

Control with Matlab

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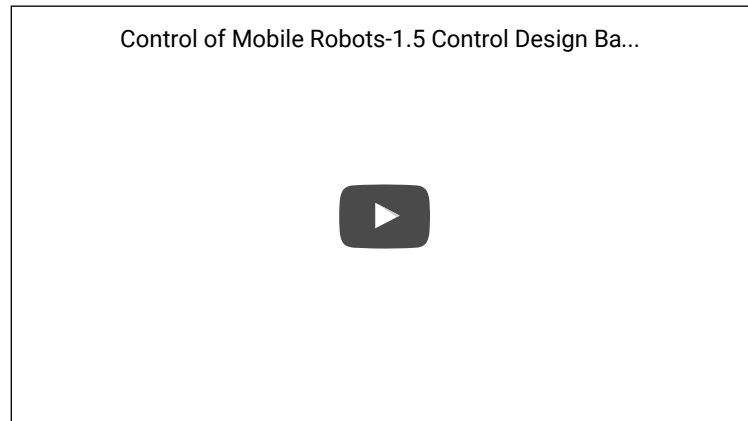
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1. PID Control

- 1.5 Control Design Basics | Control of Mobile Robots

In [1]:

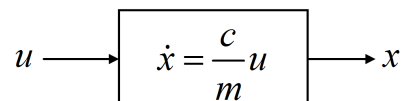
```
%%html
<iframe src="https://www.youtube.com/embed/DJuo9kLdr4M?list=PLp8ijpvp8iCvFDYdcXqqYU5Ibl_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```



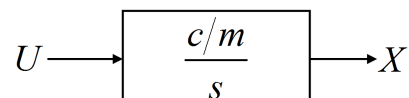
For the given car model

$$\dot{x} = \frac{c}{m}u \quad \text{for the velocity of a car, } x$$

In a block diagram



in a Laplace transform



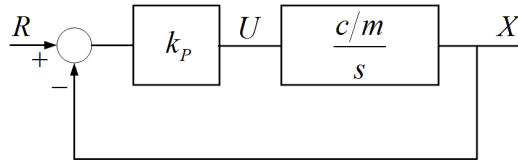
We want to achieve

$$x \rightarrow r \quad \text{as} \quad t \rightarrow \infty \quad (e = r - x \rightarrow 0)$$

Attempt 2: P Regulator

$$u = ke$$

- small error yields small control signals
- nice and smooth
- so-called proportional regulation (P regulator)



In [2]:

```
c = 1;
m = 1;

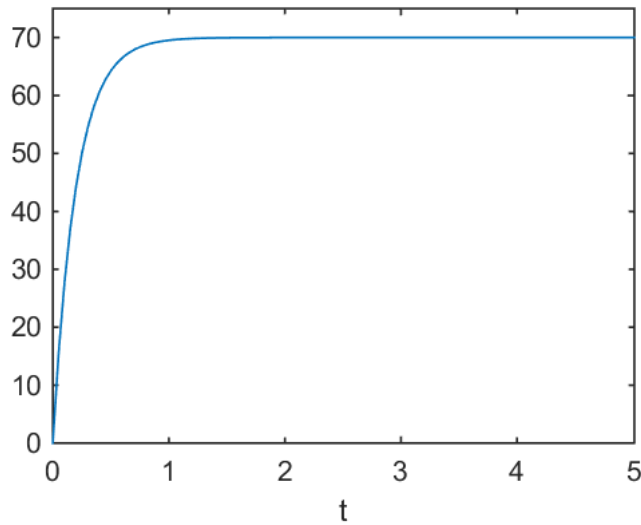
G = tf(c/m,[1 0]);

k = 5;
C = k;

Gcl = feedback(C*G,1,-1);

t = linspace(0,5,100);
r = 70*ones(size(t)); % reference

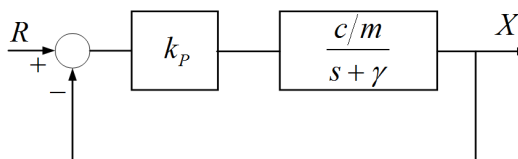
x0 = 0;
[y,tout] = lsim(Gcl,r,t,x0);
plot(tout,y), xlabel('t'), ylim([0,75])
```



Out[2]:

What if the true system is:

$$\dot{x} = \frac{c}{m}u - \gamma x$$

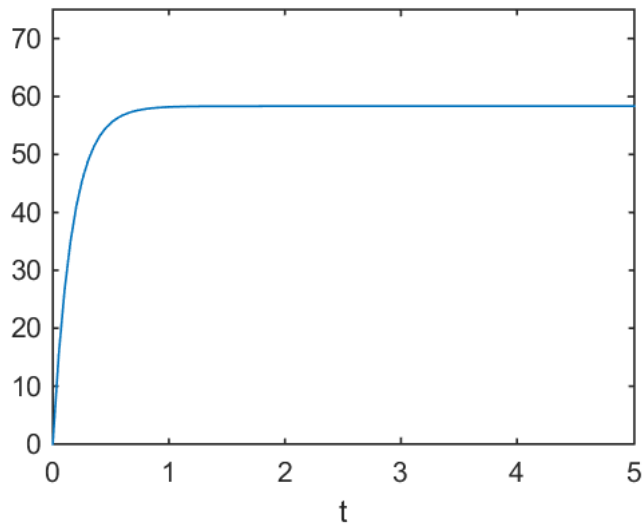


In [3]:

```
gamma = 1;
Gtr = tf(c/m,[1 gamma]);
C = k;

Gcl = feedback(C*Gtr,1,-1);

x0 = 0;
t = linspace(0,5,100);
r = 70*ones(size(t));
[y,tout] = lsim(Gcl,r,t,x0);
plot(tout,y), xlabel('t'), ylim([0,75])
```



Out[3]:

Attempt 3

- 1.6 Performance Objectives | Control of Mobile Robots
- 1.7 PID Control | Control of Mobile Robots

In [4]:

```
%%html
<iframe src="https://www.youtube.com/embed/cQhqx65kLfM?list=PLp8ijpvp8iCvFDYdcXqqYU5Ib1_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

Control of Mobile Robots-1.6 Performance Obje...



In [5]:

```
%%html
<iframe src="https://www.youtube.com/embed/Mk1yGHj4zxw?list=PLp8ijpvp8iCvFDYdcXqqYU5Ibl_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

Control of Mobile Robots-1.7 PID Control



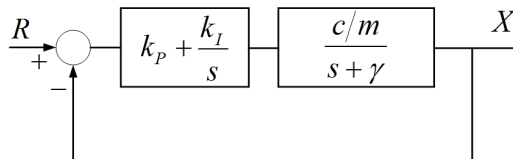
$$u = ke + \gamma \frac{m}{c} x$$

However, all of sudden we have to know all these physical parameters that we typically do not know - not robust !!!

Attempt 4: PI Regulators

- Stability (BIBO)
- Tracking
- Robustness

$$u(t) = k_P e(t) + k_I \int_0^t e(\tau) d\tau$$

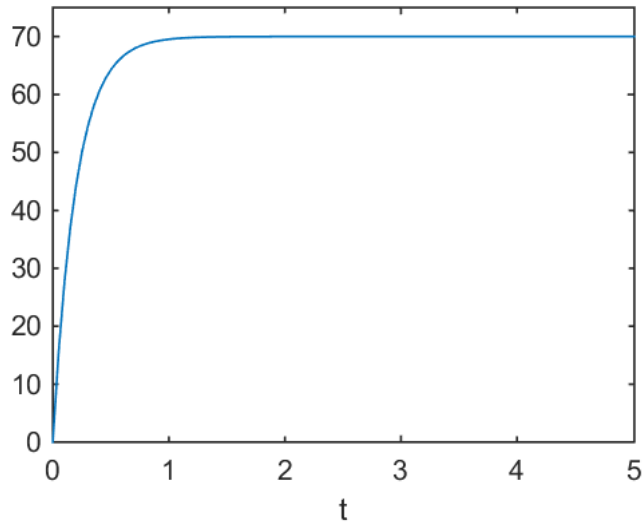


In [6]:

```
Gtr = tf(c/m,[1 gamma]);
kP = 5;
kI = 5;
C = tf([kP kI],[1 0]);

Gcl = feedback(C*Gtr,1,-1);

x0 = 0;
t = linspace(0,5,100);
r = 70*ones(size(t));
[y,tout] = lsim(Gcl,r,t,x0);
plot(tout,y), xlabel('t'), ylim([0,75])
```



Out[6]:

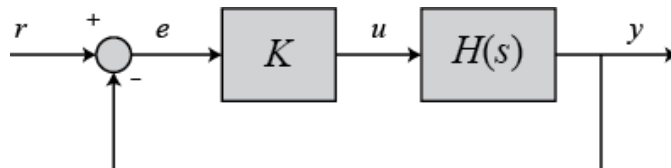
2. Root Locus

- The Root Locus Method by Brian Douglas
- from [umich control](http://ctms.engin.umich.edu/CTMS/index.php?example=Introduction§ion=ControlRootLocus) (<http://ctms.engin.umich.edu/CTMS/index.php?example=Introduction§ion=ControlRootLocus>)

In [7]:

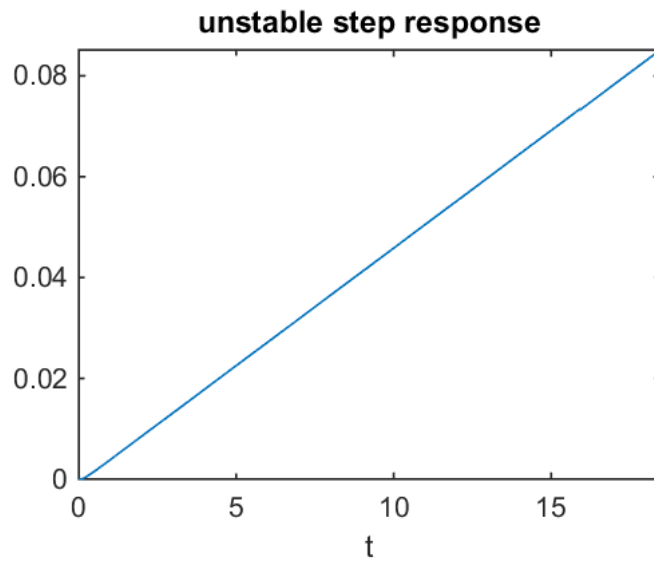
```
%%html
<iframe src="https://www.youtube.com/embed/CRvVDoQJjYI?list=PLUMWjy5jgHK1NC52DXXrriwihVrYZKqjk"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

The Root Locus Method - Introduction



In [8]:

```
s = tf('s');  
sys = (s + 7)/(s*(s + 5)*(s + 15)*(s + 20));  
[y,tout] = step(sys); % unstable  
plot(tout,y), axis tight, xlabel('t'), title('unstable step response')
```



Out[8]:

In [9]:

```
pole(sys)  
G = sys;
```

Out[9]:

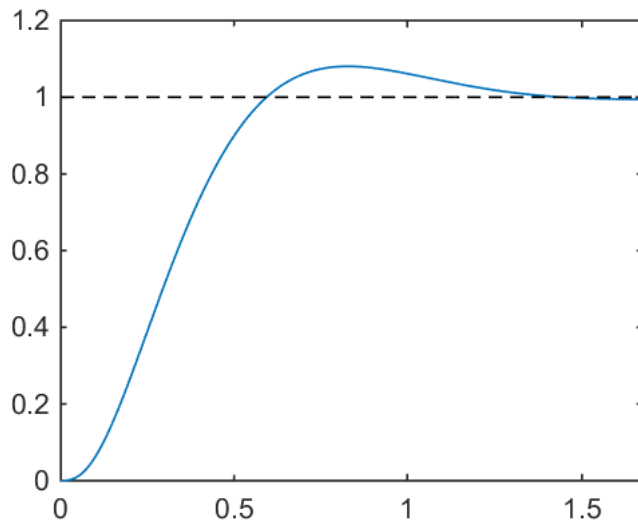
```
ans =  
  
      0  
-20.0000  
-15.0000  
-5.0000
```

```

In [10]: %k = 10;
          k = 800

          Gc1 = feedback(k*G,1,-1)
          pole(Gc1)
          [y,tout] = step(Gc1);
          plot(tout,y,tout,ones(size(tout)),'k--'), axis tight, ylim([0,1.2])

```



```

Out[10]: k =
          800

          Gc1 =
              800 s + 5600
          -----
          s^4 + 40 s^3 + 475 s^2 + 2300 s + 5600

          Continuous-time transfer function.

          ans =

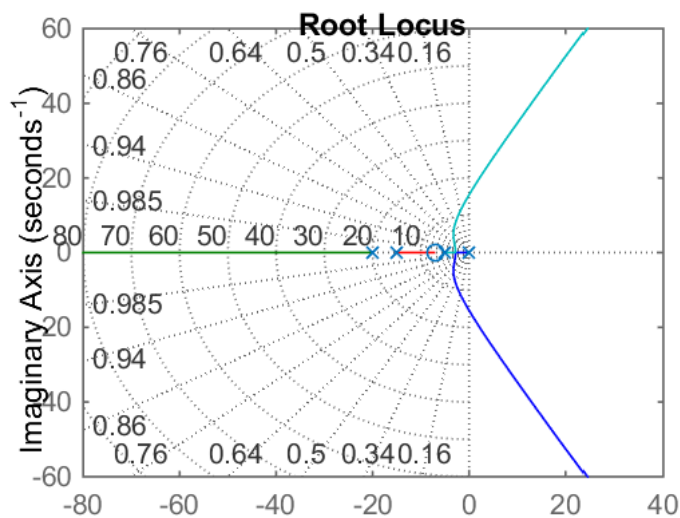
          -23.5466 + 0.0000i
          -10.1226 + 0.0000i
          -3.1654 + 3.6708i
          -3.1654 - 3.6708i

```

```

In [11]: rlocus(G), grid

```



```

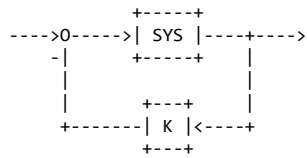
Out[11]:

```

```
In [12]: help rlocus
```

```
Out[12]: RLOCUS  Evans root locus.
```

RLOCUS(SYS) computes and plots the root locus of the single-input, single-output LTI model SYS. The root locus plot is used to analyze the negative feedback loop



and shows the trajectories of the closed-loop poles when the feedback gain K varies from 0 to Inf. RLOCUS automatically generates a set of positive gain values that produce a smooth plot.

RLOCUS(SYS,K) uses a user-specified vector K of gain values.

RLOCUS(SYS1,SYS2,...) draws the root loci of several models SYS1,SYS2,... on a single plot. You can specify a color, line style, and marker for each model, for example:

```
rlocus(sys1,'r',sys2,'y:',sys3,'gx').
```

[R,K] = RLOCUS(SYS) or R = RLOCUS(SYS,K) returns the matrix R of complex root locations for the gains K. R has LENGTH(K) columns and its j-th column lists the closed-loop roots for the gain K(j).

See RLOCUSPLOT for additional graphical options for root locus plots.

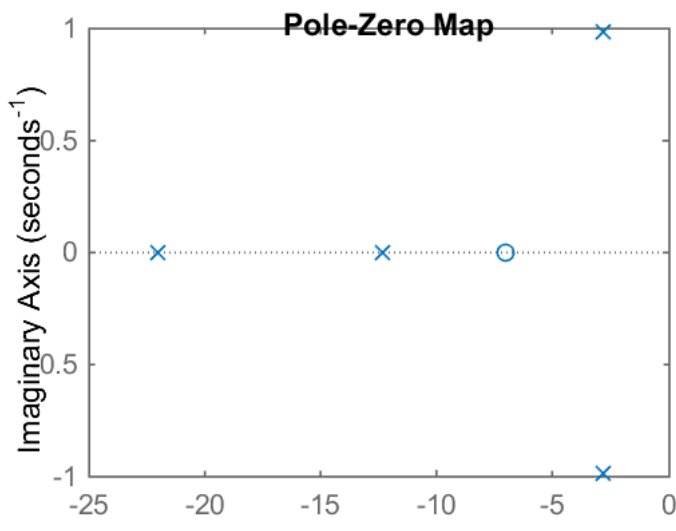
See also RLOCUSPLOT, SISOTOOL, POLE, ISSISO, LTI.

Overloaded methods:

DynamicSystem/rlocus
resppack.ltisource/rlocus

Reference page in Help browser
doc rlocus

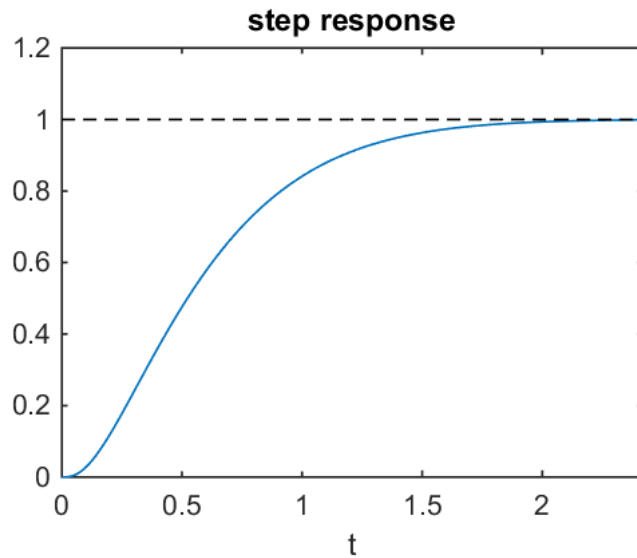
```
In [13]: K = 350;  
sys_cl = feedback(K*sys,1,-1) % negative feedback  
pzmap(sys_cl)
```



```
Out[13]: sys_cl =  
  
          350 s + 2450  
-----  
s^4 + 40 s^3 + 475 s^2 + 1850 s + 2450  
  
Continuous-time transfer function.
```



```
In [14]: [y,tout] = step(sys_cl);
plot(tout,y,tout,ones(size(tout)), 'k--'),
axis tight, ylim([0,1.2]), xlabel('t'), title('step response')
```



Out[14]:

3. State Space Representation

- from 3.1 A Simple Robot | Control of Mobile Robots
- from 3.2 State Space Models | Control of Mobile Robots

```
In [15]: %%html
<iframe src="https://www.youtube.com/embed/kQNUpNh6nBc?list=PLp8ijpvp8iCvFDYdcXqqYU5Ib1_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

3.1 A Simple Robot



In [16]:

```
%%html
<iframe src="https://www.youtube.com/embed/W6AU0yj5bFA?list=PLp8ijpvp8iCvFDYdcXqqYU5Ibl_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

3.2 State Space Models



Controlling a point mass

Given a point mass on a line whose acceleration is directly controlled:

$$\ddot{p} = u$$

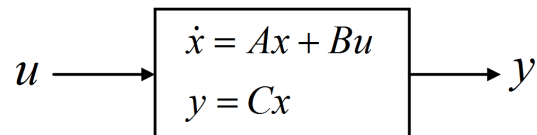
want to write this on a compact/general form

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= u\end{aligned}$$

on a state space form

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

$$y = p = x_1 = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$



3.1. The car model

If we care about/can measure the velocity:

$$A = -\gamma, \quad B = \frac{c}{m}, \quad C = 1$$

If we care about/can measure the position we have the same general equation with different matrices:

$$A = \begin{bmatrix} 0 & 1 \\ 0 & -\gamma \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ \frac{c}{m} \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

In [17]:

```
% system in ss
c = 1;
m = 1;
gamma = 1;

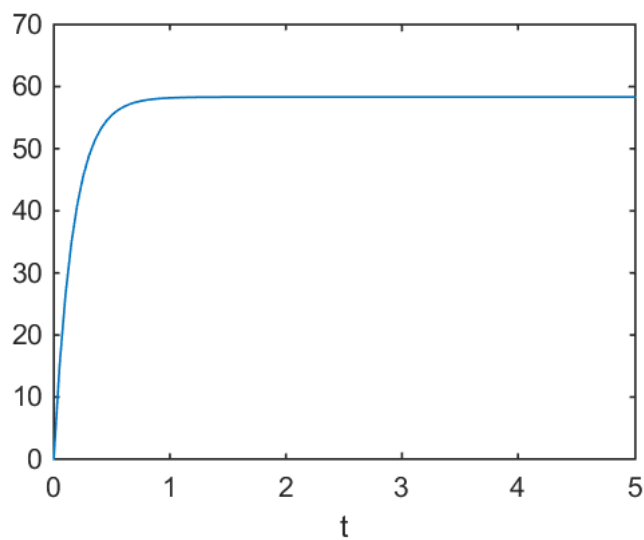
A = -gamma;
B = c/m;
C = 1;
D = 0;

Gss = ss(A,B,C,D);

% P controller
k = 5;
C = k;

% close loop
Gcl = feedback(C*Gss,1,-1);

x0 = 0;
t = linspace(0,5,100);
r = 70*ones(size(t));
[y,tout] = lsim(Gcl,r,t,x0);
plot(tout,y), xlabel('t'), ylim([0,70])
```



Out[17]:

3.2. Back to the World's Simplest Robot (Output Feedback)

- from 3.7 Output Feedback | Control of Mobile Robots

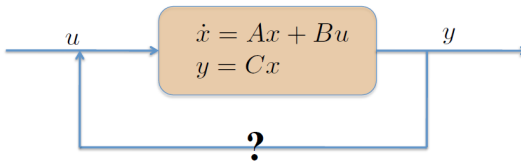
In [18]:

```
%%html
<iframe src="https://www.youtube.com/embed/HmqOnsRH73w?list=PLp8ijpvp8iCvFDYdcXqqYU5Ib1_a0qwjr"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

3.7 Output Feedback



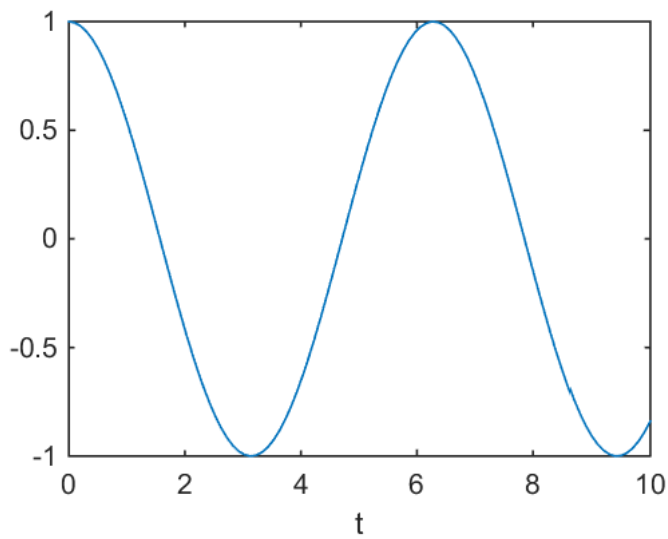
$$u = -y$$



In [19]: % to move towards the origin
% u = -y

```
A = [0 1; 0 0];
B = [0 1]';
C = [1 0];
D = 0;
G = ss(A,B,C,D);

K = 1;
Gcl = feedback(G,K,-1);
x0 = [1 0]';
t = linspace(0,10,100);
r = zeros(size(t));
[y,tout] = lsim(Gcl,r,t,x0);
plot(tout,y), xlabel('t')
```



Out[19]:

In [20]: eig(Gcl)

Out[20]: ans =
0.0000 + 1.0000i
0.0000 - 1.0000i

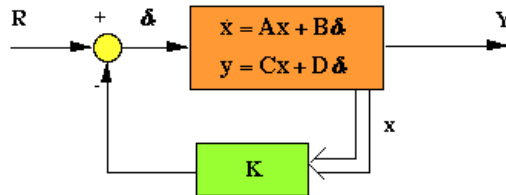
4. State Feedback

- from 3.8 State Feedback | Control of Mobile Robots

In [21]:

```
%%html
<iframe src="https://www.youtube.com/embed/y15IiJOYQps?list=PLciAw3uhNCiD3dkLTPJgHoMnsu8XgCt1m"
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

3.8 State Feedback | Control of Mobile Robots



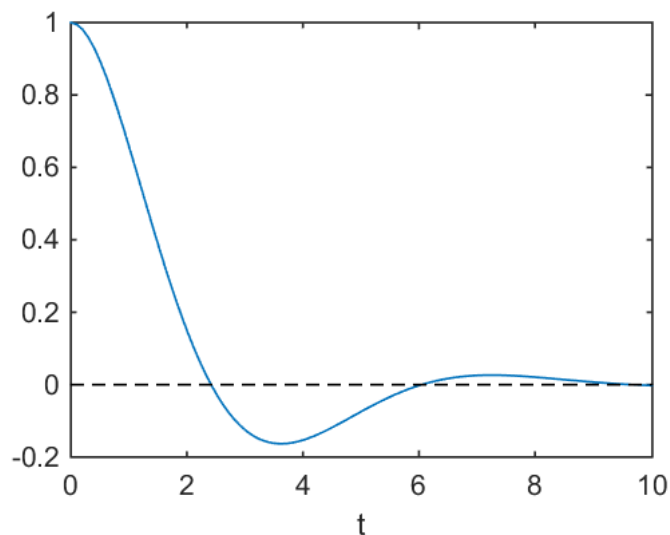
To move forwards origin, $R = 0$

In [22]:

```
A = [0 1; 0 0];
B = [0 1]';
C = [1 0];
D = 0;
G = ss(A,B,C,D);

k1 = 1;
k2 = 1;
K = [k1 k2];
Gc1 = ss(A-B*K,B,C,D);

x0 = [1 0]';
t = linspace(0,10,100);
r = zeros(size(t));
[y,tout] = lsim(Gc1,r,t,x0);
plot(tout,y,tout,zeros(size(tout)),'k--'), xlabel('t')
```



Out[22]:

In [23]:

```
eig(Gc1)
```

Out[23]:

```
ans =
```

```
-0.5000 + 0.8660i  
-0.5000 - 0.8660i
```

Eigenvalues Matter

- It is clear that some eigenvalues are better than others. Some cause oscillations, some make the system respond too slowly, and so forth ...
- In the next module we will see how to select eigenvalues and how to pick control laws based on the output rather than the state.

4.1. Pole Placement

- from 4.1 Stabilizing the Point Mass | Control of Mobile Robots
- from 4.2 Pole Placement | Control of Mobile Robots

In [24]:

```
%%html  
<iframe src="https://www.youtube.com/embed/S4wZTmEnbrY?list=PLp8ijpvp8iCvFDYdcXqqYU5Ibl_a0qwjr"  
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

4.1 Stabilizing the Point Mass



In [25]:

```
%%html  
<iframe src="https://www.youtube.com/embed/5tWhOK8Klo0?list=PLp8ijpvp8iCvFDYdcXqqYU5Ibl_a0qwjr"  
width="560" height="315" frameborder="0" allowfullscreen></iframe>
```

4.2 Pole Placement

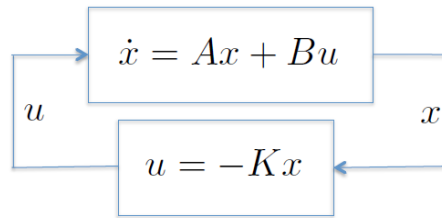


back to the point-mass, again

$$u = -Kx \rightarrow \dot{x} = (A - BK)x$$
$$A - BK = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} k_1 & k_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -k_1 & -k_2 \end{bmatrix}$$
$$\begin{vmatrix} 0 & 1 \\ -k_1 & -k_2 \end{vmatrix} = \lambda^2 + \lambda k_2 + k_1$$

Desired Eigenvalues: let's pick both eigenvalues at -1

$$(\lambda + 1)(\lambda + 1) = \lambda^2 + 2\lambda + 1$$
$$k_1 = 2, k_2 = 1$$



Pick the control gains such that the eigenvalues (poles) of the closed loop system match the desired eigenvalues

Questions

- Is this always possible? (No)
- How should we pick the eigenvalues? (Mix of art and science)

$$\dot{x} = \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u$$

$$A - BK = \begin{bmatrix} 2 - k_1 & -k_2 \\ 1 - k_1 & 1 - k_2 \end{bmatrix}$$

$$\varphi = \lambda^2 + \lambda(-3 + k_1 + k_2) + 2 - k_1 - k_2$$

Suppose

$$\varphi = (\lambda + 1)^2 = \lambda^2 + \lambda(-3 + k_1 + k_2) + 2 - k_1 - k_2$$

Let's pick both eigenvalues at -1

$$-3 + k_1 + k_2 = 2 \quad \text{and} \quad 2 - k_1 - k_2 = 1$$

→ no k_1 and k_2 exist

What's at play here is a lack of controllability, i.e., the effect of the input is not sufficiently rich to influence the system enough

In [26]:

```
A = [2 0;
      1 -1];

B = [1 1]';
C = [1 0];

P = [-0.5 + 1j, -0.5 - 1j];
%P = [-0.1 + 1j, -0.1 - 1j];
%P = [-0.5, -1];
%P = [-5, -4];

K = place(A,B,P)
```

Out[26]:

```
K =

    2.6250    -0.6250
```

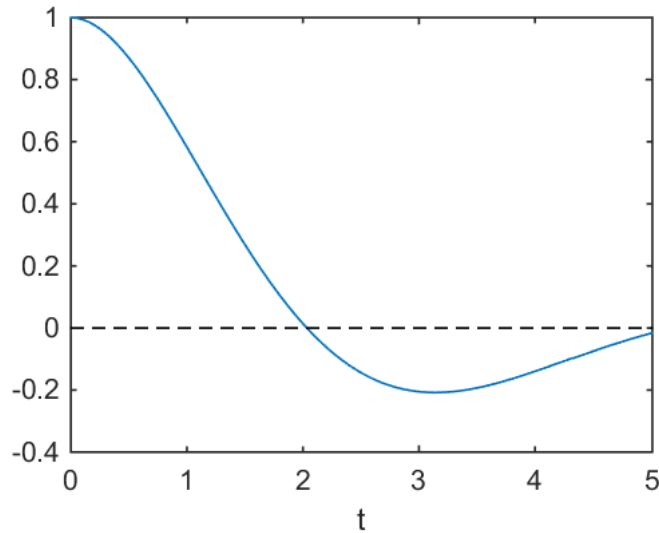
$$\dot{x} = Ax + Bu = Ax - BKx = (A - BK)x$$

In [27]:

```
x0 = [1 1]';
Gcl = ss(A-B*K,B,C, 0);

t = linspace(0,5,100);
u = zeros(size(t));

[y, tout] = lsim(Gcl,u,t,x0);
plot(tout,y,tout,zeros(size(tout)),'k--'), xlabel('t')
```



Out[27]:

4.2. Controllability

- When can we place the eigenvalues using state feedback?
- When is B matrix (the actuator configuration) rich enough so that we can make the system do whatever we want it to do?
- The answer revolves around the concept of controllability

Given a discrete-time system

$$x_{k+1} = Ax_k + Bu_k$$

We would like to drive this system in n steps to a particular target state x^*

$$\begin{aligned}x_1 &= Ax_0 + Bu_0 = Bu_0 \\x_2 &= Ax_1 + Bu_1 = ABu_0 + Bu_1 \\x_3 &= Ax_2 + Bu_2 = A^2Bu_0 + ABu_1 + Bu_2 \\&\vdots \\x_n &= A^{n-1}Bu_0 + \dots + Bu_{n-1}\end{aligned}$$

We want to solve

$$x^* = \begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix} \begin{bmatrix} u_{n-1} \\ \vdots \\ u_1 \\ u_0 \end{bmatrix}$$

This is possible for any target state if and only if

$$\text{rank} \begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix} = n$$

Example 1

$$\dot{x} = \begin{bmatrix} 2 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u$$


```
In [28]: A = [2 0;
            1 1];
        B = [1 1]';

        G = ctrb(A,B)
        rank(G)
```

```
Out[28]: G =

         1     2
         1     2

ans =

         1
```

Example 2

$$\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

```
In [29]: A = [0 1;
            0 0];
        B = [0 1]';

        G = ctrb(A,B)
        rank(G)
```

```
Out[29]: G =

         0     1
         1     0

ans =

         2
```

```
In [30]: %%javascript
$.getScript('https://kmahelona.github.io/ipython_notebook_goodies/ipython_notebook_toc.js')
```