

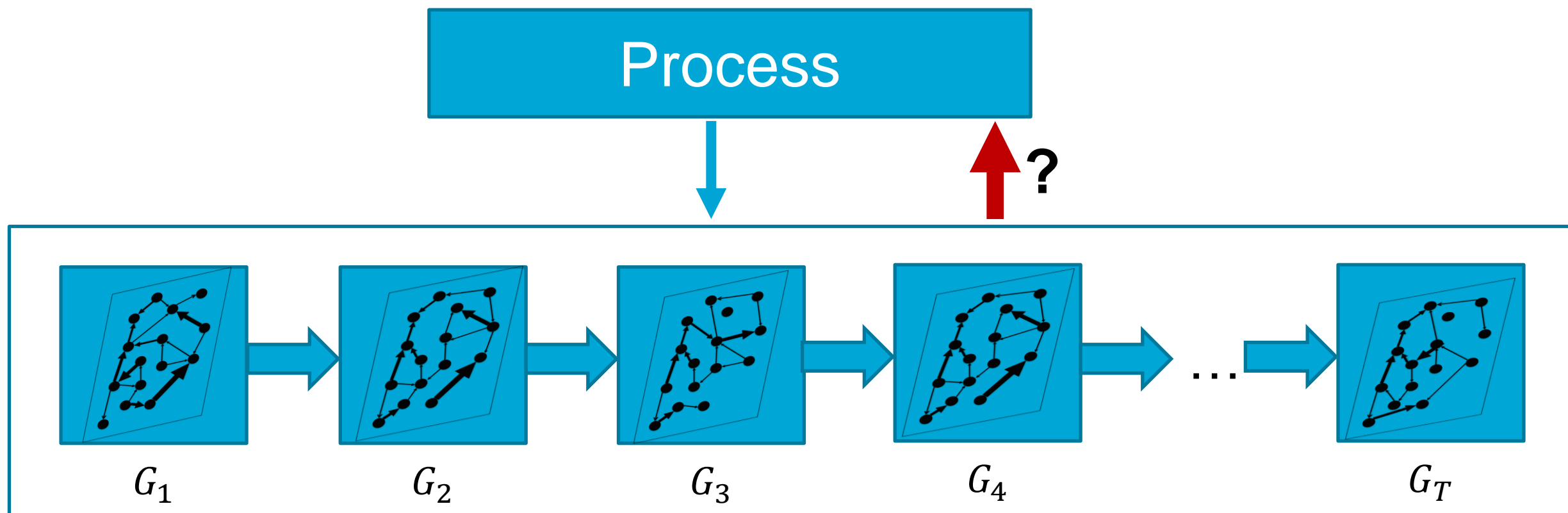
# “System Identification for Temporal Networks”

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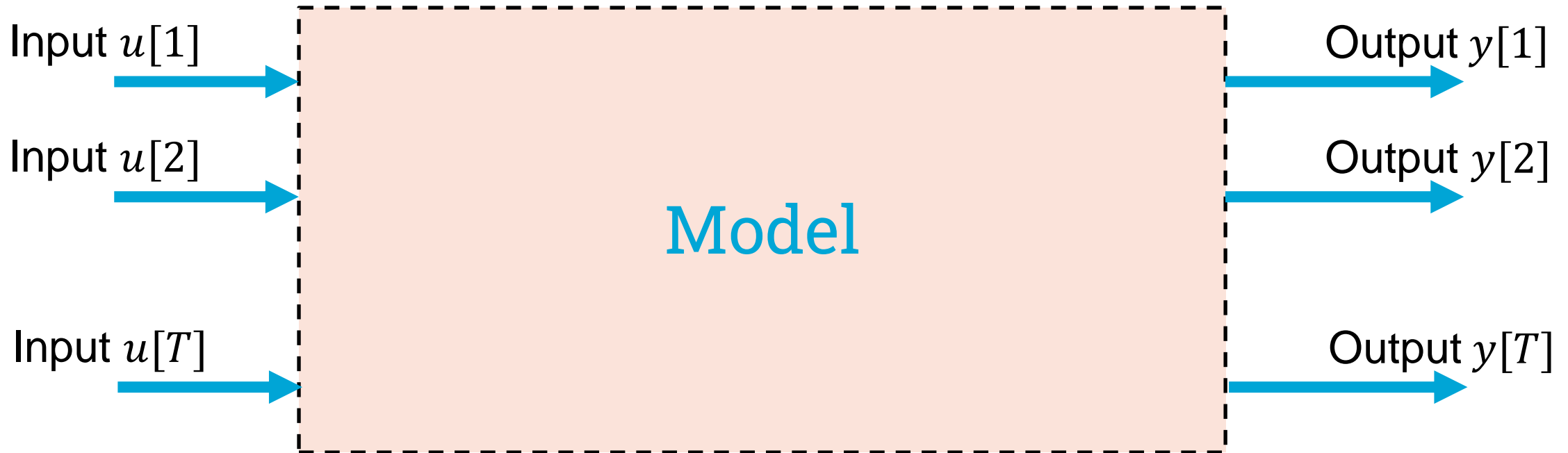
# Outline

- 1) Problem Statement;
- 2) Linear Time-Invariant (LTI) State-Space Model;
- 3) Application of the LTI Model to Time-Evolving Networks;
  - Periodic Graph Dynamics;
  - Non-Periodic Graph Dynamics.

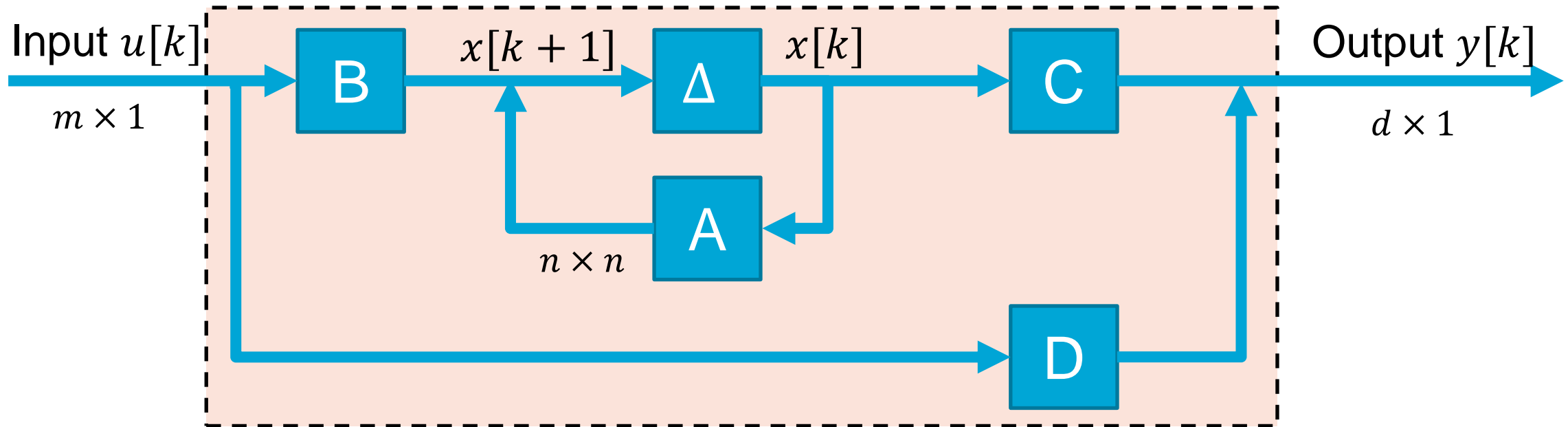
# Problem Statement



# Linear Time-Invariant (LTI) State Space Model



# Linear Time-Invariant (LTI) State Space Model



- $x[k]$  – state vector at discrete time  $k$ ;
- $\Delta$  – delay;

$$\begin{bmatrix} x[k+1] \\ y[k] \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} \cdot \begin{bmatrix} x[k] \\ u[k] \end{bmatrix} = Q \cdot \begin{bmatrix} x[k] \\ u[k] \end{bmatrix}$$

# Subspace Method for System Identification

**Compact form of the equation is**

$$Y_{1,s,N} = \Gamma_s X_{1,N} + H_s U_{1,s,N}$$

where  $Y_{1,s,N}$ ,  $U_{1,s,N}$  - block Hankel matrices:

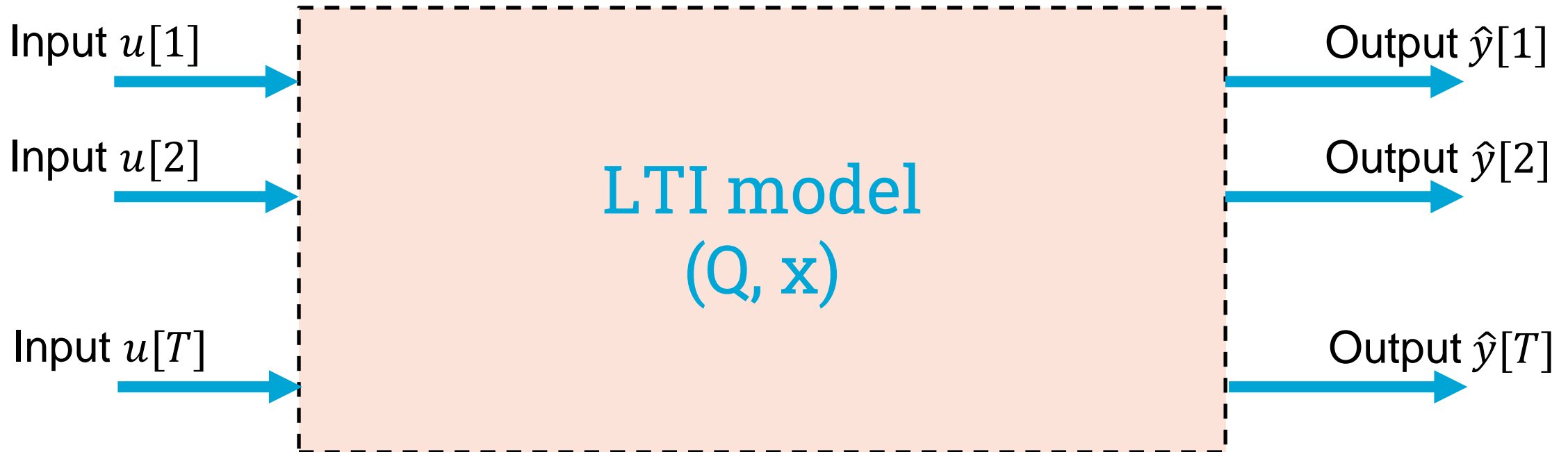
$$Y_{1,s,N} = \begin{pmatrix} y[1] & y[2] & \cdots & y[N] \\ y[2] & y[3] & \cdots & y[N+1] \\ \vdots & \vdots & \ddots & \vdots \\ y[s] & y[s+1] & \cdots & y[N+s] \end{pmatrix}, X_{1,N} = (x[1] \quad \cdots \quad x[N]), U_{1,s,N} = \begin{pmatrix} u[1] & u[2] & \cdots & u[N] \\ u[2] & u[3] & \cdots & u[N+1] \\ \vdots & \vdots & \ddots & \vdots \\ u[s] & u[s+1] & \cdots & u[N+s] \end{pmatrix},$$

$s, N$  – parameters and  $\Gamma_s, H_s$  - observability and controllability matrices:

$$\Gamma_s = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{s-1} \end{bmatrix}, \quad H_s = \begin{bmatrix} D & 0 & \cdots & 0 \\ CB & D & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ CA^{s-2}B & CA^{s-1}B & \cdots & D \end{bmatrix}.$$

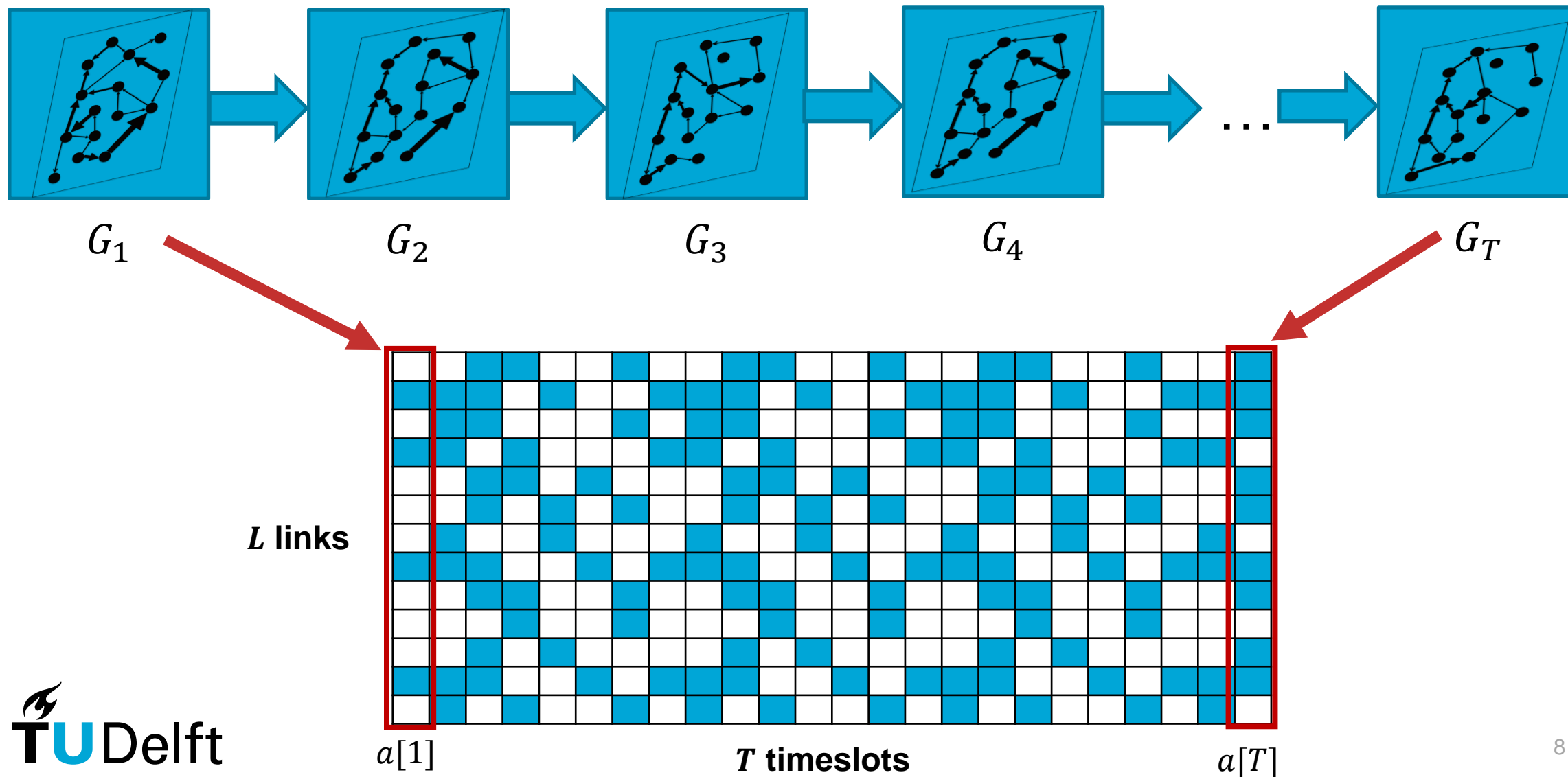
N4SID Algorithm: a **N**umerical algorithm for **s**ubspace **s**tate **s**pace **s**ystem **i**dentification, developed by Van Overschee & De Moor (1994).

# Linear Time-Invariant (LTI) State Space Model



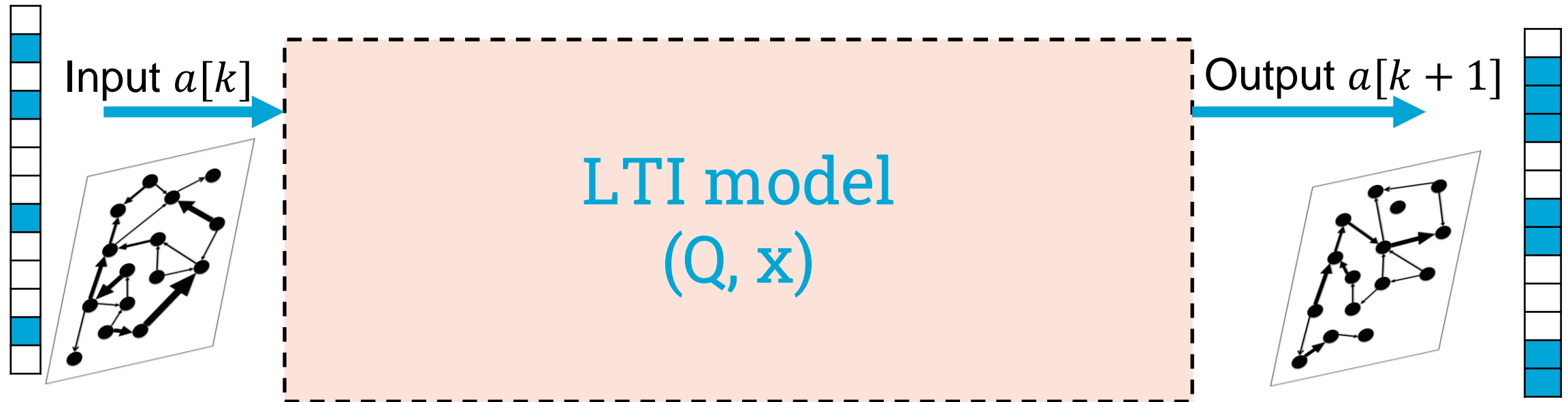
$$\begin{bmatrix} x[k+1] \\ \hat{y}[k] \end{bmatrix} = Q \cdot \begin{bmatrix} x[k] \\ u[k] \end{bmatrix}$$

# Graph Representation





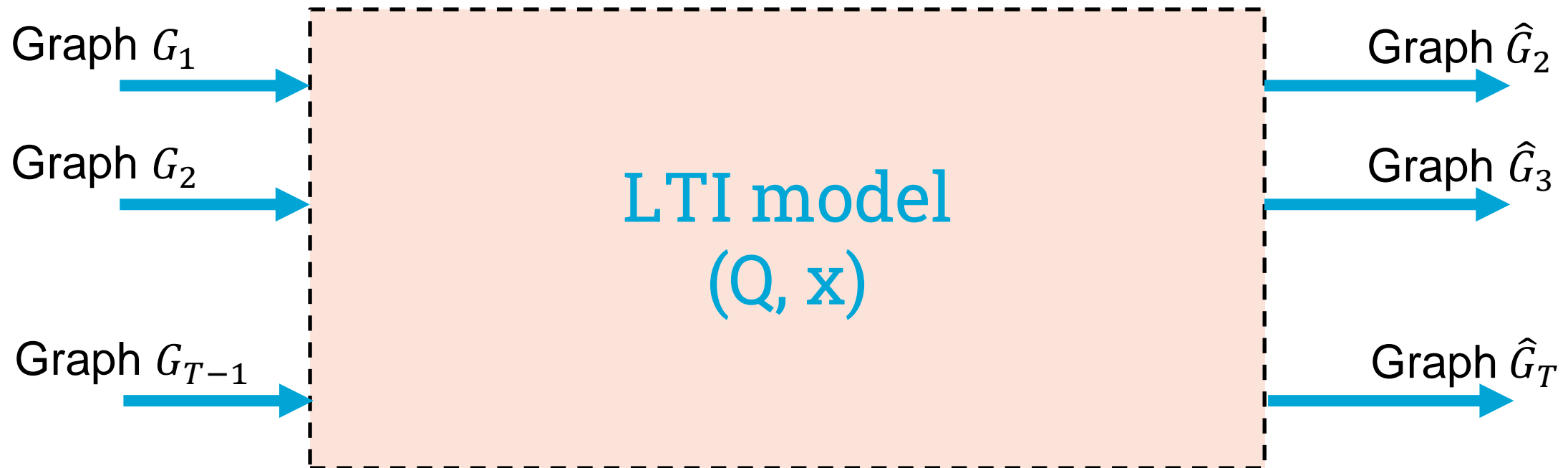
# Application to Time-Evolving Graphs



**Input:** Graph  $G_k$ ,  $L \times 1$  binary vector  $a[k]$ .

**Output:** Graph  $G_{k+1}$ ,  $L \times 1$  binary vector  $a[k + 1]$ .

# Linear Time-Invariant (LTI) State Space Model

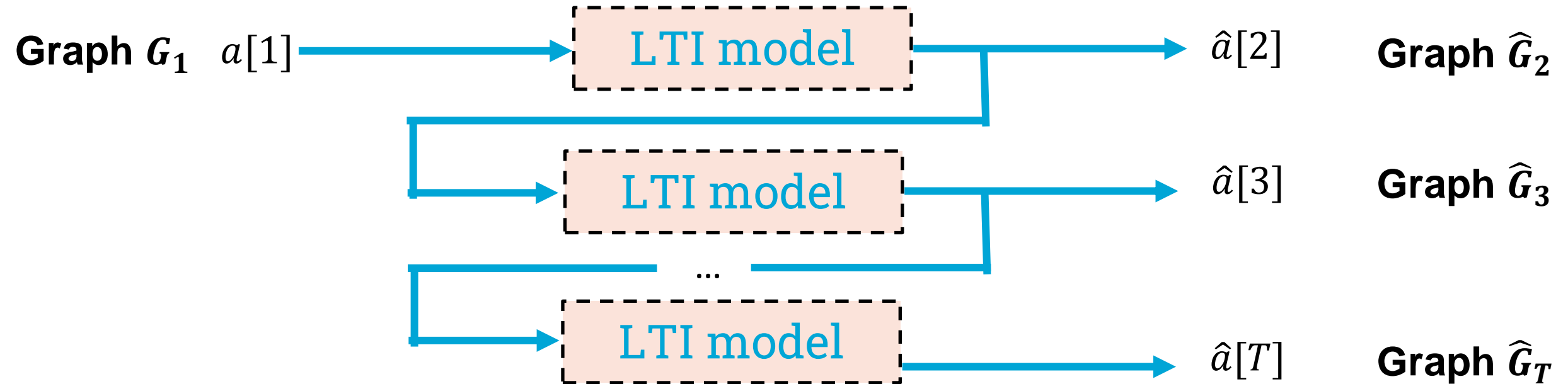


**Input:** Graph  $G_k$ ,  $L \times 1$  binary vector.

**Output:** Graph  $\hat{G}_{k+1}$ ,  $L \times 1$  binary vector.

# Application to Time-Evolving Graphs

## *Subspace Graph Generator (SG-Gen)*



$k$ -th output:

$$\begin{bmatrix} x[k+1] \\ \hat{a}[k+1] \end{bmatrix} = Q^k \begin{bmatrix} x[1] \\ a[1] \end{bmatrix}$$

# Performance of the Model

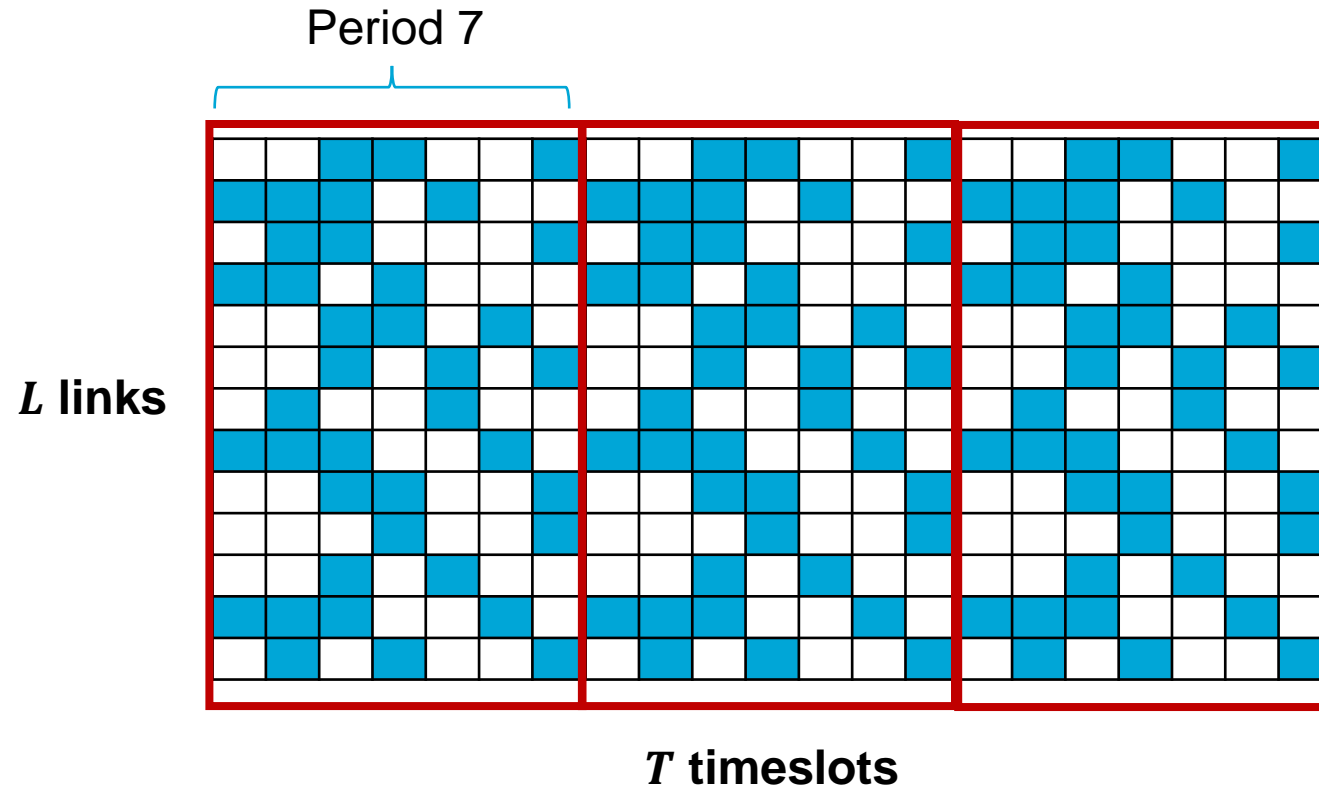
## Mean square error (MSE):

$$MSE(a, \hat{a}) = \frac{1}{T-1} \sum_{k=1}^{T-1} \sum_{i=1}^L (a_i[k] - \hat{a}_i[k])^2,$$

- $a[k]$  – real vector corresponding to the graph  $G_{k+1}$ ;
- $\hat{a}[k]$  – estimated vector corresponding to the graph  $G_{k+1}$ .
- $T$  – number of graphs,  $L$  – dimension of  $a[k]$  and  $\hat{a}[k]$ .

# Periodic Graph Dynamics

# Periodic Graph Dynamics



# Application to Periodic Graph Dynamics

- **Observation:** ANY periodic graph dynamic can be modelled accurately by SG-Gen:

$$\begin{bmatrix} x[k+1] \\ a[k+1] \end{bmatrix} = Q^k \begin{bmatrix} x[1] \\ a[1] \end{bmatrix}$$

Intuition:

Let  $v[k] = \begin{bmatrix} x[k] \\ a[k] \end{bmatrix}$  

If  $v[k]$  is periodic with period  $p$ , the SG-gen model can be rewritten as

$$[v[2] \ v[3] \ \dots v[p] \ v[1]] = Q \cdot [v[1] \ v[2] \ \dots v[p-1] \ v[p]].$$

# Application to Periodic Graph Dynamics

$$[v[2] \ v[3] \ \dots v[p] \ v[1]] = Q \cdot [v[1] \ v[2] \ \dots v[p-1] \ v[p]].$$

**Lemma.** If  $v[1], \dots, v[p]$  are linearly independent in  $\mathbb{R}^p$ , then any periodic graph sequence can be modelled by the LTI model with

$$Q = [v[1] \ v[2] \ \dots v[p-1] \ v[p]]^{-1} \cdot [v[2] \ v[3] \ \dots v[p] \ v[1]]$$



# Application to Periodic Graph Dynamics

$$[v[2] \ v[3] \ \dots v[p] \ v[1]] = Q \cdot [v[1] \ v[2] \ \dots v[p-1] \ v[p]].$$

**Is it possible to map vectors  $v[1], \dots, v[p]$  onto  $\mathbb{R}^r$  with  $r < p$ ?**

- **YES** (for some graph dynamics).

# Application to Periodic Graph Dynamics

**Theorem.** Let  $v[1], \dots, v[p]$  be an  $r \times 1$  vectors with  $r \leq p$ . Then the minimal order  $r$  of the system matrix  $Q$  is defined as

$$r = \text{rank} \begin{bmatrix} a[1] & a[2] & \cdots & a[p-1] & a[p] \\ a[2] & a[3] & & a[p] & a[1] \\ \vdots & & \ddots & & \vdots \\ a[p] & a[1] & \cdots & a[p-2] & a[p-1] \end{bmatrix}_{p \cdot L \times p},$$

where  $a[1], \dots, a[p]$  - input vectors corresponding to  $G_1, \dots, G_p$ .

**Algorithm:** Linear Periodic Graph Generator (LPG-gen)

# Non-Periodic Graph Dynamics

# Non-Periodic Graph Dynamics

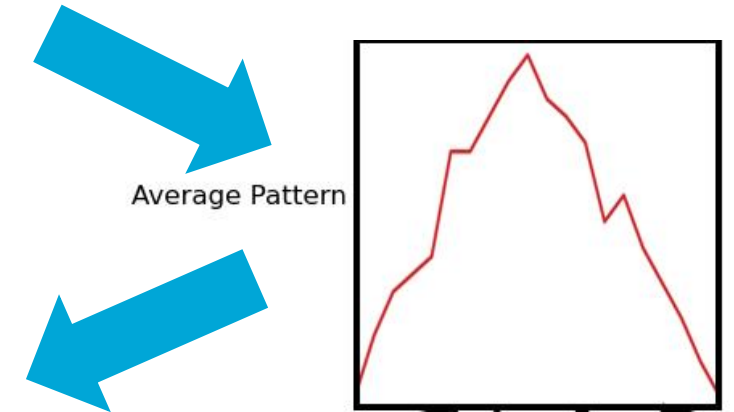
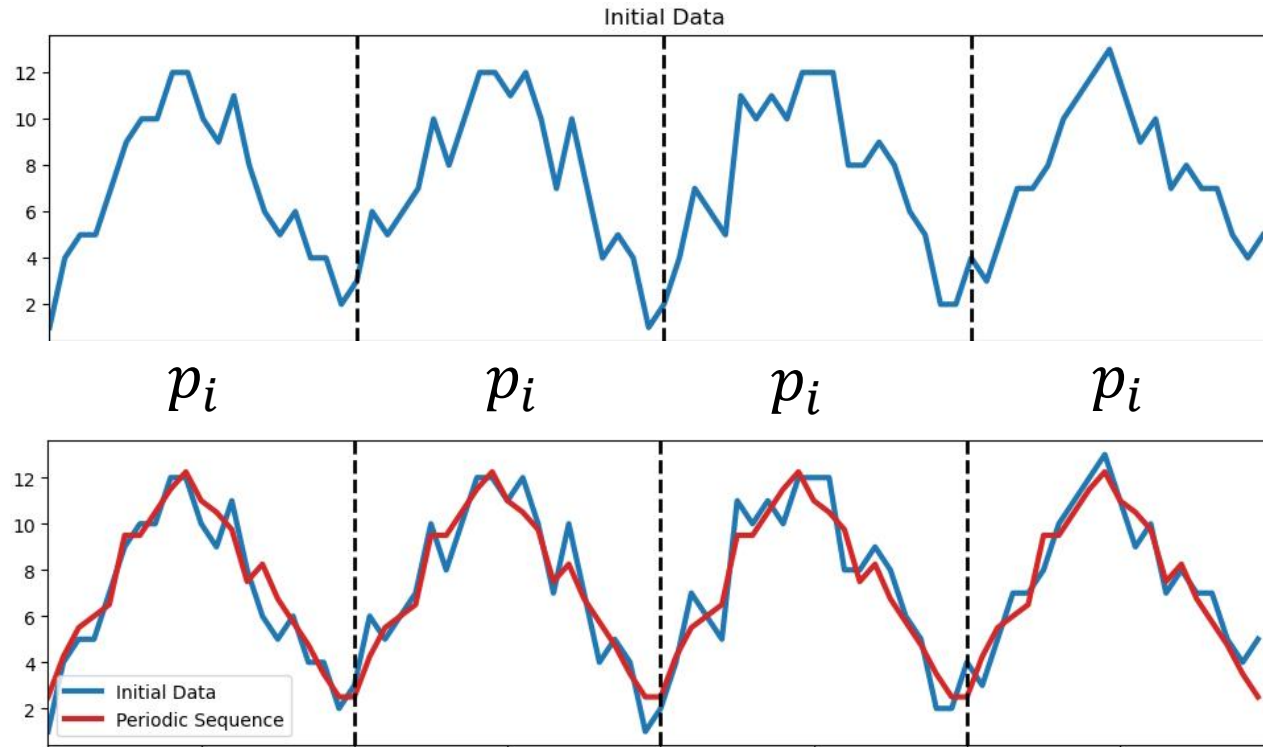
What if the graph sequence  $G_1, \dots, G_T$  is not periodic?

- **Linear Graph Generator (LG-gen).**

Idea:

- $G_1^{(i)}, G_2^{(i)}, \dots, G_T^{(i)}$  a  $i$ -th periodic graph sequence with period  $p_i$ ;
- *Construct*  $l$  periodic graph sequences such that for  $\forall k = 1, \dots, T$ 
$$G_k \approx G_k^{(1)} + G_k^{(2)} + \dots + G_k^{(l)}.$$
- Each periodic graph sequence is modelled by LPG-gen.

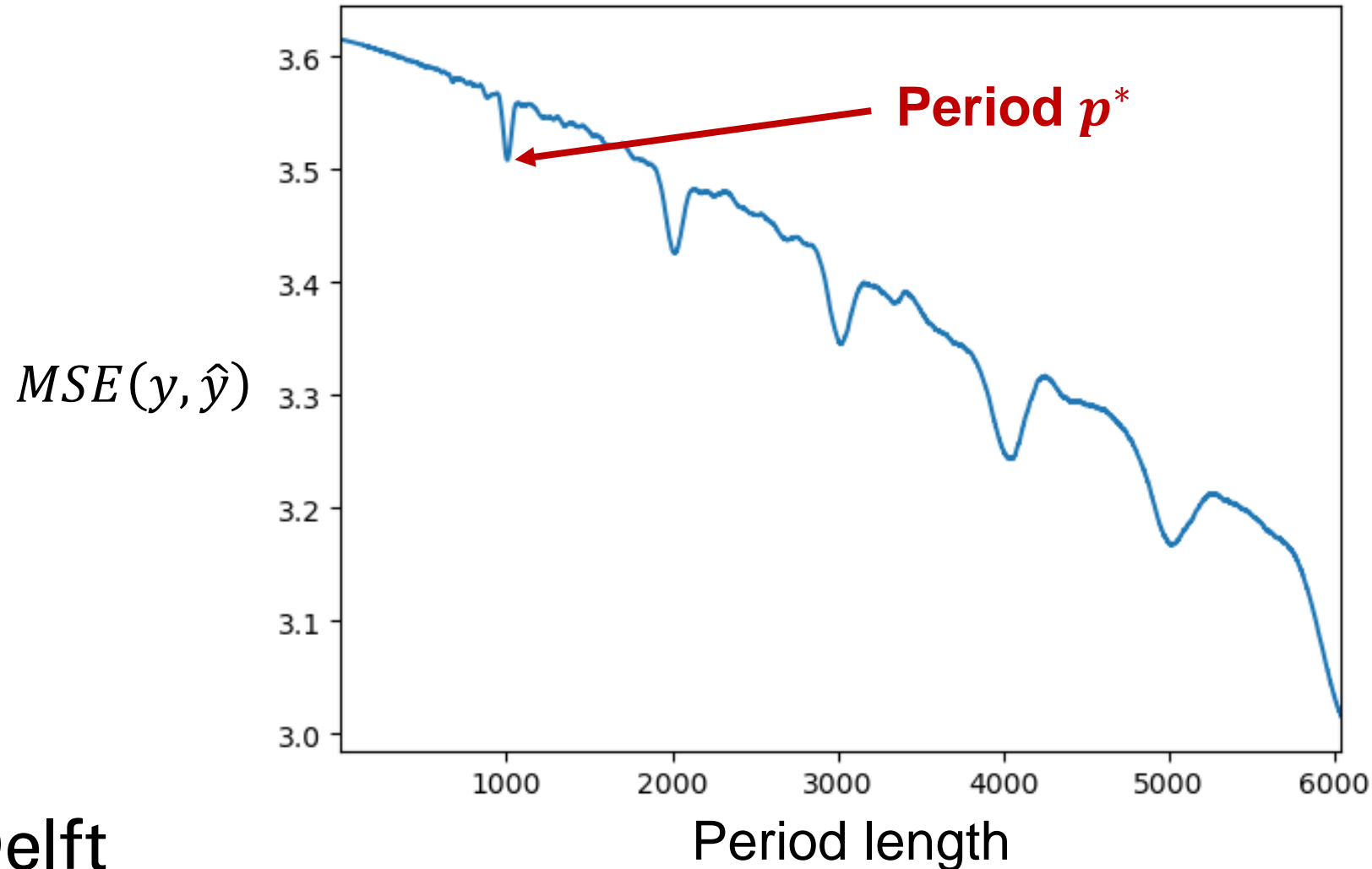
# LG-gen: Periodic Transform



$$x_i = x_{i-1} - \pi(x_{i-1}, \mathcal{P}(x_{i-1}, \mathcal{P}_{p_i})),$$

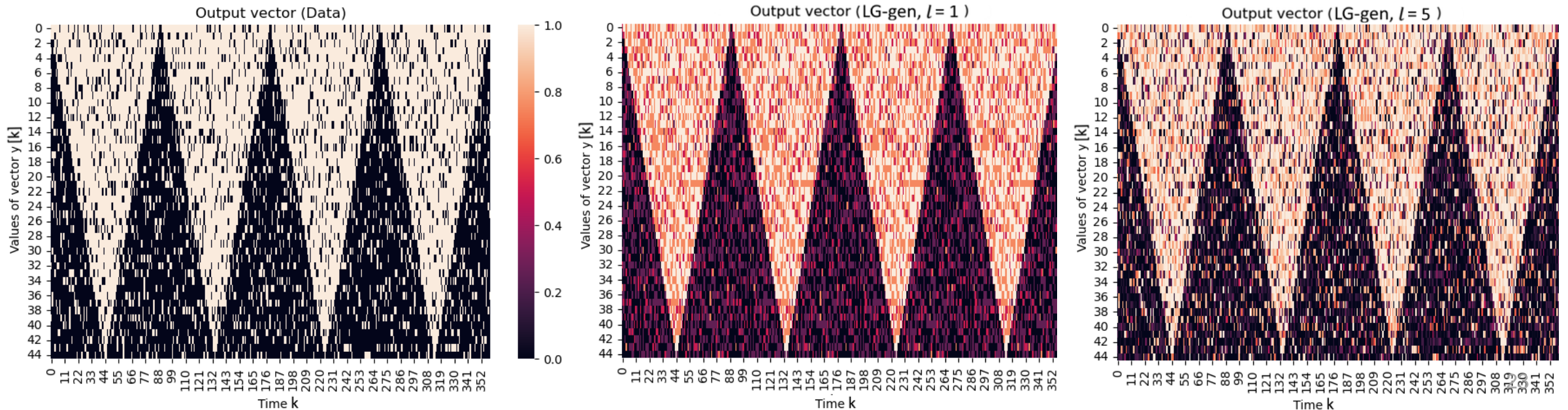
- $\pi(x_{i-1}, \mathcal{P}(x_{i-1}, \mathcal{P}_{p_i}))$  – projection of  $x_{i-1}$  onto a  $p_i$ -periodic subspace  $\mathcal{P}_{p_i}$ ;
- $x_0 = [a[1], a[2], \dots, a[T]]$  initial sequence.

# LG-gen: Identification of Periodic Sequences



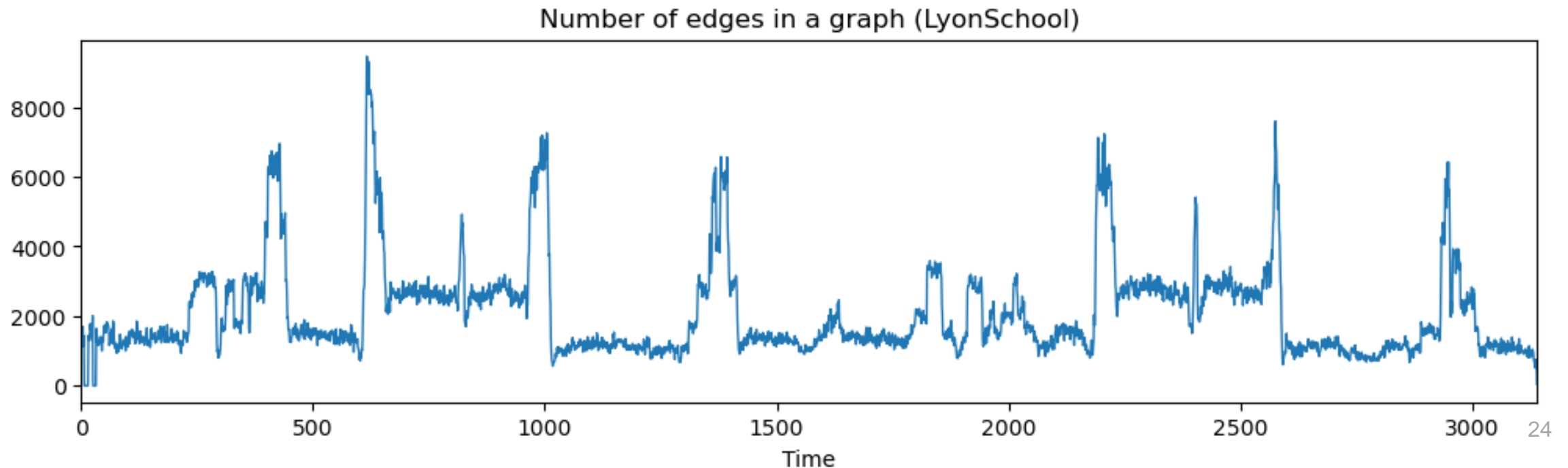
# Artificial Graph Dynamic: Results of LG-Gen

- Periodic dynamic with  $r$  random changes per time slot.
- LG-gen provides a good performance;
- 7 periodic graph sequences are sufficient to provide an ideal performance of the model, i.e.,  $G_k = \text{round} \left( \sum_{i=1}^7 G_k^{(i)} \right)$ .



# Graph Dynamic: LyonSchool

- Contact events between 242 individuals (232 children and 10 teachers) during two days in October 2009.
- The total number of contacts is 6,594,492 (29,161 unique links).



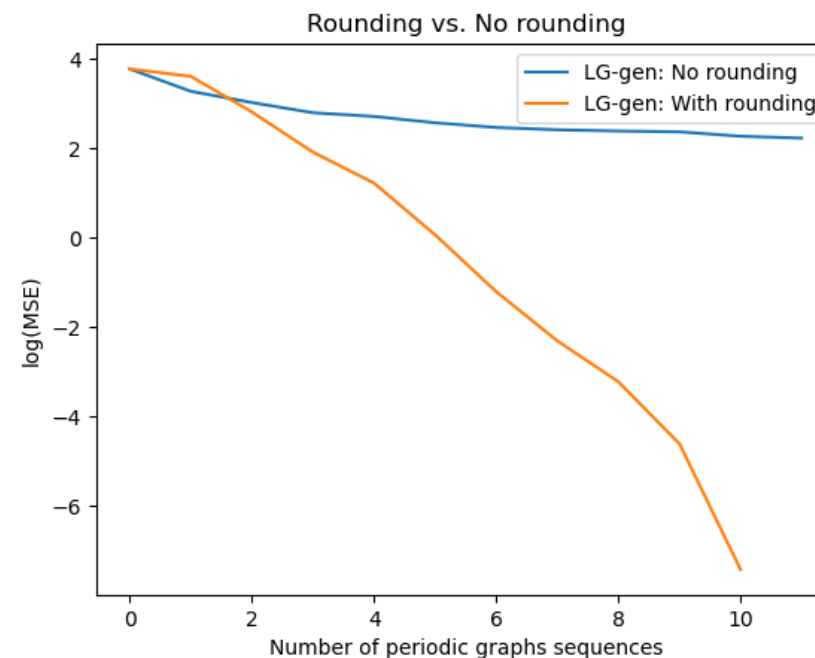


# Graph Dynamic 4: Results of LG-Gen

Type of model	Number of periodic graphs sequences l										
	1	2	3	4	5	6	7	8	9	10	11
No rounding	26.5	20.6	16.4	15.1	13.1	11.8	11.2	10.9	10.7	9.7	9.3
With rounding	37	16.7	6.8	3.4	1.07	0.3	0.1	0.04	0.01	0.0006	0

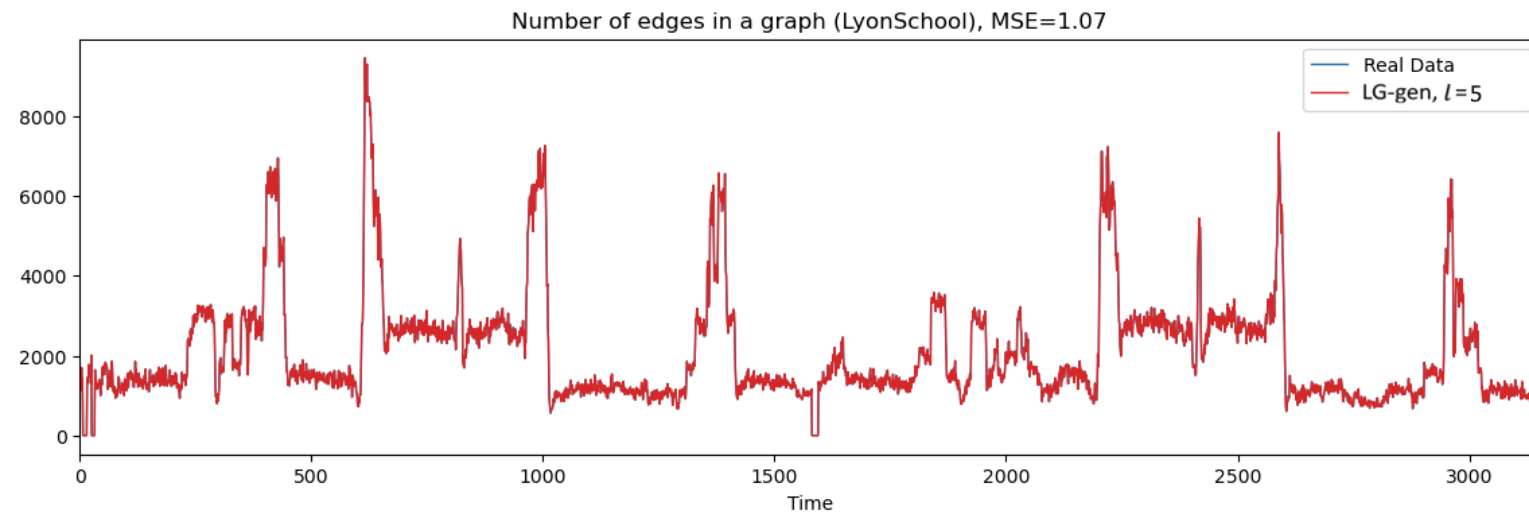
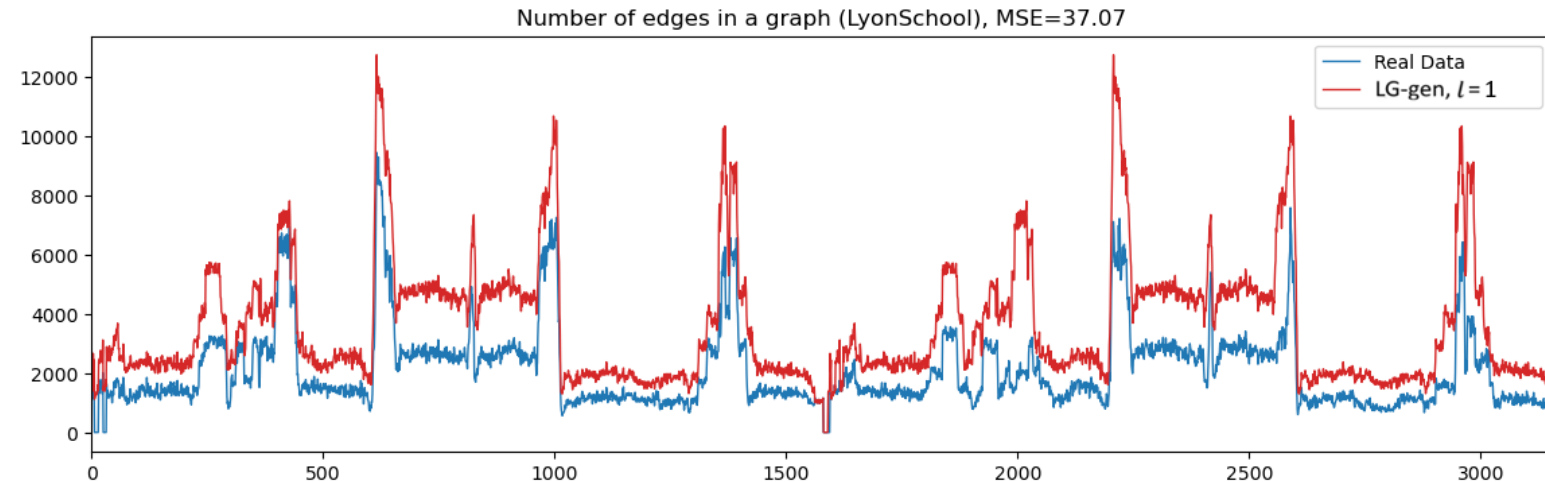
No rounding:  $G_k \approx G_k^{(1)} + G_k^{(2)} + \dots + G_k^{(l)}$

With rounding:  $G_k \approx \text{round} \left( G_k^{(1)} + G_k^{(2)} + \dots + G_k^{(l)} \right)$



# Graph Dynamic 4: Results of LG-Gen

- For  $l = 5$ , the MSE is 1.07 ( $\approx 1$  incorrect contact per timestamp).
- For  $l = 11$ , the LG-gen generates the LyonSchool network exactly.



# Results

- We modeled the graph dynamics as a *linear process*.
- Any periodic graph sequence can be modelled accurately by the LTI model.
- We proposed two algorithms, called LPG-gen and LG-gen, to model periodic and non-periodic graph sequences and discuss their efficiency.
- LPG-gen and LG-gen algorithms provided a good performance on various artificial and real graph sequences.

Thank you!