

## Chairing #Explanation

The ***Lyapunov*** function is part of the ***Lyapunov Stability Theory***, which describes a scalar function  $V$  that helps determine the stability of an equilibrium point.

Consider a system:

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$
$$\dot{\mathbf{x}} = A\mathbf{x}$$

To propose a ***Lyapunov function***, we need it to fulfill these conditions: . a)  $V(\mathbf{x}) > 0, \forall \mathbf{x} \in X, X \rightarrow R^+$  b)  $V(\mathbf{x}^{eq}) = 0$  c)  $\frac{d}{dt}V(\mathbf{x}) < 0$  for all  $X$  except  $\mathbf{x}^{eq}$  . A common proposal for  $V(\mathbf{x})$  is:

$$V(\mathbf{x}) = \frac{1}{2} \|\mathbf{x}\|_P^2$$

Where  $\|\mathbf{x}\|_P^2$  represents the **quadratic** form associated with the positive definite matrix  $P$ . We can decompose the **quadratic Euclidean norm** into its **[[Quadratic Norm Representation/matrix notation]]** as follows:

$$V(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T P \mathbf{x}, P \neq 0$$

a)  $V(\mathbf{x}) > 0, \forall \mathbf{x} \in X, X \rightarrow R^+$  From the definition of the **[[Positive Definite Matrix]]**, we know that for this to be achieved, we need  $P$  to be a positive definite matrix.

Only if  $P > 0$

b)  $V(\mathbf{x}^{eq}) = 0$  Similarly, if the **equilibrium point** is 0, then its simple to see that this condition is achieved.

$$V(\mathbf{x}^{eq}) = \frac{1}{2} (\mathbf{x}^{eq})^T P \mathbf{x}^{eq} = 0$$

If the **equilibrium point** is different than zero, then:

c)  $\frac{d}{dt}V(\mathbf{x}) < 0$  for all  $X$  except  $\mathbf{x}^{eq}$  Because of the *chain rule*, we get the *time derivative* of  $V(\mathbf{x})$  as:

$$\frac{d}{dt}V(\mathbf{x}) = \langle \nabla_x V(\mathbf{x}), \frac{d}{dt}\mathbf{x} \rangle$$

From **[[Matrix Calculus Derivative Rules]]**, we get that the derivative or *gradient* of  $V(\mathbf{x})$  is  $P\mathbf{x}$ , so the inner product becomes:

$$\frac{d}{dt}V(\mathbf{x}) = \langle P\mathbf{x}, A\mathbf{x} \rangle$$

Considering both  $Px$  &  $Ax$  are  $2 \times 1$  vectors, we need to transpose the first one to be able to apply the inner product, basically:

$$\frac{d}{dt}V(x) = (Px) \cdot (Ax) = (Px)^T(Ax)$$

Considering that:

$$(Px)^T = x^T P^T$$

$$\frac{d}{dt}V(x) = x^T P^T Ax = 0.5x^T P^T Ax + 0.5x^T P^T Ax$$

Because  $x^T P^T Ax$  is a **scalar**, its transpose is equal to itself, so:

$$(x^T P^T Ax)^T = x^T A^T P x$$

Then we can do:

$$\frac{d}{dt}V(x) = 0.5x^T P^T Ax + 0.5x^T A^T P x$$

$$\frac{d}{dt}V(x) = 0.5x^T \underbrace{(P^T A + A^T P)}_{-Q} x$$

So we arrive at the conclusion that  $Q$  must be symmetric and **greater** than 0 to guarantee **real eigenvalues**.