



Mo April 8 at 1315-1430 lecture  
 Mo April 15 at 1315-1600 lecture and exercises  
 Fr April 26 at 0815-1100 lecture and exercises  
 We May 8 at 0915-1200 lecture and exercises  
 We May 15 at 0915-1200 lecture and exercises  
 We May 22 at 0915-1000 lecture  
 We May 29 at 0915-1200 exercises

## A Course of Six Lectures

1. Introduction  
Fixed modes, Team theory, Witsenhausen's counterexample
2. Partial nestedness and quadratic invariance  
Control with information delays  
Example: Tele-operation
3. Dual decomposition  
The saddle algorithm  
Example: The Internet protocol
4. Distributed MPC  
Example: Water Supply Network
5. Distributed control of positive systems. Consensus algorithms
6. Spatially invariant systems.

## Spatially invariant systems

$$\begin{aligned} \frac{\partial}{\partial t} \psi(x, t) &= [A\psi](x, t) + [B\psi](x, t) \\ y(x, t) &= [C\psi](x, t) + [D\psi](x, t) \end{aligned}$$

The variable  $x = (x_1, \dots, x_d)$  is called the spatial variable. The components  $x_k$  could be integers, real numbers or, more generally, elements of a locally compact abelian group.

The operators  $A, B, C, D$  are assumed to be translation invariant, e.g.  $AT_x = T_x A$  for every translation  $T_x$ .

## Example 2 — Spring connected bodies

In an infinite string of bodies connected by springs, let  $p_i$  be the position of body  $i$ , which is subject to a control force  $u_i$  and a disturbance  $w_i$ . Then

$$\begin{aligned} \frac{d^2 p_i}{dt^2}(t) &= \frac{1}{2} [p_{i+1}(t) + p_{i-1}(t) - 2p_i(t)] + u_i(t) + w_i(t) \\ i &= 0, \pm 1, \pm 2, \dots \end{aligned}$$

Here the "spatial" variable  $i$  belongs to the set of integers.

## Lecture 6

- **Spatially invariant systems**
  - Separation of spatial frequencies
  - Approximation by spatial truncation

## Example 1 — String of vehicles

In an infinite string of vehicles, let  $p_i$  be the position of vehicle  $i$  relative to vehicle  $i - 1$ . Let  $u_i$  and  $w_i$  be control force and disturbance acting on vehicle  $i$ . Then

$$\begin{aligned} \frac{d^2 p_i}{dt^2}(t) &= u_i(t) - u_{i-1}(t) + w_i(t) - w_{i-1}(t) \\ i &= 0, \pm 1, \pm 2, \dots \end{aligned}$$

Here the "spatial" variable  $i$  belongs to the set of integers.

## Example 3

Consider the system

$$\begin{bmatrix} \dot{\psi}_1 \\ \dot{\psi}_2 \\ \dot{\psi}_3 \\ \dot{\psi}_4 \end{bmatrix} = \begin{bmatrix} \psi_2 + 2\psi_4 \\ \psi_1 + 2\psi_3 \\ \psi_4 + 2\psi_2 \\ \psi_3 + 2\psi_1 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \\ 0 & 2 & 0 & 1 \\ 2 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \end{bmatrix}$$

The system can also be written as

$$\dot{\psi}_{(i,j)} = \psi_{(i+1,j)} + 2\psi_{(i+1,j+1)}$$

where  $i, j \in \{0, 1\}$  and addition is taken modulo 2. Here the spatial variables are  $i$  and  $j$ .

## Lecture 6

- Spatially invariant systems
- **Separation of spatial frequencies**
- Approximation by spatial truncation

### Example 2 — Spring connected bodies

$$\frac{d^2 p_i}{dt^2}(t) = \frac{1}{2}[p_{i+1}(t) + p_{i-1}(t) - 2p_i(t)] + u_i(t) + w_i(t)$$

$$\frac{d^2 \hat{p}}{dt^2}(t) = \underbrace{\frac{1}{2}(\lambda + \lambda^{-1} - 2)}_{\alpha_\lambda} \hat{p}(\lambda, t) + \hat{u}(\lambda, t) + \hat{w}(\lambda, t)$$

With  $\psi = (p, \frac{dp}{dt})$  we get  $\frac{d}{dt}\hat{\psi} = \begin{bmatrix} 0 & 1 \\ \alpha_\lambda & 0 \end{bmatrix} \hat{\psi} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} (\hat{u} + \hat{w})$ .

The cost  $\sum_i \int_0^\infty [p_i(t)^2 + (\frac{dp_i}{dt}(t))^2 + u_i(t)^2] dt$  is minimized by  $\hat{u}(t) = -(K * \psi)(t)$  where  $\hat{K}(\lambda) = [\hat{P}_{12} \quad \hat{P}_{22}]$  and

$$\hat{P}(\lambda) = \begin{bmatrix} \hat{P}_{11} & \hat{P}_{12} \\ \hat{P}_{12} & \hat{P}_{22} \end{bmatrix} \text{ where } \begin{cases} \hat{P}_{12}(\lambda) = \alpha_\lambda + \sqrt{\alpha_\lambda^2 + 1} \\ \hat{P}_{22}(\lambda) = \sqrt{2\hat{P}_{12} + 1} \\ \hat{P}_{11}(\lambda) = (\hat{P}_{12} - \alpha_\lambda)\hat{P}_{22} \end{cases}$$

### Example 3

$$\begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \end{bmatrix} = \underbrace{\begin{bmatrix} \psi_2 + 2\psi_4 \\ \psi_1 + 2\psi_3 \\ \psi_4 + 2\psi_2 \\ \psi_3 + 2\psi_1 \end{bmatrix}}_A = \begin{bmatrix} 0 & 1 & 0 & 2 \\ 1 & 0 & 2 & 0 \\ 0 & 2 & 0 & 1 \\ 2 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \psi_1 \\ \psi_2 \\ \psi_3 \\ \psi_4 \end{bmatrix}$$

The two spatial dimensions given by the permutations

$$(\psi_1, \psi_2, \psi_3, \psi_4) \rightarrow (\psi_2, \psi_1, \psi_4, \psi_3) \rightarrow (\psi_1, \psi_2, \psi_3, \psi_4)$$

$$(\psi_1, \psi_2, \psi_3, \psi_4) \rightarrow (\psi_3, \psi_4, \psi_1, \psi_2) \rightarrow (\psi_1, \psi_2, \psi_3, \psi_4)$$

have corresponding state transformation matrices

$$S = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 & 0 & 0 \\ 1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & -1 \end{bmatrix} \quad T = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$$

which diagonalize  $TSAS^{-1}T^{-1} = \text{diag}\{3, -3, -1, 1\}$ . Design controllers for each state separately and transform back!

### Localization of Controller

Control synthesis done independently for different  $\lambda$  gives

$$\hat{u}(\lambda, t) = -\hat{K}(\lambda)\hat{\psi}(\lambda, t)$$

Transforming back from frequency domain gives

$$u(t) = -(K * \psi)(t)$$

Interestingly optimal controllers have a natural degree of decentralization, reflected in exponential decay of the convolution kernel. In particular, when the spatial coordinate is an integer, analytic extension of  $\hat{K}(\lambda)$  outside the unit circle to  $R^{-1} \leq |\lambda| \leq R$  implies that  $K(n)$  decays exponentially as  $R^{-|n|}$ .

Hence, a distributed controller can be obtained by truncation.

## Separation of spatial frequencies

Fourier transform in the spatial dimension gives

$$\frac{d}{dt}\hat{\psi}(\lambda, t) = \hat{A}(\lambda)\hat{\psi}(\lambda, t) + \hat{B}(\lambda)\hat{u}(\lambda, t)$$

$$y(\lambda, t) = \hat{C}(\lambda)\hat{\psi}(\lambda, t) + \hat{D}(\lambda)\hat{u}(\lambda, t)$$

Control synthesis can be done independently for different values of the “spatial frequency”  $\lambda$ . This gives

$$\hat{u}(\lambda, t) = -\hat{K}(\lambda)\hat{\psi}(\lambda, t)$$

Transforming back from frequency domain gives

$$u(t) = -(K * \psi)(t)$$

where the convolution kernel  $K$  is obtained from  $\hat{K}$  by inverse Fourier transform.

### Discrete Fourier Transform

The discrete Fourier transform for  $N$  states is given by the state transformation matrix  $[F]_{kl} = \frac{1}{\sqrt{N}}e^{i2\pi kl}$  where  $k, l = 0, 1, \dots, N-1$ .

In particular, for  $N = 2$

$$F = F^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

## Lecture 6

- Spatially invariant systems
- Separation of spatial frequencies
- **Approximation by spatial truncation**

### Summary

- ▶ Spatial invariance is inherited by optimal controllers
- ▶ Distributed controllers can be obtained by spatial truncation