

Lecture 14

Robust Stabilization of Time-Delay Systems

We will look at the class of uncertain time delay affine-controlled systems where a delay is accepted in state variables as well as the control action separately. It will be demonstrated that the Attractive Ellipsoid Method application allows for the creation of a feedback that enables any state trajectory of the controlled system to be converged to an ellipsoid, whose "size" is determined by the parameters of the applied feedback. Finally, we provide a numerical approach for calculating these parameters that yields the "smallest" zone-convergence for all conceivable controlled trajectories. An overview of stability conditions in terms of the Lyapunov matrix for time-delay systems may be found in [37].

14.1 Affine systems with a delay in state variables

14.1.1 System description and problem formulation

Let us consider the time delay control system of the form

$$\dot{x}(t) = f(x(t), x(t - \tau), t) + Bu(t), \quad (14.1)$$

or, in quasi-linear format,

$$\boxed{\dot{x}(t) = Ax(t) + A_1x(t - \tau) + Bu(t) + \xi(x(t), x(t - \tau), t)} \quad (14.2)$$

with the initial conditions

$$x(s) = \varphi(s), s \in [-\tau, 0] \quad (14.3)$$

where

$x \in \mathbb{R}^n$ is the vector of the system state,

$A, A_1 \in \mathbb{R}^{n \times n}$ are the system matrices,

$u \in \mathbb{R}^m$ is the vector of control inputs,

$B \in \mathbb{R}^{n \times m}$ is the matrix of the control gains,

the pair (A, B) is controllable,

$\tau > 0$ is the constant *state* time delay assumed to be known,

the vector-valued function $\xi(x(t), x(t - \tau), t)$, describing the unknown part of the model, is defined as

$$\xi(x(t), x(t - \tau), t) := f(x(t), x(t - \tau), t) - Ax(t) + A_1x(t - \tau),$$

which assumed to be bounded as

$$\boxed{\begin{aligned} \|\xi(x(t), x(t - \tau), t)\|^2 &\leq \\ c_0 + x^\top(t)Q_x x(t) + x^\top(t - \tau)Q_\tau x(t - \tau) \end{aligned}} \quad (14.4)$$

with the positive definite symmetric matrices $Q_x, Q_h \in \mathbb{R}^{n \times n}$ (supposed to be given),

the matrix B has a full $\text{rank}(B) = m \leq n$.

Problem 14.1 *We need to design the control action $u \in \mathbb{R}^m$ as a linear feedback*

$$\boxed{\begin{aligned} u &= Kx(t) + K_\tau x(t - \tau), \\ K, K_\tau &\in \mathbb{R}^{m \times n}, \end{aligned}} \quad (14.5)$$

which stabilizes all possible trajectories $\{x(t)\}_{t \geq -\tau}$ in some bounded region, containing origin in the space \mathbb{R}^n , and to make this region as small as possible.

14.1.2 Lyapunov-Krasovskii's functional and stability analysis

Let us defined the Lyapunov-Krasovskii's functional as follows:

$$\left. \begin{aligned} V(x(t), t) &:= x^\top(t)Px(t) + \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s)P_1x(s) ds, \\ h > 0, \quad P, P_1 &\in \mathbb{R}^{n \times n} \text{ are positive definite.} \end{aligned} \right\} \quad (14.6)$$

and calculate its derivative over the trajectories of the system (14.2). We get

$$\begin{aligned} \dot{V}(x(t), t) &= 2x^\top(t)P\dot{x}(t) + x^\top(t)P_1x(t) \\ &\quad - e^{-h\tau} x^\top(t-\tau)P_1x(t-\tau) - h \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s)P_1x(s) ds \\ &= 2x^\top(t)P(Ax(t) + A_1x(t-\tau) \\ &\quad + B[Kx(t) + K_\tau x(t-\tau)] + \xi(x(t), x(t-\tau), t)) \\ &\quad + x^\top(t)P_1x(t) - e^{-h\tau} x^\top(t-\tau)P_1x(t-\tau) \\ &\quad - h \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s)P_1x(s) ds + 2x^\top(t)P\xi(x(t), x(t-\tau), t). \end{aligned}$$

This expression in new variables

$$z(t) = \begin{pmatrix} x(t) \\ x(t-\tau) \\ \xi(x(t), x(t-\tau), t) \end{pmatrix}$$

can be rewritten as

$$\left. \begin{aligned} z^\top(t) &\left[\begin{array}{ccc} P(A+BK) + (A+BK)^\top P + P_1 & P(A_1+BK_\tau) & P \\ (A_1+BK_\tau)^\top P & -e^{-h\tau}P_1 & 0_{n \times n} \\ P & 0_{n \times n} & -\varepsilon I_{n \times n} \end{array} \right] z(t) \\ &\quad - h \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s)P_1x(s) ds + \varepsilon \|\xi(x(t), x(t-\tau), t)\|^2. \end{aligned} \right\} \quad (14.7)$$

Using the upper bound property (14.4)

$$\|\xi(x(t), x(t-\tau), t)\|^2 \leq c_0 + x^\top(t) Q_x x(t) + x^\top(t-\tau) Q_\tau x(t-\tau)$$

we may represent (14.7) as differential inequality

$$\left. \begin{aligned} \dot{V} &\leq \varepsilon c_0 - h \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s) P_1 x(s) ds + \\ z^\top &\left[\begin{array}{cc} P(A+BK) + (A+BK)^\top P & P(A_1+BK_\tau) \\ + P_1 + \varepsilon Q_x + \alpha P & \\ (A_1+BK_\tau)^\top P & -e^{-h\tau} P_1 + \varepsilon Q_\tau \end{array} \right] z. \end{aligned} \right\} \quad (14.8)$$

Adding and subtracting the term αV in the right-hand side of (14.8) we get

$$\left. \begin{aligned} \dot{V} &\leq -\alpha V + \varepsilon c_0 - (h - \alpha) \int_{s=t-\tau}^t e^{h(s-t)} x^\top(s) P_1 x(s) ds \\ &\quad + z^\top(t) W_{\alpha, \varepsilon}(P, P_1, K, K_\tau) z(t) \end{aligned} \right\} \quad (14.9)$$

where

$$\left. \begin{aligned} W_{\alpha, \varepsilon, h}(P, P_1, K, K_\tau) = \\ \left[\begin{array}{cc} P(A+BK) + (A+BK)^\top P & P(A_1+BK_\tau) \\ + P_1 + \varepsilon Q_x + \alpha P & \\ (A_1+BK_\tau)^\top P & -e^{-h\tau} P_1 + \varepsilon Q_\tau \end{array} \right] \end{aligned} \right\} \quad (14.10)$$

Now we are ready to formulate the following theorem.

Theorem 14.1 *If for the given matrices $P > 0, P_1 > 0, K, K_\tau$ and positive constants α, ε, h we have*

$$\boxed{W_{\alpha, \varepsilon, h}(P, P_1, K, K_\tau) < 0, h \geq \alpha,} \quad (14.11)$$

then the following stabilization property holds:

$$\boxed{\limsup_{t \rightarrow \infty} V(x(t), t) \leq \frac{\varepsilon c_0}{\alpha}.} \quad (14.12)$$

Proof. It follows directly from (14.9), if the conditions (14.11) are fulfilled.

■

Corollary 14.1 *The attractive ellipsoid*

$$\mathcal{E}_0(P_{attr}) := \{x \in \mathbb{R}^{2n} : x^\top P_{attr} x < 1\}$$

for such system corresponds to the attractivity property

$$\left. \begin{array}{l} \limsup_{t \rightarrow \infty} x^\top(t) P_{attr} x(t) \leq 1, \\ \text{with} \\ P_{attr} = \frac{\alpha}{\varepsilon c_0} P. \end{array} \right\} \quad (14.13)$$

Proof. It follows from the inequality

$$x^\top(t) P x(t) \leq V(x(t), t)$$

and (14.12). ■

14.1.3 Optimal feedback parameters

Following the same optimization scheme as in the previous lectures, we may formulate the problem of feedback parameters optimization in the following way:

$$\left. \begin{array}{l} \text{tr} \left\{ \frac{\varepsilon}{\alpha} P^{-1} \right\} \rightarrow \inf_{P>0, P_1>0, K, K_\tau, \alpha>0, \varepsilon>0, h>0} \\ \text{subject to the constraints (14.11)} \end{array} \right\} \quad (14.14)$$

As usual, let us transform matrix $W_{\alpha, \varepsilon, h}(P, P_1, K, K_\tau)$ into

$$W_T = T^\top W_{\alpha, \varepsilon, h}(P, P_1, K, K_\tau) T$$

with $T = \begin{bmatrix} P^{-1} & 0_{n \times n} \\ 0_{n \times n} & P_1^{-1} \end{bmatrix}$, which leads to

$$\left. \begin{array}{l} W_T = \begin{bmatrix} P^{-1} W_{T,11} P^{-1} & P^{-1} W_{T,12} P_1^{-1} \\ P_1^{-1} W_{T,21} P^{-1} & P_1^{-1} W_{T,22} P_1^{-1} \end{bmatrix} = \\ \left[\begin{array}{cc} (A+BK)P^{-1} + P^{-1}(A+BK)^\top + P^{-1}P_1P^{-1} + \varepsilon P^{-1}Q_x P^{-1} + \alpha P^{-1} & (A_1+BK_\tau)P_1^{-1} \\ P_1^{-1}(A_1+BK_\tau)^\top & -e^{-h\tau}P_1^{-1} + \varepsilon P_1^{-1}Q_\tau P_1^{-1} \end{array} \right] \end{array} \right\} \quad (14.15)$$

The terms $P^{-1}P_1P^{-1}$, $P^{-1}Q_xP^{-1}$ and $P_1^{-1}Q_\tau P_1^{-1}$ in the diagonal blocks may be estimated as

$$P^{-1}P_1P^{-1} < H_1, \quad P^{-1}Q_xP^{-1} < H_1, \quad P_1^{-1}Q_\tau P_1^{-1} < H_3,$$

which by the Schur's complement equivalently expressed as the following matrix inequalities:

$$\begin{bmatrix} H_1 & P^{-1} \\ P^{-1} & P_1^{-1} \end{bmatrix} > 0, \quad \begin{bmatrix} H_2 & P^{-1} \\ P^{-1} & Q_x^{-1} \end{bmatrix} > 0, \quad \begin{bmatrix} H_3 & P^{-1} \\ P^{-1} & Q_\tau^{-1} \end{bmatrix} > 0 \quad (14.16)$$

In new variables

$$X = P^{-1}, \quad Y = KP^{-1}, \quad X_1 = P_1^{-1}, \quad Y_1 = K_\tau P_1^{-1} \quad (14.17)$$

the matrix inequalities $\bar{W}_T < 0$ and (14.16) looks as LMI's:

$$\bar{W}_T = \begin{bmatrix} AX + BY + \\ X^\top A^\top + Y^\top B^\top + & A_1 X_1 + BY_1 \\ H_1 + \varepsilon H_2 + \alpha X & \\ \\ X_1^\top A_1^\top + Y_1^\top B^\top & -e^{-h\tau} X_1 + \varepsilon H_3 \end{bmatrix} < 0 \quad (14.18)$$

and

$$\begin{bmatrix} H_1 & X \\ X & X_1 \end{bmatrix} > 0, \quad \begin{bmatrix} H_2 & X \\ X & Q_x^{-1} \end{bmatrix} > 0, \quad \begin{bmatrix} H_3 & X \\ X & Q_\tau^{-1} \end{bmatrix} > 0 \quad (14.19)$$

So, the optimization problem (14.14) in the new variables (14.17) may be formulated as a matrix optimization problem with LMI's constraints:

$$\left. \begin{array}{l} \text{tr} \left\{ \frac{\varepsilon}{\alpha} X \right\} \rightarrow \inf \\ \text{subject to the constraints (14.18)} \\ \text{and (14.19)} \end{array} \right\} \quad (14.20)$$

If X^* , Y^* , X_1^* and Y^* are the solutions of the optimization problem (14.20), then the optimal feedback parameters of the feedback stabilizer are

$$K^* = Y^* (X^*)^{-1}, \quad K_\tau^* = Y_1^* (X_1^*)^{-1}. \quad (14.21)$$

14.2 Affine systems with a delay in control actions

14.2.1 System description

Here we will consider the following time-delay system

$$\dot{x}(t) = f(x(t), t) + Bu(t - \tau) + \eta(x(t), t), \quad (14.22)$$

with external perturbations $\eta(x(t), t)$, or, in quasi-linear format,

$$\left. \begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t - \tau) + \xi(x(t), t), \\ A &\in R^{n \times n}, \quad B \in R^{n \times m}, \\ \xi(x(t), t) &:= f(x(t), t) - Ax(t) + \eta(x(t), t) \end{aligned} \right\} \quad (14.23)$$

with the initial conditions

$$x(s) = \varphi(s), \quad s \in [-\tau, 0].$$

Let us suppose that

$$\|\xi(x(t), t)\|^2 \leq c_0 + x^\top(t) Q_x x(t), \quad (14.24)$$

and additionally that the matrix A is Hurwitz (stable), the pair $\{A, B\}$ is controllable and the matrix B has a full rank m , i.e.,

$$\text{rank}(B) = m \leq n, \quad B^\top B > 0.$$

14.2.2 Prediction approach and unavoidable stabilization error

To stabilize the time delay control system (14.23) the *prediction approach* (see [18], [19], [20], and [21]) is used. The typical prediction equation for the system (14.23) has the form

$$y(t) = e^{A\tau} x(t) + \int_{-\tau}^0 e^{-sA} Bu(t + s) ds. \quad (14.25)$$

Obviously, knowing the control function $u(t)$ on the time interval $[t - \tau, t)$ is required to calculate the *prediction variable* $y(t)$. This information is expected to be acceptable and usable for control design. It is simple to verify that the prediction variable $y(t)$ obeys the following delay-free equation:

$$\dot{y}(t) = Ay(t) + Bu(t) + e^{A\tau} \xi(t). \quad (14.26)$$

According to *the predictor method* the stabilization of the original system (14.23) can be ensured by designing the stabilizing controller for the prediction system (14.26).

Lemma 14.1 *For the processes (14.23) and (14.25) the following relations holds:*

$$\boxed{x(t + \tau) = y(t) + \int_0^\tau e^{(\tau-s)A} \xi(t + s) ds.} \quad (14.27)$$

Proof. Using the formula for the general solution of the system (14.23) we obtain

$$x(t + \tau) = e^{A\tau} x(t) + \int_t^{t+\tau} e^{(t+\tau-s)A} Bu(s - \tau) ds +$$

$$\int_t^{t+\tau} e^{(t+\tau-s)A} \xi(s) ds = e^{A\tau} x(t) +$$

$$\int_{-\tau}^0 e^{-sA} Bu(t + s) ds + \int_0^\tau e^{(\tau-s)A} \xi(t + s) ds,$$

from which the equality (14.27) follows. ■

This lemma describes dependence of the original system state $x(t)$ on the predictor variable $y(t)$ and the uncertain term $\xi(x(t), t)$. The integral term in the right-hand side (14.27) obviously does not depend on control inputs and the predictor variables, but it is linear functional of $\xi(x(t), t)$. So, it defines an *unavoidable stabilization error* of the system (14.23),

According to this lemma the original system state $x(t)$ is dependent on the predictor variable $y(t)$ and the uncertain term $\xi(x(t), t)$. The integral term on the right-hand side evidently is independent of the predictor variables and control inputs, but it is a linear function of $\xi(x(t), t)$. As a result, it specifies an *unavoidable system stabilization error* (14.23)

$$\boxed{w_\tau(t) := \int_0^\tau e^{(\tau-s)A} \xi(t + s) ds,} \quad (14.28)$$

produced by the prediction technique, namely,

$$\boxed{x(t + \tau) = y(t) + w_\tau(t).}$$

Therefore, minimization of the attractive set for the original system (14.23) can be provided by the design of the appropriate controller for the prediction system 14.26.

Remark 14.1 *Unfortunately, under uncertainty presence (when $\xi(x(t), t) \neq 0$) the property $y(t) \rightarrow 0$ does not imply $x(t) \rightarrow 0$. Therefore, the time delay control system (14.23) may **only be practically stabilized** in some pre-determined zone (attractive set).*

So, now *our aim* is to develop the predictor-based control design scheme, which minimizes (in some sense) the attractive set of the system (14.23). For this purpose we will use the *Attractive Ellipsoid Method* (AEM) as it is described in the previous lectures.

Assume that for the system (14.26) the standard proportional feedback controller

$$\boxed{u(t) = Ky(t)} \quad (14.29)$$

is applied.

14.2.3 Attractive ellipsoid

Introduce the "*energetic function*"

$$\left. \begin{aligned} V_\tau(x(t+\tau), y(t)) &= x^\top(t+\tau) P_x x(t+\tau) + \\ &+ y^\top(t) P_y y(t) + \beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds, \\ 0 < P_x \in \mathbb{R}^{n \times n}, \quad 0 < P_y \in \mathbb{R}^{n \times n}, \quad \beta > 0, \quad h > 0. \end{aligned} \right\} \quad (14.30)$$

Theorem 14.2 *If for the system (14.23) there exist positive definite matrices P_x , P_y and positive constants α , ε , β , h such that*

$$\left[\begin{array}{cccc} 0 > W_\tau = & & & \\ \begin{bmatrix} P_x A + A^\top P_x + \\ \alpha I_{n \times n} + \\ (1 + \beta e^{h\tau}) Q_x \end{bmatrix} & P_x B K & P_x & 0_{n \times n} \\ (P_x B K)^\top & P_y (A + B K) + \\ & (A + B K)^\top P_y & 0_{n \times n} & P_y e^{A\tau} \\ & + \alpha I_{n \times n} & & \\ P_x & 0_{n \times n} & -\varepsilon I_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & e^{A^\top \tau} P_y & 0_{n \times n} & -\varepsilon I_{n \times n} \end{array} \right] \quad (14.31)$$

and

$$\boxed{\varepsilon \leq \beta, \alpha \leq h,} \quad (14.32)$$

then the property

$$\boxed{\limsup_{t \rightarrow \infty} V_\tau(x(t+\tau), y(t)) \leq 2c_0 \frac{\varepsilon}{\alpha}} \quad (14.33)$$

is guaranteed.

Proof. For $V_\tau(x(t+\tau), y(t))$ we have

$$\left. \begin{aligned} V_\tau(x(t+\tau), y(t)) &= x^\top(t+\tau) P_x x(t+\tau) + \\ & y^\top(t) P_y y(t) + \beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds = \\ & \left(\begin{array}{c} x(t+\tau) \\ y(t) \end{array} \right)^\top \left[\begin{array}{cc} P_x & 0_{n \times n} \\ 0_{n \times n} & P_y \end{array} \right] \left(\begin{array}{c} x(t+\tau) \\ y(t) \end{array} \right) + \\ & \beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds, \end{aligned} \right\} \quad (14.34)$$

and calculate its derivative:

$$\begin{aligned} \dot{V}_\tau(x(t+\tau), y(t)) &= \\ & 2 \left(\begin{array}{c} x(t+\tau) \\ y(t) \end{array} \right)^\top \left[\begin{array}{cc} P_x & 0_{n \times n} \\ 0_{n \times n} & P_y \end{array} \right] \left(\begin{array}{c} \dot{x}(t+\tau) \\ \dot{y}(t) \end{array} \right) - \\ & h\beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds + \beta e^{h\tau} x^\top(t+\tau) Q_x x(t+\tau) - \\ & -\beta x^\top(t) Q_x x(t) = \\ & 2 \left(\begin{array}{c} x(t+\tau) \\ y(t) \end{array} \right)^\top \left(\begin{array}{c} P_x A x(t+\tau) + P_x B K y(t) + P_x \xi(t+\tau) \\ P_y (A + B K) y(t) + P_y e^{A\tau} \xi(t) \end{array} \right) + \\ & -h\beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds + \beta e^{h\tau} x^\top(t+\tau) Q_x x(t+\tau) - \\ & -\beta x^\top(t) Q_x x(t), \end{aligned}$$

which can be represented as

$$\begin{aligned}
 \dot{V}_\tau(x(t+\tau), y(t)) = & \\
 z(t)^\top & \begin{bmatrix} P_x A + A^\top P_x + \beta e^{h\tau} Q_x & P_x B K & P_x & 0_{n \times n} \\ (P_x B K)^\top & P_y (A + B K) + (A + B K)^\top P_y & 0_{n \times n} & P_y e^{A\tau} \\ P_x & 0_{n \times n} & -\varepsilon I_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & e^{A^\top \tau} P_y & 0_{n \times n} & -\varepsilon I_{n \times n} \end{bmatrix} z(t) \\
 & + \varepsilon \left(\|\xi(t+\tau)\|^2 + \|\xi(t)\|^2 \right) - \beta x^\top(t) Q_x x(t) \\
 & - h\beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds,
 \end{aligned}$$

where

$$z(t) = \begin{pmatrix} x(t+\tau) \\ y(t) \\ \xi(t+\tau) \\ \xi(t) \end{pmatrix}$$

Using the property (14.24), the right-hand side of the last differential equation can be estimated as

$$\left. \begin{aligned} & \dot{V}_\tau(x(t+\tau), y(t)) \leq z(t)^\top W_\tau z(t) \\ & -\alpha V_\tau(x(t+\tau), y(t)) + 2\varepsilon c_0 + (\varepsilon - \beta) x^\top(t) Q_x x(t) \\ & - (h - \alpha) \beta \int_{s=t}^{t+\tau} e^{h(s-t)} x^\top(s) Q_x x(s) ds \end{aligned} \right\} \quad (14.35)$$

If $W_\tau < 0$, $\varepsilon \leq \beta$ and $\alpha \leq h$, then from (14.35) we get

$$\dot{V}_\tau(x(t+\tau), y(t)) \leq -\alpha V_\tau(x(t+\tau), y(t)) + 2\varepsilon c_0$$

implying (14.33). ■

Corollary 14.2 *In view of the inequality*

$$x^\top(t+\tau) P_x x(t+\tau) \leq V_\tau(x(t+\tau), y(t))$$

we may conclude that

$$\limsup_{t \rightarrow \infty} x^\top(t) P_x x(t) \leq 2c_0 \frac{\varepsilon}{\alpha},$$

and, as the result,

$$\left. \begin{aligned} \limsup_{t \rightarrow \infty} x^\top(t) P_{attr} x(t) &\leq 1, \\ P_{attr} &= \frac{\alpha}{2c_0\varepsilon} P_x \end{aligned} \right\} \quad (14.36)$$

14.2.4 Minimal attractive ellipsoid for the original system

Following the standard technique, used in the previous lectures, instead of the matrix nonlinear constraint $W_\tau < 0$ we may consider the equivalent constraint

$$W_{\tau,T} := T^\top W_\tau T < 0$$

with

$$T = \begin{bmatrix} P_x^{-1} & 0_{n \times n} & 0_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & P_y^{-1} & 0_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & 0_{n \times n} & I_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & 0_{n \times n} & 0_{n \times n} & I_{n \times n} \end{bmatrix} = \text{diag} [P_x^{-1}, P_y^{-1}, I_{n \times n}, I_{n \times n}].$$

We get

$$0 > W_{\tau,T} = \begin{bmatrix} AP_x^{-1} + P_x^{-1}A^\top + P_x^{-1}[\alpha I_{n \times n} + (1 + \beta e^{h\tau})Q_x]P_x^{-1} & BKP_y^{-1} & I_{n \times n} & 0_{n \times n} \\ (BKP_y^{-1})^\top & (A + BK)P_y^{-1} + [(A + BK)P_y^{-1}]^\top + \alpha(P_y^{-1})^2 & 0_{n \times n} & e^{A\tau} \\ I_{n \times n} & 0_{n \times n} & -\varepsilon I_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & e^{A^\top \tau} & 0_{n \times n} & -\varepsilon I_{n \times n} \end{bmatrix}.$$

In new variables

$$X = P_x^{-1}, \quad Y = P_y^{-1}, \quad Z = KP_y^{-1},$$

and using the upper estimates for quadratic elements in diagonal blocks

$$P_x^{-1} [\alpha I_{n \times n} + (1 + \beta e^{h\tau})Q_x] P_x^{-1} < Q_1,$$

$$(P_y^{-1})^2 < Q_2,$$

which by the Schur's complement can be represented as LMI's

$$\begin{bmatrix} Q_1 & P_x^{-1} \\ P_x^{-1} & [\alpha I_{n \times n} + (1 + \beta e^{h\tau}) Q_x]^{-1} \end{bmatrix} > 0 \quad (14.37)$$

and

$$\begin{bmatrix} Q_2 & P_y^{-1} \\ P_y^{-1} & I_{n \times n} \end{bmatrix} > 0, \quad (14.38)$$

we are able to conclude that $W_{\tau,T} < \bar{W}_{\tau,T}$ where

$$\bar{W}_{\tau,T} = \begin{bmatrix} AX + X^\top P_x^{-1} + Q_1 & BZ & I_{n \times n} & 0_{n \times n} \\ (BZ)^\top & AY + BZ + [AY + BZ]^\top + \alpha Q_2 & 0_{n \times n} & e^{A\tau} \\ I_{n \times n} & 0_{n \times n} & -\varepsilon I_{n \times n} & 0_{n \times n} \\ 0_{n \times n} & e^{A^\top \tau} & 0_{n \times n} & -\varepsilon I_{n \times n} \end{bmatrix} \quad (14.39)$$

We are ready to formulate the following result.

Theorem 14.3 *The optimal feedback matrix K^* , minimazing the attractive ellipsoid $\mathcal{E}_0(P_{attr})$, is equal to*

$$K^* = Z^* (Y^*)^{-1}, \quad (14.40)$$

where Z^* and Y^* are the solution of the following matrix optimization problem

$$\begin{aligned} \text{tr} \{P_{attr}^{-1}\} = 2c_0 \text{tr} \left\{ \frac{\varepsilon}{\alpha} X \right\} \rightarrow \inf_{X>0, Y>0, Z, Q_1>0, Q_2>0, \varepsilon>0, \alpha>0} \\ \text{subject to LMI's constraints} \\ \bar{W}_{\tau,T} < 0, \\ \begin{bmatrix} Q_1 & X \\ X & [\alpha I_{n \times n} + (1 + \beta e^{h\tau}) Q_x]^{-1} \end{bmatrix} > 0, \\ \begin{bmatrix} Q_2 & Y \\ Y & I_{n \times n} \end{bmatrix} > 0, \end{aligned} \quad (14.41)$$

Proof. It follows from the estimate $W_{\tau,T} < \bar{W}_{\tau,T} < 0$, the representation (14.36) for P_{attr} and the LMI's (14.37) and (14.38) given in new variable X, Y . ■

14.2.5 Example and Exercise

Example 14.1 For the dynamic system

$$\left. \begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t - \tau) + \xi(t), \\ u(t) &= 0 \text{ for } t \in [-\tau, 0] \end{aligned} \right\} \quad (14.42)$$

with

$$\left. \begin{aligned} A &= \begin{pmatrix} -1 & 2 & 0.4 \\ -1.5 & -0.7 & 2 \\ 0.5 & -0.6 & -1 \end{pmatrix}, \quad B = \begin{pmatrix} 0.3 \\ 0 \\ 1.2 \end{pmatrix}, \quad \tau = 0.5, \\ \xi(t) &= \begin{pmatrix} 0.0028 \cos(0.6t) - 0.0879 \sin(0.6t) \\ 0.0499 \cos(0.6t) + 0.0049 \sin(0.6t) \end{pmatrix} \end{aligned} \right\} \quad (14.43)$$

design a feedback controller using the **predictive approach**.

Solution 14.1 It is easy to show that the upper bound (14.24)

$$\|\xi(x(t), t)\|^2 \leq c_0 + x^\top(t) Q_x x(t)$$

is valid for

$$c_0 = 0.0103, \quad Q_x = 0_{n \times n}.$$

The obtained numerical solution of the constrained optimization problem (14.41) is

$$\begin{aligned} X^* &= \begin{bmatrix} 0.6280 & -0.2592 & -0.0494 \\ -0.2592 & 0.9754 & -0.2716 \\ -0.0494 & -0.2716 & 0.2775 \end{bmatrix}, \\ Y^* &= \begin{bmatrix} 0.5233 & -0.0260 & -0.0761 \\ -0.0260 & 0.5073 & -0.1453 \\ -0.0761 & -0.1453 & 0.1892 \end{bmatrix}, \\ Z^* &= [0.1072 \quad 0.0861 \quad -0.1798] \\ \alpha^* &= 0.7, \quad \beta^* = 7, \quad \varepsilon^* = 3.1, \quad h = 2, \end{aligned}$$

which leads to

$$K^* = [0.055 \quad -0.712 \quad -0.1952]$$

and

$$P_{attr} = \begin{bmatrix} 1.2772 & 0 & 0 \\ 0 & 1.2772 & 0 \\ 0 & 0 & 1.2772 \end{bmatrix}.$$

The states behavior $\begin{pmatrix} x_1(t) & x_2(t) & x_3(t) \end{pmatrix}$ with the initial point

$$\begin{pmatrix} x_1(0) = 0, & x_2(0) = -1, & x_3(0) = 3 \end{pmatrix}$$

in the single and 3D-format are presented in figures 14.1 and 14.2.

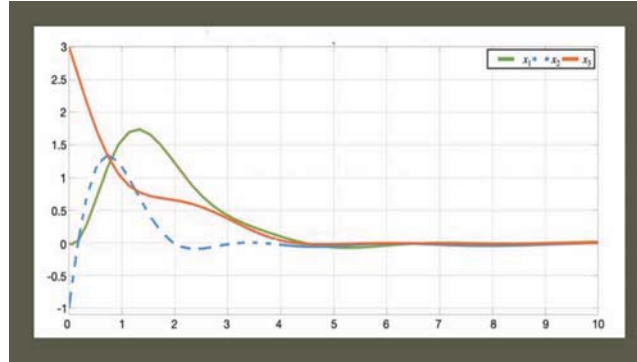


Figure 14.1: Trajectories of the controlled system.

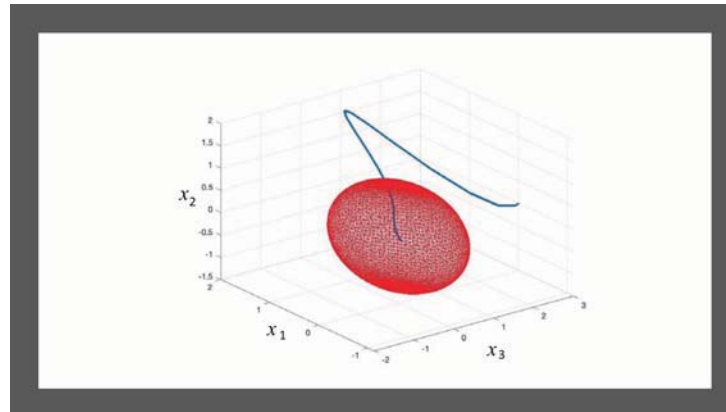


Figure 14.2: Convergence into the attractive ellipsoid.

Exercise 14.1 For the dynamic system

$$\left. \begin{aligned} \dot{x}(t) &= Ax(t) + Bu(t - \tau) + \xi(t), \\ u(t) &= 0 \text{ for } t \in [-\tau, 0] \end{aligned} \right\} \quad (14.44)$$

with

$$\left. \begin{aligned} A &= \begin{pmatrix} -1 & 2 \\ -1.5 & -0.7 \end{pmatrix}, \quad B = \begin{pmatrix} 3 \\ 1 \end{pmatrix}, \quad \tau = 0.25, \\ \xi(t) &= \begin{pmatrix} 0.0028 \cos(0.6t) \\ 0.0049 \sin(1.1t) \end{pmatrix} \end{aligned} \right\} \quad (14.45)$$

design a feedback controller using the **predictive approach**.