

“System Identification for Temporal Networks”

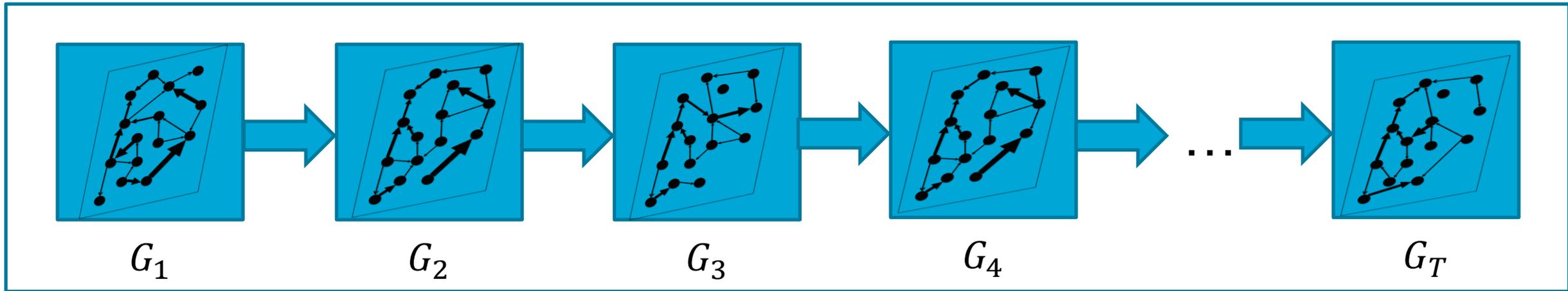
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Outline

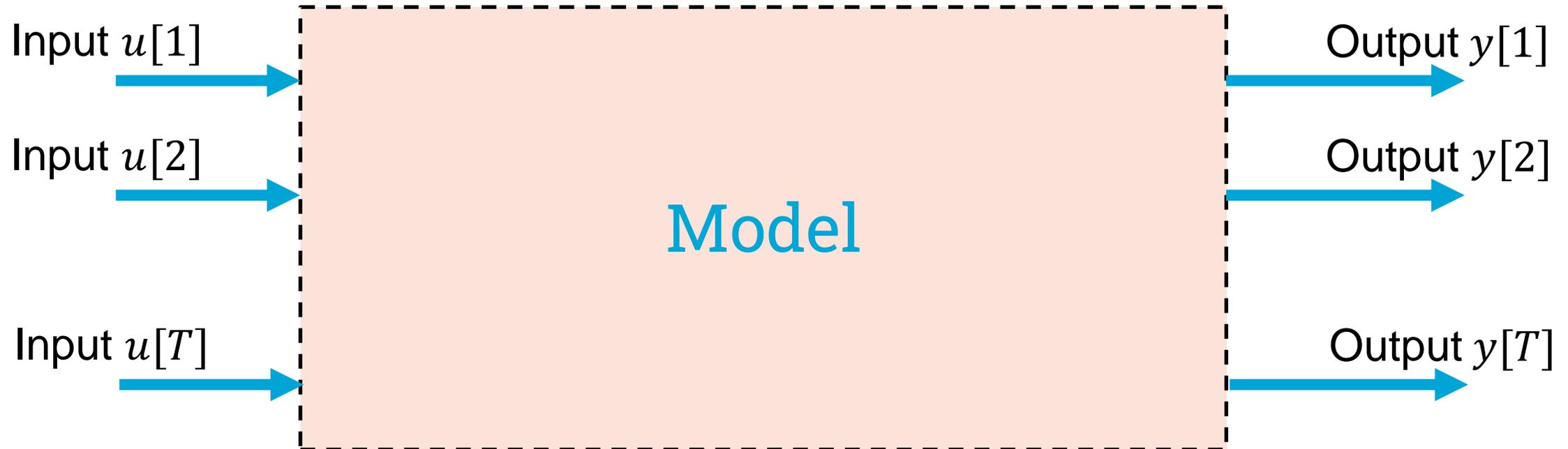
- 1) Problem Statement;
- 2) Linear Time-Invariant (LTI) State-Space Model;
- 3) Application of the LTI Model to Time-Evolving Networks;
 - Periodic Graph Dynamics;
 - Non-Periodic Graph Dynamics.

Problem Statement

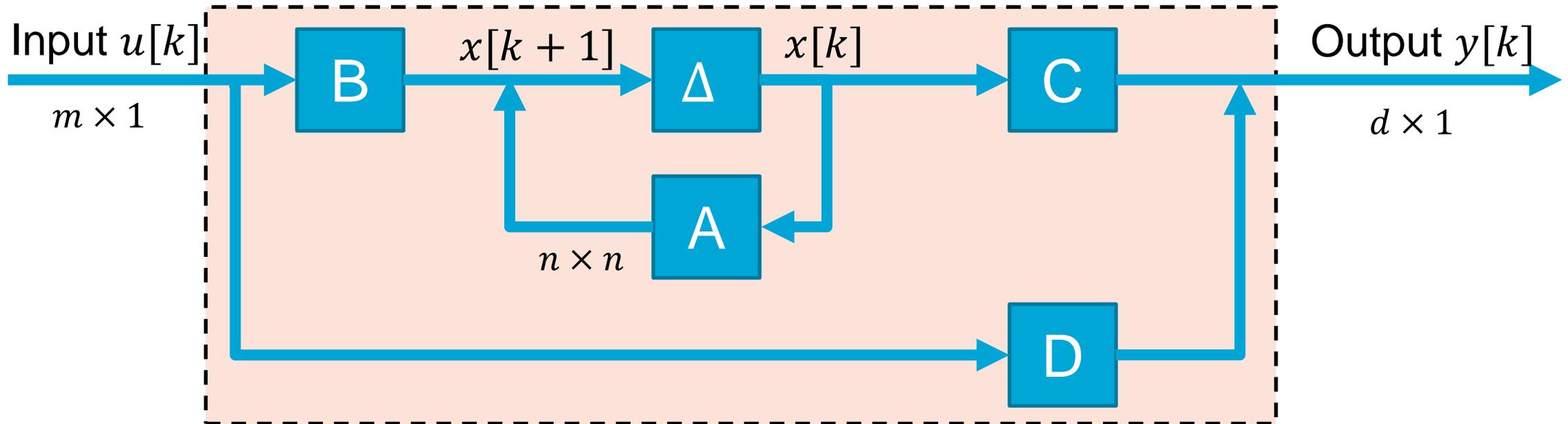
Process



Linear Time-Invariant (LTI) State Space Model



Linear Time-Invariant (LTI) State Space Model



- $x[k]$ – state vector at discrete time k ;
- Δ – 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 $Y_{i,s,N}, U_{i,s,N}$ - block Hankel matrices:

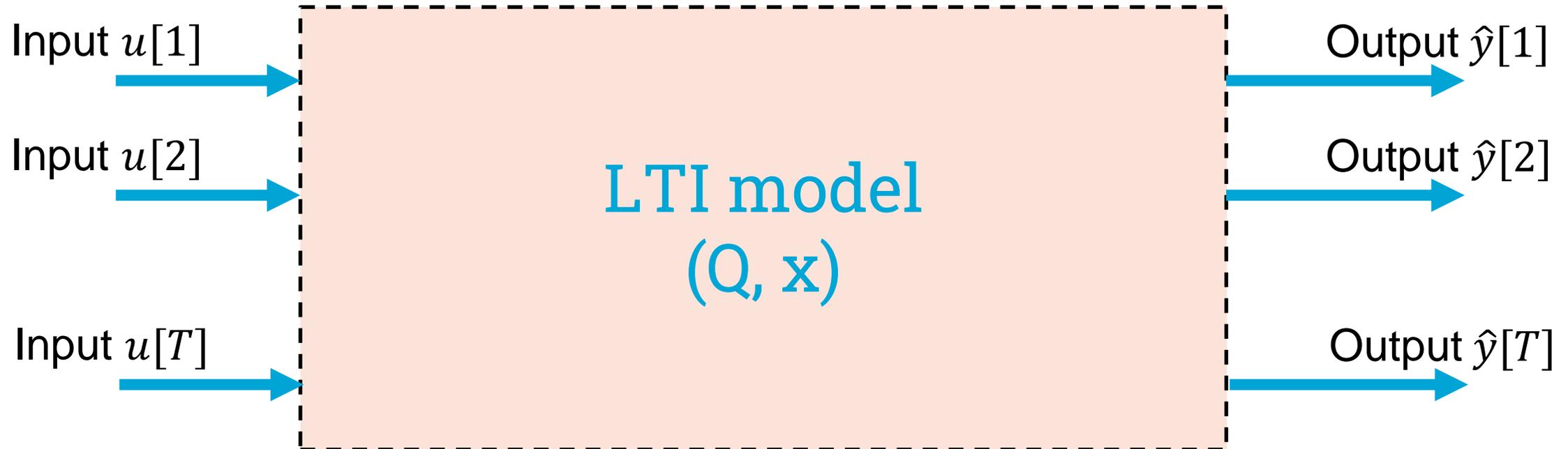
$$Y_{1,s,N} = \begin{pmatrix} y[1] & y[2] & \cdots & y[N] \\ y[2] & y[3] & \cdots & y[N+1] \\ \vdots & \vdots & \ddots & \vdots \\ y[s] & y[s+1] & \cdots & y[N+s] \end{pmatrix}, X_{1,N} = (x[1] \quad \cdots \quad x[N]), U_{1,s,N} = \begin{pmatrix} u[1] & u[2] & \cdots & y[N] \\ u[2] & u[3] & \cdots &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

s, N – parameters and Γ_s, H_s - observability and controllability matrices:

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

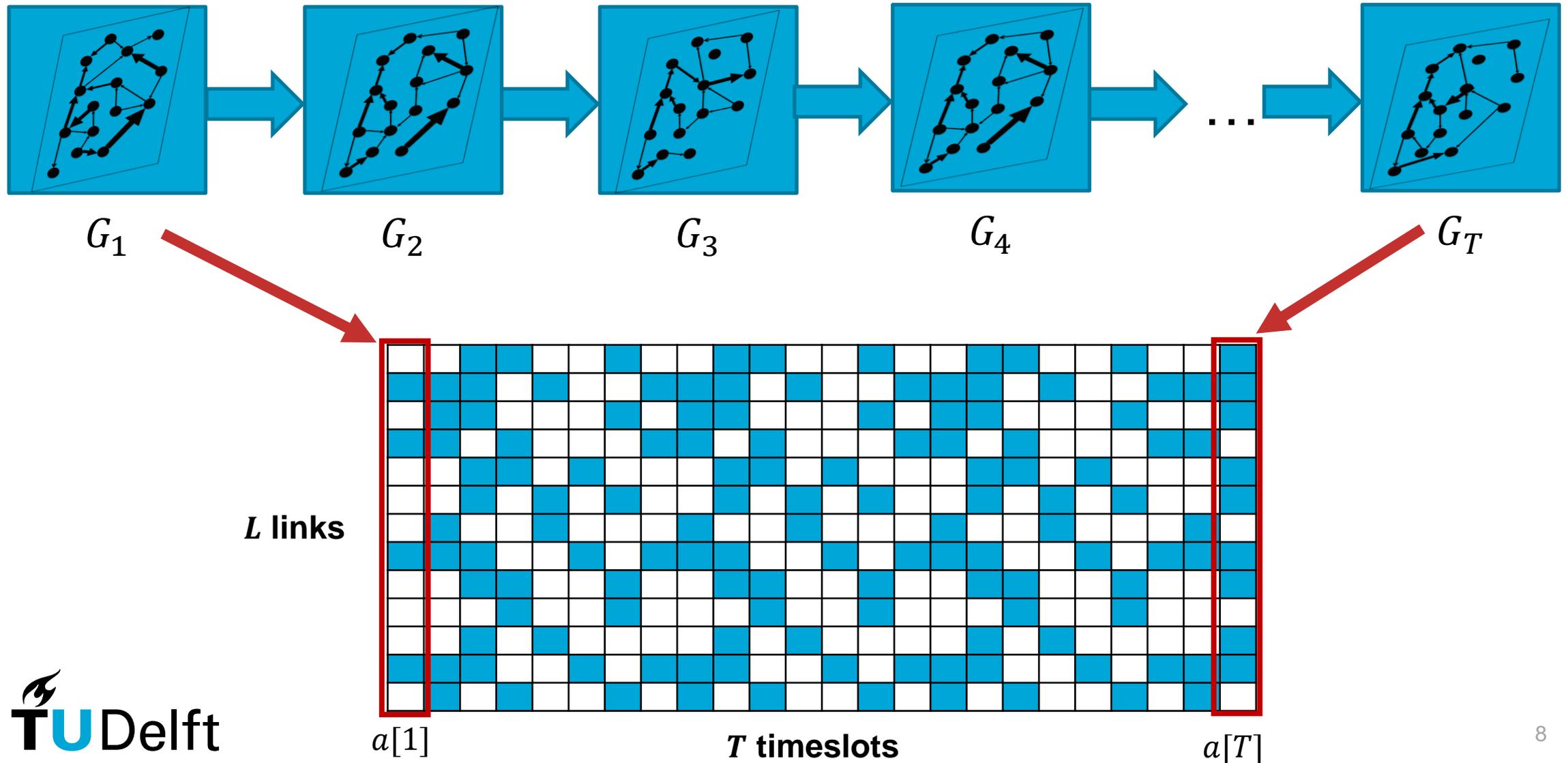
N4SID Algorithm: a **N**umerical algorithm for **s**ubspace **s**tate **s**pace **s**ystem **i**dentification, developed by Van Overschee & De Moor (1994).

Linear Time-Invariant (LTI) State Space Model

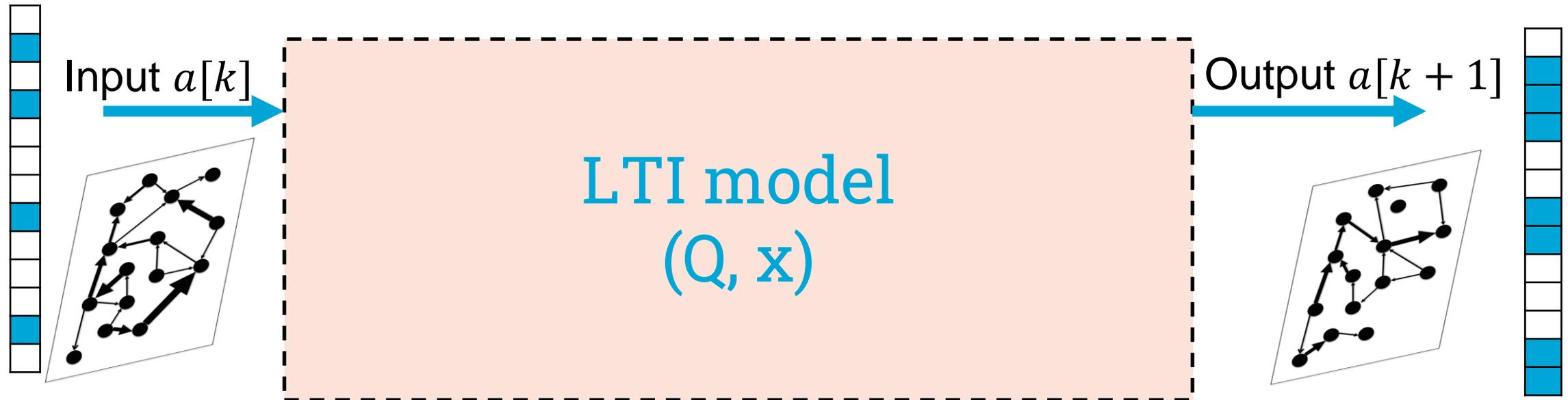


$$\begin{bmatrix} x[k+1] \\ \hat{y}[k] \end{bmatrix} = Q \cdot \begin{bmatrix} x[k] \\ u[k] \end{bmatrix}$$

Graph Representation



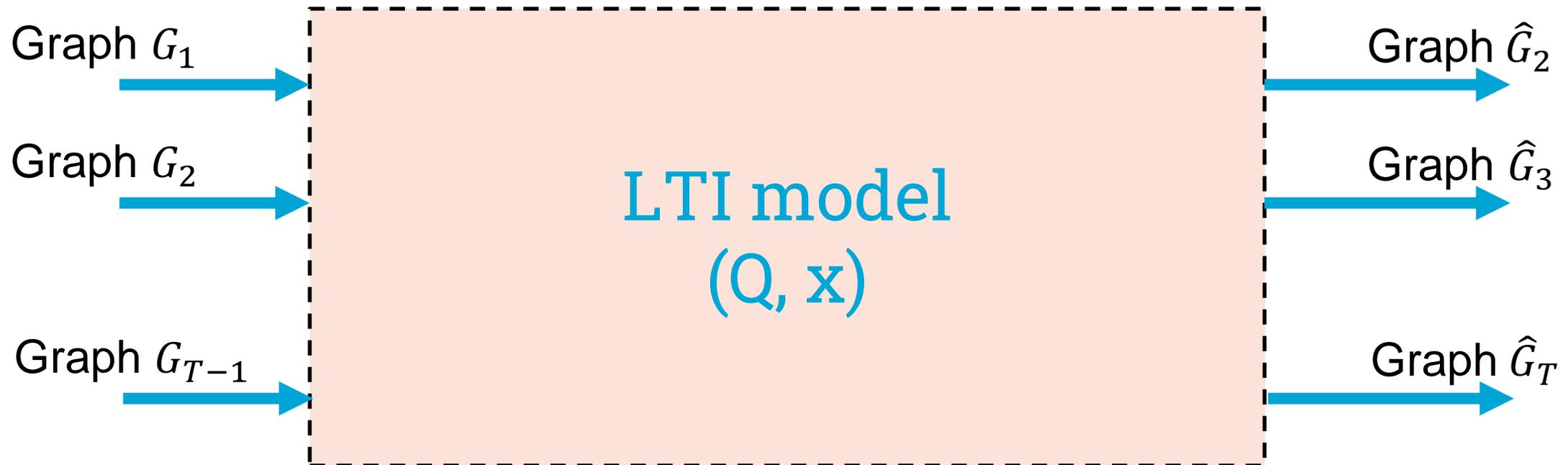
Application to Time-Evolving Graphs



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

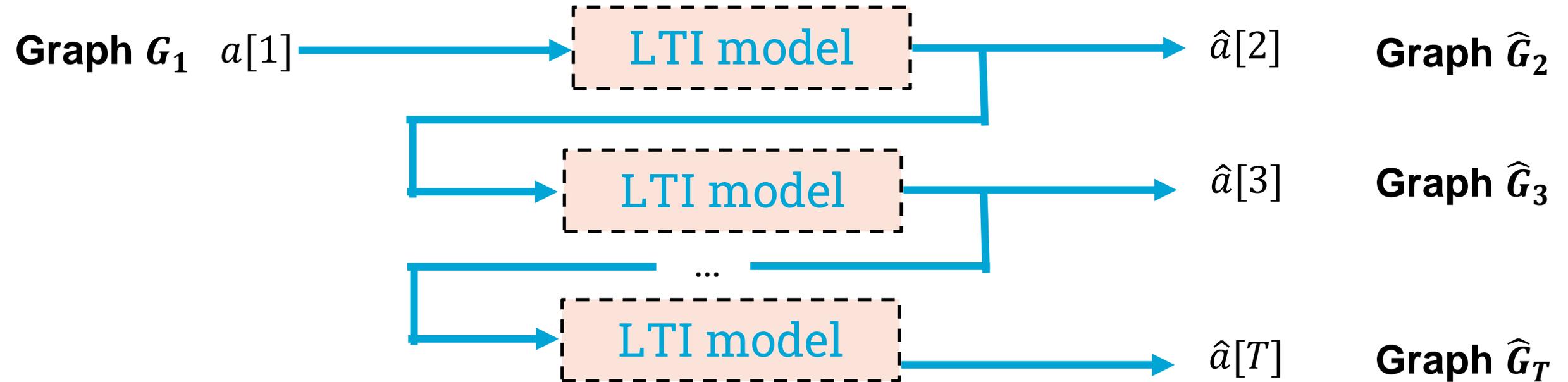


Input: Graph G_k , $L \times 1$ binary vector.

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

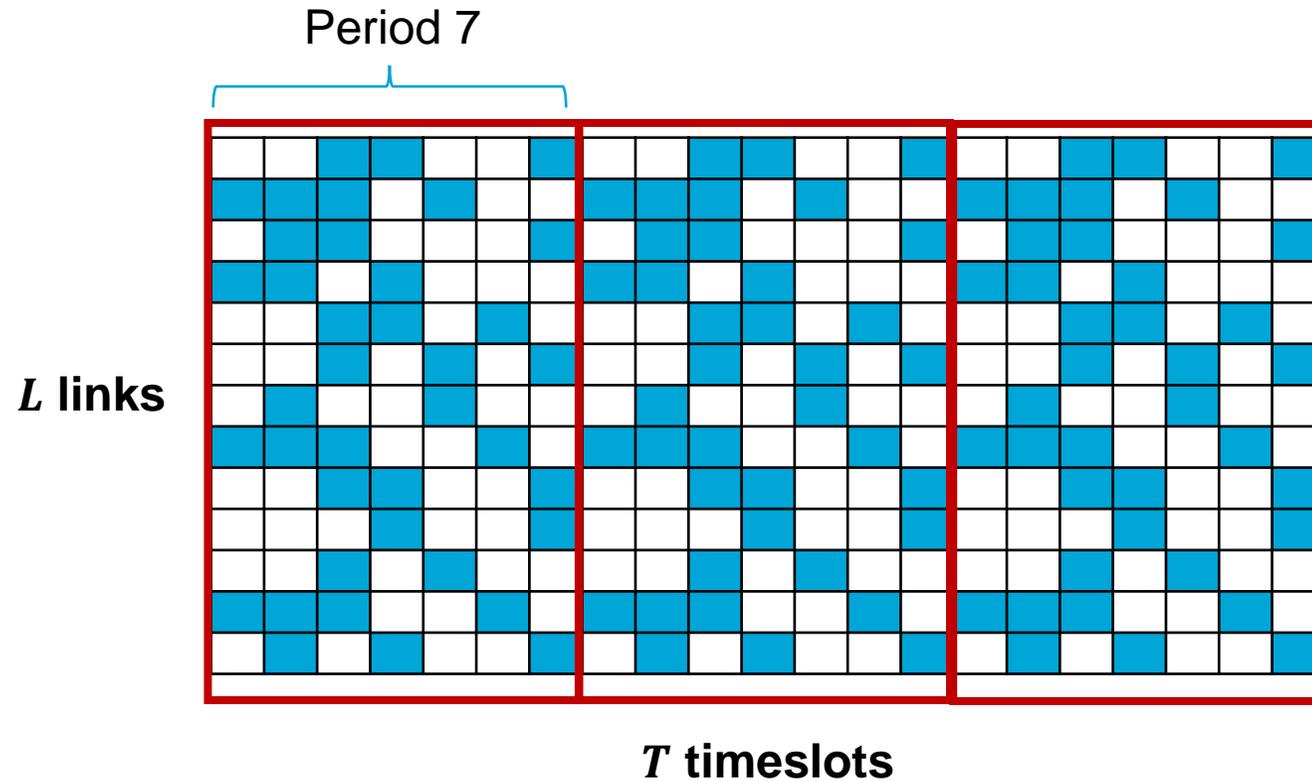
Mean square error (MSE):

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

- $a[k]$ – 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

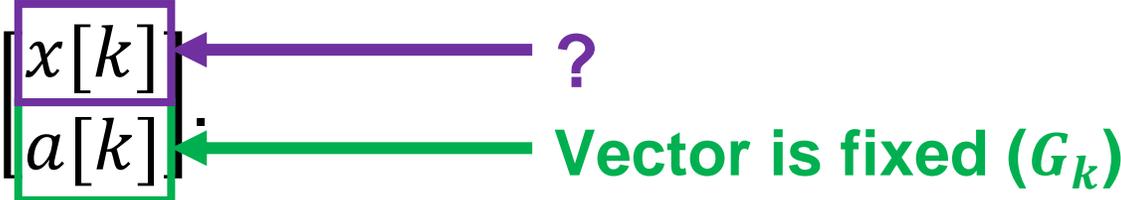


Application to Periodic Graph Dynamics

- **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

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

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

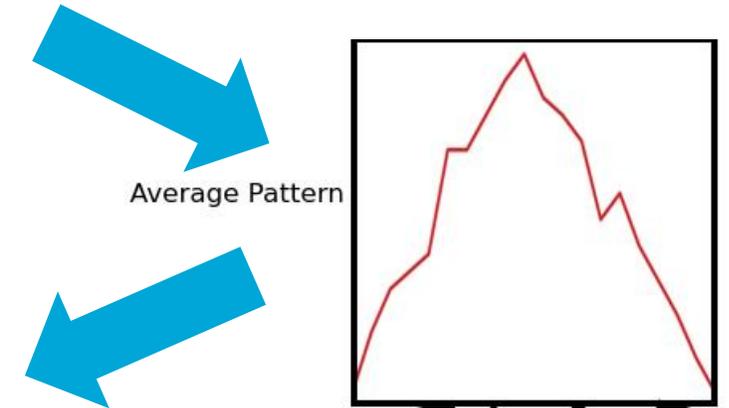
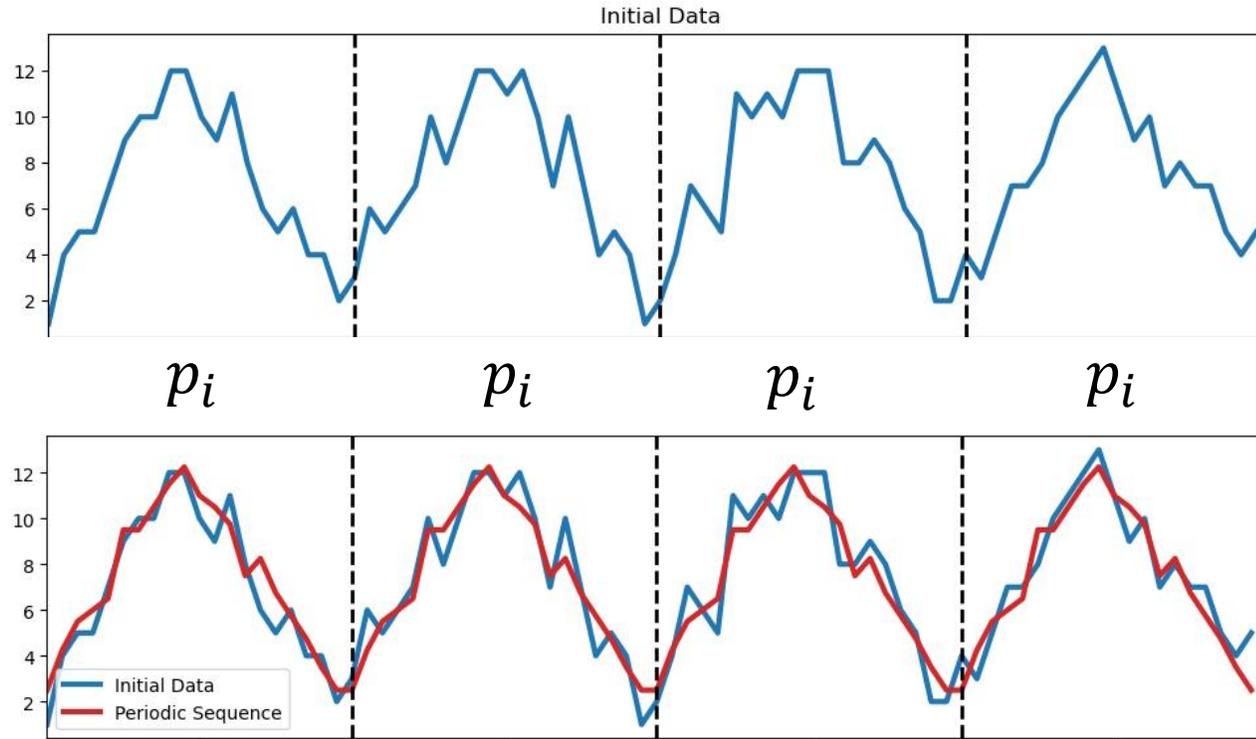
What if the graph sequence G_1, \dots, G_T is not periodic?

- **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.

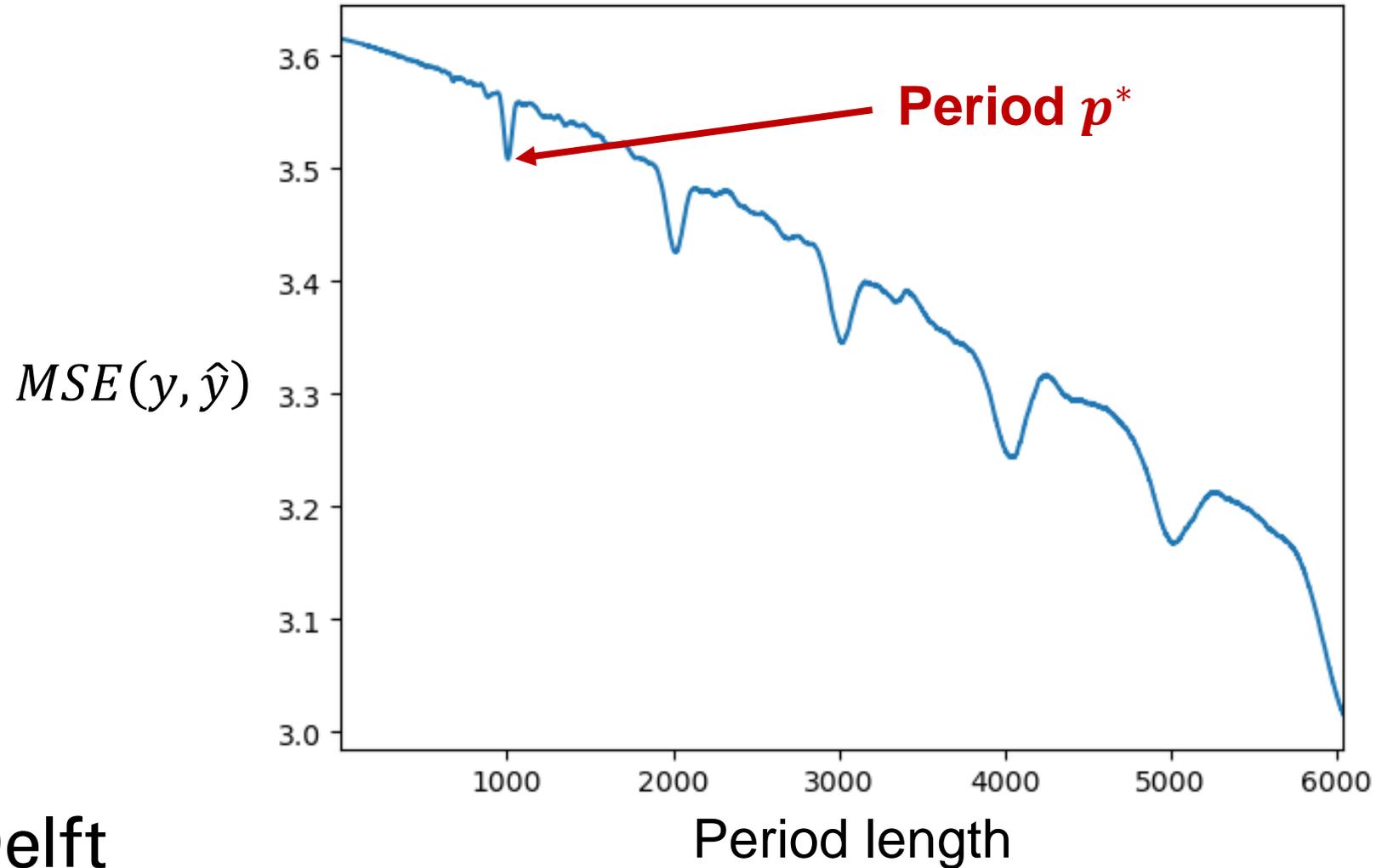
LG-gen: Periodic Transform



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

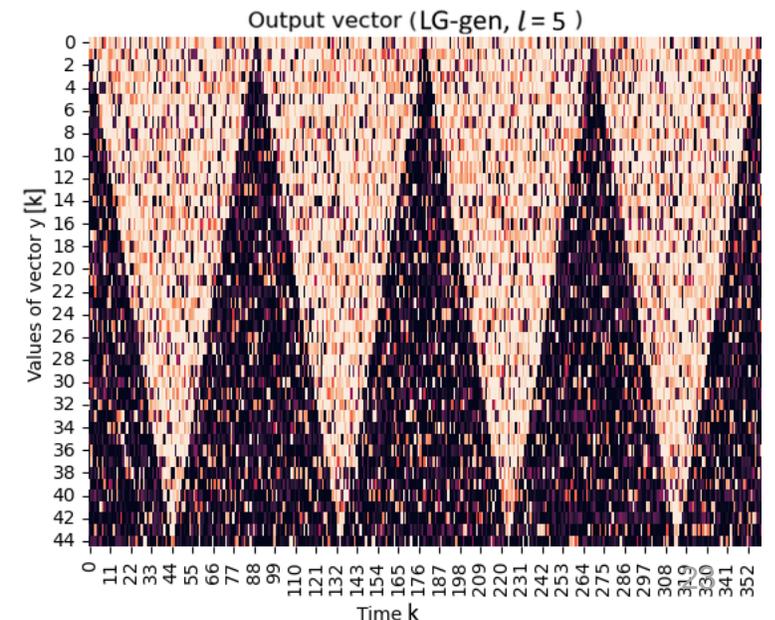
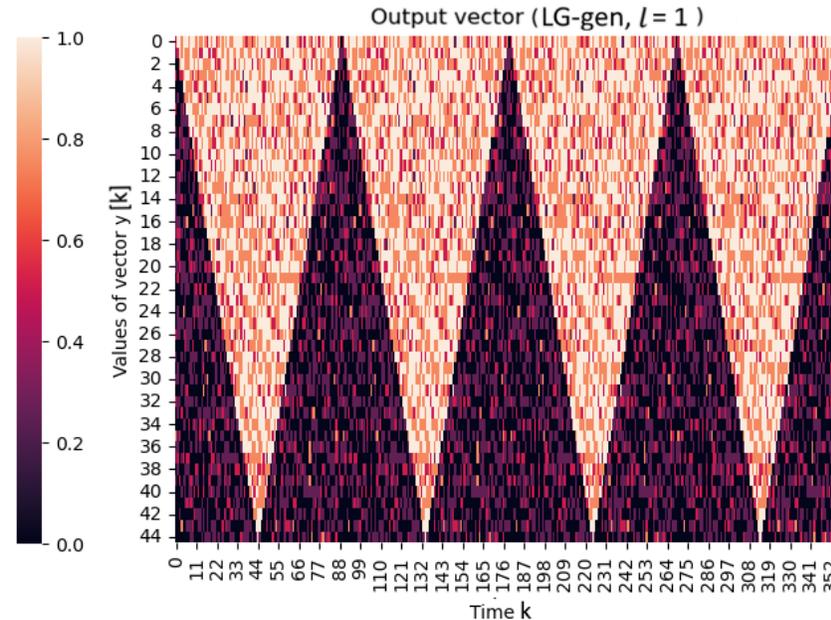
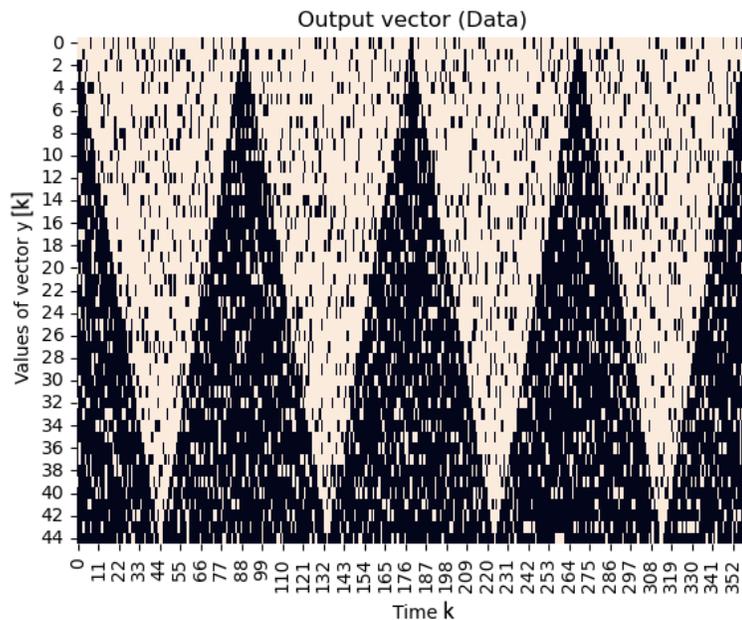
- $\pi(x_{i-1}, \mathcal{P}(x_{i-1}, \mathcal{P}_{p_i}))$ - projection of x_{i-1} onto a p_i -periodic subspace \mathcal{P}_{p_i} ;
- $x_0 = [a[1], a[2], \dots, a[T]]$ initial sequence.

LG-gen: Identification of Periodic Sequences



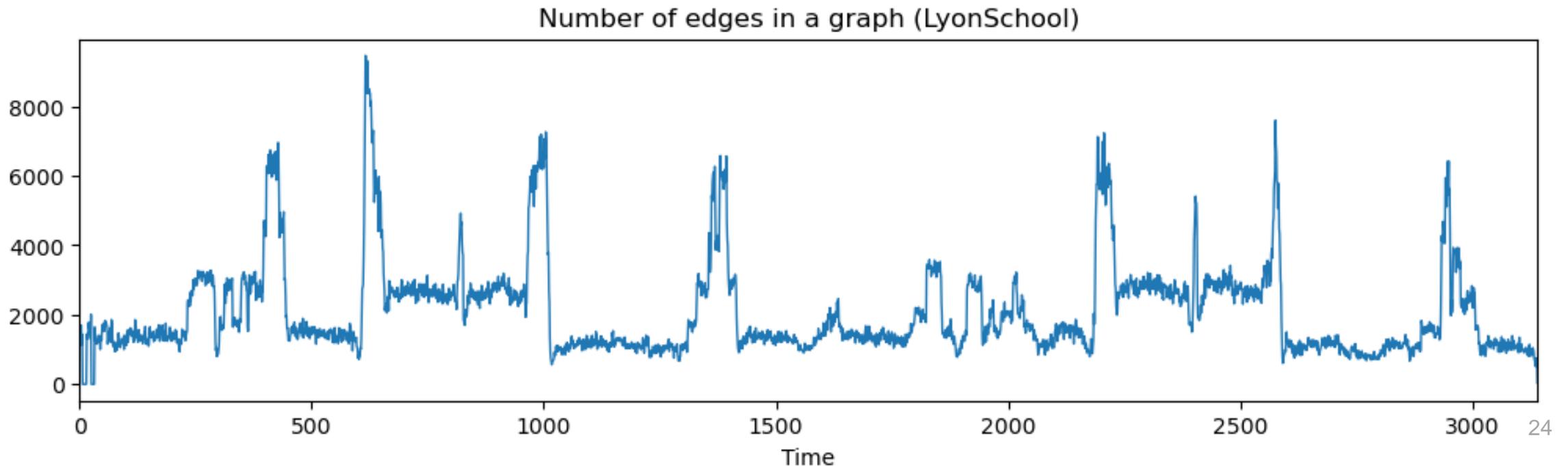
Artificial Graph Dynamic: Results of LG-Gen

- 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).

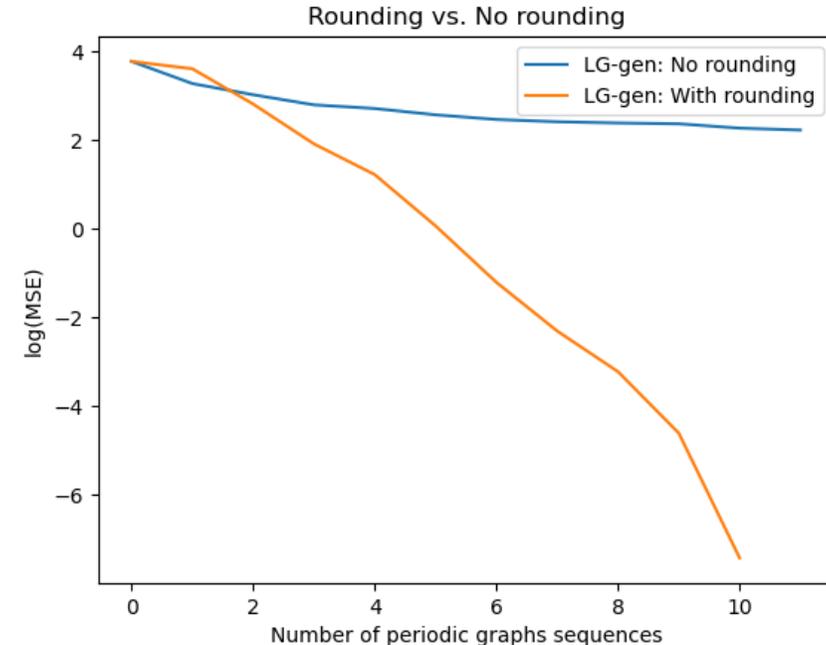


Graph Dynamic 4: Results of LG-Gen

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

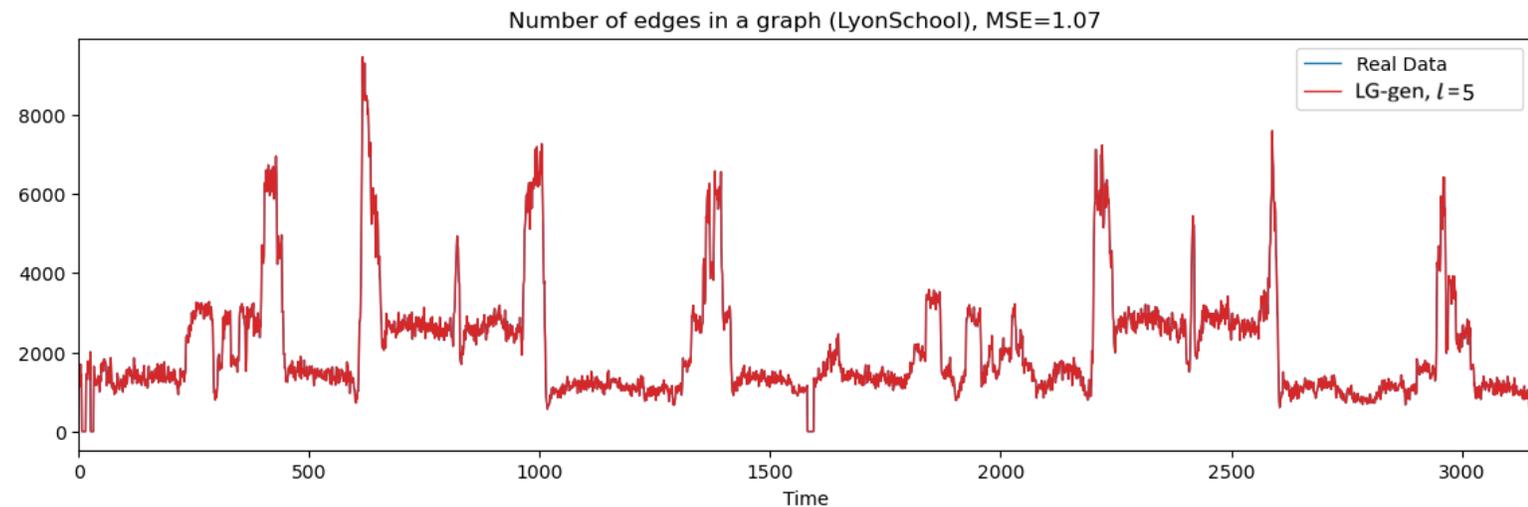
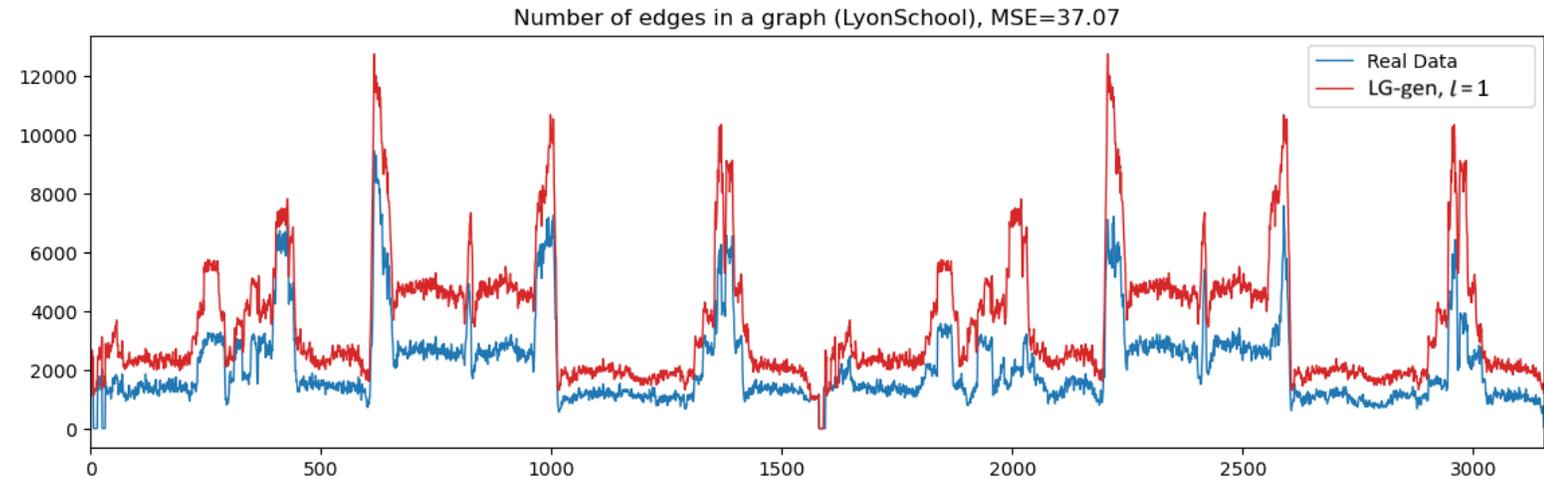
No rounding: $G_k \approx G_k^{(1)} + G_k^{(2)} + \dots + G_k^{(l)}$

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 (≈ 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!