Modeling Temporal Morphological Systems via Lattice Dynamical Systems

#### Barrera, Dougherty, Gubitoso and Hirata

#### **BIOINFO-USP**

University of São Paulo, Brazil - Texas A & M University

# Outline

- Introduction
- Lattices
- Operator Representation
- Dynamical Systems
- A model for system identification
- Identification of Boolean Dynamical Systems
- Conclusion

# Introduction



#### Sequential Switching Circuits



 $\phi_1(\mathbf{x}[i-5], \mathbf{x}[i-4], \mathbf{x}[i-3], \mathbf{x}[i-2], \mathbf{x}[i-1], \mathbf{x}[i]) = \overline{x}_1[i-3] \cdot \overline{x}_1[i-2] \cdot \overline{x}_1[i-1] \cdot \overline{x}_2[i-5] \cdot \overline{x}_2[i-3] \cdot \overline{x}_2[i-1]$ 

 $\phi_2(\mathbf{x}[i-5], \mathbf{x}[i-4], \mathbf{x}[i-3], \mathbf{x}[i-2], \mathbf{x}[i-1], \mathbf{x}[i]) = \overline{x}_1[i-4] \cdot \overline{x}_2[i-5] \cdot \overline{x}_2[i-4] \cdot \overline{x}_2[i-3] \cdot \overline{x}_2[i-2] \cdot \overline{x}_2[i-1]$ 

# Mathematical Morphology

- studies operators between complete lattices, what includes switching functions
- lattice operators are decomposed in terms of simple morphological operators: erosion, dilation, anti-erosion, anti-dilation
- Any lattice operator can be decomposed in a canonical morphological representation

## Lattice Dynamical Systems

- We introduce the notion of Finite Lattice Dynamical System (FLDS)
- Give a representation for FLDSs, based on canonical morphological representations
- Develop a theory for statistical identification of FLDSs,

## Lattices

#### **Boolean Lattice**



#### **Function Lattice**

2	0	1	2	2	2
1	0	1	2	2	2
0	-1	1	2	2	2
-1	-1	1	1	1	1
-2	-2	-1	-1	-1	-1
	-2	-1	0	1	2

## **Operator Representation**

#### Intervals

• Let  $a, b \in Fun[W,L], a \le b \text{ iff } a(x) \le b(x), x \in W$ 



• Interval  $[a,b] = \{u \in \operatorname{Fun}[W,L]: a \le u \le b\}$ 

# **Binary Sup-generating**

• Sup-generating operator:  $\lambda_{a,b}(u) = 1 \Leftrightarrow u \in [a,b]$ 





[a,b]

 $\lambda_{a,b}$ 

# Kernel

Kernel of  $\psi$  at *y*:  $K(\psi)(y) = \{u \in Fun[W,L]: y \le \psi(u)\}$ 

2	0	1	2	2	2
1	0	1	2	2	2
0	-1	1	2	2	2
-1	-1	1	1	1	1
-2	-2	-1	-1	-1	-1

-2 -1 0 1 2



### Basis

Basis of  $\psi$  at y: B( $\psi$ ) is the set of maximal intervals contained in K( $\psi$ )



## **Canonical Representation**

 $\psi(u) = \bigcup \left\{ y \in M : \bigcup \left\{ \lambda_{a,b}(u) : [a,b] \in B(\psi)(y) \right\} = 1 \right\}$ 



 $\psi(-1,-1) = 1$ 

# **Dynamical Systems**

#### Finite Lattice Dinamical System

 $\begin{aligned} \mathbf{x} &: T \to \mathcal{L}^n & \mathbf{y} : T \to \mathcal{L}^m & \mathbf{x}[t] \in \mathcal{L}^n \\ \mathbf{u} &: T \to \mathcal{L}^n \end{aligned}$ 

$$\mathbf{x}[t+1] = \Phi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N], \mathbf{u}[t-N], ..., \mathbf{u}[t], ..., \mathbf{u}[t+N])$$
$$\mathbf{y}[t] = \Psi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N], \mathbf{u}[t-N], ..., \mathbf{u}[t], ..., \mathbf{u}[t+N])$$

$$S(\Phi_t, \Psi_t) \qquad \qquad \Phi_t : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}^n$$
$$\Psi_t : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}^m$$

#### Representation

$$\mathbf{x}_{j}[t+1] = \phi_{t,j}(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N], \mathbf{u}[t-N], ..., \mathbf{u}[t], ..., \mathbf{u}[t+N])$$
$$\mathbf{y}_{k}[t] = \psi_{t,k}(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N], \mathbf{u}[t-N], ..., \mathbf{u}[t], ..., \mathbf{u}[t+N])$$

$$\mathbf{x}_{j}[t] \in \mathcal{L}$$
  
 $\phi_{t,j} : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}$   
 $\psi_{t,k} : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}$ 

The component functions have canonical morphological representations

# System: input-output

$$\mathbf{y}[t] = S_{(\mathbf{x}[0],...,\mathbf{x}[N],...,\mathbf{x}[2N])}(\Phi_t, \Psi_t)(\mathbf{u}[t-N],...,\mathbf{u}[t],...,\mathbf{u}[t+N])$$

$$\mathbf{y} = S_{(\mathbf{x}[0],\dots,\mathbf{x}[N],\dots,\mathbf{x}[2N])}(\Phi_t,\Psi_t)(\mathbf{u})$$

#### Filter

$$\mathbf{u}: T \to \mathcal{L}^n \qquad \qquad \mathbf{y}: T \to \mathcal{L}^m$$

 $\Gamma:\mathcal{L}^{(2N+1)n}\to\mathcal{L}^m$ 

 $\mathbf{y}[t] = \Gamma_t(\mathbf{u}[t-N], ..., \mathbf{u}[t], ..., \mathbf{u}[t+N])$ 

For example, processing of motion images.

# Input-free systems

$$\mathbf{x}[t+1] = \Phi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N])$$

$$\mathbf{y}[t] = \Psi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], ..., \mathbf{x}[t+N])$$

# Causal systems

$$\mathbf{x}[t+1] = \Phi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], \mathbf{u}[t-N], ..., \mathbf{u}[t])$$

$$\mathbf{y}[t] = \Psi_t(\mathbf{x}[t-N], ..., \mathbf{x}[t], \mathbf{u}[t-N], ..., \mathbf{u}[t])$$

# Time translation invariant systems

$$\begin{split} \Phi : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}^n & \Phi_t = \Phi \\ \Psi : \mathcal{L}^{2(2N+1)n} \to \mathcal{L}^m & \Psi_t = \Psi \end{split}$$

#### Causal, time translation invariant



A model for system identification



#### System Error

 $Er(S(\Phi,\Psi)) = E\left[l(S_{(\mathbf{X}[0],...,\mathbf{X}[N])}(\Phi,\Psi)(\mathbf{U}[t-N],...,\mathbf{U}[t-1],\mathbf{U}[t]),\mathbf{I}[t])\right]$ 

 $l:\mathcal{L}^m\times\mathcal{L}^m\to\Re^+$ 

# **Stationary Conditions**

$$P(\mathbf{x}[t-N], ..., \mathbf{x}[t-1], \mathbf{x}[t], \mathbf{u}[t-N], ..., \mathbf{u}[t-1], \mathbf{u}[t], \mathbf{i}[t]) = p$$

$$\begin{split} Er\big(S(\Phi,\Psi)\big) &= E\Big[l\big(S_{(\mathbf{X}[0],...,\mathbf{X}[N])}(\Phi,\Psi)(\mathbf{U}[t-N],...,\mathbf{U}[t-1],\mathbf{U}[t]),\mathbf{I}[t]\big)\Big] \\ &= \sum_{(\mathbf{x}[t-N],...,\mathbf{x}[t],\mathbf{u}[t-N],...,\mathbf{u}[t],\mathbf{i}[t])\in\mathcal{L}^{2(N+1)n}\times\mathcal{L}^{m}} l\big(S_{(\mathbf{x}[t-N],...,\mathbf{x}[t])}(\Phi,\Psi)(\mathbf{u}[t-N],...,\mathbf{u}[t]),\mathbf{i}[t]\big)\times p(\mathbf{x}[t-N],...,\mathbf{x}[t],\mathbf{u}[t-N],...,\mathbf{u}[t],\mathbf{i}[t]) \end{split}$$

# Component Error

$$Er_k\big[S(\Phi,\Psi)\big] = E\Big[l_k\big(S_{(\mathbf{X}[0],...,\mathbf{X}[N])}(\Phi,\Psi)(\mathbf{U}[t-N],...,\mathbf{U}[t-1],\mathbf{U}[t])_k,\mathbf{I}_k[t]\big)\Big].$$

$$l_k: \mathcal{L} imes \mathcal{L} o \Re^+$$

#### **Additive Loss Function**

$$l = \sum_{k=1}^{m} c_k l_k \qquad c_k \in \Re^+$$

$$Er(S(\Phi,\Psi)) = \sum_{k=1}^{m} Er_k \left[ S(\Phi,\Psi) \right]$$

$$e_{MAE}(\mathbf{a}, \mathbf{b}) = \sum_{k=1}^{m} |\mathbf{a}_k - \mathbf{b}_k| \qquad e_{MAE} = \sum_{k=1}^{m} e_{kMAE}$$
$$\mathbf{a}, \mathbf{b} \in \{0, 1\}^m \qquad e_{kMAE}(a, b) = |a - b|$$

## Independence Condition

- Under additive loss function optimize the system error is equivalent to optimize the system components error
- The problem of system identification is reduced to a family of problems of lattice operator design.

Identification of Boolean Dynamical Systems

# A Boolean System



$$\mathbf{x}_{1}[t+1] = 1 \iff \begin{cases} \mathbf{x}_{1}[t] = 0 \\ \text{and} \\ \left[ \left( (\mathbf{x}_{3}[t] = 1 \text{ or } \mathbf{x}_{3}[t-1] = 1 \text{ or } \mathbf{x}_{3}[t-2] = 1 \right) \text{ and} \\ (\mathbf{x}_{4}[t] = 1 \text{ or } \mathbf{x}_{4}[t-1] = 1 \text{ or } \mathbf{x}_{4}[t-2] = 1 ) \right) \\ \text{or} \\ \left( \mathbf{x}_{3}[t] = \mathbf{x}_{3}[t-1] = \mathbf{x}_{3}[t-2] = \mathbf{x}_{3}[t-3] = \mathbf{x}_{3}[t-4] = 0 \text{ and} \\ \mathbf{x}_{4}[t] = \mathbf{x}_{4}[t-1] = \mathbf{x}_{4}[t-2] = \mathbf{x}_{4}[t-3] = \mathbf{x}_{4}[t-4] = 0 \right) \end{array} \right]$$

#### System simulation



# System identification: system error



# System identification: transition error





# Conclusion

- Introduced the notion of Finite Lattice Dynamical System
- Proposed a model for FLDS identification
- Under additive condition, system identification reduces to a family of problems of lattice operator design
- A Boolean example was presented
- This perspective unifies theories such as switching theory and discrete automatic control