



Prof.Dr.

Linear Dynamic Systems

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State space description of a Linear Dynamic System

$$\dot{\boldsymbol{x}}(t) = \boldsymbol{F}(t)\boldsymbol{x}(t) + \boldsymbol{G}(t)\boldsymbol{w}(t) + \boldsymbol{L}(t)\boldsymbol{s}(t)$$
(3.1)

- x(t) Set of random variables describing the linear system (the state vector)
- $m{w}(t)$ Random forcing function $m{s}(t)$ Deterministic control input
- F(t) square matrix
- G(t),L(t) matrices (not necessarily square!)

Here we will consider only linear dynamic models without control input:

$$\dot{\boldsymbol{x}}(t) = \boldsymbol{F}(t)\boldsymbol{x}(t) + \boldsymbol{G}(t)\boldsymbol{w}(t)$$
(3.2)

For given initial conditions $x_0(t)$, the general solution to equ.(3.2) can be written:

$$\boldsymbol{x}(t) = \boldsymbol{\Phi}(t, t_0) \boldsymbol{x}(t_0) + \int_{t_0}^{t} \boldsymbol{\Phi}(t, t') \boldsymbol{G}(t') \boldsymbol{w}(t') dt'$$
(3.3)

The general solution is the sum of the the solution of the homogeneous equation and a particular solution of the non-homogeneous equation!

 $\Phi(t,t_0)$ is called the state transition matrix. The following relations hold for this matrix:

$$\frac{d}{dt}\Phi(t,t_0) = F(t)\Phi(t,t_0)
\Phi(t_2,t_0) = \Phi(t_2,t_1)\Phi(t_1,t_0)
\Phi(t,t) = \Phi(t,t_0)\Phi(t_0,t) = I \Rightarrow \Phi^{-1}(t,t_0) = \Phi(t_0,t)$$
(3.4)

Up to now $\Phi(t, t_0)$ is still unknown!

Transition matrix for stationary systems: in stationary systems, the matrix F in equs. (3.1) and (3.2) is time-invariant. Stationary systems can often be used to replace approximatively more complex systems over short time periods. A general Taylor expansion gives:

$$\mathbf{x}(t) = \mathbf{x}(t_0) + \dot{\mathbf{x}}(t_0)(t - t_0) + \frac{1}{2!}\ddot{\mathbf{x}}(t_0)(t - t_0)^2 + \dots$$
(3.5)

From the homogeneous part of equ. (3.2) we can replace

$$\dot{\boldsymbol{x}}(t_0) = \boldsymbol{F}\boldsymbol{x}(t_0)
\ddot{\boldsymbol{x}}(t_0) = \dot{\boldsymbol{F}}\boldsymbol{x}(t_0) + \boldsymbol{F}\dot{\boldsymbol{x}}(t_0) = \boldsymbol{F}\boldsymbol{F}\boldsymbol{x}(t_0) = \boldsymbol{F}^2\boldsymbol{x}(t_0)
\vdots
\boldsymbol{x}^{(n)}(t_0) = \boldsymbol{F}^n\boldsymbol{x}(t_0)$$
(3.6)

Substituting equ. (3.6) into (3.5):

$$x(t) = x(t_0) + Fx(t_0)(t - t_0) + \frac{F^2}{2!}x(t_0)(t - t_0)^2 + \dots$$

$$x(t) = \left[I + F(t - t_0) + \frac{F^2}{2!}(t - t_0)^2 + \dots\right]x(t_0)$$
 (3.7)

The term in square brackets is by definition the matrix exponential

$$e^{A} = I + A + \frac{A^{2}}{2!} + \frac{A^{3}}{3!} + \dots$$

$$e^{F(t-t_{0})} = I + F(t-t_{0}) + \frac{F^{2}(t-t_{0})^{2}}{2!} + \dots$$
(3.8)

For stationary systems, the state transition matrix depends only on the time interval (t-t0) and the matrix ${m F}$

$$\Phi(t, t_0) = e^{F(t - t_0)}$$
(3.9)

We are now in a position to discretize the continuous system of equ. (3.2):

$$\dot{\boldsymbol{x}}(t) \quad = \quad \boldsymbol{F}(t)\boldsymbol{x}(t) + \boldsymbol{G}(t)\boldsymbol{w}(t)$$

$$x(t_n) = \Phi(t_n, t_{n-1})x(t_{n-1}) + u(t_n)$$
or
 $x_n = \Phi(t_n, t_{n-1})x_{n-1} + u_n$
(3.10)

with

$$\boldsymbol{u}_n = \int_{t_{n-1}}^{t_n} \boldsymbol{\Phi}(t, t') \boldsymbol{G}(t') \boldsymbol{w}(t') dt'$$
(3.11)

For stationary systems, the state transition matrix is computed from equ. (3.9). The discretization of equ. (3.10) holds also for non-stationary systems; but then the state transition matrix cannot be computed from equ. (3.9)!

You will find the examples discussed in this lecture as Jupyter notebook under

https://github.com/spacegeodesy/ParameterEstimationDynamicSystems/blob/master/example03.ipynb