

### 4.5.2 Delay-Dependent Condition

In the last section, we have considered the delay-independent condition for the stability of the time delay system (4.47). In this section, we further present a delay-dependent condition. For this purpose, we first rewrite the system as

$$\begin{aligned}\dot{x}(t) &= Ax(t) + A_d x(t-d) \\ &= (A + A_d)x(t) - A_d (x(t) - x(t-d)) \\ &= (A + A_d)x(t) - A_d \int_{-d}^0 \dot{x}(t+s) ds \\ &= (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s) + A_d x(t-d+s)] ds. \quad (4.51)\end{aligned}$$

The main result in this section is given as follows.

$$\begin{aligned}\dot{x}(t) &= Ax(t) + A_d x(t-d) \\ &= (A + A_d)x(t) - A_d (x(t) - x(t-d)) \\ &= (A + A_d)x(t) - A_d \int_{-d}^0 \dot{x}(t+s) ds \\ &= (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s) + A_d x(t-d+s)] ds\end{aligned}$$

**Theorem 4.9** The time-delay system (4.47) is uniformly asymptotically stable if there exist a symmetric positive definite matrix  $X$  and a scalar  $0 < \beta < 1$ , such that

$$\begin{bmatrix} \Phi(X) & \bar{d}XA^T & \bar{d}XA_d^T \\ \bar{d}AX & -\bar{d}\beta I & 0 \\ \bar{d}A_dX & 0 & -\bar{d}(1-\beta)I \end{bmatrix} < 0, \quad (4.52)$$

$$\begin{bmatrix} \Phi(X) & d_mXA^T & d_mXA_d^T \\ d_mAX & -d_m\beta I & 0 \\ d_mA_dX & 0 & -d_m(1-\beta)I \end{bmatrix} < 0$$

$$\Phi(X) = X(A + A_d)^T + (A + A_d)X + d_mA_dA_d^T$$

PROOF:

$$V(x_t) = x^T Px + \frac{1}{\beta} \int_{-d}^0 \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta ds + \frac{1}{1-\beta} \int_{-d}^0 \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta ds$$

$$V(x_t) = T_1 + T_2 + T_3$$

$$V(x_t) = \underbrace{x^T Px}_{T_1} + \underbrace{\frac{1}{\beta} \int_{-d}^0 \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta ds}_{T_2} + \underbrace{\frac{1}{1-\beta} \int_{-d}^0 \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta ds}_{T_3}$$

$$\frac{d}{dt} V(x_t) = \frac{d}{dt} T_1 + \frac{d}{dt} T_2 + \frac{d}{dt} T_3$$

$$\frac{d}{dt} T_1 = \frac{d}{dt} x^T Px = \dot{x}^T Px + x^T P \dot{x}$$

$$= \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s) + A_d x(t-d+s)] ds \right]^T Px + x^T P \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s) + A_d x(t-d+s)] ds \right]$$

$$= \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s)] ds - A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]^T Px + x^T P \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s)] ds - A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]$$

$$= \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s)] ds - A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]^T Px + x^T P \left[ (A + A_d)x(t) - A_d \int_{-d}^0 [Ax(t+s)] ds - A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]$$

$$\dot{T}_{1r} = [x^T (A + A_d)^T Px + x^T P (A + A_d)x]$$

Now focus on the green terms

$$\dot{T}_{1g} = \left[ -A_d \int_{-d}^0 [Ax(t+s)] ds \right]^T Px + x^T P \left[ -A_d \int_{-d}^0 [Ax(t+s)] ds \right]$$

Since each of these two terms is scalar, each can be transposed. Let's transpose the first term.

$$\begin{aligned}\dot{T}_{1g} &= x^T P \left[ -A_d \int_{-d}^0 [Ax(t+s)] ds \right] + x^T P \left[ -A_d \int_{-d}^0 [Ax(t+s)] ds \right] \\ \dot{T}_{1g} &= -2x^T P A_d \left[ \int_{-d}^0 [Ax(t+s)] ds \right]\end{aligned}$$

Now focus on the blue terms,

$$\dot{T}_{1b} = \left[ -A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]^T P x + x^T P \left[ -A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right]$$

Since each of these two terms is scalar, each can be transposed. Let's transpose the first term.

$$\begin{aligned}\dot{T}_{1b} &= x^T P \left[ -A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right] + x^T P \left[ -A_d \int_{-d}^0 [A_d x(t-d+s)] ds \right] \\ \dot{T}_{1b} &= -2x^T P A_d \left[ \int_{-d}^0 [A_d x(t-d+s)] ds \right]\end{aligned}$$

Now let us focus on the derivative of T2 wrt t

$$\begin{aligned}\dot{T}_2 &= \frac{d}{dt} \left[ \frac{1}{\beta} \int_{-d}^0 \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta ds \right] \\ \dot{T}_2 &= \frac{1}{\beta} \int_{-d}^0 \frac{d}{dt} \left( \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta \right) ds\end{aligned}$$

Using Leibnitz integral formulas,

$$\begin{aligned}\frac{d}{dt} \int_a^b f(t, x) dx &= \int_a^b \left( \frac{\partial}{\partial t} f(t, x) \right) dx \\ \frac{d}{dt} \int_{a(t)}^{b(t)} f(t, x) dx &= \int_{a(t)}^{b(t)} \left( \frac{\partial}{\partial t} f(t, x) \right) dx + f(t, b(t)) \frac{db(t)}{dt} - f(t, a(t)) \frac{da(t)}{dt} \\ \frac{d}{dt} \left( \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta \right) &= \int_{t+s}^t \left( \frac{d}{dt} x^T(\theta) A^T A x(\theta) \right) d\theta + \left( x^T(\theta) A^T A x(\theta) \right) \Big|_{\theta=t} (1) - \left( x^T(\theta) A^T A x(\theta) \right) \Big|_{\theta=t+s} (1) \\ \frac{d}{dt} \left( \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta \right) &= (x^T(t) A^T A x(t)) - (x^T(t+s) A^T A x(t+s))\end{aligned}$$

$$\begin{aligned}
\dot{T}_2 &= \frac{1}{\beta} \int_{-d}^0 \frac{d}{dt} \left( \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta \right) ds \\
\dot{T}_2 &= \frac{1}{\beta} \int_{-d}^0 [ (x^T(t) A^T A x(t)) - (x^T(t+s) A^T A x(t+s)) ] ds \\
\dot{T}_2 &= \frac{1}{\beta} \int_{-d}^0 [ (x^T(t) A^T A x(t)) ] ds - \frac{1}{\beta} \int_{-d}^0 [ (x^T(t+s) A^T A x(t+s)) ] ds \\
\dot{T}_2 &= \frac{d}{\beta} (x^T(t) A^T A x(t)) - \frac{1}{\beta} \int_{-d}^0 [ (x^T(t+s) A^T A x(t+s)) ] ds
\end{aligned}$$

Now let us focus on the derivative of T3 wrt t

$$\begin{aligned}
\dot{T}_3 &= \frac{d}{dt} \left[ \frac{1}{1-\beta} \int_{-d}^0 \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta ds \right] \\
\dot{T}_3 &= \frac{1}{1-\beta} \int_{-d}^0 \frac{d}{dt} \left( \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta \right) ds
\end{aligned}$$

By using Leibnitz's integral formulae,

$$\begin{aligned}
\frac{d}{dt} \int_{a(t)}^{b(t)} f(t, x) dx &= \int_{a(t)}^{b(t)} \left( \frac{\partial}{\partial t} f(t, x) \right) dx + f(t, b(t)) \frac{db(t)}{dt} - f(t, a(t)) \frac{da(t)}{dt} \\
\frac{d}{dt} \left( \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta \right) &= \int_{t-d+s}^t \left( \frac{d}{dt} x^T(\theta) A_d^T A_d x(\theta) \right) d\theta + (x^T(\theta) A_d^T A_d x(\theta)) \Big|_{\theta=t} (1) - (x^T(\theta) A_d^T A_d x(\theta)) \Big|_{\theta=t-d+s} (1)
\end{aligned}$$

$$\frac{d}{dt} \left( \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta \right) = (x^T(t) A^T A x(t)) - (x^T(t-d+s) A^T A x(t-d+s))$$

$$\begin{aligned}
\dot{T}_3 &= \frac{1}{1-\beta} \int_{-d}^0 \frac{d}{dt} \left( \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta \right) ds \\
\dot{T}_3 &= \frac{1}{1-\beta} \int_{-d}^0 [ (x^T(t) A_d^T A_d x(t)) - (x^T(t-d+s) A_d^T A_d x(t-d+s)) ] ds \\
\dot{T}_3 &= \frac{1}{1-\beta} \int_{-d}^0 [ (x^T(t) A_d^T A_d x(t)) ] ds - \frac{1}{1-\beta} \int_{-d}^0 [ (x^T(t-d+s) A_d^T A_d x(t-d+s)) ] ds
\end{aligned}$$

$$\dot{T}_3 = \frac{d}{1-\beta} \left( x^T(t) A_d^T A_d x(t) \right) - \frac{1}{1-\beta} \int_{-d}^0 \left[ \left( x^T(t-d+s) A_d^T A_d x(t-d+s) \right) \right] ds$$

As a result of this,

$$V(x_t) = \underbrace{x^T P x}_{\dot{T}_1} + \underbrace{\frac{1}{\beta} \int_{-d}^0 \int_{t+s}^t x^T(\theta) A^T A x(\theta) d\theta ds}_{\dot{T}_2} + \underbrace{\frac{1}{1-\beta} \int_{-d}^0 \int_{t-d+s}^t x^T(\theta) A_d^T A_d x(\theta) d\theta ds}_{\dot{T}_3}$$

$$\frac{d}{dt} V(x_t) = \dot{T}_1 + \dot{T}_2 + \dot{T}_3$$

$$\begin{aligned} \frac{d}{dt} V(x_t) &= [x^T(A + A_d)^T P x + x^T P(A + A_d)x] - 2x^T P A_d \left[ \int_{-d}^0 [Ax(t+s)] ds \right] - 2x^T P A_d \left[ \int_{-d}^0 [A_d x(t-d+s)] ds \right] \\ &+ \left[ \frac{d}{\beta} (x^T(t) A^T A x(t)) - \frac{1}{\beta} \int_{-d}^0 [(x^T(t+s) A^T A x(t+s))] ds \right] \\ &+ \left[ \frac{d}{1-\beta} (x^T(t) A_d^T A_d x(t)) - \frac{1}{1-\beta} \int_{-d}^0 [(x^T(t-d+s) A_d^T A_d x(t-d+s))] ds \right] \end{aligned}$$

And

$$\begin{aligned} \frac{d}{dt} V(x_t) &= [x^T(A + A_d)^T P x + x^T P(A + A_d)x] - 2x^T P A_d \left[ \int_{-d}^0 [Ax(t+s)] ds \right] - 2x^T P A_d \left[ \int_{-d}^0 [A_d x(t-d+s)] ds \right] \\ &+ \left[ \frac{d}{\beta} (x^T(t) A^T A x(t)) - \frac{1}{\beta} \int_{-d}^0 [(x^T(t+s) A^T A x(t+s))] ds \right] \\ &+ \left[ \frac{d}{1-\beta} (x^T(t) A_d^T A_d x(t)) - \frac{1}{1-\beta} \int_{-d}^0 [(x^T(t-d+s) A_d^T A_d x(t-d+s))] ds \right] \end{aligned}$$

Now let us focus on the red-terms,

$-2x^T P A_d \left[ \int_{-d}^0 [Ax(t+s)] ds \right]$	<p><b>Lemma 2.1</b> Let <math>X, Y \in \mathbb{R}^{m \times n}</math>, <math>F \in \mathbb{S}^m</math>, <math>F &gt; 0</math>, and <math>\delta &gt; 0</math> be a scalar, then</p> $X^T F Y + Y^T F X \leq \delta X^T F X + \delta^{-1} Y^T F Y. \quad (2.1)$ <p>Particularly, when <math>X = x</math> and <math>Y = y</math> are vectors, the aforementioned inequality reduces to</p> $2x^T F y \leq \delta x^T F x + \delta^{-1} y^T F y. \quad (2.2)$ <p style="text-align: right;"><b>29</b></p>
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$$\begin{aligned} 2 \underbrace{\left( \underbrace{-x^T P A_d}_{\mathbb{X}^T} \right) \left( \underbrace{\frac{d}{F} \left[ \int_{-d}^0 [Ax(t+s)] ds \right]}_{\mathbb{Y}} \right)}_{\mathbb{Y}^T} &\leq \underbrace{\left[ \frac{\beta}{\delta} \left( \underbrace{-x^T P A_d}_{\mathbb{X}^T} \right) \left( \underbrace{-x^T P A_d}_{\mathbb{X}} \right)^T \right]}_{\mathbb{Y}^T} + \underbrace{\left[ \frac{1}{\delta^{-1}} \left( \underbrace{\frac{1}{d} \int_{-d}^0 [Ax(t+s)] ds}_{\mathbb{Y}^T} \right)^T \right]}_{\mathbb{Y}^T} \underbrace{\left( \underbrace{\frac{d}{F} \left[ \int_{-d}^0 [Ax(t+s)] ds \right]}_{\mathbb{Y}} \right)}_{\mathbb{Y}} \\ 2 \underbrace{\left( \underbrace{-x^T P A_d}_{\mathbb{X}^T} \right) \left( \underbrace{\frac{d}{F} \left[ \int_{-d}^0 [Ax(t+s)] ds \right]}_{\mathbb{Y}} \right)}_{\mathbb{Y}^T} &\leq \beta d x^T P A_d A_d^T P x + \frac{d}{\beta} \underbrace{\left[ \int_{-d}^0 \frac{1}{d} [Ax(t+s)] ds \right]^T}_{\mathbb{Y}^T} \underbrace{\left[ \int_{-d}^0 \frac{1}{d} [Ax(t+s)] ds \right]}_{\mathbb{Y}} \end{aligned}$$

<p>Now by using Cauchy-Schwarz</p> $\left( \int_a^b f(x) dx \right)^2 \leq (b-a) \int_a^b [f(x)]^2 dx$	$\left( \frac{d}{\beta} \right) \underbrace{\left[ \int_{-d}^0 \frac{1}{d} [Ax(t+s)] ds \right]^T}_{\mathbb{Y}^T} \underbrace{\left[ \int_{-d}^0 \frac{1}{d} [Ax(t+s)] ds \right]}_{\mathbb{Y}}$
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	$\leq \left(\frac{d}{\beta}\right) d \int_{-d}^0 \left(\frac{1}{d}\right)^2 [x^T(t+s)A^T A x(t+s)] ds$ $\leq \left(\frac{1}{\beta}\right) \int_{-d}^0 [x^T(t+s)A^T A x(t+s)] ds$
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Therefore,

$$2 \underbrace{\left( \underbrace{-x^T P A_d}_{\mathbb{X}^T} \right) \underbrace{\left( \frac{d}{\beta} \int_{-d}^0 [A x(t+s)] ds \right)}_{\mathbb{Y}}}_{\mathbb{F}} \leq \beta d x^T P A_d A_d^T P x + \left(\frac{1}{\beta}\right) \int_{-d}^0 [x^T(t+s)A^T A x(t+s)] ds$$

$$2(-x^T P A_d)(d) \left[ \frac{1}{d} \int_{-d}^0 [A x(t+s)] ds \right] - \left(\frac{1}{\beta}\right) \int_{-d}^0 [x^T(t+s)A^T A x(t+s)] ds \leq \beta d x^T P A_d A_d^T P x$$

And the red terms are eliminated,

$$\begin{aligned} \frac{d}{dt} V(x_t) &= [x^T(A + A_d)^T P x + x^T P(A + A_d)x] - 2x^T P A_d \left[ \int_{-d}^0 [A x(t+s)] ds \right] - 2x^T P A_d \left[ \int_{-d}^0 [A_d x(t-d+s)] ds \right] \\ &\quad + \left[ \frac{d}{\beta} (x^T(t)A^T A x(t)) - \frac{1}{\beta} \int_{-d}^0 [(x^T(t+s)A^T A x(t+s))] ds \right] \\ &\quad + \left[ \frac{d}{1-\beta} (x^T(t)A_d^T A_d x(t)) - \frac{1}{1-\beta} \int_{-d}^0 [(x^T(t-d+s)A_d^T A_d x(t-d+s))] ds \right] \end{aligned}$$

Now let us focus on the blue-terms,

$$-2x^T P A_d \left[ \int_{-d}^0 [A_d x(t-d+s)] ds \right] \leq d(1-\beta)x^T P A_d A_d^T P x + \frac{1}{1-\beta} \int_{-d}^0 [(x^T(t-d+s)A_d^T A_d x(t-d+s))] ds$$

This inequality is obtained by the same technique applied to the red-terms.

Therefore,

$$\begin{aligned} \frac{d}{dt} V(x_t) &\leq [x^T(A + A_d)^T P x + x^T P(A + A_d)x] + \beta d x^T P A_d A_d^T P x + \left[ \frac{d}{\beta} (x^T(t)A^T A x(t)) \right] + d(1-\beta)x^T P A_d A_d^T P x + \left[ \frac{d}{1-\beta} (x^T(t)A_d^T A_d x(t)) \right] \\ \frac{d}{dt} V(x_t) &\leq x^T [P(A + A_d) + (A + A_d)^T P] x + x^T (\beta d P A_d A_d^T P) x + x^T \left[ \frac{d}{\beta} A^T A \right] x + x^T [d(1-\beta) P A_d A_d^T P] x + x^T \left[ \frac{d}{1-\beta} A_d^T A_d \right] x \\ \frac{d}{dt} V(x_t) &\leq x^T [P(A + A_d) + (A + A_d)^T P] x + x^T \left[ \frac{d}{\beta} A^T A \right] x + x^T [d P A_d A_d^T P] x + x^T \left[ \frac{d}{1-\beta} A_d^T A_d \right] x \\ \frac{d}{dt} V(x_t) &\leq x^T [P(A + A_d) + (A + A_d)^T P] x + x^T \left[ \frac{d}{\beta} A^T A + d P A_d A_d^T P + \frac{d}{1-\beta} A_d^T A_d \right] x \end{aligned}$$

And since  $0 < d \leq d_m$   $\boxed{d_m := d_{max}}$

$$\begin{aligned} \frac{d}{dt} V(x_t) &\leq x^T [P(A + A_d) + (A + A_d)^T P] x + x^T \left[ \frac{d_m}{\beta} A^T A + d_m P A_d A_d^T P + \frac{d_m}{1-\beta} A_d^T A_d \right] x \\ \frac{d}{dt} V(x_t) &\leq x^T \left[ P(A + A_d) + (A + A_d)^T P + \frac{d_m}{\beta} A^T A + d_m P A_d A_d^T P + \frac{d_m}{1-\beta} A_d^T A_d \right] x \end{aligned}$$

For this term to be negative-definite, the RHS-matrix must be negative-definite

$$\left[ P(A + A_d) + (A + A_d)^T P + \frac{d_m}{\beta} A^T A + d_m P A_d A_d^T P + \frac{d_m}{1-\beta} A_d^T A_d \right] < 0$$

By pre- and post-multiplying the LHS, and denoting  $\boxed{X := P^{-1}}$

$$\begin{aligned}
& \left[ P(A + A_d) + (A + A_d)^T P + d_m P A_d A_d^T P + \frac{d_m}{\beta} A^T A + \frac{d_m}{1 - \beta} A_d^T A_d \right] < 0 \\
& \left[ X P (A + A_d) X + X (A + A_d)^T P X + X d_m P A_d A_d^T P X + X \frac{d_m}{\beta} A^T A X + X \frac{d_m}{1 - \beta} A_d^T A_d X \right] < 0 \\
& \left[ (A + A_d) X + X (A + A_d)^T + d_m A_d A_d^T + \frac{d_m}{\beta} X A^T A X + \frac{d_m}{1 - \beta} X A_d^T A_d X \right] < 0
\end{aligned}$$

Now define,  $\boxed{\Phi(X) := (A + A_d)X + X(A + A_d)^T + d_m A_d A_d^T}$

$$\begin{aligned}
& \left[ \Phi(X) + \frac{d_m}{\beta} X A^T A X + \frac{d_m}{1 - \beta} X A_d^T A_d X \right] < 0 \\
& \left[ \Phi(X) + d_m X A^T \left( \frac{1}{d_m \beta} \right) A X d_m + d_m X A_d^T \left( \frac{1}{d_m (1 - \beta)} \right) A_d X d_m \right] < 0 \\
& d_m X A^T \left( \frac{1}{d_m \beta} \right) A X d_m + d_m X A_d^T \left( \frac{1}{d_m (1 - \beta)} \right) A_d X d_m = \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \begin{bmatrix} \left( \frac{1}{d_m \beta} \right) & 0 \\ 0 & \left( \frac{1}{d_m (1 - \beta)} \right) \end{bmatrix} \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} \\
& \Phi(X) + \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \begin{bmatrix} \left( \frac{1}{d_m \beta} \right) & 0 \\ 0 & \left( \frac{1}{d_m (1 - \beta)} \right) \end{bmatrix} \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} < 0 \\
& \Phi(X) - \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \begin{bmatrix} \left( \frac{1}{-d_m \beta} \right) & 0 \\ 0 & \left( \frac{1}{-d_m (1 - \beta)} \right) \end{bmatrix} \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} < 0 \\
& \Phi(X) - \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \begin{bmatrix} -d_m \beta & 0 \\ 0 & -d_m (1 - \beta) \end{bmatrix}^{-1} \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} < 0 \\
& \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} < 0 \leftrightarrow \begin{bmatrix} A_{11} & 0 \\ 0 & A_{22} - A_{21} [A_{11}]^{-1} A_{12} \end{bmatrix} < 0 \leftrightarrow \begin{bmatrix} A_{11} - A_{12} [A_{22}]^{-1} A_{21} & 0 \\ 0 & A_{22} \end{bmatrix} < 0
\end{aligned}$$

$ \begin{aligned} & \begin{bmatrix} A_{11} - A_{12} [A_{22}]^{-1} A_{21} & 0 \\ 0 & A_{22} \end{bmatrix} < 0 \\ & A_{11} := \Phi(X) \\ & A_{12} := \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \\ & A_{21} := \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} \\ & A_{22} := \begin{bmatrix} -d_m \beta & 0 \\ 0 & -d_m (1 - \beta) \end{bmatrix} \end{aligned} $	$ \begin{bmatrix} A_{11} - A_{12} [A_{22}]^{-1} A_{21} & 0 \\ 0 & A_{22} \end{bmatrix} < 0 \leftrightarrow \begin{bmatrix} \Phi(X) & \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \\ \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} & \begin{bmatrix} -d_m \beta I & 0 \\ 0 & -d_m (1 - \beta) I \end{bmatrix} \end{bmatrix} < 0 $
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Finally, the delay-dependent LMI condition is given by

$$\begin{bmatrix} \Phi(X) & \begin{bmatrix} d_m X A^T & d_m X A_d^T \end{bmatrix} \\ \begin{bmatrix} d_m A X \\ d_m A_d X \end{bmatrix} & \begin{bmatrix} -d_m \beta I & 0 \\ 0 & -d_m (1 - \beta) I \end{bmatrix} \end{bmatrix} < 0$$

Therefore, the LMI conditions are given as,

For a given  $\boxed{d_m > 0}$ , Find 
$$\begin{array}{c}
X > 0 \\
0 < \beta < 1 \\
\begin{bmatrix} \Phi(X) & d_m X A^T & d_m X A_d^T \\ d_m A X & -d_m \beta I & 0 \\ d_m A_d X & 0 & -d_m (1 - \beta) I \end{bmatrix} < 0
\end{array}$$

## Delay Independent LMI Condition

<p><b>4.5 Time-Delay Systems</b></p> <p>The problem of stability analysis for a time-delay system can be stated as follows.</p> <p><b>Problem 4.3</b> Given matrices <math>A, A_d \in \mathbb{R}^{n \times n}</math>, check the stability of the following linear time-delay system</p> $\begin{cases} \dot{x}(t) = Ax(t) + A_d x(t-d) \\ x(t) = \phi(t), \quad t \in [-d, 0], \quad 0 < d \leq \bar{d}, \end{cases} \quad (4.47)$ <p>where  <math>\phi(t)</math> is the initial condition  <math>d</math> represents the time-delay  <math>\bar{d}</math> is a known upper bound of <math>d</math></p>	<p><b>4.5.1 Delay-Independent Condition</b></p> <p>The following theorem gives a sufficient condition for the stability problem in terms of an LMI.</p> <p><b>Theorem 4.8</b> The system (4.47) is asymptotically stable if there exist two symmetric matrices <math>P, S \in \mathbb{S}^n</math>, such that</p> $\begin{cases} P > 0 \\ \begin{bmatrix} A^T P + PA + S & PA_d \\ A_d^T P & -S \end{bmatrix} < 0. \end{cases} \quad (4.48)$
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PROOF:

$$V(x_t) = x^T(t)Px(t) + \int_{t-d}^t x^T(\tau)Sx(\tau)d\tau, \quad \boxed{x_t := x(t+\theta), \theta \in [-d, 0]}$$

Let us obtain time-derivative of Lyap-fcn wrt the time

$$\begin{aligned} \frac{d}{dt}V(x_t) &= \dot{x}^T Px + x^T P \dot{x} + \frac{d}{dt} \left[ \int_{t-d}^t x^T(\tau)Sx(\tau)d\tau \right] \\ &= [Ax + A_d x(t-d)]^T Px + x^T P [Ax + A_d x(t-d)] + \frac{d}{dt} \left[ \int_{t-d}^t x^T(\tau)Sx(\tau)d\tau \right] \\ &= [Ax + A_d x(t-d)]^T Px + x^T P [Ax + A_d x(t-d)] + \frac{d}{dt} \left[ \int_{t-d}^t x^T(\tau)Sx(\tau)d\tau \right] \end{aligned}$$

By using Leibnitz's integral formulae,

$$\begin{aligned} \frac{d}{dt} \int_{a(t)}^{b(t)} f(t, x) dx &= \int_{a(t)}^{b(t)} \left( \frac{\partial}{\partial t} f(t, x) \right) dx + f(t, b(t)) \frac{db(t)}{dt} - f(t, a(t)) \frac{da(t)}{dt} \\ \frac{d}{dt} \left[ \int_{t-d}^t x^T(\tau)Sx(\tau)d\tau \right] &= \left[ \int_{t-d}^t \left( \frac{d}{dt} x^T(\tau)Sx(\tau) \right) d\tau \right] + \left( x^T(\tau)Sx(\tau) \right) \Big|_{\tau=t} - \left( x^T(\tau)Sx(\tau) \right) \Big|_{\tau=t-d} \quad (1) \\ &= (x^T(t)Sx(t)) - (x^T(t-d)Sx(t-d)) \end{aligned}$$

and

$$\frac{d}{dt}V(x_t) = [Ax + A_d x(t-d)]^T Px + x^T P [Ax + A_d x(t-d)] + [(x^T(t)Sx(t)) - (x^T(t-d)Sx(t-d))]$$



$$\frac{d}{dt}V(x_t) = x^T[A^TP + PA + S]x + x^TPA_dx(t-d) + x^T(t-d)A_d^TPx - x^T(t-d)Sx(t-d)$$

$$\frac{d}{dt}V(x_t) = \begin{bmatrix} x(t) \\ x(t-d) \end{bmatrix}^T \begin{bmatrix} A^TP + PA + S & PA_d \\ A_d^TP & -S \end{bmatrix} \begin{bmatrix} x(t) \\ x(t-d) \end{bmatrix}$$

And since  $\boxed{\frac{d}{dt}V(x_t) < 0}$  is required for the stability, RHS matrix must be negative-definite,

$$\begin{bmatrix} A^TP + PA + S & PA_d \\ A_d^TP & -S \end{bmatrix} < 0$$

Therefore, the conditions are given as

$$\boxed{\begin{matrix} P > 0 \\ \begin{bmatrix} A^TP + PA + S & PA_d \\ A_d^TP & -S \end{bmatrix} < 0 \end{matrix}}$$