

Analysis of Linear Control Systems

Topics:

- Properties of linear dynamic systems
- Time-domain analysis of linear systems
- Numerical simulation of linear system
- Root-locus of linear systems
- Frequency-domain analysis of linear systems
- Introduction to model Reduction techniques

Properties of linear dynamic systems

- Linear systems obey the superposition principle
- Stability analysis
 - Bounded-input bounded-output (BIBO) stability
 - A system is stable if all its poles have negative real-parts
 - Poles are the roots of the denominator of the system's transfer function $G(s)$
 - Zeros (transmission zeros) are the roots of the numerator of the system's transfer function $G(s)$
 - Poles on the imaginary axis with multiplicity one are critically stable
 - Poles on the imaginary axis with multiplicity more than one are unstable
- Matlab commands
 - Find system poles using commands `pole()` and `eig()`
 - Find system zeros using command `zero()`
 - Sketch system's poles and zeros using command `pzmap()`
 - Use command `isstable()` to check system's stability, returns 1=stable, 0=unstable

Example: Check the stability of system $G(s) = \frac{s^3+7s^2+24s+24}{s^4+10s^3+35s^2+50s+24}$

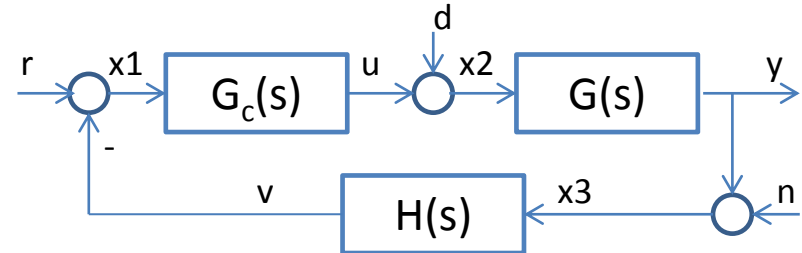
```
>> G=tf([1,7,24,24],[1,10,35,50,24]); eig(G); pzmap(G);
```

Or

```
>> s=tf('s'); G=(s^3+7*s^2+24*s+24)/(s^4+10*s^3+35*s^2+50*s+24); isstable(G);
```

Properties of linear dynamic systems ...

Internal stability



- The system shown is internally stable iff:
 - 1) The transfer function of $1+H(s)G(s)G_c(s)$ has no zeros in RHP
 - 2) The loop transfer function $H(s)G(s)G_c(s)$ has no pole-zero cancellation in RHP

Matlab command: **intstable()**

- Syntax: `[V,c]=intstable(G,Gc,H);`
 - If system is internally stable , $V=0$ and c is empty
 - If system is I/O unstable, $V=1$ and c holds the unstable closed-loop poles
 - If system is I/O stable, but not internally stable, $V=2$ and c holds the cancelled unstable poles
- Matlab command **minreal()** gives the simplified transfer function, after pole-zero cancellation, which may not be internally stable

Properties of linear dynamic systems ...

Controllability and observability analysis

- **Controllability:**
- The state $x_i(t)$ is said to be controllable if there exists an input that in finite time can drive it to any specified value $x_i(t_f)$ from the initial value $x_i(0)$
- The system is fully controllable if all its states are controllable
 - Full controllability of the system depends on the A and B matrices of its state-space model
 - An nth-order system is fully controllable if its controllability matrix $T_c = [B, AB, \dots, A^{n-1}B]$ has full rank

Matlab commands: **ctrb()**, **ctrbf()**, **rank()**

- $T_c = \text{ctrb}(A, B)$; returns the controllability matrix T_c
- $[A_c, B_c, C_c, T_c] = \text{ctrbf}(A, B, C)$; returns the equivalent state-space matrices (A_c, B_c, C_c) in the stair-case form $A_c = \begin{bmatrix} \hat{A}_{\bar{c}} & 0 \\ \hat{A}_{21} & \hat{A}_c \end{bmatrix}$, $B_c = \begin{bmatrix} 0 \\ \hat{B}_c \end{bmatrix}$, $C_c = [\hat{C}_{\bar{c}} \quad \hat{C}_c]$, where $(\hat{A}_c, \hat{B}_c, \hat{C}_c)$ represent the controllable subsystem $G(s) = \hat{C}_c(sI - \hat{A}_c)^{-1}\hat{B}_c + D$
- $\text{rank}(T_c)$; generates the rank of the matrix T_c

Properties of linear dynamic systems ...

- Observability

- The state $x_i(t)$ is said to be observable if for any $t_f > 0$, the initial state $x_i(0)$ can be determined from the time history of the input $u(t)$ and output $y(t)$ in the interval $[0, t_f]$
- The system is fully observable if all system states are observable
 - Full observability of the system depends on the A and C matrices of its state-space model

- An nth-order system is fully observable if its observability matrix $T_o = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix}$ has full rank
 - Dual of controllability

Matlab commands: **obsv()**, **obsvf()**

- $T_o = \text{obsv}(A, C)$; returns the observability matrix T_o
- $[A_o, B_o, C_o, T_o] = \text{obsvf}(A, B, C)$; returns the equivalent state-space matrices (A_o, B_o, C_o) in the stair-case form $A_o = \begin{bmatrix} \hat{A}_{\bar{o}} & \hat{A}_{12} \\ 0 & \hat{A}_o \end{bmatrix}$, $B_o = \begin{bmatrix} \hat{B}_{\bar{o}} \\ \hat{B}_o \end{bmatrix}$, $C_o = [0 \quad \hat{C}_o]$, where $(\hat{A}_o, \hat{B}_o, \hat{C}_o)$ represent the observable subsystem $G(s) = \hat{C}_o (sI - \hat{A}_o)^{-1} \hat{B}_o + D$

Properties of linear dynamic systems ...

Controllability and observability Gramians

- Controllability and observability Gramians W_c and W_o show how controllable and observable a system is, where $W_c = \int_0^\infty e^{At} B B^T e^{A^T t} dt$ and $W_o = \int_0^\infty e^{A^T t} C^T C e^{At} dt$
 - W_c and W_o satisfy the Lyapunov equations $AW_c + W_c A^T = -BB^T$ and $A^T W_o + W_o A = -C^T C$
 - W_c and W_o are positive definite if and only if (A,B) is controllable and (A,C) is observable
 - The singular values of W_c indicate the contribution of the input signal to each state
 - The singular values of W_o indicate the contribution of each state to the output signal

Matlab commands: **lyap()**, **svd()**, **gram()**

- $W_c = \text{lyap}(A, B*B')$; returns the controllability Gramian matrix W_c
- $W_o = \text{lyap}(A', C'*C)$; returns the observability Gramian matrix W_o
- $[U, S, V] = \text{svd}(W_c)$; produces a diagonal matrix S , of the same dimension as W_c and with nonnegative diagonal elements in decreasing order, and unitary matrices U and V so that $W_c = U*S*V'$
- $W = \text{gram}(G, \text{type})$; returns the gramian matrix W for a system with state-space model G , where type is 'c' or 'o' for controllability or observability Gramians

Properties of linear dynamic systems ...

Other Matlab commands:

- **kalmdec()**; produces Kalman decomposition of a given system
- **timmomt()**; produces time moments M_i of a given system, where $M_i = \int_0^\infty t^i g(t) dt$, and $g(t)$ is the impulse response of the system $G(s)$
- **markovp()**; produces the Markov parameters d_i of a given system, where $d_0 = CB + D$ and $d_i = CA^i B$, $i=1,2,\dots$

Norm measures of signals and systems

- Norm measures of signals
 - L_p -norm defines the size of a signal $u(t)$ as $\|u(t)\|_p = \left(\int_{-\infty}^\infty |u(t)|^p dt\right)^{1/p}$, where p is a positive integer
 - The L_1 -norm is $\|u(t)\|_1 = \int_{-\infty}^\infty |u(t)| dt$
 - The L_2 -norm is $\|u(t)\|_2 = \left(\int_{-\infty}^\infty |u(t)|^2 dt\right)^{1/2}$, (the measure of signal power)
 - The L_∞ -norm is $\|u(t)\|_\infty = \sup_t |u(t)|$, (the least upper-bound of $|u(t)|$)

Properties of linear dynamic systems ...

Norm measures of systems

- The size of a system $G(s)$ is generally measured in H_2 -norm and H_∞ -norm
 - The H_2 -norm is $\|G(s)\|_2 = \left(\frac{1}{2\pi j} \int_{-j\infty}^{j\infty} |G(j\omega)|^2 d\omega \right)^{1/2} = \left(\int_0^\infty \text{tr}(g(t)^T g(t)) dt \right)^{1/2}$
 - (square-root of the integral-squared of the impulse-response $g(t)$ of the system)
 - The H_∞ -norm is $\|G(s)\|_\infty = \sup_{u(t) \neq 0} \frac{\|y(t)\|_2}{\|u(t)\|_2} = \sup_{\omega} |G(j\omega)|$
 - (peak-value of the magnitude of the frequency-response of the system)
- Properties of L and H norms:
 - $\|y(t)\|_2 \leq \|G(s)\|_\infty \|u(t)\|_2$
 - $\|y(t)\|_\infty \leq \|G(s)\|_2 \|u(t)\|_\infty$
 - $\|G_1(s)G_2(s)\|_\infty \leq \|G_1(s)\|_\infty \|G_2(s)\|_\infty$

Matlab command: **norm()**

- `norm(X,P)`; X is a matrix, P is norm type 1, 2, inf, fro (Frobenius)
- `norm(V,P)`; V is a vector, P is norm type (1, 2, ..., inf (max), and -inf (min))
- `norm(G)`; H_2 -norm of $G(s)$
- `norm(G,inf)`; H_∞ -norm of $G(s)$

Time-Domain Analysis of Linear Systems

Analytical solutions to continuous-time responses

- Consider the system $\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$, with initial condition $x(0) = x_0$
 - The system response to an arbitrary input $u(t)$, for $t \geq 0$, is:
 - $$\begin{cases} x(t) = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau \\ y(t) = C\left(e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau\right) + Du(t) \end{cases}$$
 - Laplace transform of the solution is: $\begin{cases} X(s) = (sI - A)^{-1}(x(0) + BU(s)) \\ Y(s) = C(sI - A)^{-1}(x(0) + BU(s)) + DU(s) \end{cases}$

Analytical solutions to discrete-time responses

- Consider the system $\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) + Du(k) \end{cases}$, with sample-time T and $x(0) = x_0$
 - The system response to a sampled arbitrary input $u(k)$, for $k = 0, 1, 2, \dots$, is:
 - $$\begin{cases} x(k+1) = e^{AT}x(k) + \int_0^T e^{A\tau}Bd\tau u(k) \\ y(k) = C\left(e^{AT}x(k) + \int_0^T e^{A\tau}Bd\tau u(k)\right) + Du(k) \end{cases}, \text{ where } y(k) = \mathfrak{z}^{-1}(G(z)U(z))$$

Numerical Simulation of Linear Systems

Second-order system analysis

- Consider a second-order linear system $G(s) = \frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$

– The step response of the system is:

1. If $\zeta = 0$, $y(t) = 1 - \cos(\omega_n t)$

2. If $0 < \zeta < 1$, $y(t) = 1 - e^{-\zeta\omega_n t} \frac{1}{\sqrt{1-\zeta^2}} \sin(\omega_d t + \theta)$

where $\omega_d = \omega_n \sqrt{1-\zeta^2}$ and $\theta = \tan^{-1} \frac{\sqrt{1-\zeta^2}}{\zeta}$

3. If $\zeta = 1$, $y(t) = 1 - (1 + \omega_n t) e^{-\omega_n t}$

4. If $\zeta > 1$, $y(t) = 1 - \frac{\omega_n}{2\sqrt{\zeta^2-1}} \left(\frac{e^{\lambda_1 t}}{\lambda_1} - \frac{e^{\lambda_2 t}}{\lambda_2} \right)$

where $\lambda_1 = -\zeta - \sqrt{\zeta^2 - 1}$ and $\lambda_2 = -\zeta + \sqrt{\zeta^2 - 1}$

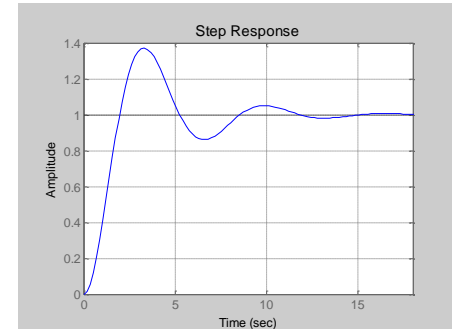
More concisely:

– if $\zeta \neq 1$, $y(t) = 1 - \omega_n e^{-\zeta\omega_n t} \left(\frac{\cosh(\omega_d t)}{\omega_n} - \frac{\zeta \sinh(\omega_d t)}{\omega_d} \right)$

Numerical Simulation of Linear Systems

Qualitative specifications in step responses

- Second-order linear system $G(s) = \frac{\omega_n^2}{s^2 + 2\zeta\omega_n s + \omega_n^2}$
 - Specifications of the step response (shown):
 - **Steady-state value, y_{ss}** : $y_{ss} = \lim_{s \rightarrow 0} sG(s) \frac{1}{s} = G(0)$
 - **Rise-time, t_r** : the time for the response to go from 10% to 90% of y_{ss}
 - **Settling-time, t_s** : the time when $y(t)$ enters and stays in the range $y_{ss} \pm \Delta y$
 - **Overshoot (maximum peak), M_p** : percent overshoot $M_p = \frac{y_{max} - y_{ss}}{y_{ss}} \times 100\%$

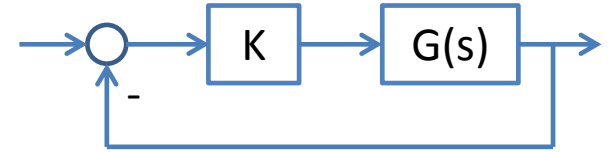


Matlab commands: **step()**, **dcgain()**, **impulse()**, **lsim()**, **grid**

- `step(G)`; draws the step response of $G(s)$
- `[y,t]=step(G)`; evaluate the step response of $G(s)$ and not draw it
- `K=dcgain(G)`; calculate the steady state value y_{ss} of the output $y(t)$
- `impulse(G)`; draw the impulse response of $G(s)$
- `lsim(G,u,t)`; draw the response of $G(s)$ to an arbitrary input $u(t)$
- `grid`; superimpose a grid on the plot

Root Locus of Linear Systems

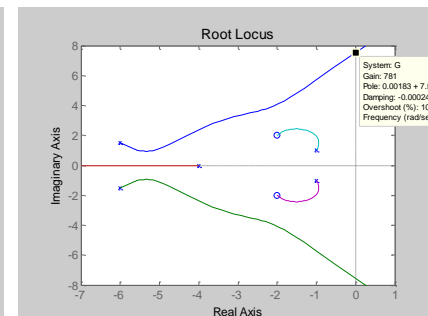
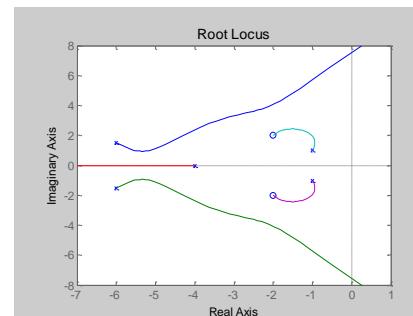
Unity feedback system



- Sketch the location of closed-loop poles
 - Roots location of $1 + KG(s) = 0$, where $-\infty < K < \infty$

Matlab commands: **rlocus()**, **grid**

- `rlocus(G)`; draws the root locus
- `rlocus(G,K)`; draw root-locus for a given gain vector K
- `[R,K]=rlocus(G)`; evaluate the closed-loop pole location R, not draw
- `rlocus(G1,'-',G2,'-.b'G3,':r')`; draw root-locus for several models
- Example: Draw the root-locus, where $G(s) = \frac{s^2+4s+8}{s^5+18s^4+120.3s^3+357.7s^2+478.5s+306}$
 - `>> num=[1,4,8]; den=[1,18,120.3,375.5,478.5,306];`
 - `G=tf(num,den); rlocus(G);`
 - Right click on the jw-axis crossing
 - Critical gain $k \approx 781$



Frequency-Domain Analysis of Linear Systems

Frequency-domain plots of $G(s)$, $s=j\omega$

- Real and imaginary part representation

$$G(j\omega) = P(\omega) + jQ(\omega)$$

- Nyquist plot draws $Q(\omega) = \text{Im}(G(j\omega))$ v. $P(\omega) = \text{Re}(G(j\omega))$

- Magnitude and Phase representation in separate plots

$$G(j\omega) = A(\omega)e^{-j\phi(\omega)}$$

- Bode plot draws the magnitude in decibels ($M(\omega) = 20\log(A(\omega))$) and phase in degrees versus the frequency in log scale

- Magnitude and Phase representation in a single plot

$$G(j\omega) = A(\omega)e^{-j\phi(\omega)}$$

- Nichols chart draws the magnitude versus phase

Matlab commands: **nyquist**, **bode()**, **nichols()**,

- `nyquist(G)`; draws the Nyquist plot
- `[R,I,K]=nyquist(G)`; evaluate the Nyquist data, not draw
- `bode(G)`; draw the Bode plot of $G(s)$
- `[A,phi,omega]=bode(G)`; evaluate the Bode data, not draw
- `nichols(G)`; draw the Nichols chart of $G(s)$
- `[A,phi,omega]=nichols(G)`; evaluate the Nichols data, not draw

Frequency-Domain Analysis of Linear Systems

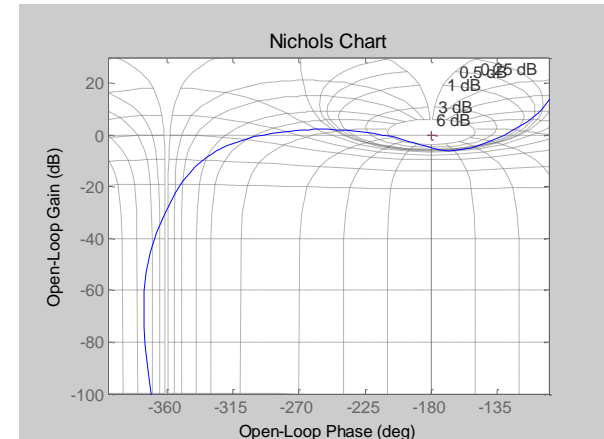
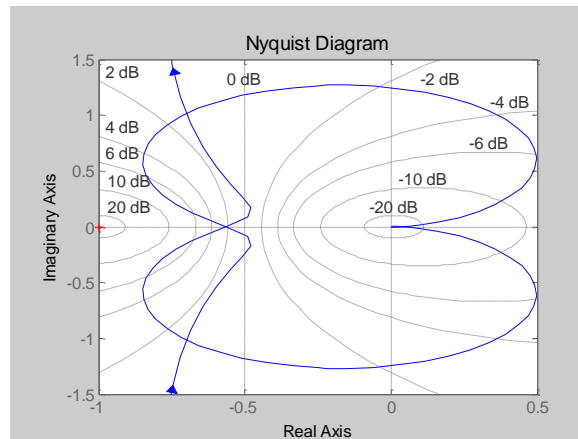
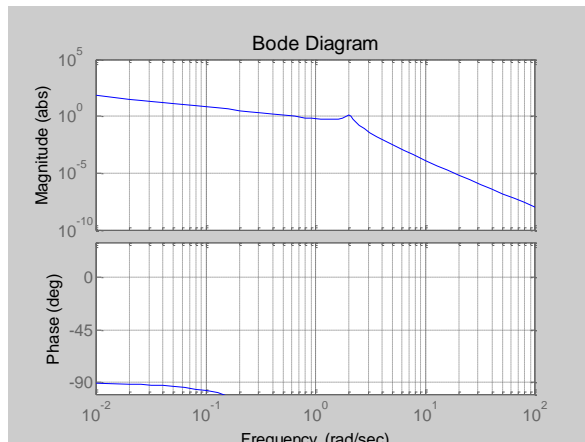
Example: Draw the Bode plot, Nyquist plot and Nichols chart for

$$G(s) = \frac{s+8}{s(s^2+0.2s+4)(s+1)(s+3)}$$

Matlab code:

```
close all; s=tf('s'); G=(s+8)/(s*(s^2+0.2*s+4)*(s+1)*(s+3));  
figure(1), bode(G);  
figure(2), nyquist(G); set(gca,'Ylim',[-1.5,1.5]);  
figure(3), nichols(G); set(gca,'Xlim',[-400,-100],'Ylim',[-100,30]);
```

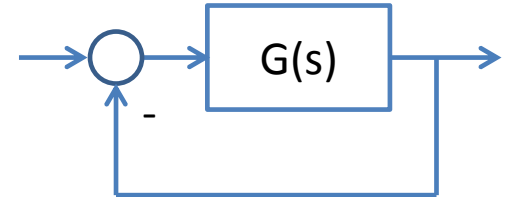
You may right-click on any of the figures and modify its properties (labels, limits, units, style, etc)



Frequency-Domain Analysis of Linear Systems

Stability analysis in frequency-domain

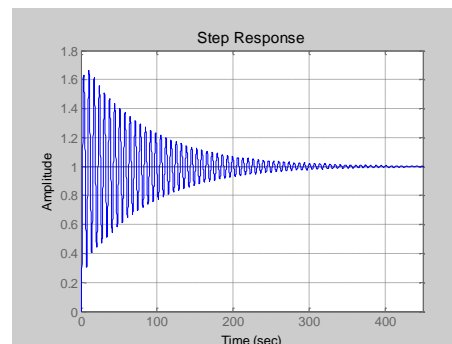
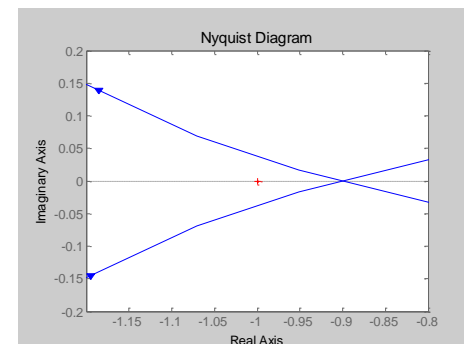
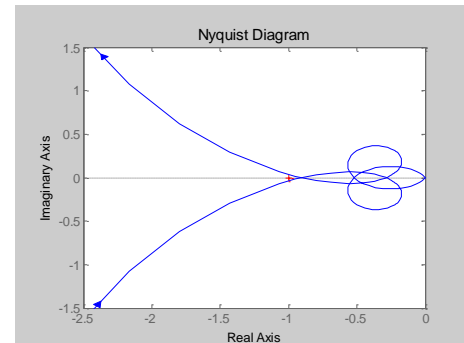
- The closed-loop system is stable if:
 - The Nyquist plot of $G(s)$ encircles (counter-clockwise) the point $(-1+0j)$ as many times as the number of RHP poles of $G(s)$



- Example: For $G(s) = \frac{2.7778(s^2+0.192s+1.92)}{s(s+1)^2(s^2+0.384s+2.56)}$ is the closed-loop system stable?

Matlab code:

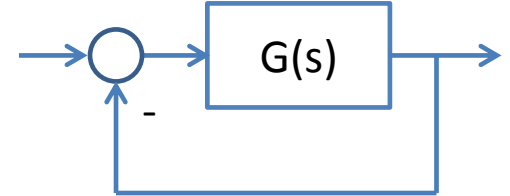
```
s=tf('s'); G=2.7778*(s^2+0.192*s+1.92)/(s*(s+1)^2*(s^2+0.384*s+2.56));  
figure(1), nyquist(G); axis([-2.5,0,-1.5,1.5]); grid;  
figure(2), nyquist(G); axis([-1.2,-0.8,-0.2,0.2]); grid;  
figure(3), step(feedback(G,1,-1));
```



Stable, but too oscillatory!

Frequency-Domain Analysis of Linear Systems

Gain and phase margins of a system



- Gain margin: $G_m = \frac{1}{A(\omega_{cg})}$
 - $A(\omega_{cg})$ and ω_{cg} are the magnitude and frequency where the Nyquist plot intersects the negative real-axis
 - The smaller the gain-margin, the faster the closed-loop system response
 - If $G_m < 1$, the closed-loop system is unstable
- Phase margin: $P_m = \phi(\omega_{cp}) + 180$
 - $\phi(\omega_{cp})$ and ω_{cp} are the phase and frequency where the Nyquist plot intersects the unit circle
 - The larger the phase-margin, the less overshoot in closed-loop system response
 - If $P_m < 0$, the closed-loop system is unstable
- Example: For $G(s) = \frac{2.7778(s^2+0.192s+1.92)}{s(s+1)^2(s^2+0.384s+2.56)}$ find the closed-loop system margins

Matlab code:

```
s=tf('s'); G=2.7778*(s^2+0.192*s+1.92)/(s*(s+1)^2*(s^2+0.384*s+2.56));
```

```
[gm,pm,wg,wp]=margin(G);
```

⇒ gm = 1.1050; pm = 2.0985; wg = 0.9621; wp = 0.9261 (closed-loop is stable)

Introduction to Model Reduction Techniques

- Pade approximation to delay terms

- A delay term $e^{-\tau s}$ can be approximated as rational transfer functions $\frac{n_p(s)}{d_p(s)}$

Matlab Command: `pade()`

- Syntax: `[np,dp]=pade(τ,n)`; where n is the order of the Pade approximation

- Example: Find the Pade approximation of a pure delay $G(s) = e^{-s}$

Matlab code:

```
tau=1;
```

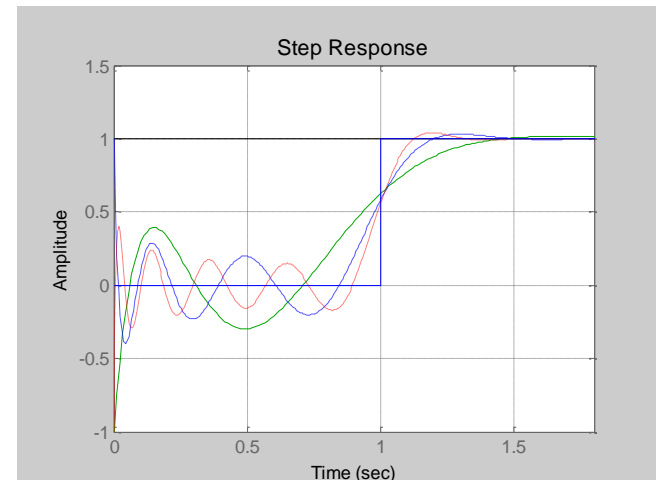
```
[n1,d1]=pade(tau,3); G1=tf(n1,d1); % 3rd-order Pade approx
```

```
[n2,d2]=pade(tau,6); G2=tf(n2,d2); % 5th-order Pade approx
```

```
[n3,d3]=pade(tau,9); G3=tf(n3,d3); % 7th-order Pade approx
```

```
step(G1,'-g',G2,'-.b',G3,':r'); % Compare plots
```

```
line([0,1,1+eps,3],[0,0,1,1]); % The step plot
```



Introduction to Model Reduction Techniques

State space model reduction

- Balanced realization method:
 - Given a system model, find its unique balanced realization, where the controllability and observability Gramians are equal ($W = W_c = W_o$)
 - Partition the balanced realization and eliminate the states with small Gramians

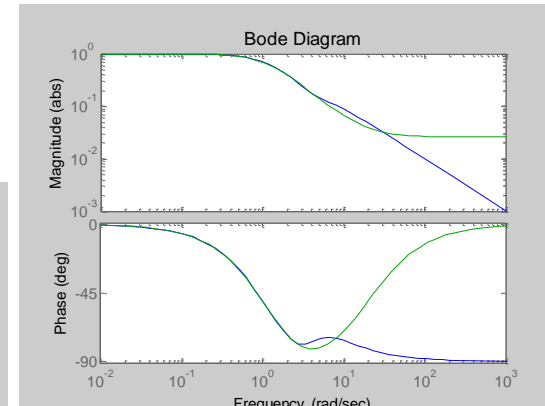
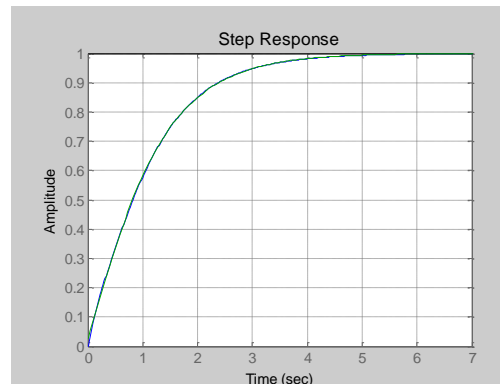
Matlab Command: **balreal()**, **modred()**

- Example: For $G(s) = \frac{s^3 + 7s^2 + 24s + 24}{s^4 + 10s^3 + 35s^2 + 50s + 24}$ find a 2nd-order approximate model

Matlab code:

```
num=[1,7,24,24]; den=[1,10,35,50,24];  
G=tf(num,den); [Gb,g,T]=balreal(ss(G));  
Gr=modred(Gb,[3,4]); zpk(Gr),  
figure(1), bode(G,Gr), grid;  
figure(2), step(G,Gr);
```

$$G_r(s) = \frac{0.025974(s+22.36)(s+4.307)}{(s+1.078)(s+2.319)}$$



Introduction to Model Reduction Techniques

State space model reduction

- Schur's balanced realization truncation method:
 - Similar to modred(), but can handle unstable systems
- Optimal Hankel norm approximation
 - Based on Hankel norm optimization technique

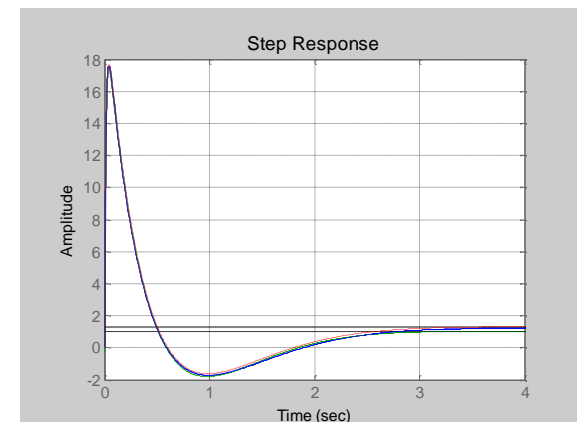
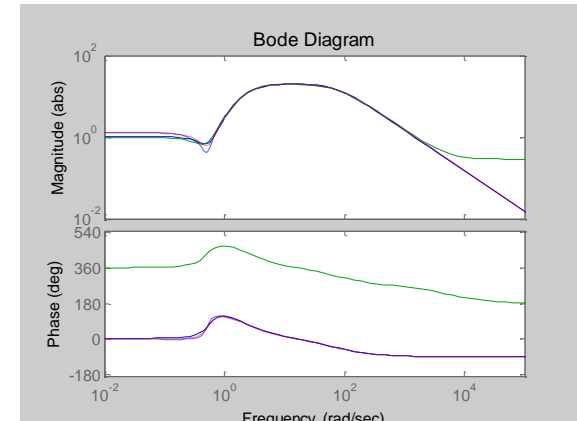
Matlab Commands: **schmr()**, **ohklmr()**

- Example: Find a 3rd-order approximate model for

$$G(s) = \frac{68.6131s^5 + 80.3787s^4 + 67.087s^3 + 29.9339s^2 + 8.8818s + 1}{0.0462s^6 + 3.5338s^5 + 16.5609s^4 + 28.4472s^3 + 21.7611s^2 + 7.6194s + 1}$$

Matlab code:

```
num=[68.6131,80.3787,67.087,29.9339,8.8818,1];
den=[0.0462,3.5338,16.5609,28.4472,21.7611,7.6194,1];
G=ss(tf(num,den)); [Gb,g,T]=balreal(G);
Gr=modred(Gb,[4,5,6]); zpk(Gr),
Gs=schmr(G,1,3); zpk(Gs),
Gh=ohklmr(G,1,3); zpk(Gh),
figure(1), bode(G,Gr, '-g',Gs,'-.b',Gh,':r'); grid;
figure(2), step(G,Gr, '-g',Gs,'-.b',Gh,':r');
```



$$G_r(s) = \frac{-0.28747(s-5398)(s^2+0.3576s+0.2519)}{(s+76.48)(s^2+3.853s+5.1111)}$$

$$G_s(s) = \frac{1485.3076(s^2+0.1789s+0.2601)}{(s+71.64)(s^2+3.881s+4.188)}$$

$$G_h(s) = \frac{1527.8048(s^2+0.2764s+0.2892)}{(s+73.93)(s^2+3.855s+4.585)}$$