

Multivariable Control: **Placement of Eigenvectors**

Prof. Guy O. Beale
Electrical and Computer Engineering Department
George Mason University
Fairfax, Virginia

Presentation Outline

- ◆ Problem Statement
- ◆ Motivating Example
- ◆ Why Eigenvectors Are Important
- ◆ How Eigenvectors Can Be Found
- ◆ Which Eigenvectors Might Be Selected
- ◆ Example Continued
- ◆ Conclusions

Problem Statement

- ◆ Given either state-space description for a system:

$$\dot{x}(t) = Ax(t) + Bu(t), \quad u(t) = -Lx(t)$$

$$x(k+1) = \Phi x(k) + \Gamma u(k), \quad u(k) = -Lx(k)$$

- ◆ Choose feedback gain L so that:
 - Eigenvalues are placed at specified locations;

$$|\lambda I - A + BL| = 0 \quad \text{or} \quad |\lambda I - \Phi + \Gamma L| = 0$$

- Good performance is achieved.

Problem Statement

- ◆ For a single-input system, L is unique for a given set of eigenvalues.
- ◆ For a multi-input system, L is not unique for a given set of eigenvalues.
- ◆ For the same eigenvalues, different L matrices will produce:
 - Different eigenvectors;
 - Different performance.

Motivating Example

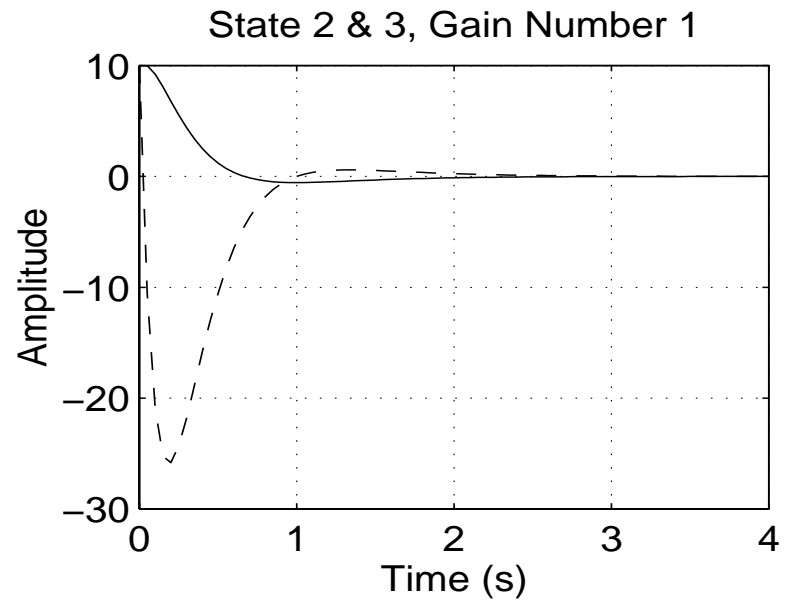
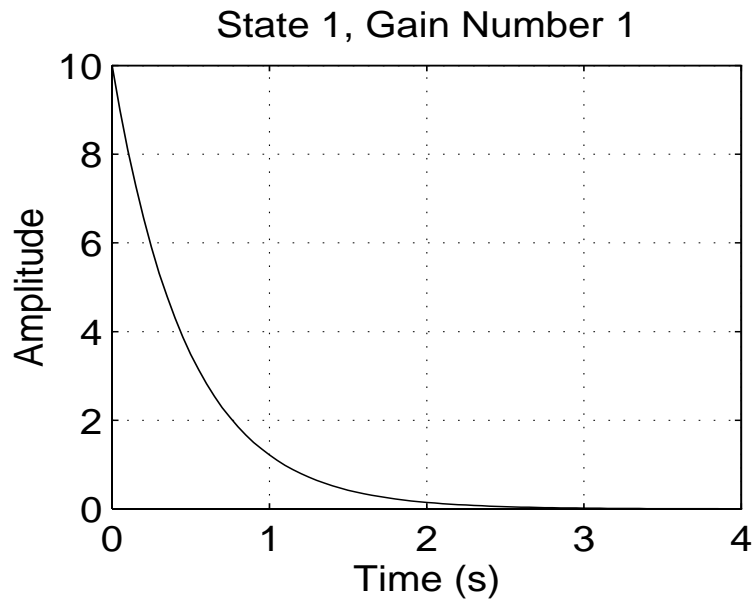
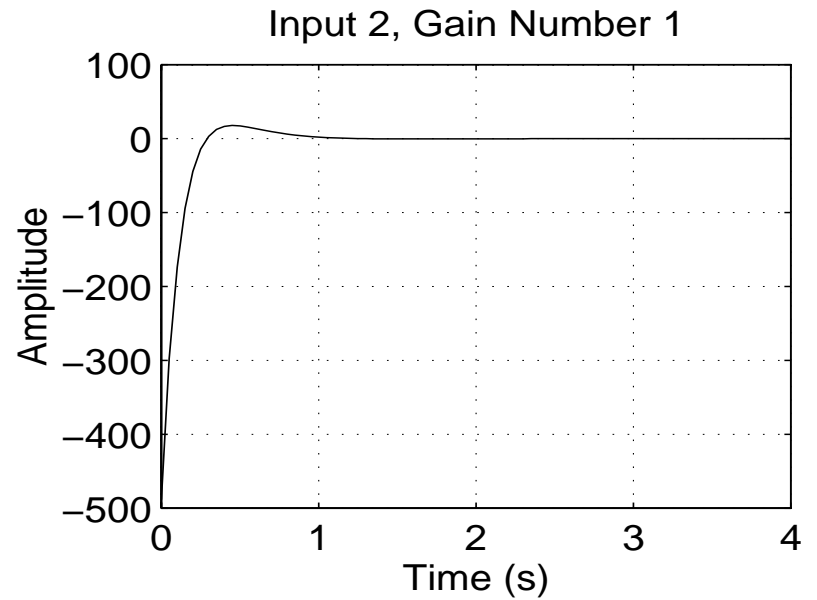
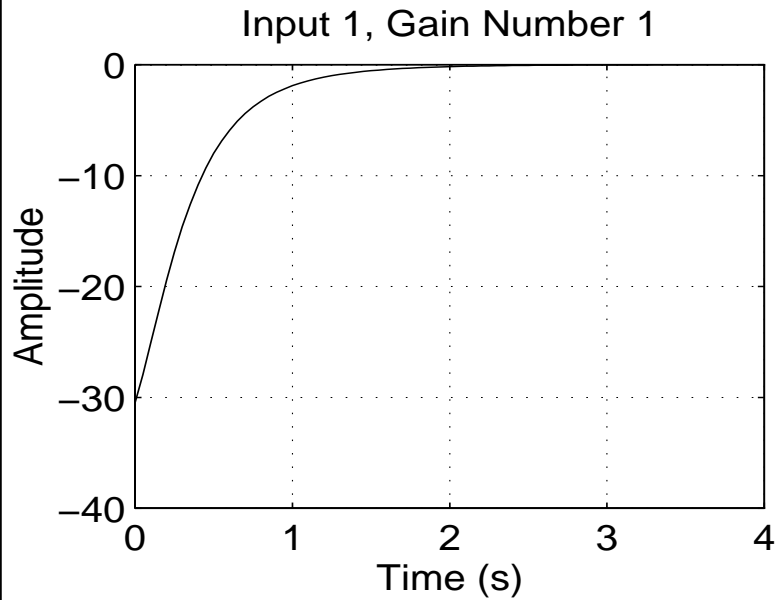
- ◆ Discretized system model with $T_{\text{samp}} = 0.05$ seconds:

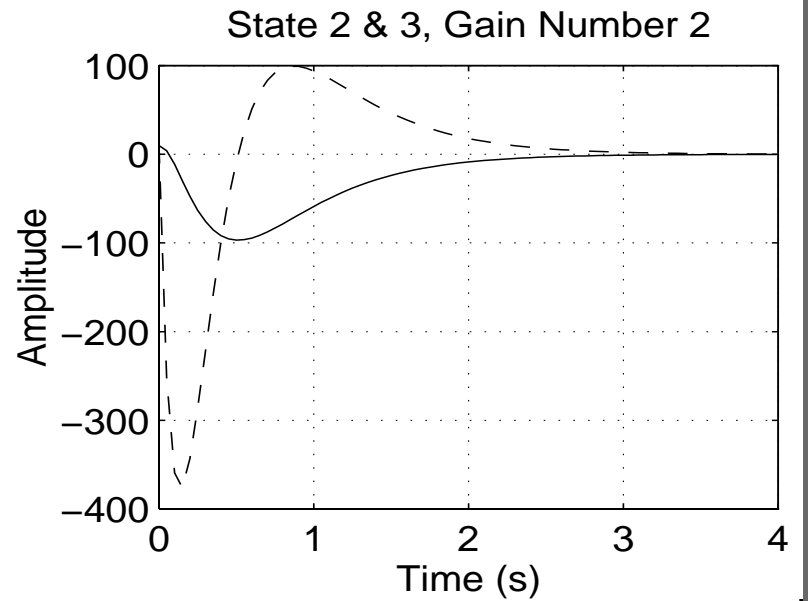
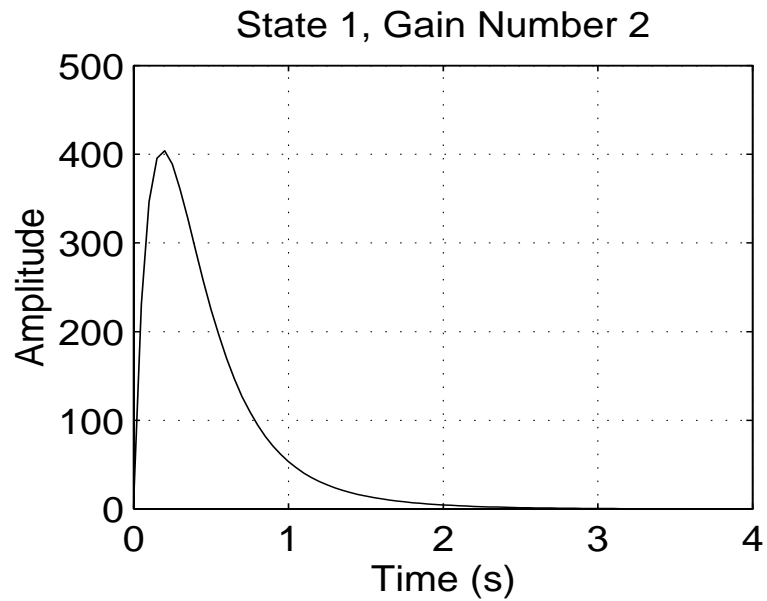
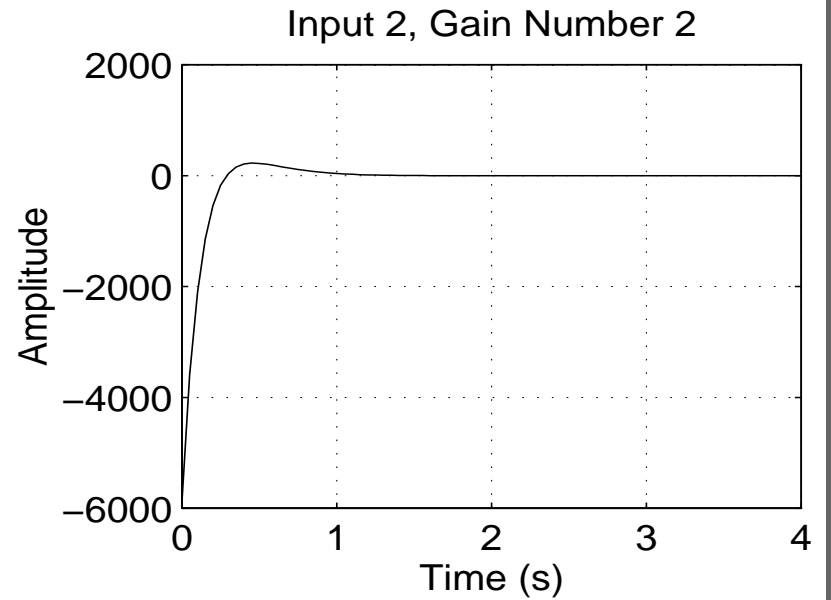
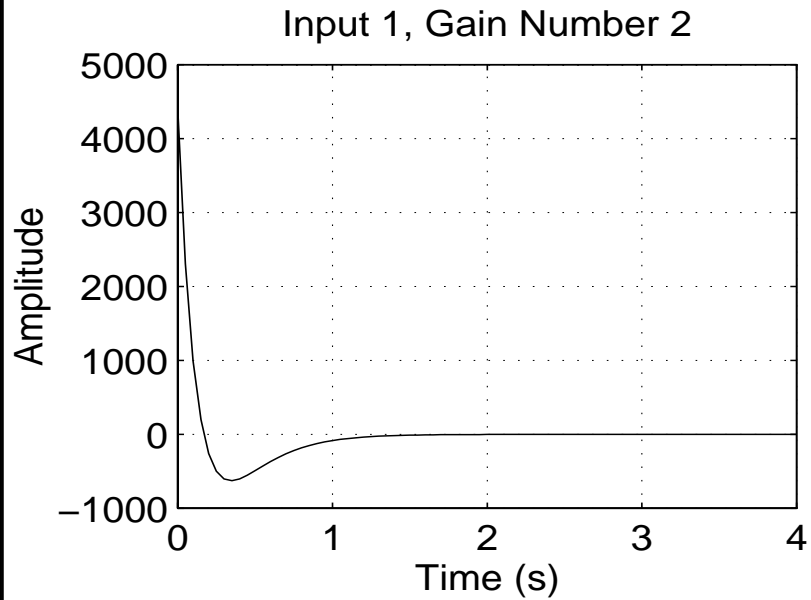
$$\Phi = \begin{bmatrix} 1.0010 & 0.0501 & 0.0012 \\ 0.0049 & 1.0050 & 0.0489 \\ 0.1954 & 0.2003 & 0.9562 \end{bmatrix}, \quad \Gamma = \begin{bmatrix} 0.0500 & 0.0000 \\ 0.0001 & 0.0012 \\ 0.0049 & 0.0489 \end{bmatrix}, \quad \lambda_{OL} = \begin{bmatrix} 0.9524 \\ 0.9045 \\ 1.1054 \end{bmatrix}$$

$$x(0) = [10 \quad 10 \quad 10]^T$$

- ◆ Desired closed-loop eigenvalues are: $\lambda_{CL} = 0.9, 0.8, 0.7$

$$L_1 = \begin{bmatrix} 2.020 & 1.002 & 0.024 \\ 11.525 & 28.568 & 8.726 \end{bmatrix}, \quad L_2 = \begin{bmatrix} -105.74 & -243.56 & -91.52 \\ 142.48 & 323.76 & 121.34 \end{bmatrix}$$





Motivating Example

- ◆ Eigenvalues are identical with each gain matrix.
- ◆ Settling times are similar for the two sets of responses.
- ◆ Some signal shapes are similar (x_2, x_3, u_2).
- ◆ L_2 produces much larger signal amplitudes than L_1 for all variables.
- ◆ Quantitative comparison between the 2 responses:

$$J = \sum_{k=0}^{80} \left[x^T(k) Q x(k) + u^T(k) R u(k) \right]$$

$$Q = I_3, \quad R = I_2$$

$$J_1 = 379,450, \quad J_2 = 84,243,000$$

Eigenvector Importance

- ◆ If the following control law is used:

$$u = -Lx, \quad A_{CL} = A - BL \quad \text{or} \quad \Phi_{CL} = \Phi - \Gamma L$$

- ◆ The closed-loop state equations are:

$$x(k) = \Phi_{CL}^k x(0), \quad x(t) = e^{A_{CL}t} x(0)$$

- ◆ If the system has real and distinct closed-loop eigenvalues λ_i :

$$A_{CL} = T\Lambda T^{-1}, \quad \Lambda = \text{diag}[\lambda_1 \quad \lambda_2 \quad \cdots \quad \lambda_n]$$
$$\Phi_{CL} = T\Lambda T^{-1}$$

Eigenvector Importance

- ◆ Right eigenvectors for the system are:

$$T = [v_1 \ v_2 \ \cdots \ v_n], \quad [\lambda_i I - A_{CL}]v_i = 0 \quad \text{or} \quad [\lambda_i I - \Phi_{CL}]v_i = 0$$

- ◆ Left eigenvectors for the system are:

$$T^{-1} = \begin{bmatrix} w_1^T \\ w_2^T \\ \vdots \\ w_n^T \end{bmatrix}, \quad w_i^T [\lambda_i I - A_{CL}] = 0 \quad \text{or} \quad w_i^T [\lambda_i I - \Phi_{CL}] = 0$$

- ◆ The closed-loop system matrices are:

$$\Phi_{CL}^k = T \Lambda^k T^{-1}, \quad e^{A_{CL}t} = T e^{\Lambda t} T^{-1}$$

Eigenvector Importance

- ◆ Solutions to the closed-loop state equations are:

$$x(k) = \sum_{i=1}^n v_i \lambda_i^k [w_i^T x(0)], \quad x(t) = \sum_{i=1}^n v_i e^{\lambda_i t} [w_i^T x(0)]$$

- ◆ The eigenvalue λ_i defines the mode of the system.
- ◆ Product of left eigenvector and initial conditions $w_i^T x(0)$ determines the amplitude of the mode.
- ◆ Right eigenvector v_i determines the shape of the mode.

Finding Eigenvectors

- ◆ Each closed-loop right eigenvector satisfies

$$[\lambda_i I - A + BL]v_i = 0 \quad \text{or} \quad [\lambda_i I - \Phi + \Gamma L]v_i = 0$$

- Eigenvector v_i must be in the null space of $\lambda_i I - A + BL$ or $\lambda_i I - \Phi + \Gamma L$
- Eigenvectors v_i and v_j must be linearly independent for $i \neq j$.
- ◆ But there is a problem: feedback matrix L is unknown!
- ◆ Defining $q_i = Lv_i$ yields

$$[\lambda_i I - A \quad \vdots \quad B] \begin{bmatrix} v_i \\ \dots \\ q_i \end{bmatrix} = 0 \quad \text{or} \quad [\lambda_i I - \Phi \quad \vdots \quad \Gamma] \begin{bmatrix} v_i \\ \dots \\ q_i \end{bmatrix} = 0$$

$$v_i = n \times 1, \quad q_i = m \times 1$$

Finding Eigenvectors

- ◆ For each λ_i , the dimension of the null space = m = number of control inputs.
- ◆ Singular value decomposition (SVD) can be used to find a basis for the null space.
- ◆ For the n closed-loop eigenvalues λ_i

$$[q_1 \quad \vdots \quad q_2 \quad \vdots \quad \cdots \quad \vdots \quad q_n] = L[v_1 \quad \vdots \quad v_2 \quad \vdots \quad \cdots \quad \vdots \quad v_n]$$

$$Q = LT \Rightarrow L = QT^{-1}$$

- ◆ That matrix L places eigenvalues and eigenvectors at values specified by λ_i and v_i , respectively.

Finding Eigenvectors

- ◆ Singular value decomposition: for each λ_i

$$S_i = [\lambda_i I - A \ : \ B] = U_i \Sigma_i (V_i)^T \quad \text{or} \quad S_i = [\lambda_i I - \Phi \ : \ \Gamma] = U_i \Sigma_i (V_i)^T$$

$$S_i = n \times (n + m), \quad U_i = n \times n$$

$$\Sigma_i = n \times (n + m), \quad V_i = (n + m) \times (n + m)$$

- ◆ Last m columns of V_i are a basis for null space of S_i

$$\begin{bmatrix} v_i \\ \cdots \\ q_i \end{bmatrix} = \sum_{k=1}^m \alpha_{i,k} V_{i,n+k}$$

- ◆ v_i and v_j must be linearly independent for $i \neq j$

Finding Eigenvectors

Procedure Summarized

- ◆ For each desired closed-loop eigenvalue λ_i :
 - Form the matrix S_i ;
 - Form the SVD of S_i to get V_i ;
 - Choose weighting factors $\alpha_{i,k}$ to get v_i and q_i .

- ◆ Compute the feedback gain matrix $L = QT^{-1}$

Selecting Eigenvectors

IF T (matrix of eigenvectors) is made to be diagonal by choice of v_i :

THEN

- ◆ A_{CL} or Φ_{CL} is diagonal;
- ◆ Response of each state variable is controlled by its own eigenvalue;
- ◆ Closed-loop system is diagonalized, even though the open-loop system was not diagonal.

Selecting Eigenvectors

- ◆ In general, T cannot be made diagonal.
- ◆ Alternatives:
 - Choose the v_i so that each eigenvector has as many entries as possible = 0 in off-diagonal positions in T , and linear independence is maintained.
 - Choose the v_i so that the eigenvectors are as orthogonal as possible, and linear independence is maintained \Rightarrow MATLAB[®] “place” function.
- ◆ Neither of these approaches is best in every application.

Example Continued

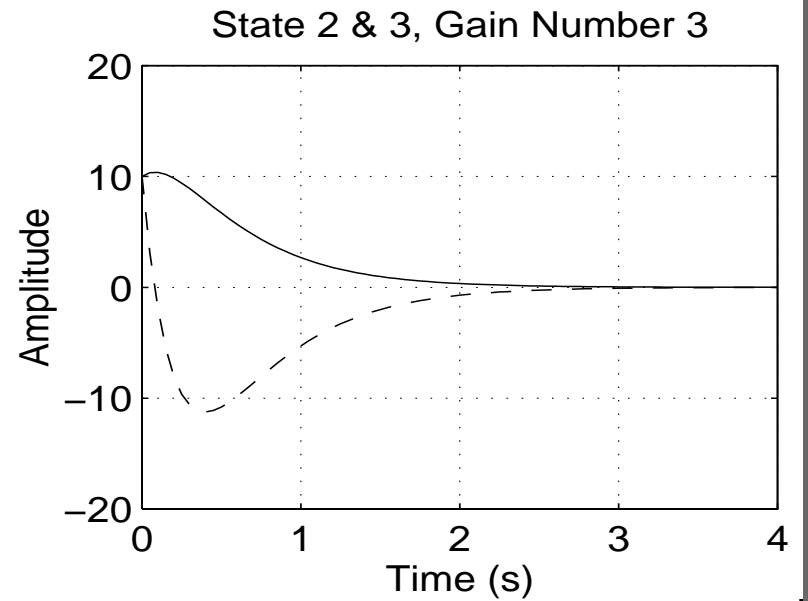
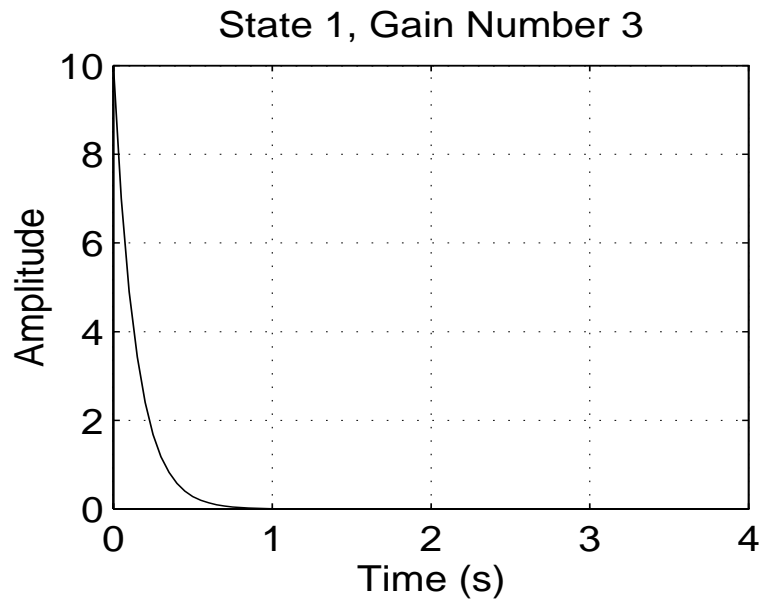
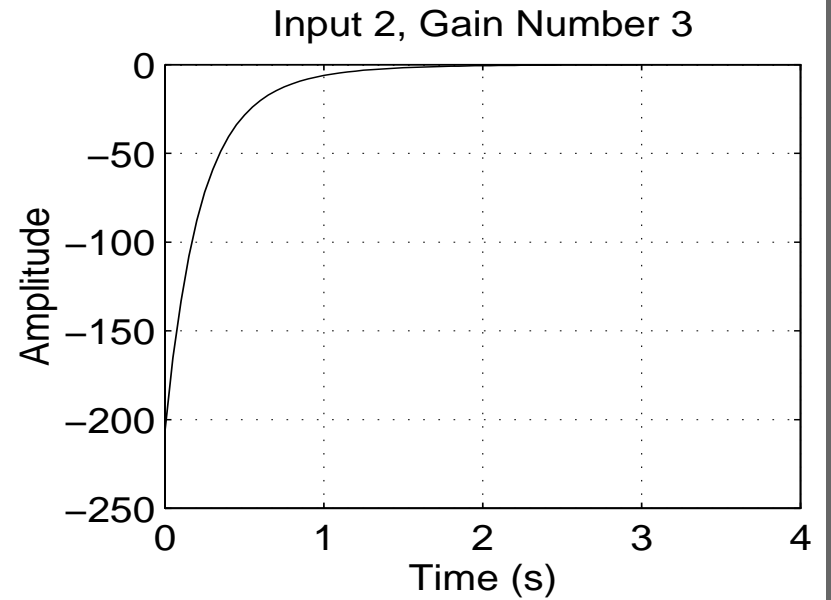
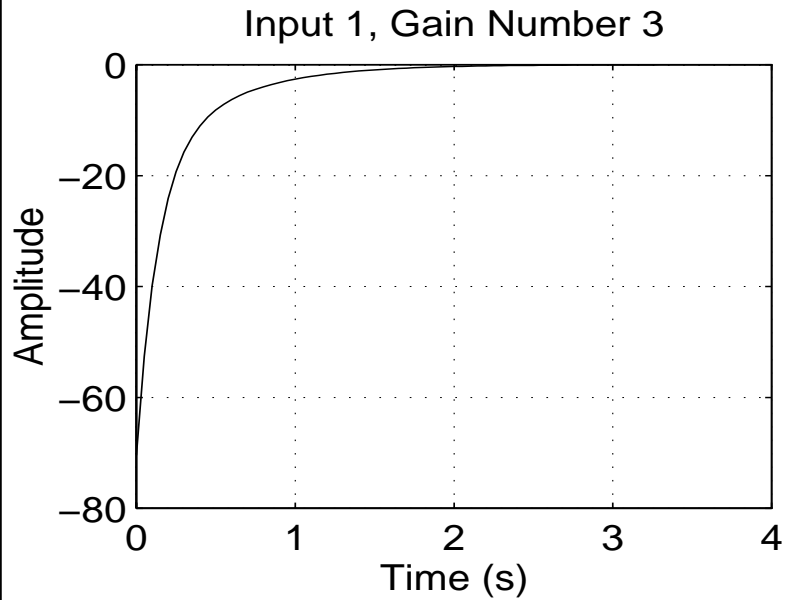
$$V_{1,4-5} = \begin{bmatrix} -0.284 & 0.000 \\ 0.255 & 0.336 \\ -0.536 & -0.709 \\ -0.332 & 0.320 \\ -0.676 & 0.531 \end{bmatrix}, \quad V_{2,4-5} = \begin{bmatrix} -0.216 & 0.000 \\ -0.057 & -0.089 \\ 0.253 & 0.397 \\ -0.920 & -0.080 \\ -0.197 & 0.910 \end{bmatrix}, \quad V_{3,4-5} = \begin{bmatrix} -0.161 & 0.000 \\ -0.015 & -0.030 \\ 0.106 & 0.209 \\ -0.980 & -0.025 \\ -0.048 & 0.977 \end{bmatrix}$$

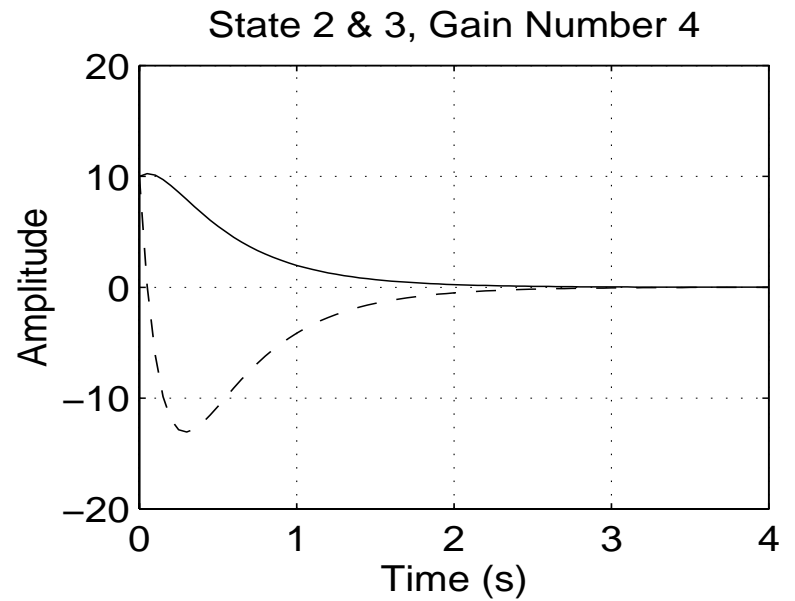
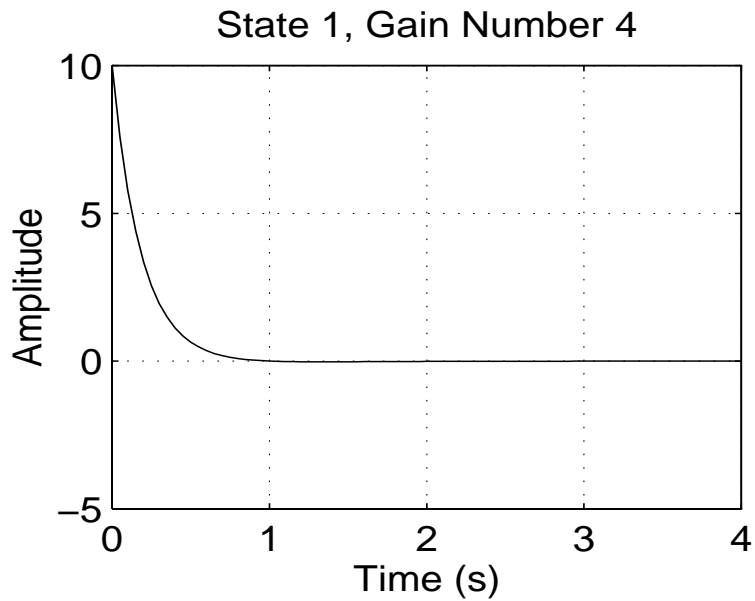
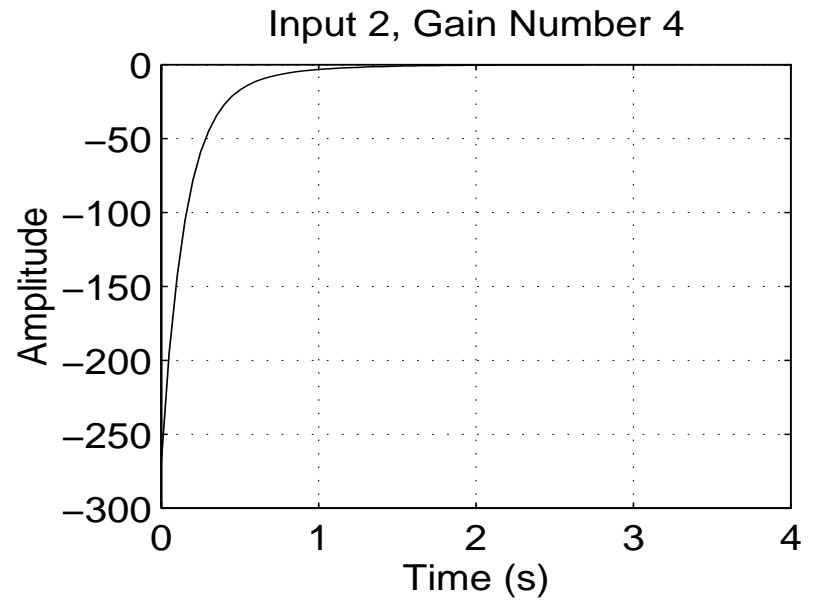
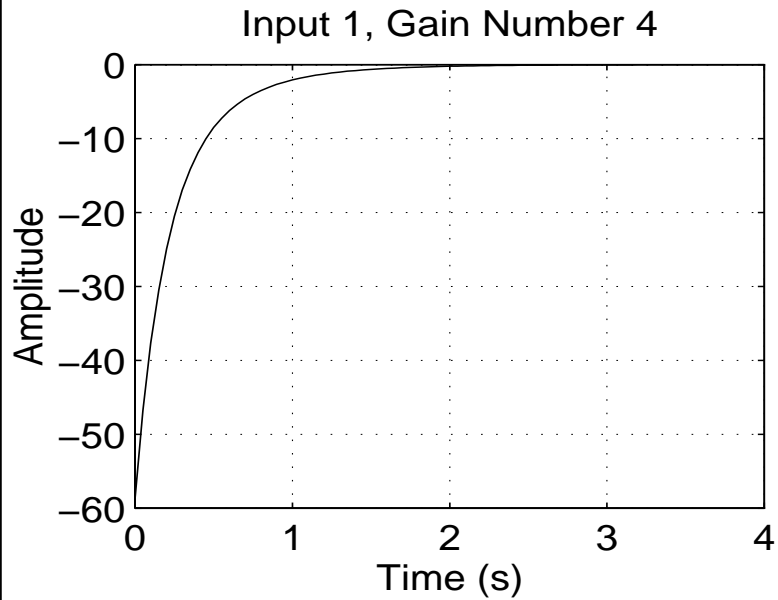
$$\lambda_1 = 0.9$$

$$\lambda_2 = 0.8$$

$$\lambda_3 = 0.7$$

- L_1 : 1st column for 0.9; 2nd column for 0.8 and 0.7.
- L_2 : sum of 1st and 2nd columns for 0.9, 0.8, and 0.7.
- L_3 : 2nd column for 0.9 and 0.8; $\alpha_{3,k}$ chosen for 0.7 to make the entry in row 3 equal to 0.
- L_4 : uses MATLAB[®] “place” function.





Example Continued

$$\Phi_{CL_1} = \begin{bmatrix} 0.9000 & 0.0000 & 0.0000 \\ -0.0091 & 0.9706 & 0.0384 \\ -0.3781 & -1.2016 & 0.5294 \end{bmatrix}, \quad \begin{aligned} J_1 &= 379,450 \\ \|L_1\|_2 &= 32.058 \end{aligned}$$

$$\Phi_{CL_2} = \begin{bmatrix} 6.2882 & 12.228 & 4.5771 \\ -0.1555 & 0.6408 & -0.0876 \\ -6.2539 & -14.438 & -4.5290 \end{bmatrix}, \quad \begin{aligned} J_2 &= 84,243,000 \\ \|L_2\|_2 &= 467.68 \end{aligned}$$

$$\Phi_{CL_3} = \begin{bmatrix} 0.7000 & 0.0000 & 0.0000 \\ 0.0002 & 0.9903 & 0.0429 \\ -0.0003 & -0.4008 & 0.7097 \end{bmatrix}, \quad \begin{aligned} J_3 &= 138,900 \\ \|L_3\|_2 &= 13.875 \end{aligned}$$

$$\Phi_{CL_4} = \begin{bmatrix} 0.7997 & -0.0287 & -0.0116 \\ 0.0001 & 0.9854 & 0.0405 \\ -0.0036 & -0.6006 & 0.6150 \end{bmatrix}, \quad \begin{aligned} J_4 &= 169,400 \\ \|L_4\|_2 &= 18.181 \end{aligned}$$

Example Continued

$$L_1 = \begin{bmatrix} 2.0200 & 1.0220 & 0.0240 \\ 11.525 & 28.568 & 8.7260 \end{bmatrix}$$

$$L_2 = \begin{bmatrix} -105.74 & -243.56 & -91.518 \\ 142.48 & 323.76 & 121.34 \end{bmatrix}$$

$$L_3 = \begin{bmatrix} 6.0200 & 1.0220 & 0.0240 \\ 3.3989 & 12.192 & 5.0379 \end{bmatrix}$$

$$L_4 = \begin{bmatrix} 4.0269 & 1.5757 & 0.2561 \\ 3.6665 & 16.219 & 6.9526 \end{bmatrix}$$

Comparison with Optimal LQR

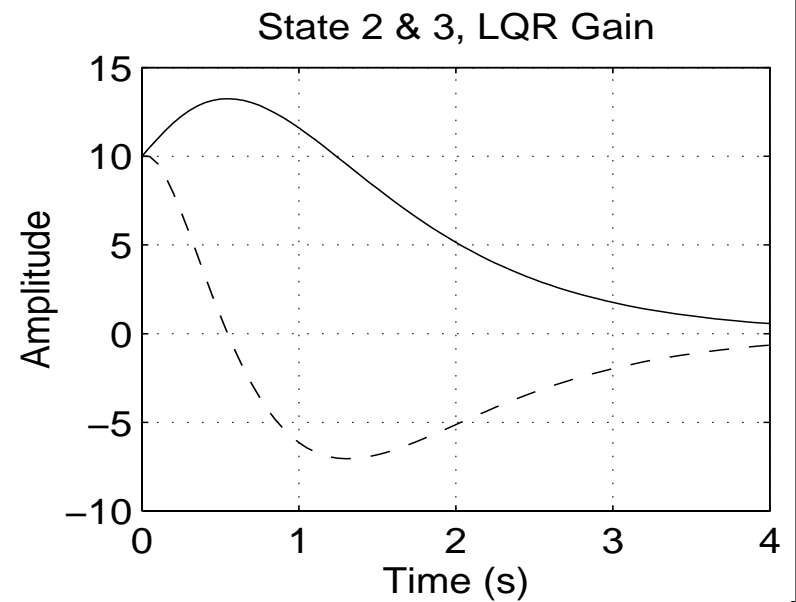
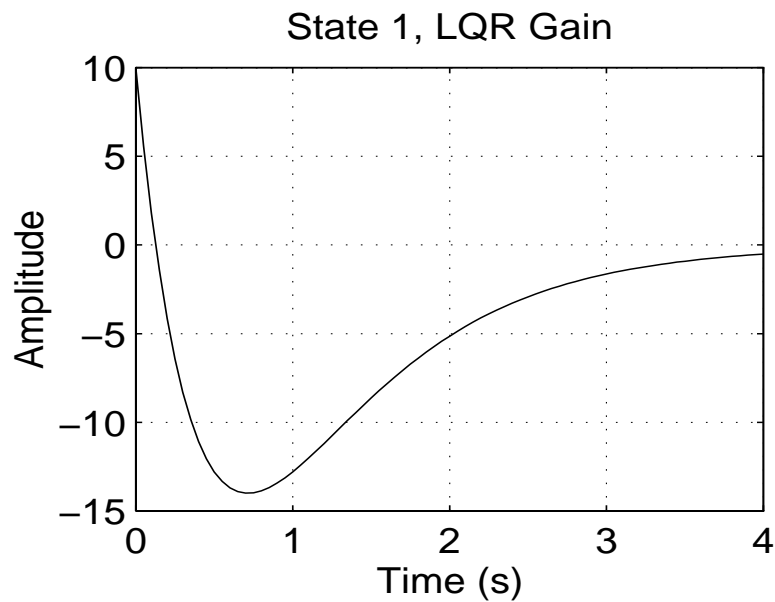
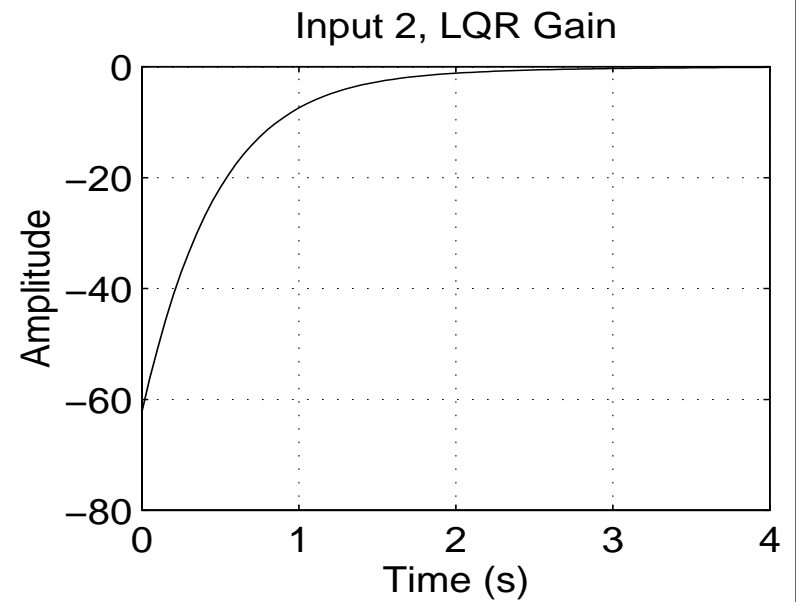
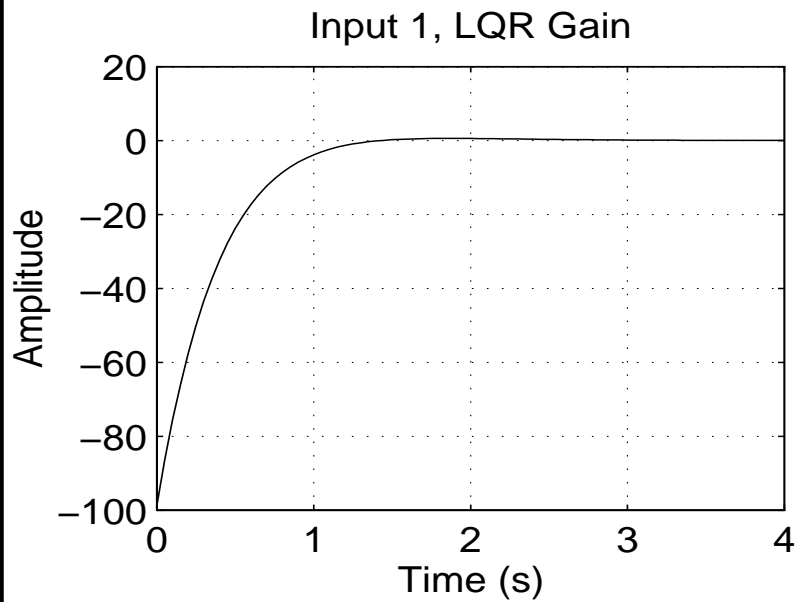
- ◆ Weighting Matrices: $Q = I_3, R = I_2$

$$\Phi_{CL_LQR} = \begin{bmatrix} 0.8352 & -0.1938 & -0.0831 \\ 0.0025 & 1.0006 & 0.0472 \\ 0.0945 & 0.0188 & 0.8858 \end{bmatrix}, \quad \begin{aligned} J_{LQR} &= 71,879 \\ \|L_{LQR}\|_2 &= 7.2451 \end{aligned}$$

$$L_{LQR} = \begin{bmatrix} 3.3158 & 4.8785 & 1.6856 \\ 1.7306 & 3.2223 & 1.2713 \end{bmatrix}$$

- ◆ Closed-loop eigenvalues:

$$\lambda_{CL_LQR} = [0.9448 \quad 0.8884 + j0.0363 \quad 0.8884 - j0.0363]$$



Summary of Results

Gain	J	J/J_{LQR}	$\ L\ _2$	$\ L\ _2/\ L_{LQR}\ _2$
L_1	379,450	5.28	32.058	4.42
L_2	84,243,000	1,172	467.68	64.6
L_3	138,900	1.93	13.875	1.92
L_4	169,400	2.36	18.181	2.51
L_{LQR}	71,879	1	7.2451	1

- ◆ Wide variation in performance is seen, even though the eigenvalues are fixed in location for $L_1 - L_4$.
- ◆ The optimal LQR solution does not guarantee the location of eigenvalues or “satisfactory” performance, only stability and optimality.

Conclusions

- ◆ Eigenvectors have a dramatic effect on closed-loop performance.
- ◆ For multi-input systems, eigenvectors should be chosen as well as eigenvalues, as far as possible.
- ◆ Eigenvectors should be as linearly independent as possible.
- ◆ The 2-norm of the feedback gain matrix seems to be a good measure of this independence.