ASEN 5044, Fall 2018

Statistical Estimation for Dynamical Systems

Lecture 3:
State Space Models
and Linear Dynamical Systems

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Overview

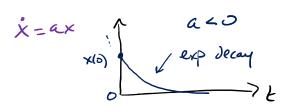
- Last time: Quick linear algebra refresher
- Today: State Space Models
 - motivation, examples
 - → (A,B,C,D) matrix parameters for continuous time (CT) linear dynamical systems
 - o state transition matrix (STM) and matrix exponential solutions

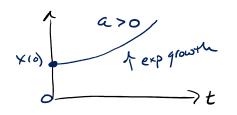
READ: Chapter 1.2-1.3 in Simon book

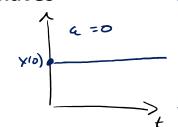
Refresher/Motivation:

$$\dot{x}(t) = ax(t) \iff x(t) = e^{at}x(0)$$

- Given a, knowing x(0) completely determines x(t) in future
- Given x(0), knowing a completely determines how x(t) **behaves**







a=±jw oscillation e Flag. w to to 70

How to generalize these insights from Scalar I. near systems (ODES) - 520 1070 to move complex systems (ODES) of vector variables & (1: hear) vector - Matrix ODES

e.g.
$$\dot{x}_{1} = f_{1}(x_{1}, x_{2}, ..., x_{n})$$

$$\dot{x}_{2} = f_{2}(x_{2}, x_{2}, ..., x_{n})$$

$$\vdots$$

$$\dot{x}_{n} = f_{n}(x_{1}, x_{2}, ..., x_{n})$$

Solutions to all these ODES are coupled generally

State Space Models

- Idea of "state variable" (state vector):
 - Completely summarize information about condition of a dynamical system (i.e. summarizes results of past events leading to present)
 - Sufficient to <u>completely</u> and <u>uniquely</u> describe system at all future times (given an input + dynamics model)

(i.e. the state is such that "knowing 'it' now is enough to tell you all about 'it' later")

Example #1: Mass with 1 Deg of Freedom

Example #2: Same Mass, Slightly Different Model

Re-arrange Dynamics to Reflect States (want d vs. f)

• Let
$$x = \begin{bmatrix} v \\ d \end{bmatrix} = \begin{bmatrix} x_i \\ x_2 \end{bmatrix} = \begin{cases} \frac{1}{2} \\ \frac{1}$$

Re-arrange Dynamics to Reflect States (want d vs. f)

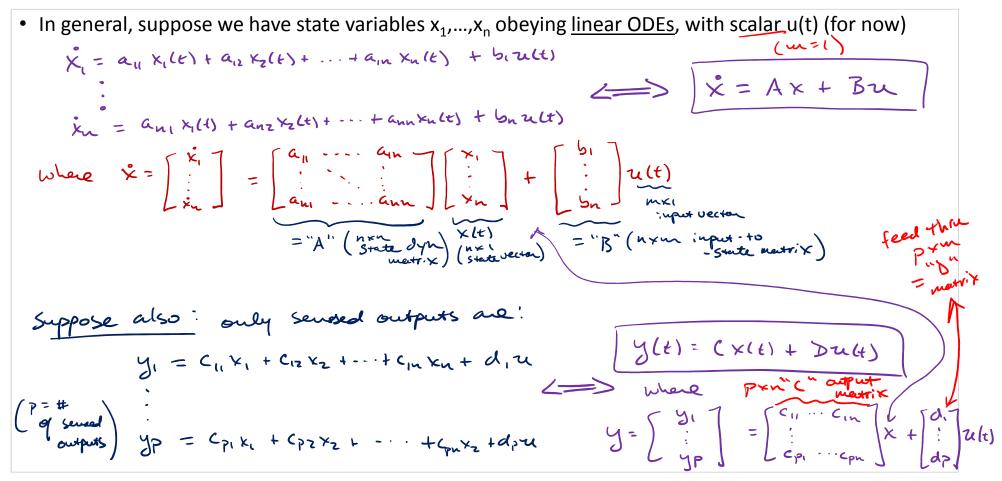
Can rewrite this all in general (Linear) matrix - vector size form:

$$\dot{x} = \begin{bmatrix} \dot{v}(t) \\ \dot{d}(t) \end{bmatrix} = \begin{bmatrix} 0.8(t) + 0.d(t) + \frac{1}{10} & F(t) \\ 1.8(t) + 0.d(t) + 0.F(t) \end{bmatrix} \quad \text{(where } x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \dot{v} \\ \dot{d} \end{bmatrix}$$

$$= \begin{bmatrix} 0.x_1 + 0.x_2 + \frac{1}{10} & U(t) \\ 1.x_1 + 0.x_2 + 0.U(t) \end{bmatrix}$$

$$= x_1 = \begin{bmatrix} x_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 \\ \dot{x}_1 \end{bmatrix} + \begin{bmatrix} 1 \\ \dot{x}_2 \end{bmatrix}$$

State Space Form of Linear Dynamical Systems



General Linear State Space Models

Can generalize this formulation for time-varying or time-invariant dynamics:

Lihean Time Varying (LTV) State space model

$$\dot{\chi}(t) = \underline{A(t)} \chi(t) + \underline{B(t)} \iota_{\lambda}(t) \quad \lambda_{\lambda}(t) \quad \lambda_{\lambda}(t)$$

. Linear Time Invantants (LTI) SS models :

$$\dot{x}(t) = A \ x(t) + Bu(t)$$

$$y(t) = C \ x(t) + Du(t)$$

$$w.f.t. + vel = u.(t) = Beiling = builties for u is $\geq 1?$

$$uu(t) = u.(t) = builties =$$$$