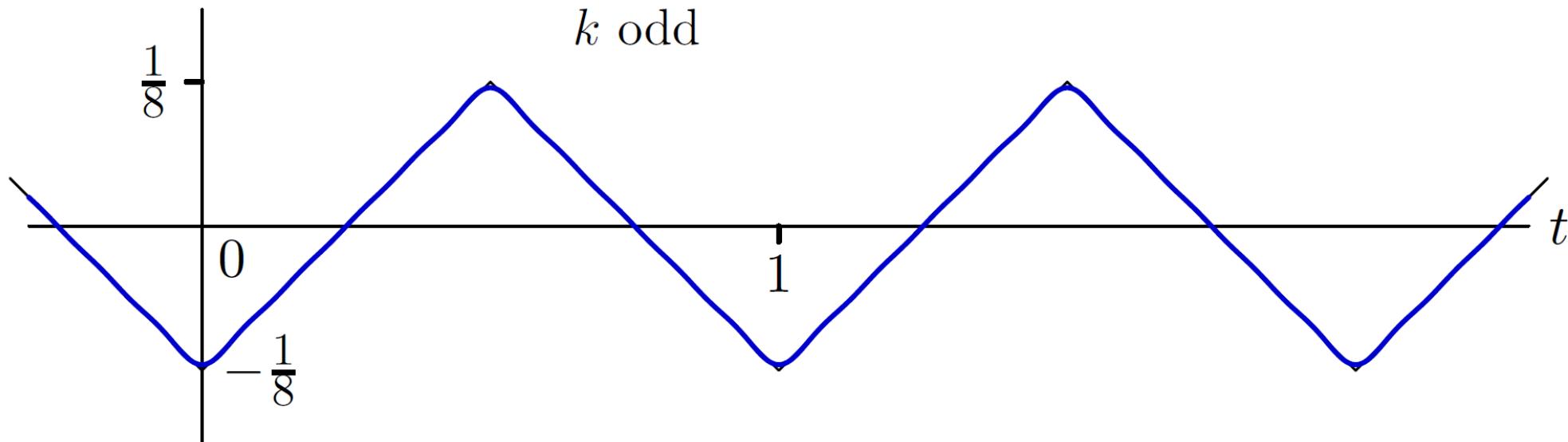
## Example: Triangle Waveform

$$\sum_{\substack{k=-7 \\ k \text{ odd}}}^7 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



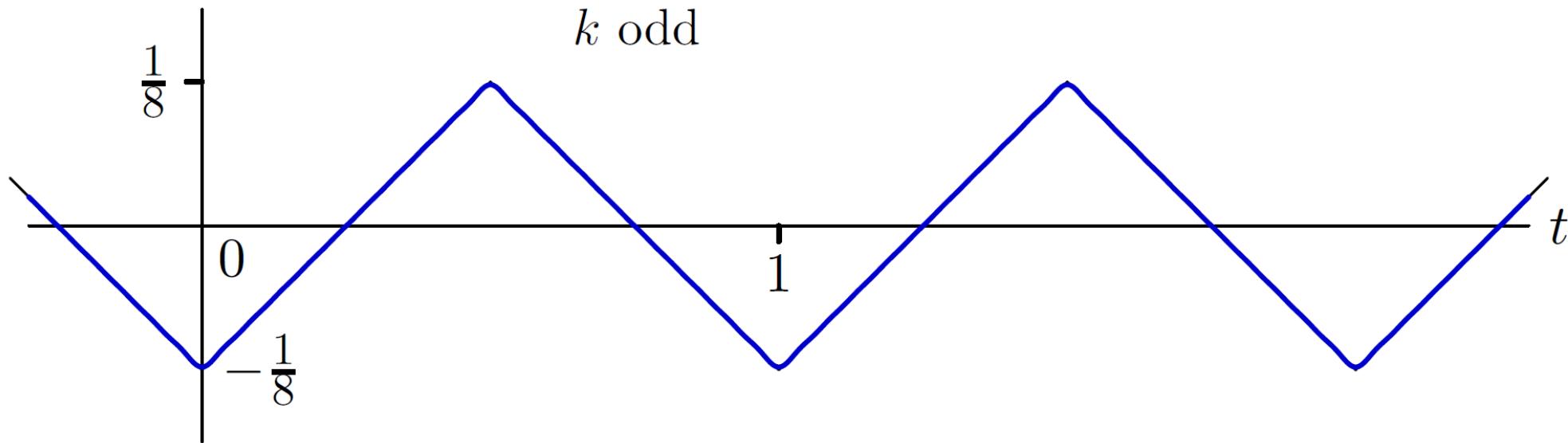
## Example: Triangle Waveform

$$\sum_{\substack{k=-9 \\ k \text{ odd}}}^9 \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



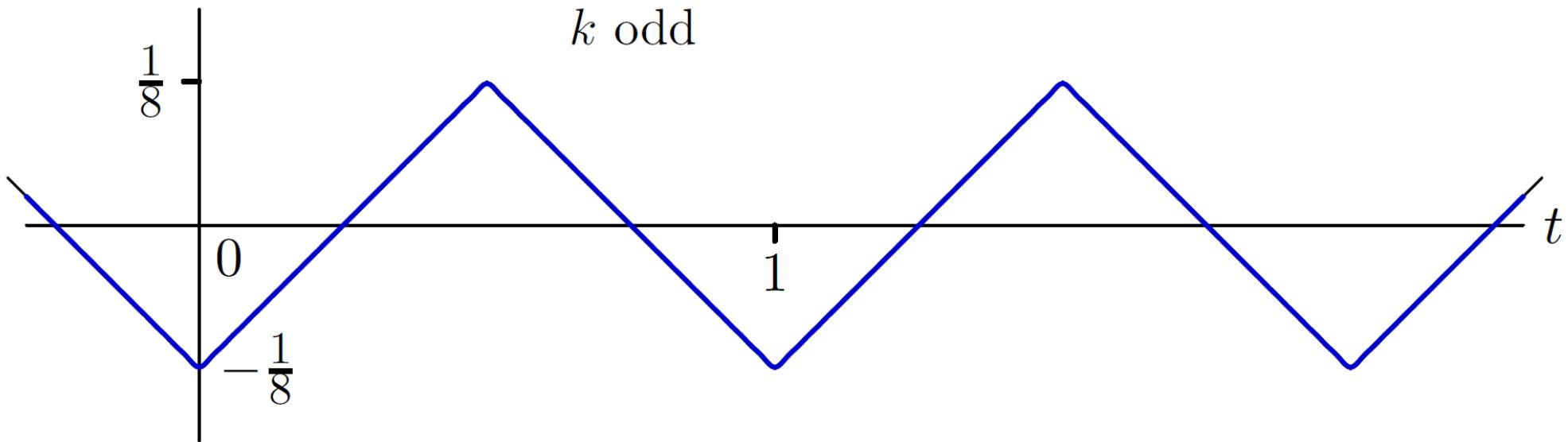
## Example: Triangle Waveform

$$\sum_{\substack{k=-19 \\ k \text{ odd}}}^{19} \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



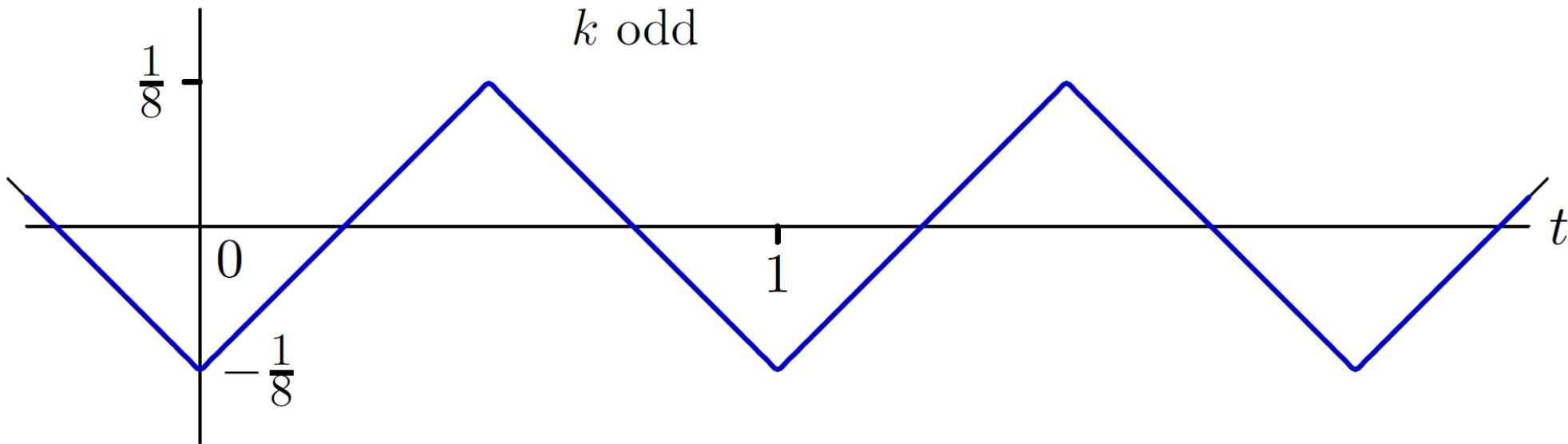
## Example: Triangle Waveform

$$\sum_{\substack{k=-29 \\ k \text{ odd}}}^{29} \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



## Example: Triangle Waveform

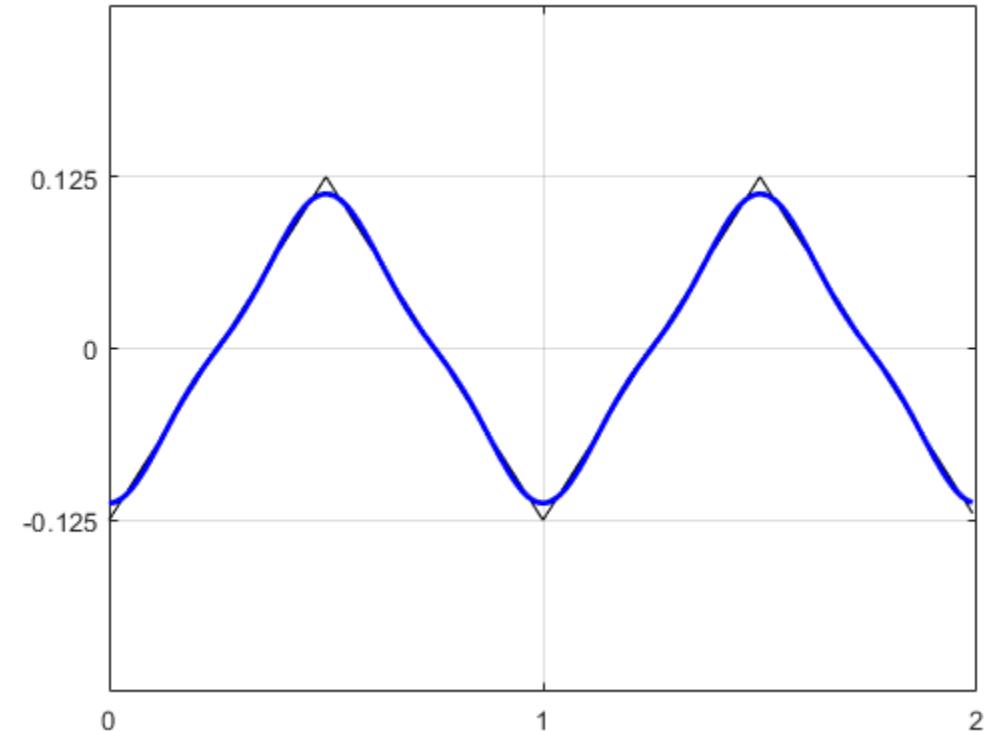
$$\sum_{\substack{k = -39 \\ k \text{ odd}}}^{39} \frac{-1}{2k^2\pi^2} e^{j2\pi kt}$$



# Example: Triangle Waveform: MATLAB

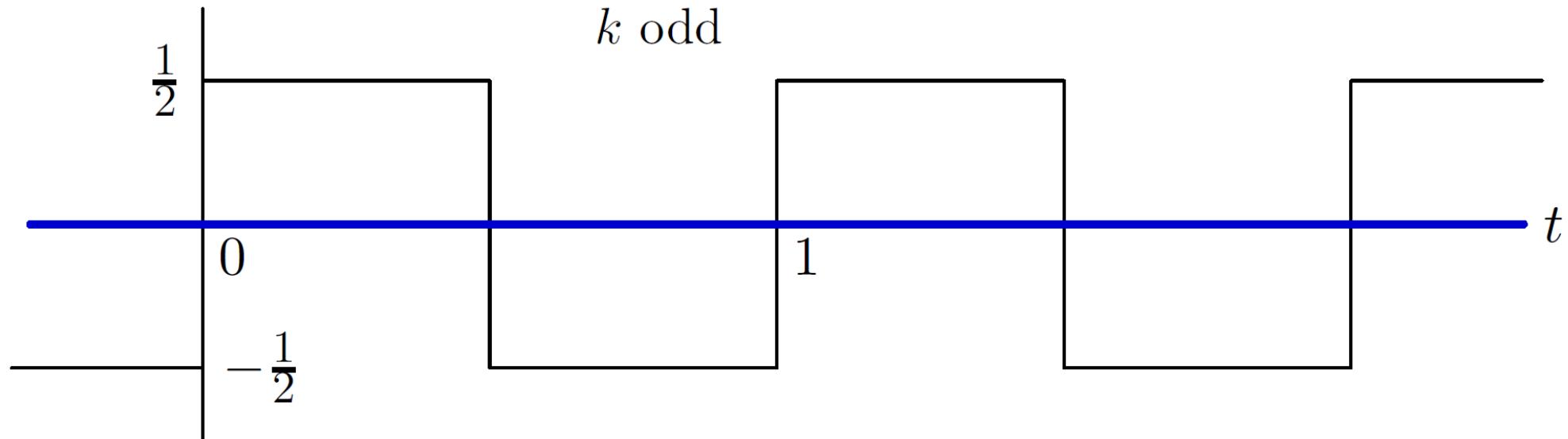
$$x(t) = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{-1}{2k^2\pi^2} e^{j2\pi kt} = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{-2}{\omega_0^2 k^2} e^{j\omega_0 kt}$$

```
T = 1;  
  
fs = 100;  
t = 0:1/fs:2*T-1/fs;  
  
w0 = 2*pi/T;  
x = 1/8*sawtooth(w0*t, 1/2);  
  
% xhat = Fourier series of triangle wave  
xhat = zeros(size(t));  
  
for k = -3:2:3  
    xhat = xhat - 2/(w0^2*k^2)*exp(1j*w0*k*t);  
end  
  
plot(t, x, 'k'), hold on  
plot(t, xhat, 'b', 'linewidth', 2), hold off  
grid on  
ylim([-2/8,2/8])  
xticks([0,1,2])  
yticks([-1/8,0,1/8])
```



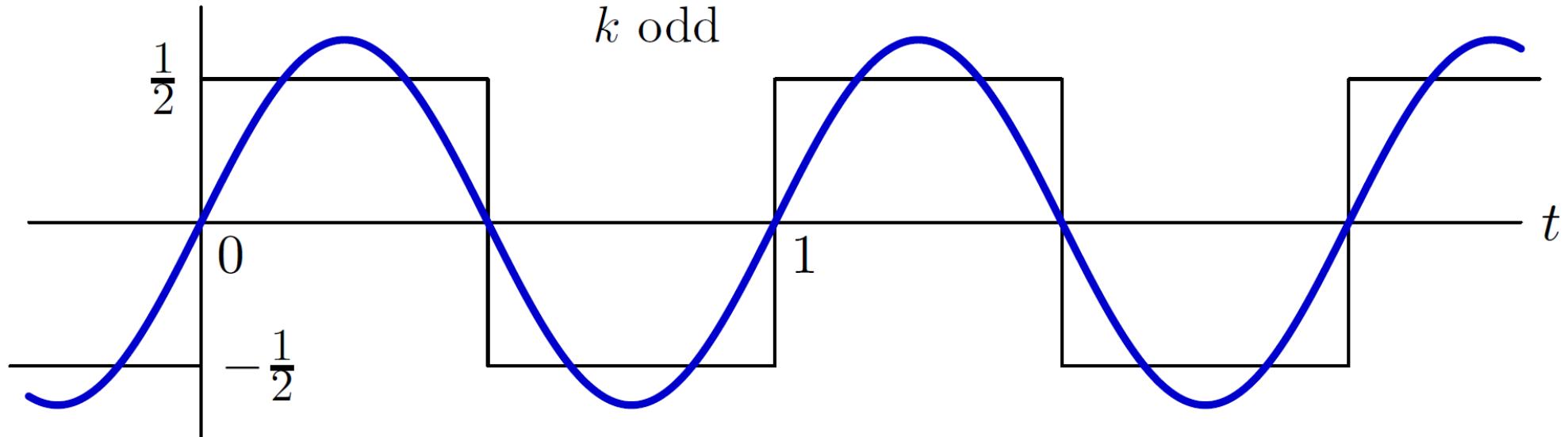
## Example: Square Waveform

$$\sum_{\substack{k=-0 \\ k \text{ odd}}}^0 \frac{1}{jk\pi} e^{j2\pi kt}$$



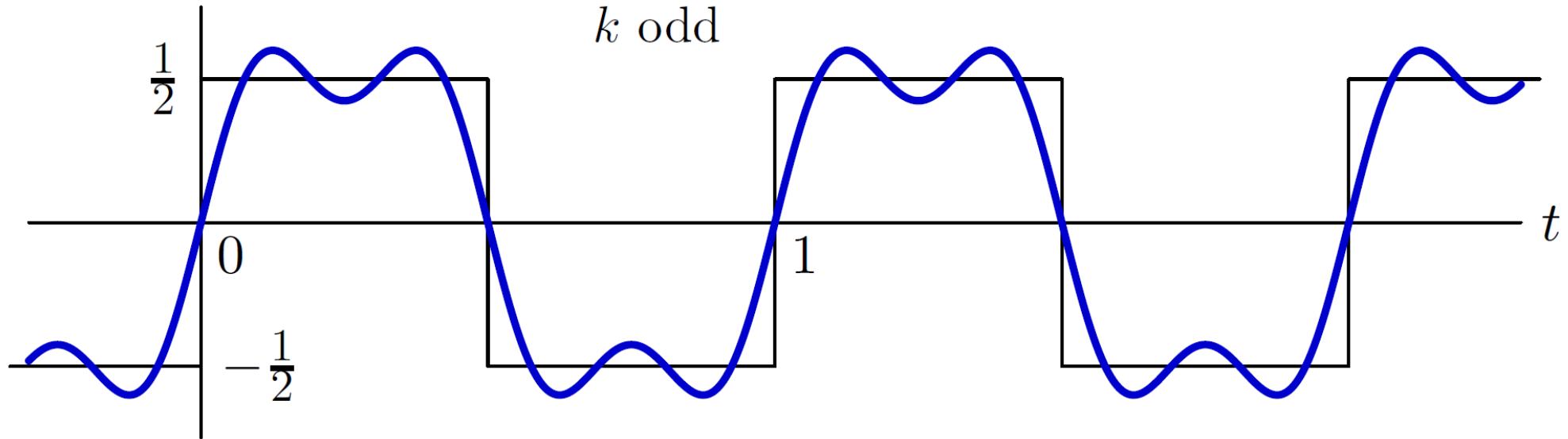
## Example: Square Waveform

$$\sum_{\substack{k = -1 \\ k \text{ odd}}}^1 \frac{1}{jk\pi} e^{j2\pi kt}$$



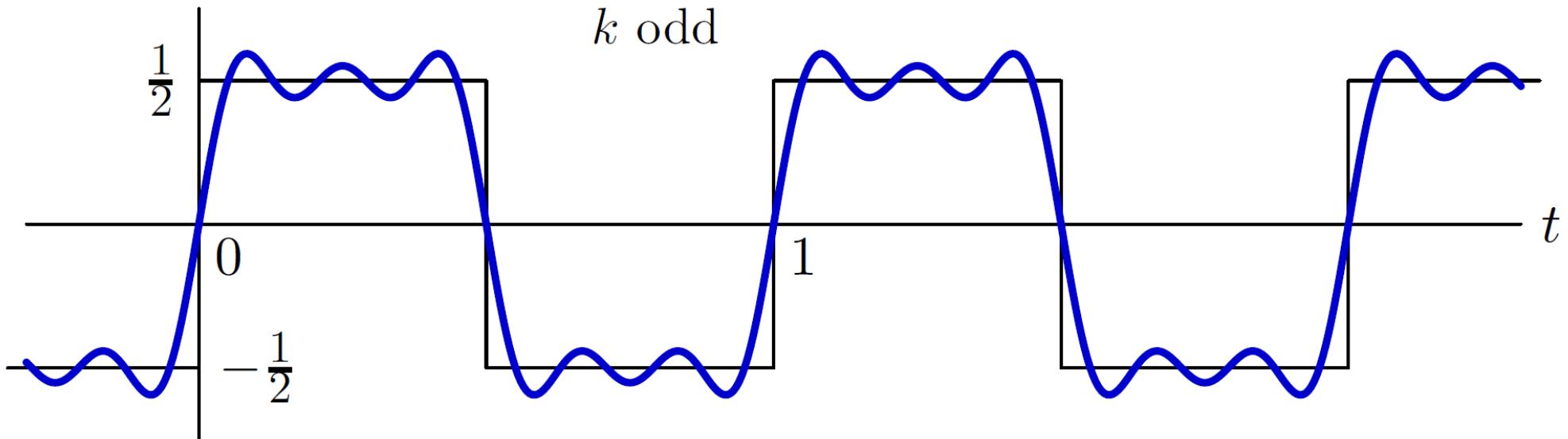
## Example: Square Waveform

$$\sum_{\substack{k=-3 \\ k \text{ odd}}}^3 \frac{1}{jk\pi} e^{j2\pi kt}$$



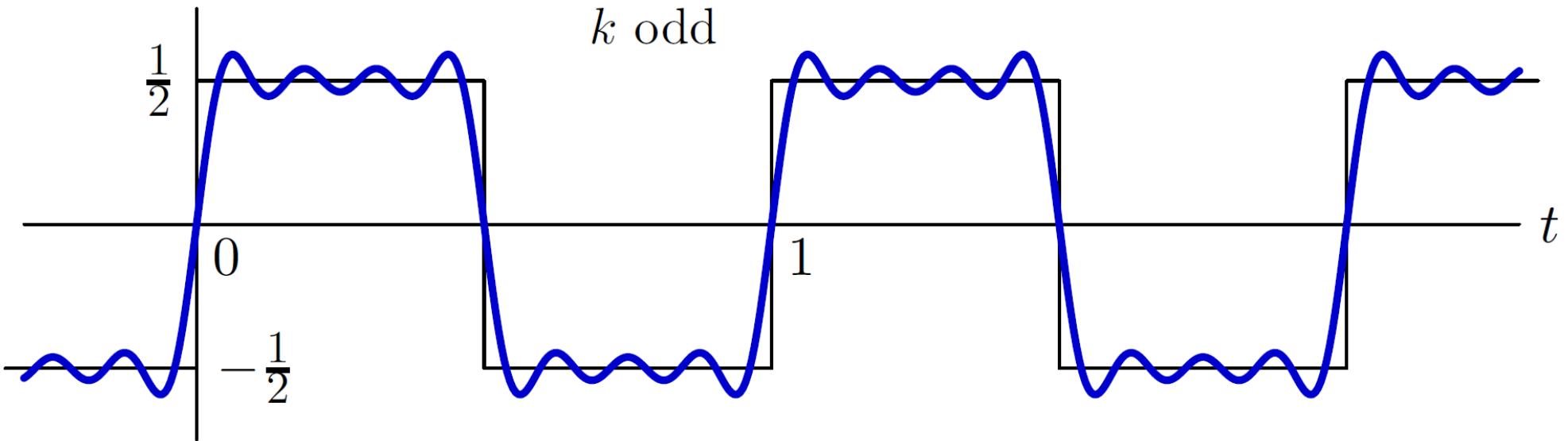
## Example: Square Waveform

$$\sum_{\substack{k=-5 \\ k \text{ odd}}}^5 \frac{1}{jk\pi} e^{j2\pi kt}$$



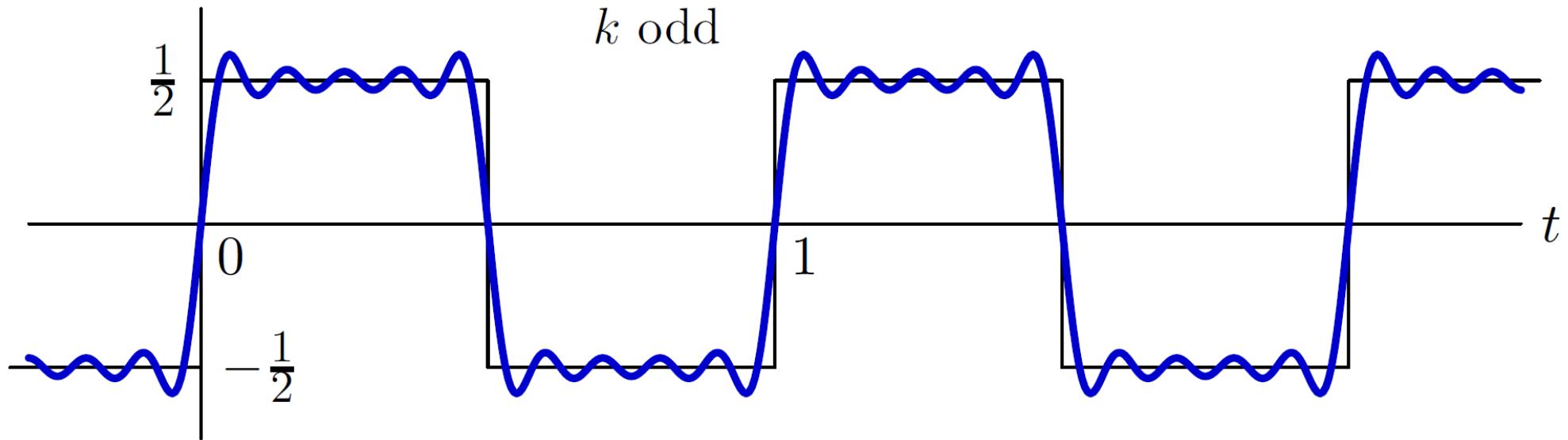
## Example: Square Waveform

$$\sum_{\substack{k=-7 \\ k \text{ odd}}}^7 \frac{1}{jk\pi} e^{j2\pi kt}$$



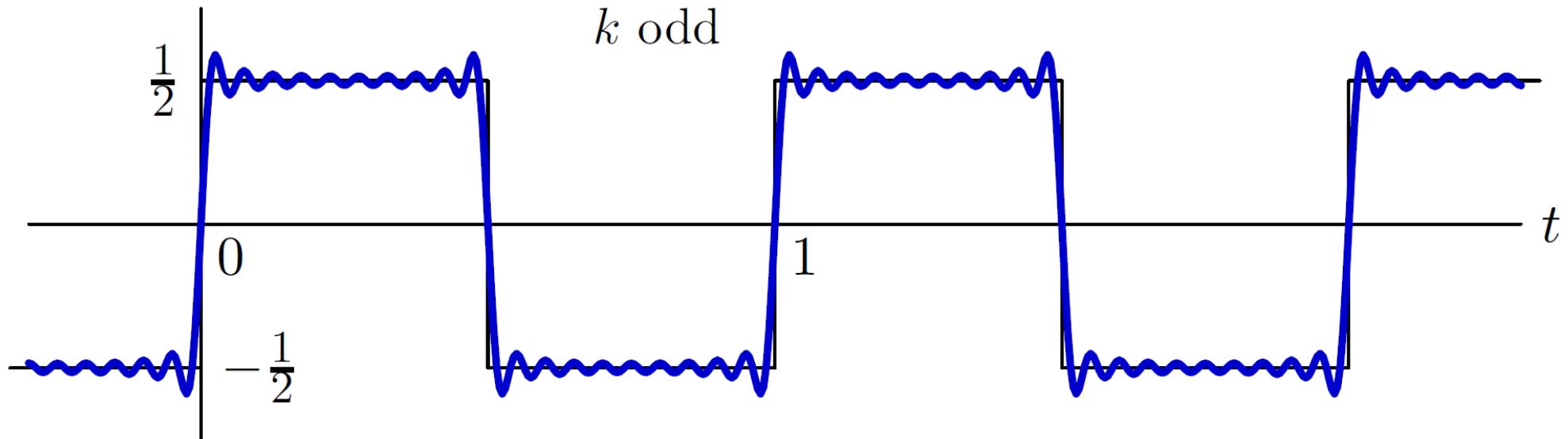
## Example: Square Waveform

$$\sum_{\substack{k=-9 \\ k \text{ odd}}}^9 \frac{1}{jk\pi} e^{j2\pi kt}$$



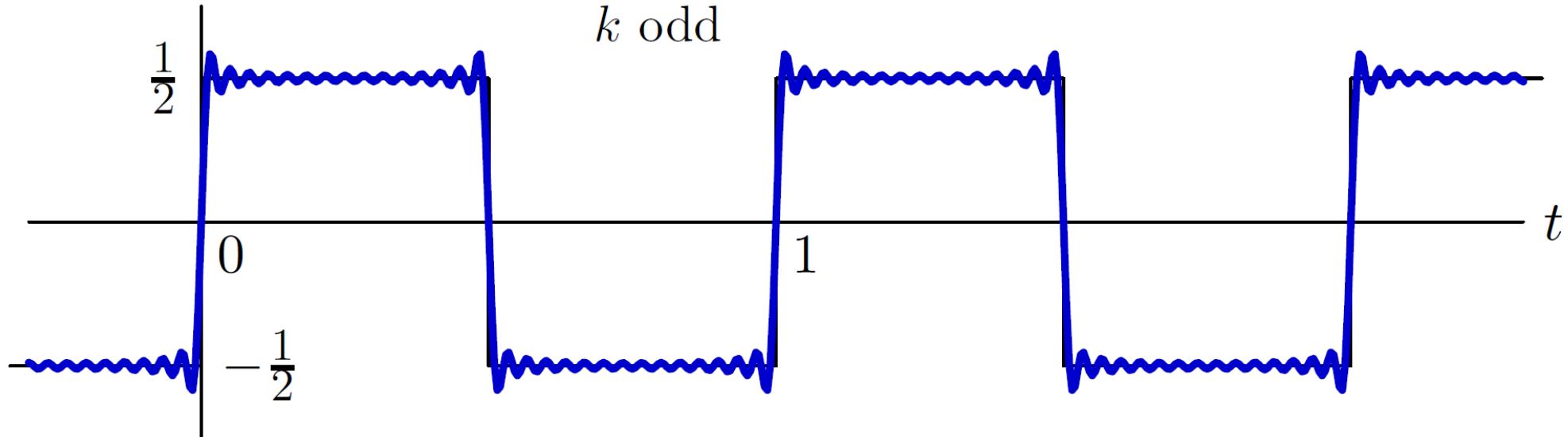
## Example: Square Waveform

$$\sum_{\substack{k = -19 \\ k \text{ odd}}}^{19} \frac{1}{jk\pi} e^{j2\pi kt}$$



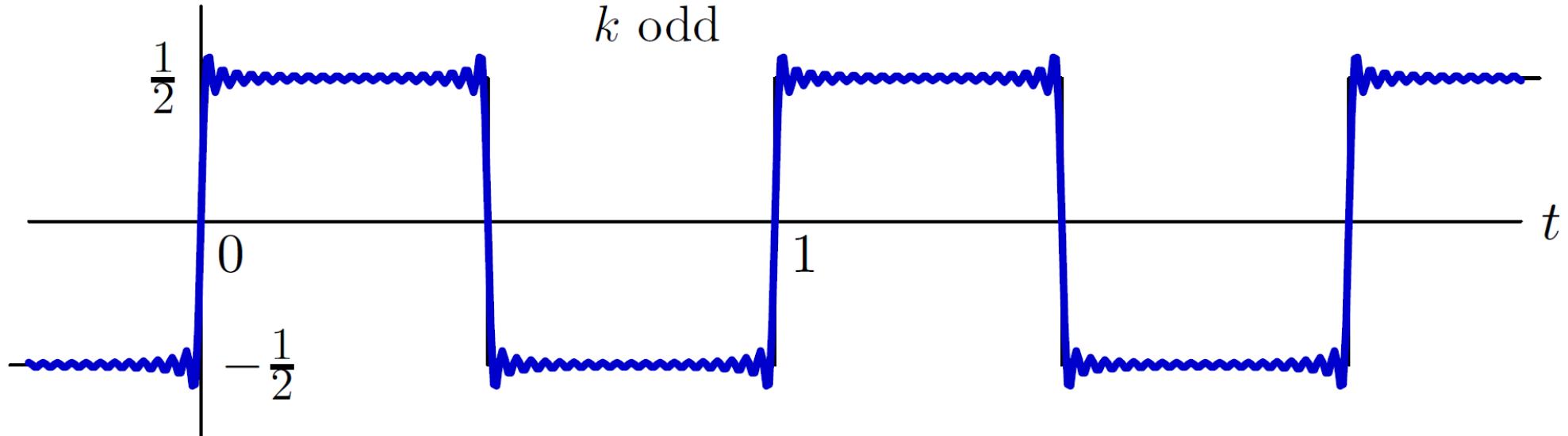
## Example: Square Waveform

$$\sum_{\substack{k = -29 \\ k \text{ odd}}}^{29} \frac{1}{jk\pi} e^{j2\pi kt}$$



## Example: Square Waveform

$$\sum_{\substack{k = -39 \\ k \text{ odd}}}^{39} \frac{1}{jk\pi} e^{j2\pi kt}$$



# Example: Square Waveform: MATLAB

$$x(t) = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{1}{jk\pi} e^{j2\pi kt} = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{2}{j\omega_0 k} e^{j\omega_0 kt}$$

```
T = 1;

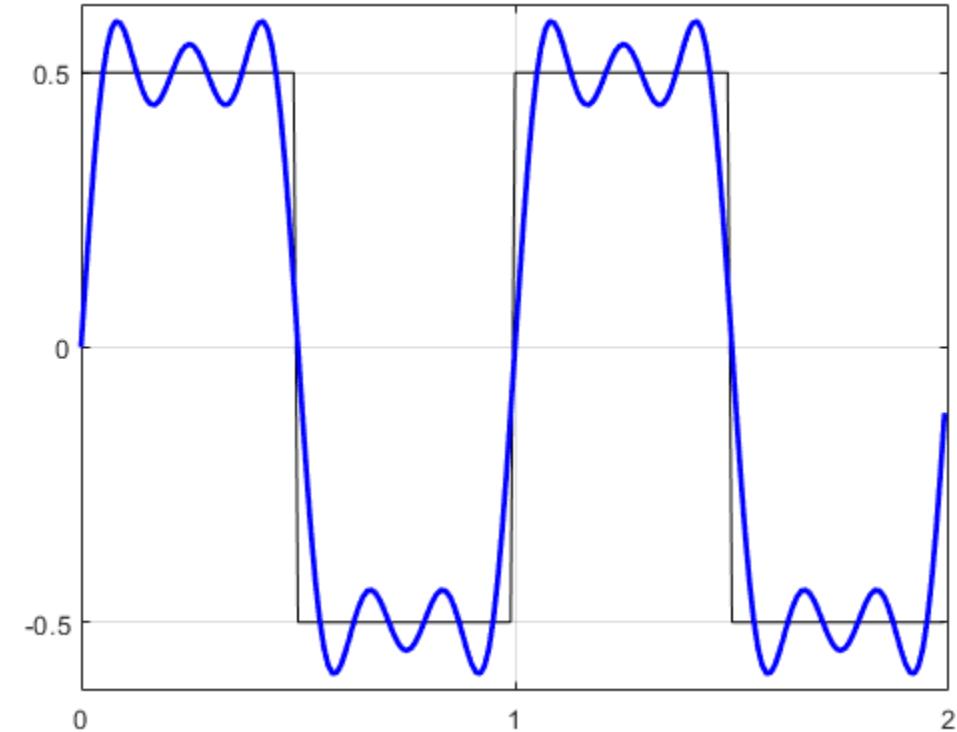
fs = 100;
t = 0:1/fs:2*T-1/fs;

w0 = 2*pi/T;
x = 1/2*square(w0*t);

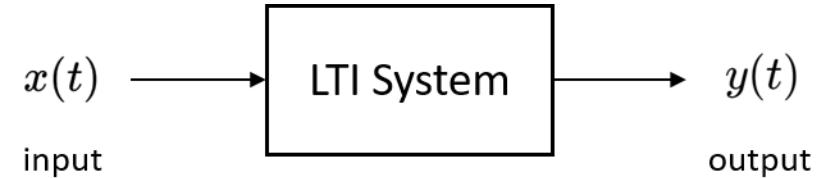
% xhat = Fourier series of square wave
xhat = zeros(size(t));

for k = -5:2:5
    xhat = xhat + 2/(1j*w0*k)*exp(1j*w0*k*t);
end

plot(t, x, 'k'), hold on
plot(t, xhat, 'b', 'linewidth', 2), hold off
grid on
ylim([-5/8,5/8])
xticks([0,1,2])
yticks([-1/2,0,1/2])
```



# Response to a Periodic Input (Filtering)



- Periodic input: Fourier series → sum of complex exponentials

$$x(t) = \sum_{k=-\infty}^{\infty} a_k e^{j\omega_0 kt}$$

- Complex exponentials: eigenfunctions of LTI system

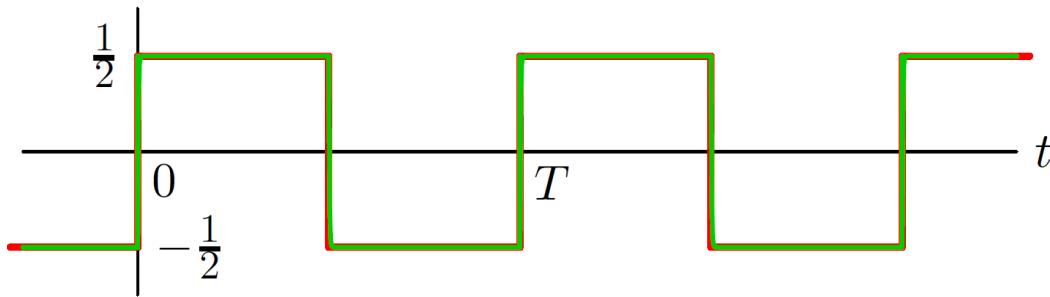
$$e^{j\omega_0 kt} \longrightarrow H(j\omega_0 k) e^{j\omega_0 kt}$$

- Output: same eigenfunctions, but amplitudes and phase are adjusted by the LTI system

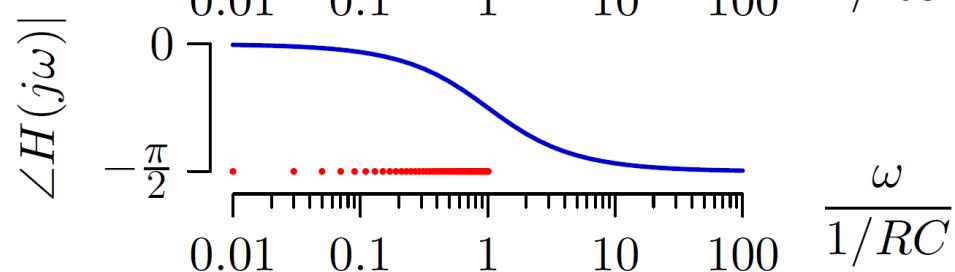
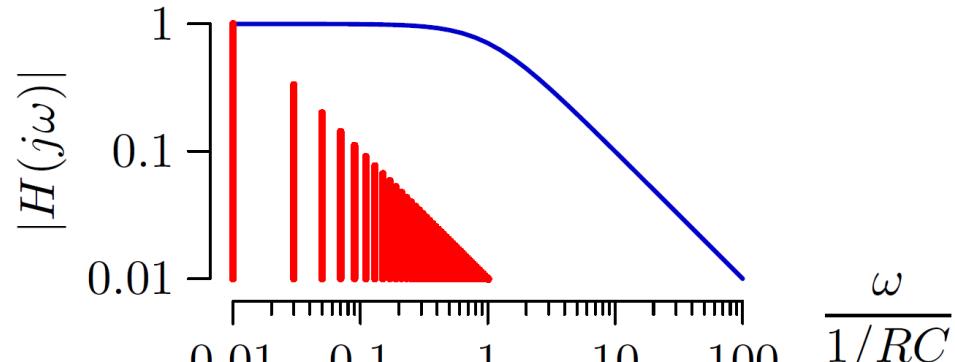
$$y(t) = \sum_{k=-\infty}^{\infty} a_k H(j\omega_0 k) e^{j\omega_0 kt}$$

- The output of an LTI system is a “filtered” version of the input

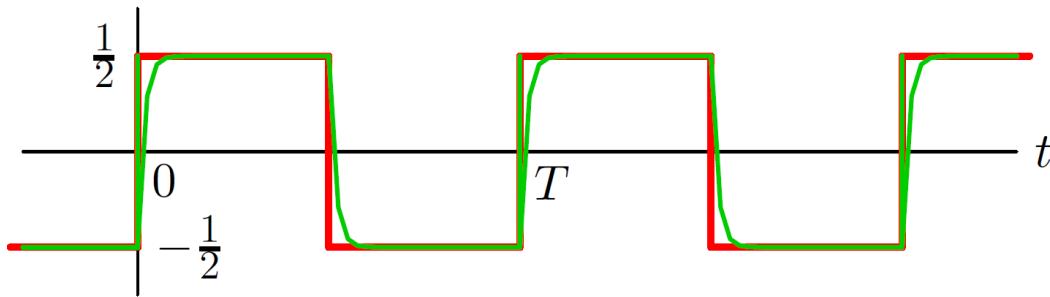
# Output is a “Filtered” Version of Input



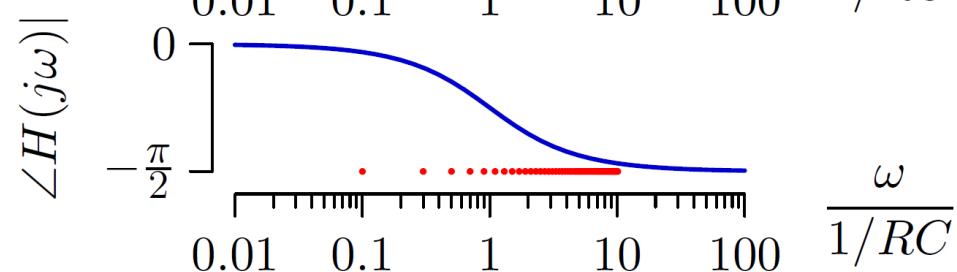
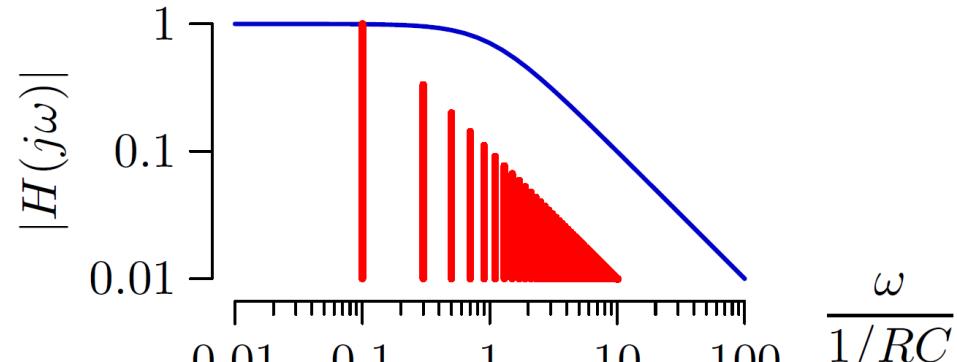
$$x(t) = \sum_{k \text{ odd}} \frac{1}{j\pi k} e^{j\omega_0 kt}; \quad \omega_0 = \frac{2\pi}{T}$$



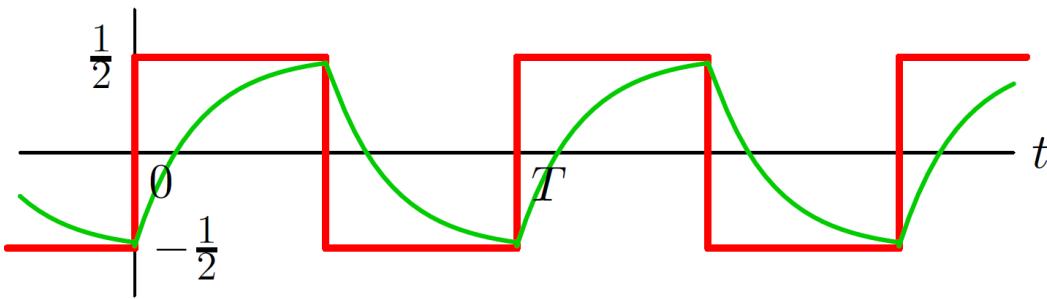
# Output is a “Filtered” Version of Input



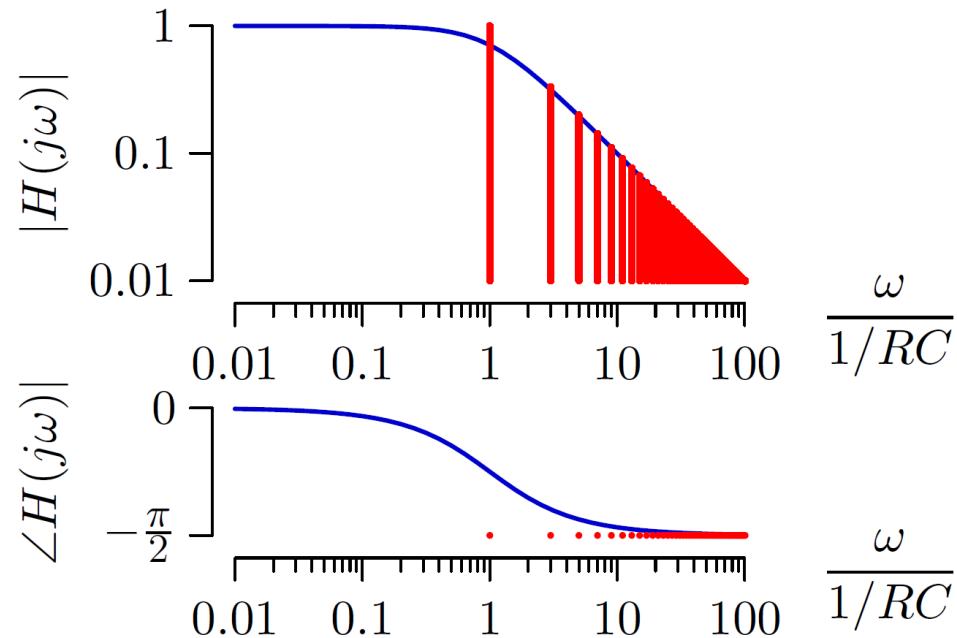
$$x(t) = \sum_{k \text{ odd}} \frac{1}{j\pi k} e^{j\omega_0 kt}; \quad \omega_0 = \frac{2\pi}{T}$$



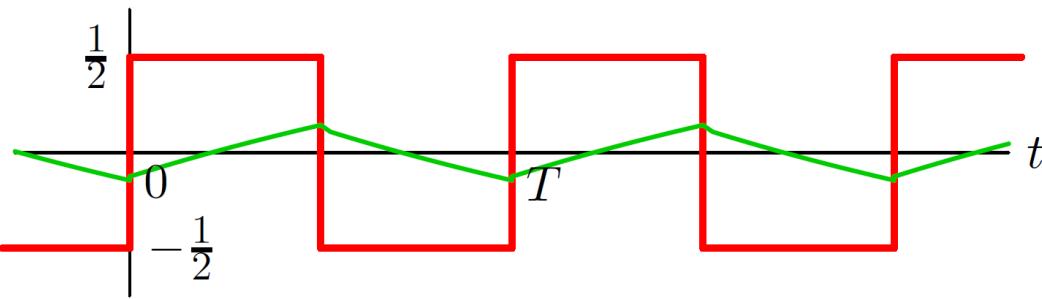
# Output is a “Filtered” Version of Input



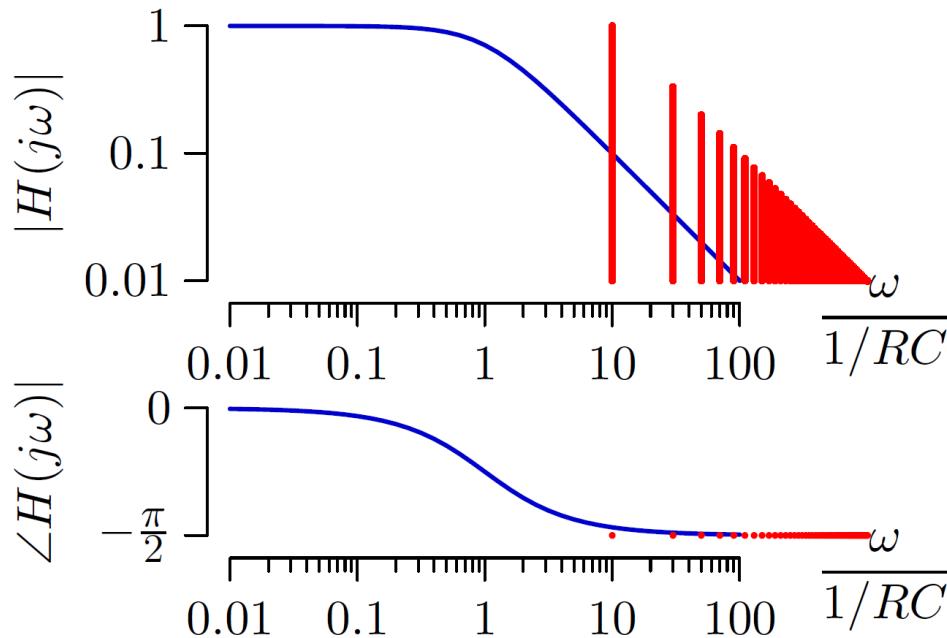
$$x(t) = \sum_{k \text{ odd}} \frac{1}{j\pi k} e^{j\omega_0 kt}; \quad \omega_0 = \frac{2\pi}{T}$$



# Output is a “Filtered” Version of Input



$$x(t) = \sum_{k \text{ odd}} \frac{1}{j\pi k} e^{j\omega_0 kt}; \quad \omega_0 = \frac{2\pi}{T}$$



# Response to a Square Wave Input: MATLAB

- Decompose a square wave to a linear combination of sinusoidal signals

$$x(t) = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{1}{jk\pi} e^{j2\pi kt} = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{2}{j\omega_0 k} e^{j\omega_0 kt}$$

- The output response of LTI

$$\dot{y} + \frac{1}{\tau} y = \frac{1}{\tau} x(t)$$

# Response to a Square Wave Input: MATLAB

- Decompose a square wave to a linear combination of sinusoidal signals

$$x(t) = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{1}{jk\pi} e^{j2\pi kt} = \sum_{k=\dots,-3,-1,1,3,\dots} \frac{2}{j\omega_0 k} e^{j\omega_0 kt}$$

- The output response of LTI

$$\dot{y} + \frac{1}{\tau} y = \frac{1}{\tau} x(t)$$

- Given input  $e^{j\omega t}$

$$\dot{y} + \frac{1}{\tau} y = \frac{1}{\tau} e^{j\omega t}$$

- $y = Ae^{j(\omega t+\phi)}$

$$\begin{aligned} j\omega A e^{j(\omega t+\phi)} + \frac{1}{\tau} A e^{j(\omega t+\phi)} &= \frac{1}{\tau} e^{j\omega t} \\ \left(j\omega + \frac{1}{\tau}\right) A e^{j\phi} &= \frac{1}{\tau} \end{aligned}$$

$$\boxed{\begin{aligned}|H(j\omega)| &= A = \frac{1}{|j\tau\omega + 1|} \\ \angle H(j\omega) &= \phi = -\angle(j\tau\omega + 1)\end{aligned}}$$

# Response to a Square Wave Input: MATLAB

- Linearity: input  $\sum a_k x_k(t)$  produces  $\sum a_k y_k(t)$

```
% xhat = Fourier series of square wave
xhat = zeros(size(t));

for k = -19:2:19
    xhat = xhat + 2/(1j*w0*k)*exp(1j*w0*k*t);
end
```

```
tau = 1/5;

yhat = zeros(size(t));

for k = -19:2:19
    w = w0*k;
    A = 1./abs(1j*tau*w + 1);
    Ph = -phase(1j*tau*w + 1);
    yhat = yhat + A*2/(1j*w0*k)*exp(1j*(w*t + Ph));
end
```

$$x(t) = \sum_{k=-\infty, -3, -1, 1, 3, \dots}^{\infty} \frac{2}{j\omega_0 k} e^{j\omega_0 kt}$$

$$y(t) = \sum_{k=-\infty}^{\infty} a_k H(j\omega_0 k) e^{j\omega_0 kt}$$

$$|H(j\omega)| = A = \frac{1}{|j\tau\omega + 1|}$$

$$\angle H(j\omega) = \phi = -\angle(j\tau\omega + 1)$$

# Response to a Square Wave Input: MATLAB

- Linearity: input  $\sum a_k x_k(t)$  produces  $\sum a_k y_k(t)$

$$\dot{y} + \frac{1}{\tau}y = \frac{1}{\tau}x(t)$$

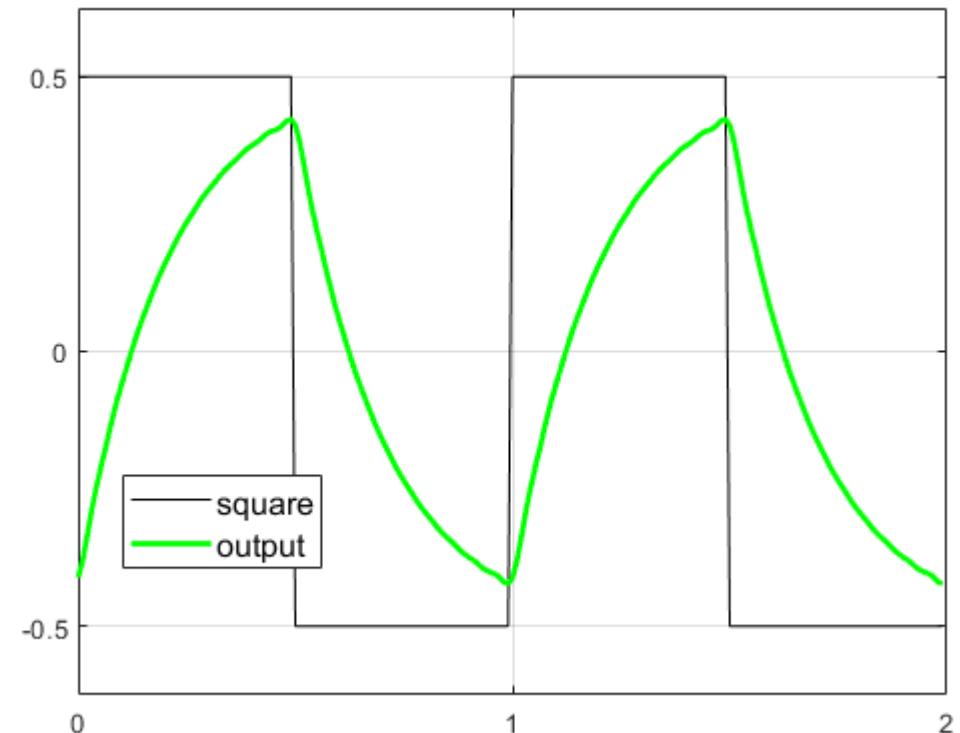
```
% xhat = Fourier series of square wave
xhat = zeros(size(t));

for k = -19:2:19
    xhat = xhat + 2/(1j*w0*k)*exp(1j*w0*k*t);
end
```

```
tau = 1/5;

yhat = zeros(size(t));

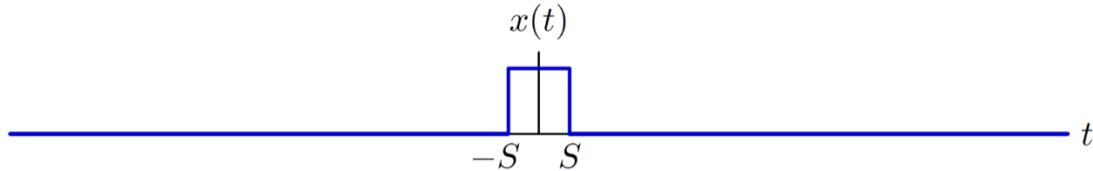
for k = -19:2:19
    w = w0*k;
    A = 1./abs(1j*tau*w + 1);
    Ph = -phase(1j*tau*w + 1);
    yhat = yhat + A*2/(1j*w0*k)*exp(1j*(w*t + Ph));
end
```



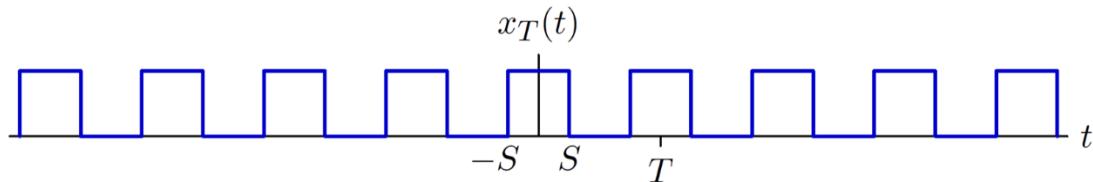
# **Response to General Input (in Frequency)**

# Response to a General Input (Aperiodic Signal) in Frequency Domain

- An aperiodic signal can be thought of as periodic with infinite period
- Let  $x(t)$  represent an aperiodic signal



- Periodic extension



$$x_T(t) = \sum_{k=-\infty}^{\infty} x(t + kT)$$

- Then

$$x(t) = \lim_{T \rightarrow \infty} x_T(t)$$

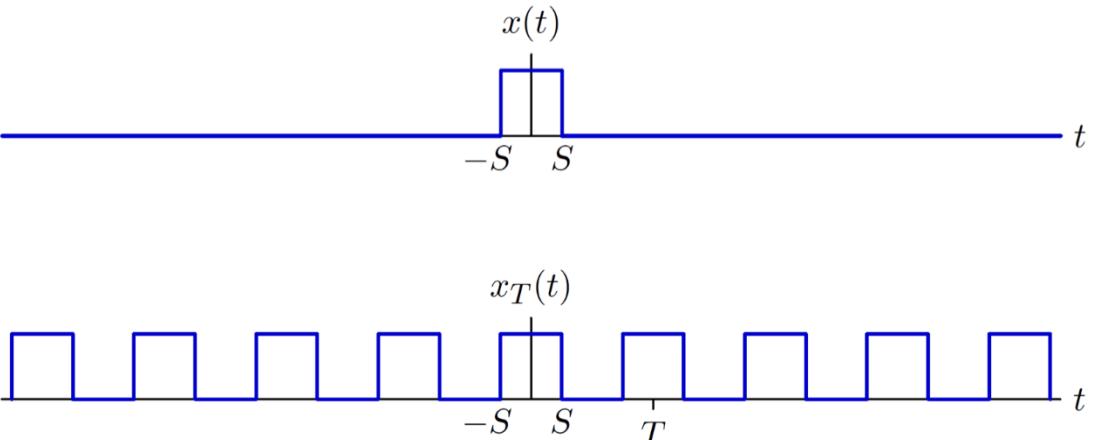
# Example: Periodic Square Wave

$$\omega_0 = \frac{2\pi}{T}$$

$$\begin{aligned}
 a_k &= \frac{1}{T} \int_{-\frac{T}{2}}^{\frac{T}{2}} x_T(t) e^{-j\omega_0 kt} dt \\
 &= \frac{1}{T} \left[ \int_{-s}^s 1 \cdot e^{-j\omega_0 kt} dt \right] = \frac{1}{T} \left[ -\frac{e^{-j\omega_0 kt}}{j\omega_0 k} \Big|_{-s}^s \right] \\
 &= -\frac{1}{T} \left[ \frac{e^{-j\omega_0 ks}}{j\omega_0 k} - \frac{e^{j\omega_0 ks}}{j\omega_0 k} \right] = \frac{1}{T} \left[ \frac{-2}{j\omega_0 k} \cdot \frac{e^{-j\omega_0 ks} - e^{j\omega_0 ks}}{2} \right] \\
 &= \frac{1}{T} \frac{-2}{j\omega_0 k} (-j \sin \omega_0 ks) = \frac{1}{T} \frac{2 \sin \omega_0 ks}{\omega_0 k} \\
 &= \frac{2}{T} \frac{\sin \omega_0 ks}{\omega_0 k}
 \end{aligned}$$

$$Ta_k = \frac{2 \sin \omega s}{\omega}, \quad \omega = k \omega_0, \quad \omega_0 = \frac{2\pi}{T}$$

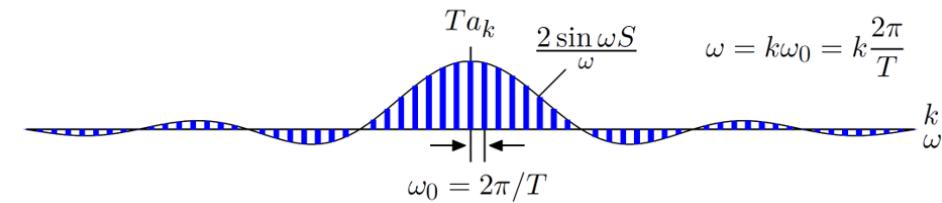
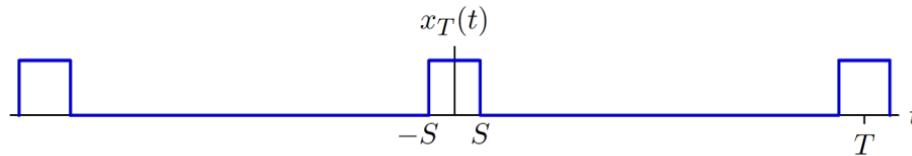
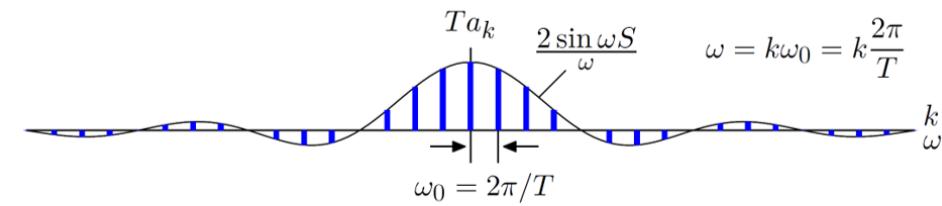
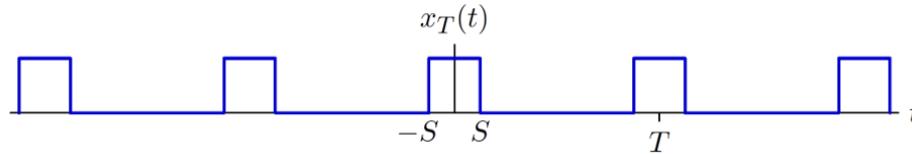
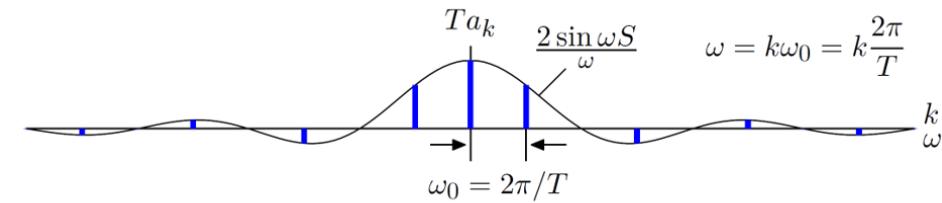
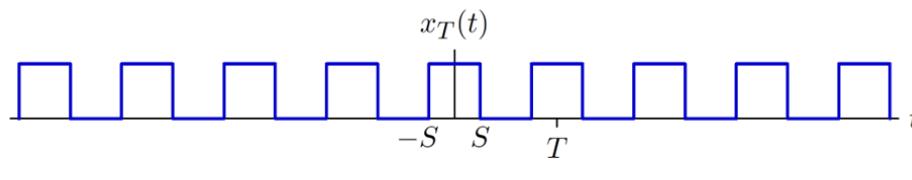
$$x(t) = \begin{cases} 1, & |t| < s \\ 0, & s < |t| < \frac{T}{2} \end{cases}$$



# Example: Periodic Square Wave

$$T a_k = \frac{2 \sin \omega s}{\omega}, \quad \omega = k \omega_0, \quad \omega_0 = \frac{2\pi}{T}$$

- Doubling period doubles # of harmonics in given frequency interval



# Example: Periodic Square Wave

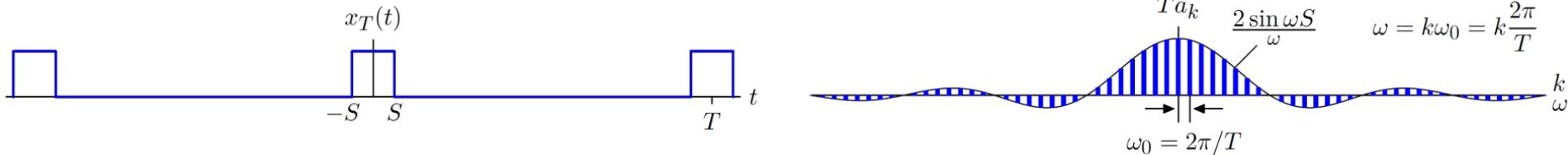
- As  $T \rightarrow \infty$ , discrete harmonic amplitudes  $\rightarrow$  a continuum  $X(j\omega)$

$$\lim_{T \rightarrow \infty} T a_k = \lim_{T \rightarrow \infty} \int_{-\frac{T}{2}}^{\frac{T}{2}} x(t) e^{-j\omega t} dt = \frac{2\sin\omega s}{\omega} = X(j\omega)$$

- As alternative way of interpreting is as samples of an envelope function, specifically

$$T a_k = \left. \frac{2\sin\omega s}{\omega} \right|_{\omega=k\omega_0}$$

- That is, with  $\omega$  thought of as a continuous variable, the set of Fourier series coefficients approaches the envelop function as  $T \rightarrow \infty$



# Fourier Transform

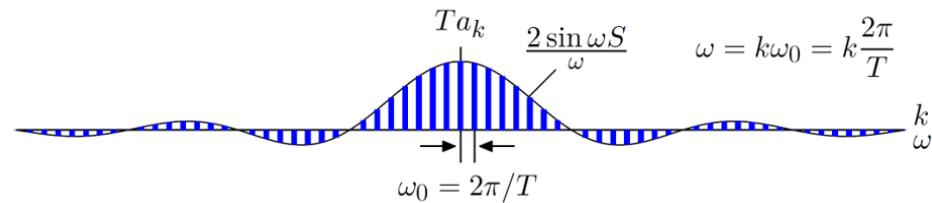
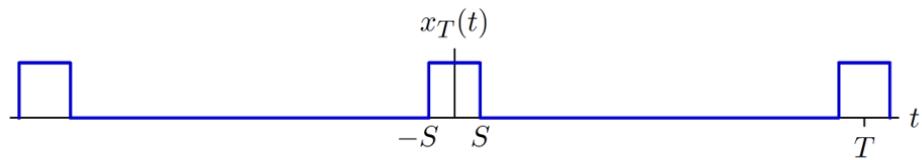
# Fourier Transform

- As  $T \rightarrow \infty$ , synthesis sum  $\rightarrow$  integral

$$\begin{aligned}x(t) &= \sum_{k=-\infty}^{\infty} a_k e^{j\omega_0 k t} = \sum_{k=-\infty}^{\infty} \frac{X(j\omega_0 k)}{T} e^{j\omega_0 k t} \\&= \sum_{k=-\infty}^{\infty} \frac{\omega_0}{2\pi} X(j\omega_0 k) e^{j\omega_0 k t},\end{aligned}$$

- Aperiodic signal has all the frequency components instead of discrete harmonic components

$$\begin{aligned}&= \frac{1}{2\pi} \sum_{k=-\infty}^{\infty} X(j\omega_0 k) e^{j\omega_0 k t} \omega_0, \quad \omega = k\omega_0, \quad \omega_0 = \frac{2\pi}{T} \\&= \frac{1}{2\pi} \int X(j\omega) e^{j\omega t} d\omega\end{aligned}$$



# Fourier Transform

- Definition: Fourier transform

$$X(j\omega) = \int_{-\infty}^{\infty} x(t)e^{-j\omega t} dt \quad \text{analysis}$$
$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega)e^{j\omega t} d\omega \quad \text{synthesis}$$

# Response to General Input

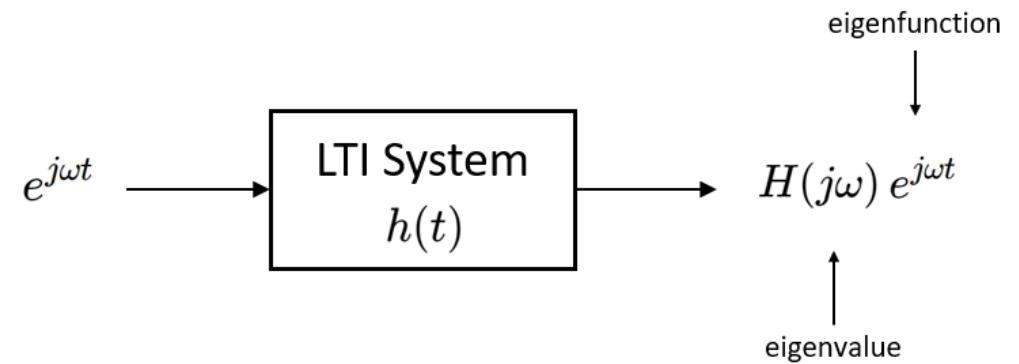
- Response to LTI system with impulse response  $h(t)$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) \underline{e^{j\omega t}} d\omega$$

$$y(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\omega) \underline{H(j\omega)} \underline{e^{j\omega t}} d\omega$$

$$y(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \underline{X(j\omega) H(j\omega)} e^{j\omega t} d\omega$$

$$x(t) * h(t) \quad \longleftrightarrow \quad X(j\omega) H(j\omega)$$

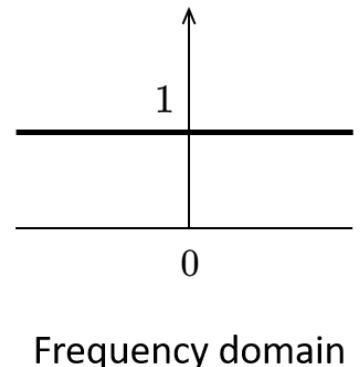
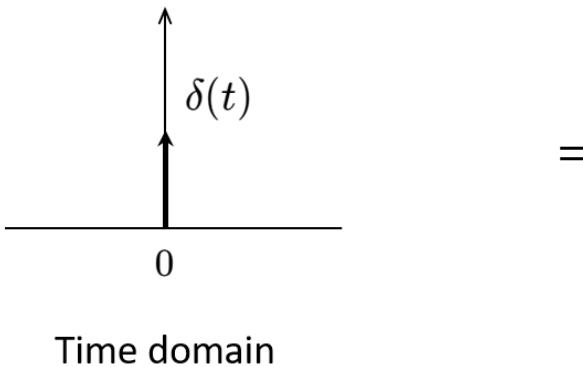


# Magic of Impulse Response

- Fourier transform of Dirac delta function

$$\delta(t) \xrightarrow{\text{LTI}} h(t)$$

$$\int_{-\infty}^{\infty} \delta(t) e^{-j\omega t} dt = 1$$

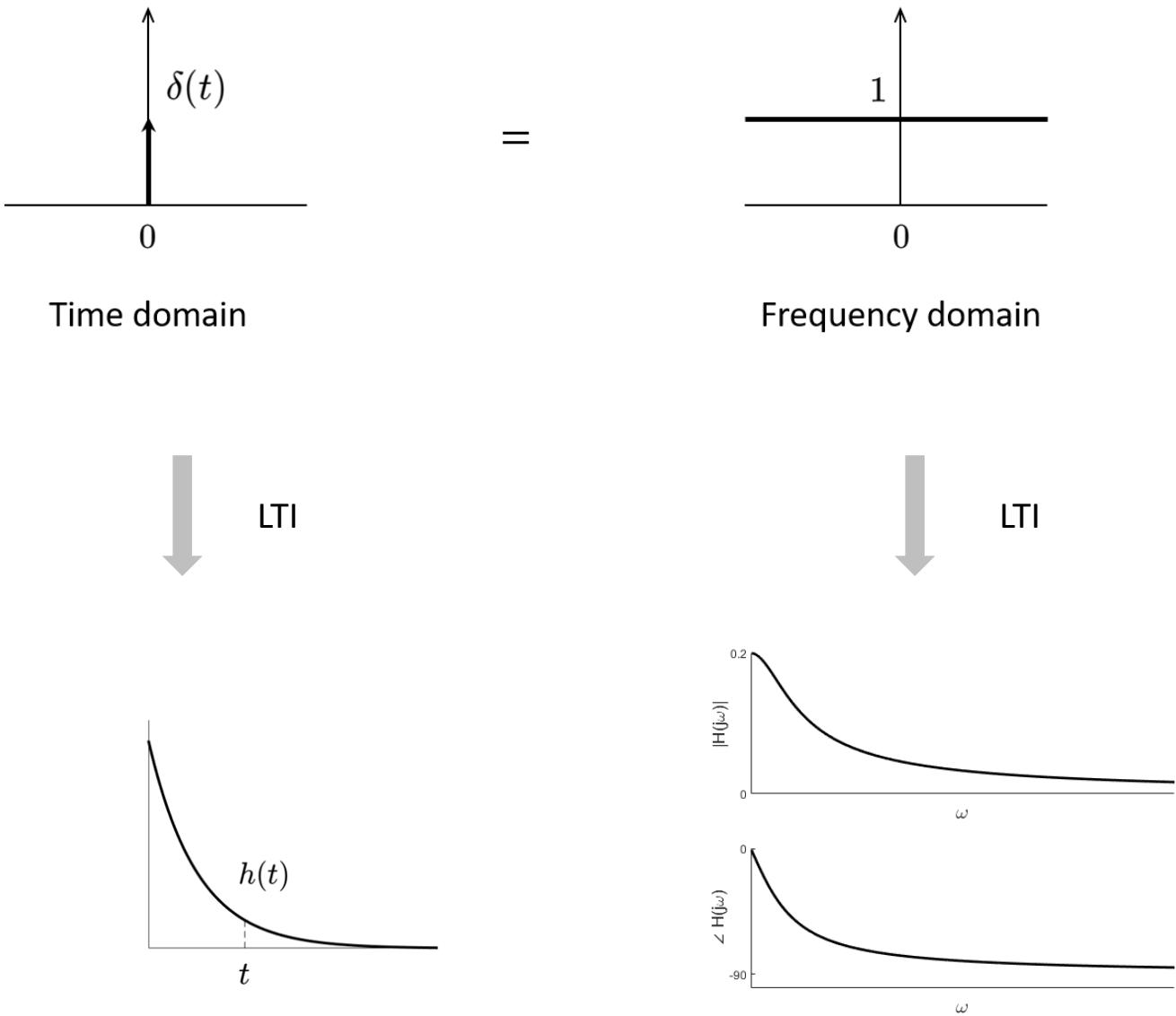


- Dirac delta function contains all the frequency components with 1
  - Convolution in time
  - Filtering in frequency

# Magic of Impulse Response

- Impulse basically excites a system with all the frequency of  $e^{j\omega t}$
- Impulse response contains the information on how much magnitude and phase are filtered via the LTI system at all the frequency

$$x(t) * h(t) \quad \longleftrightarrow \quad X(j\omega)H(j\omega)$$



# Frequency Response (Frequency Sweep)

- Frequency sweeping is another way to collect LTI system characteristics
  - (same as the impulse response)
- Given input  $e^{j\omega t}$

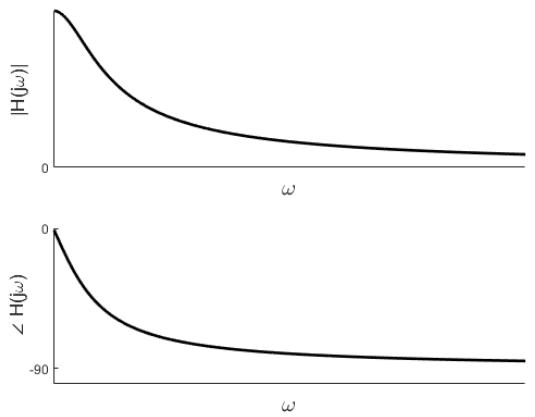
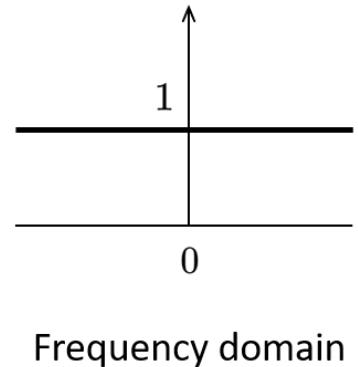
$$\dot{y} + 5y = 5e^{j\omega t}$$

- $y = Ae^{j(\omega t+\phi)}$

$$\begin{aligned} j\omega Ae^{j(\omega t+\phi)} + 5Ae^{j(\omega t+\phi)} &= 5e^{j\omega t} \\ (j\omega + 5) Ae^{j\phi} &= 5 \end{aligned}$$

$$|H(j\omega)| = A = \frac{5}{|j\omega + 5|}$$

$$\angle H(j\omega) = \phi = -\angle(j\omega + 5)$$



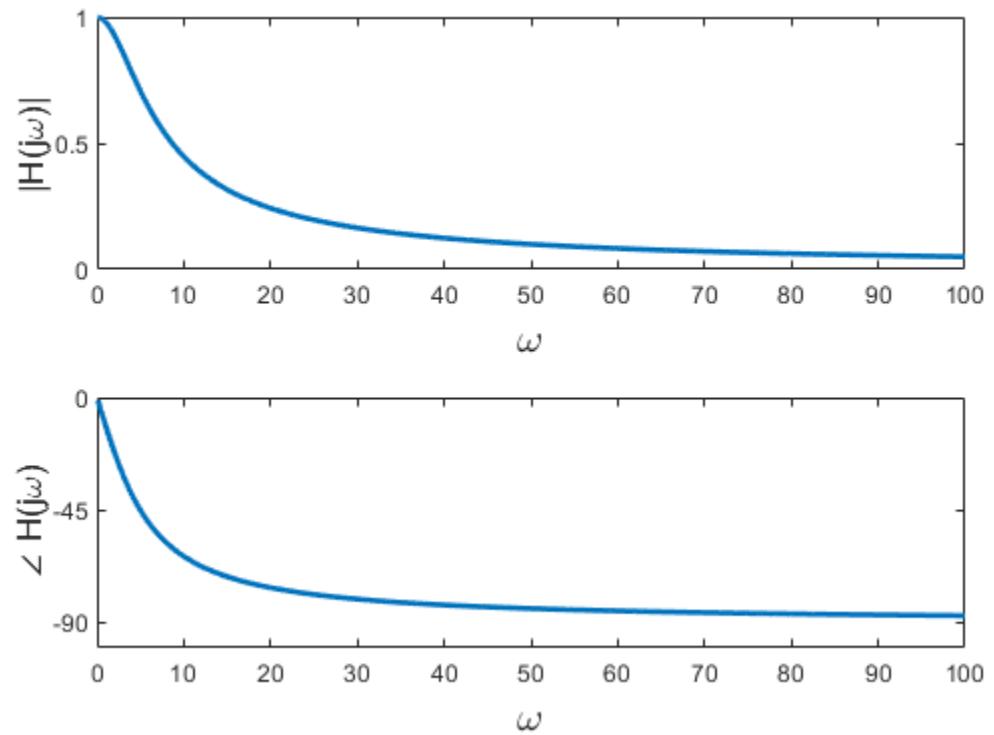
# The First Order ODE: MATLAB

$$\dot{y} + 5y = 5e^{j\omega t}$$

$$|H(j\omega)| = A = \frac{5}{|j\omega + 5|}$$

$$\angle H(j\omega) = \phi = -\angle(j\omega + 5)$$

```
w = 0.1:0.1:100;  
  
A = 5./abs(1j*w+5);  
P = -angle(1j*w+5)*180/pi;
```



## Example: The Second Order ODE

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2 y = \omega_n^2 x(t)$$

$$\ddot{y} + 2\zeta\omega_n\dot{y} + \omega_n^2 y = \omega_n^2 e^{j\Omega t}$$

$$y = A e^{j(\Omega t + \phi)} \quad (-\Omega^2 + j2\zeta\omega_n\Omega + \omega_n^2) A e^{j\phi} e^{j\Omega t} = \omega_n^2 e^{j\Omega t}$$

$$A e^{j\phi} = \frac{\omega_n^2}{-\Omega^2 + j2\zeta\omega_n\Omega + \omega_n^2} = \frac{1}{1 - \left(\frac{\Omega}{\omega_n}\right)^2 + j2\zeta\left(\frac{\Omega}{\omega_n}\right)}$$

$$A = \frac{1}{\sqrt{\left(1 - \left(\frac{\Omega}{\omega_n}\right)^2\right)^2 + 4\zeta^2\left(\frac{\Omega}{\omega_n}\right)^2}} = \frac{1}{\sqrt{(1 - \gamma^2)^2 + 4\zeta^2\gamma^2}}, \quad \left(\gamma = \frac{\Omega}{\omega_n}\right)$$

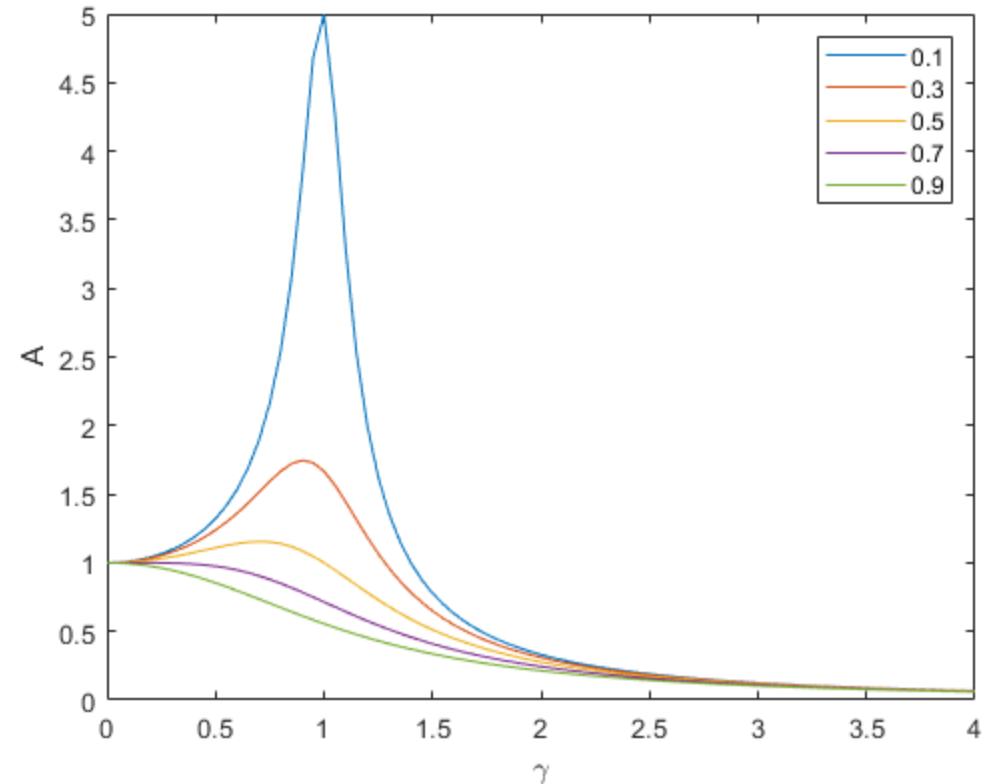
$$\phi = -\tan^{-1}\left(\frac{2\zeta\frac{\Omega}{\omega_n}}{1 - \left(\frac{\Omega}{\omega_n}\right)^2}\right) = -\tan^{-1}\left(\frac{2\zeta\gamma}{1 - \gamma^2}\right)$$

# The Second Order ODE: MATLAB

$$A = \frac{1}{\sqrt{\left(1 - \left(\frac{\Omega}{\omega_n}\right)^2\right)^2 + 4\zeta^2\left(\frac{\Omega}{\omega_n}\right)^2}} = \frac{1}{\sqrt{(1 - \gamma^2)^2 + 4\zeta^2\gamma^2}}, \quad \left(\gamma = \frac{\Omega}{\omega_n}\right)$$

$$\phi = -\tan^{-1}\left(\frac{2\zeta\frac{\Omega}{\omega_n}}{1 - \left(\frac{\Omega}{\omega_n}\right)^2}\right) = -\tan^{-1}\left(\frac{2\zeta\gamma}{1 - \gamma^2}\right)$$

```
r = 0:0.05:4;  
  
zeta = 0.1:0.2:1;  
A = [];  
for i = 1:length(zeta)  
    A(i,:) = 1./sqrt((1-r.^2).^2 + (2*zeta(i)*r).^2);  
end
```



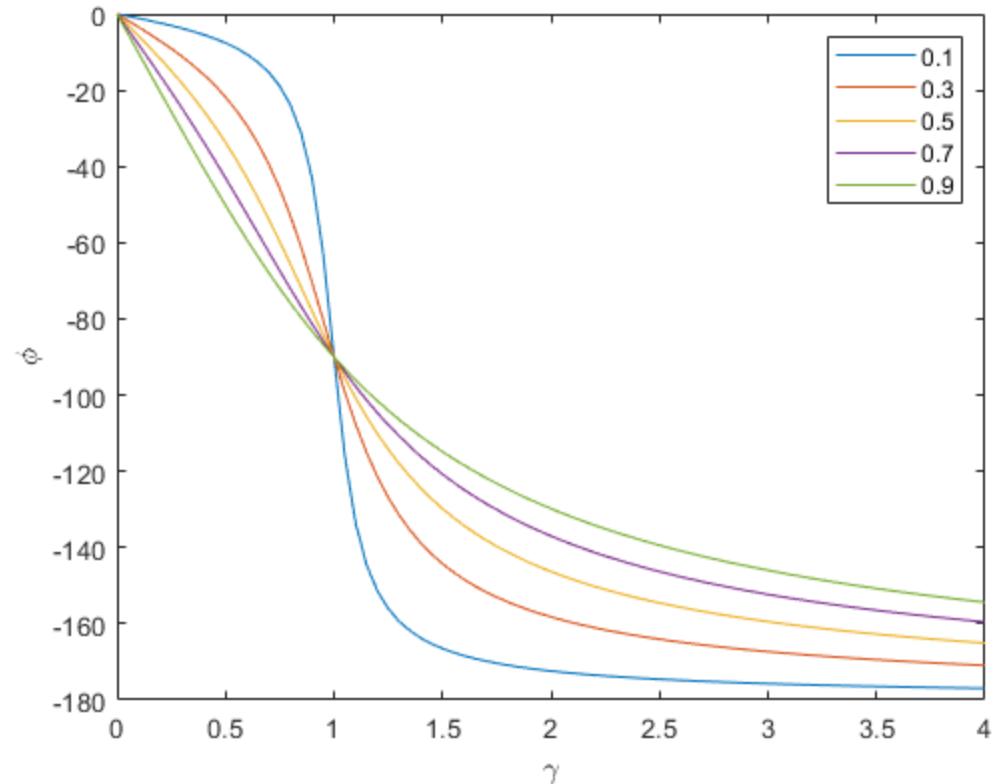
# The Second Order ODE: MATLAB

$$A = \frac{1}{\sqrt{\left(1 - \left(\frac{\Omega}{\omega_n}\right)^2\right)^2 + 4\zeta^2\left(\frac{\Omega}{\omega_n}\right)^2}} = \frac{1}{\sqrt{(1 - \gamma^2)^2 + 4\zeta^2\gamma^2}}, \quad \left(\gamma = \frac{\Omega}{\omega_n}\right)$$

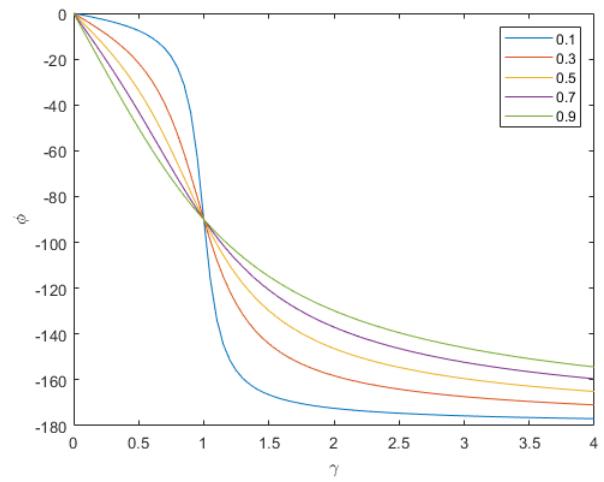
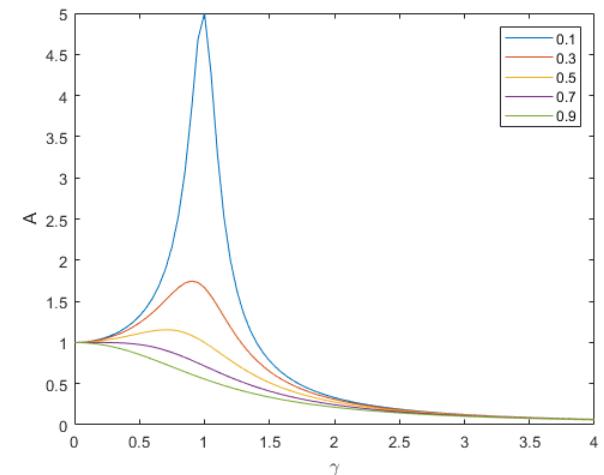
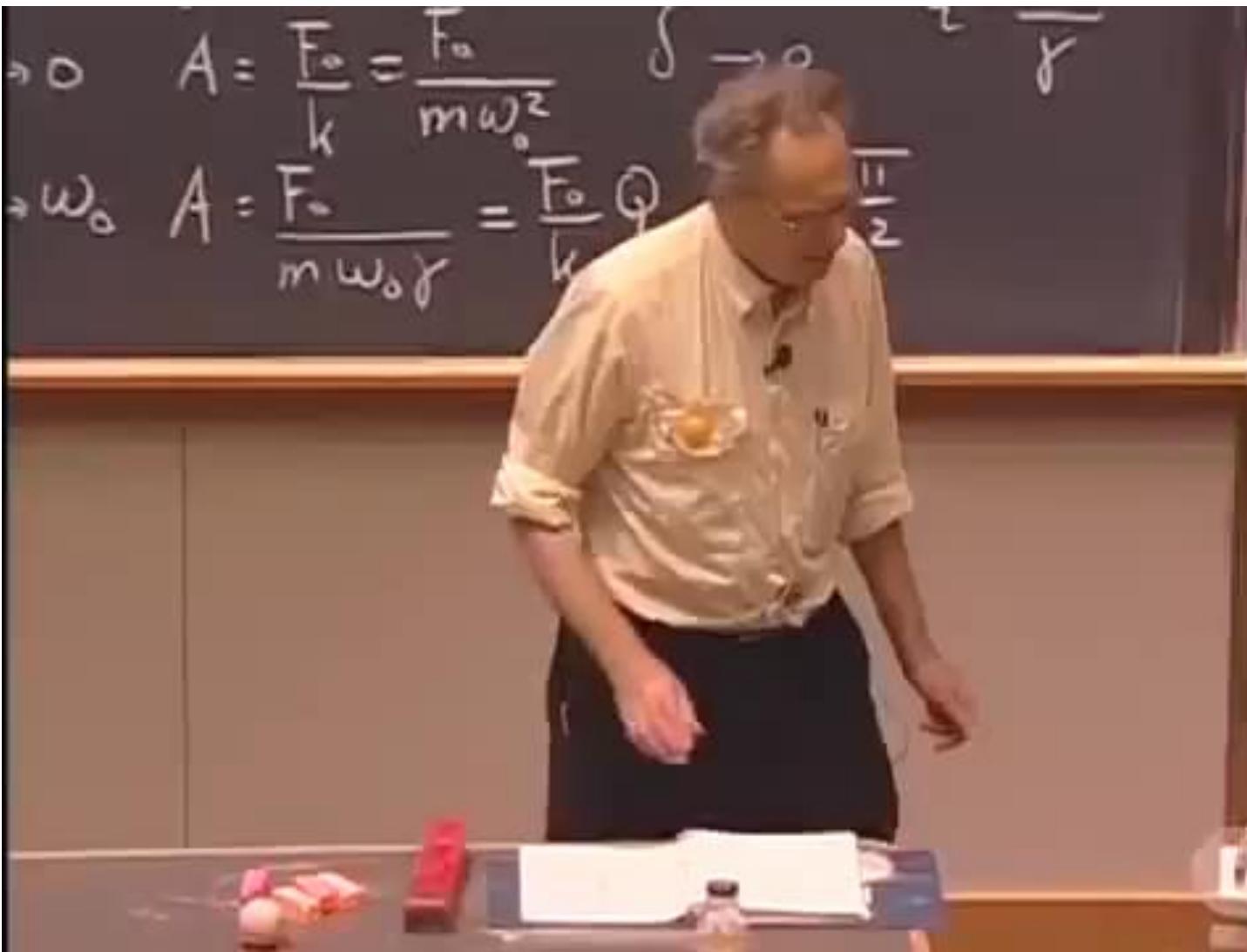
$$\phi = -\tan^{-1}\left(\frac{2\zeta\frac{\Omega}{\omega_n}}{1 - \left(\frac{\Omega}{\omega_n}\right)^2}\right) = -\tan^{-1}\left(\frac{2\zeta\gamma}{1 - \gamma^2}\right)$$

```
phi = [];
for i = 1:length(zeta)
    phi(i,:) = -atan2((2*zeta(i).*r),(1-r.^2));
end

plot(r,phi*180/pi)
xlabel('gamma')
ylabel('phi')
legend('0.1','0.3','0.5','0.7','0.9')
```

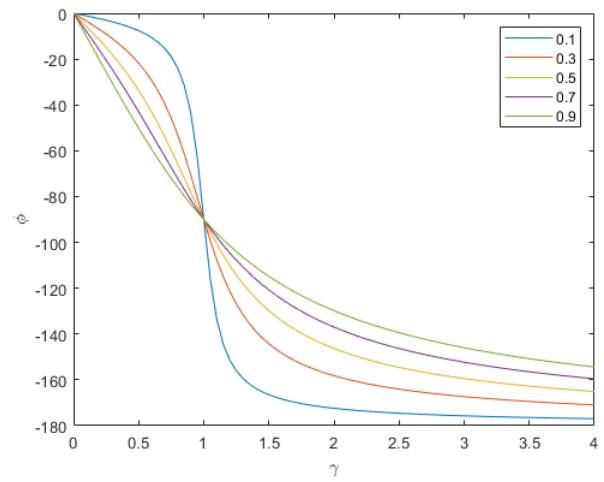
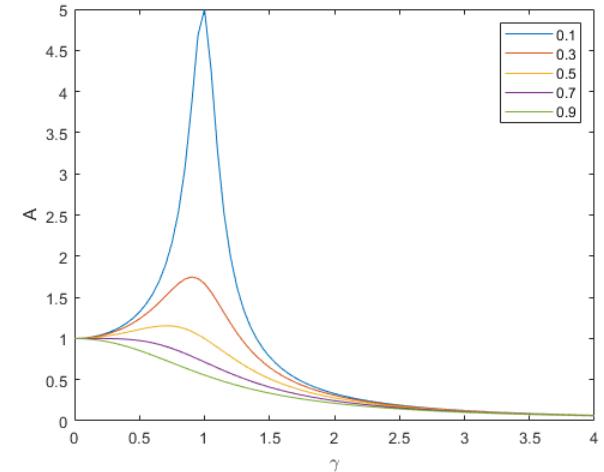


# Experiment: The Second Order ODE



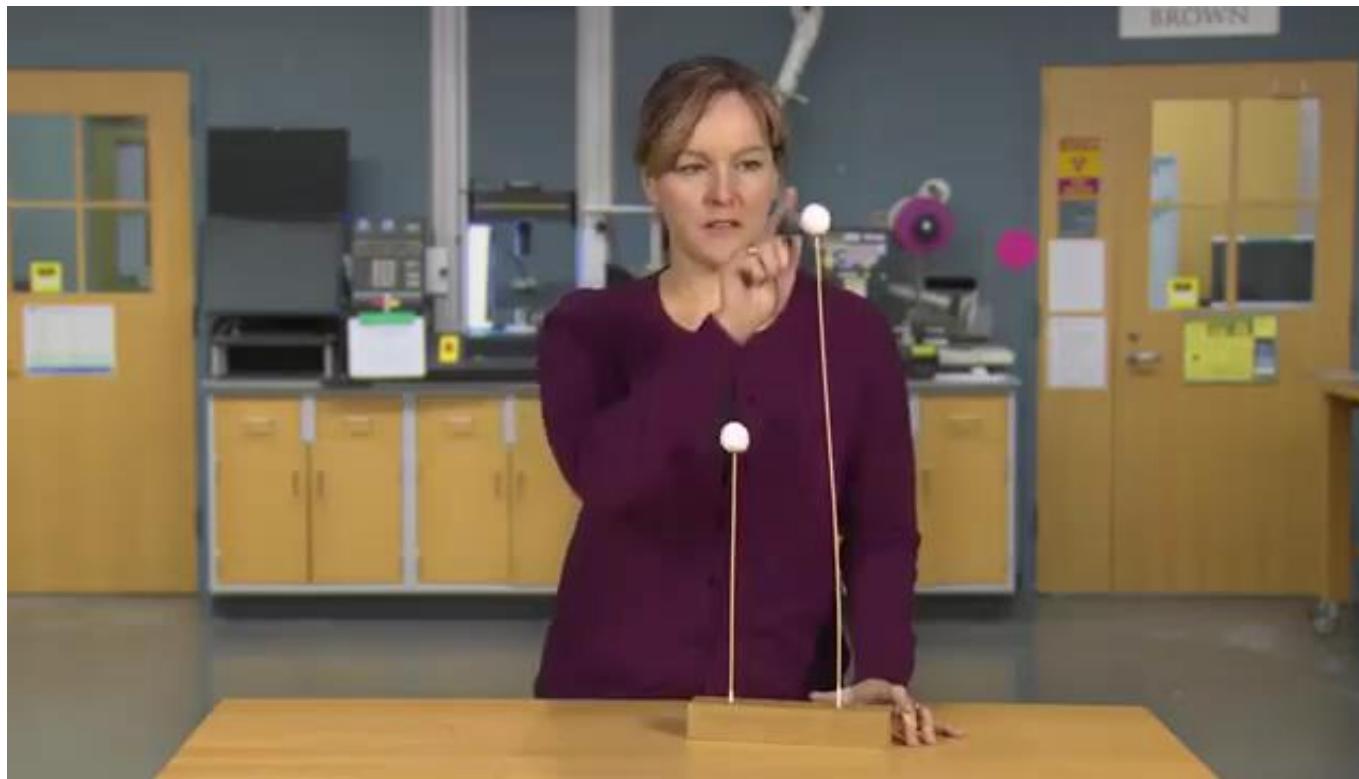
# Experiment: The Second Order ODE

$$A = \frac{\omega_0^2 - \omega^2}{\sqrt{(\omega_0^2 - \omega^2)^2 + (\omega\gamma)^2}} \quad \tan \delta = \frac{\omega\gamma}{\omega_0^2 - \omega^2} \quad x = A \cos(\omega t - \delta)$$
$$\omega \rightarrow \infty \quad A \rightarrow 0 \quad \delta = \pi$$
$$\omega \rightarrow 0 \quad A = \frac{F_0}{k} = \frac{F_0}{m\omega_0^2} \quad \delta \rightarrow 0 \quad Q =$$
$$\omega = \omega_0 \quad A = \frac{F_0}{\omega_0\gamma} = \frac{F_0}{k}Q \quad \delta \rightarrow \frac{\pi}{2}$$

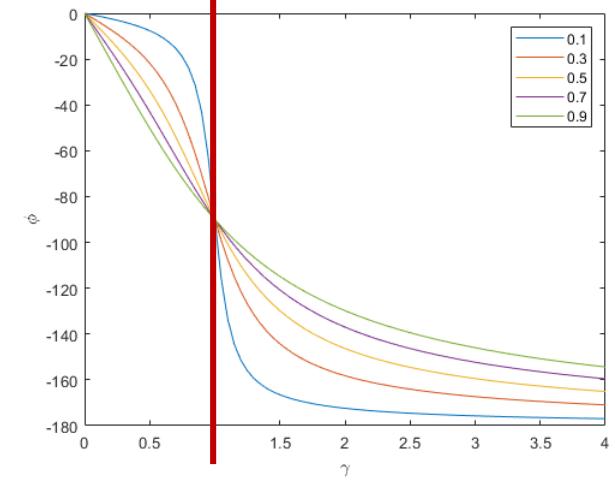
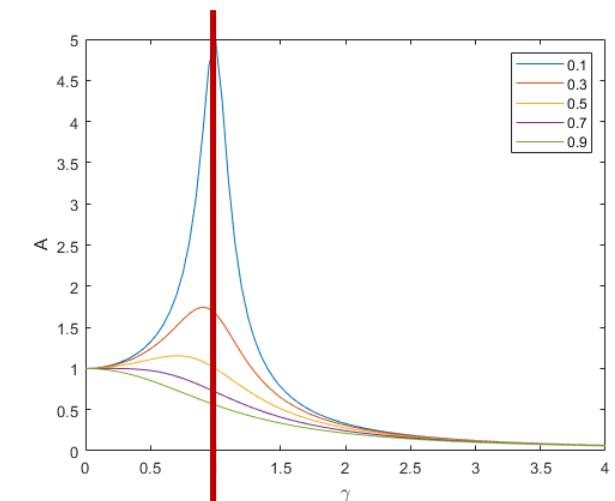


# Resonance

- Input frequency near resonance frequency
- Resonance frequency is generally different from natural frequency, but they often are close enough



Resonance frequency



# Summary

- To understand LTI system
- Impulse response
- Frequency sweep