

Background

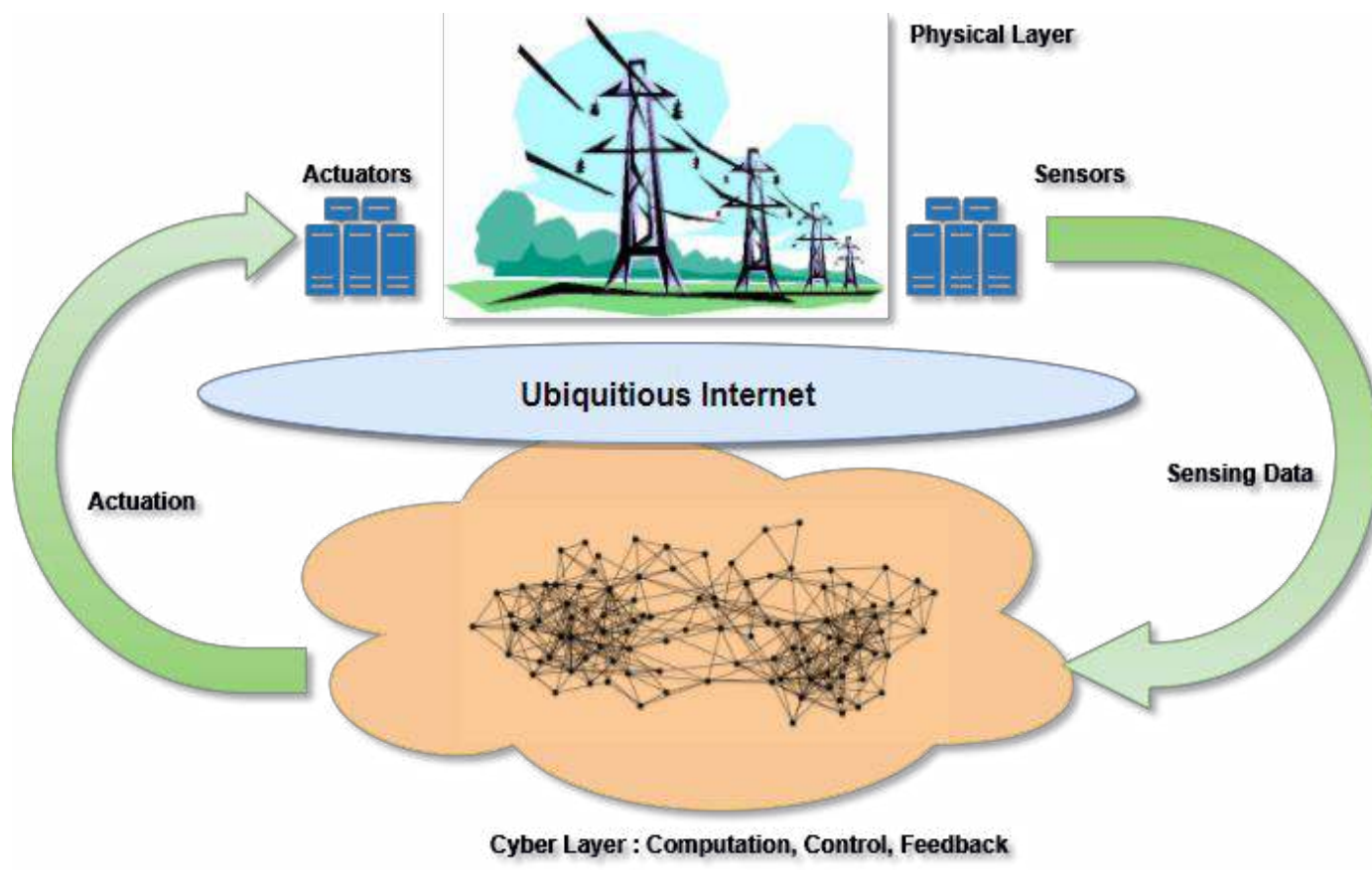
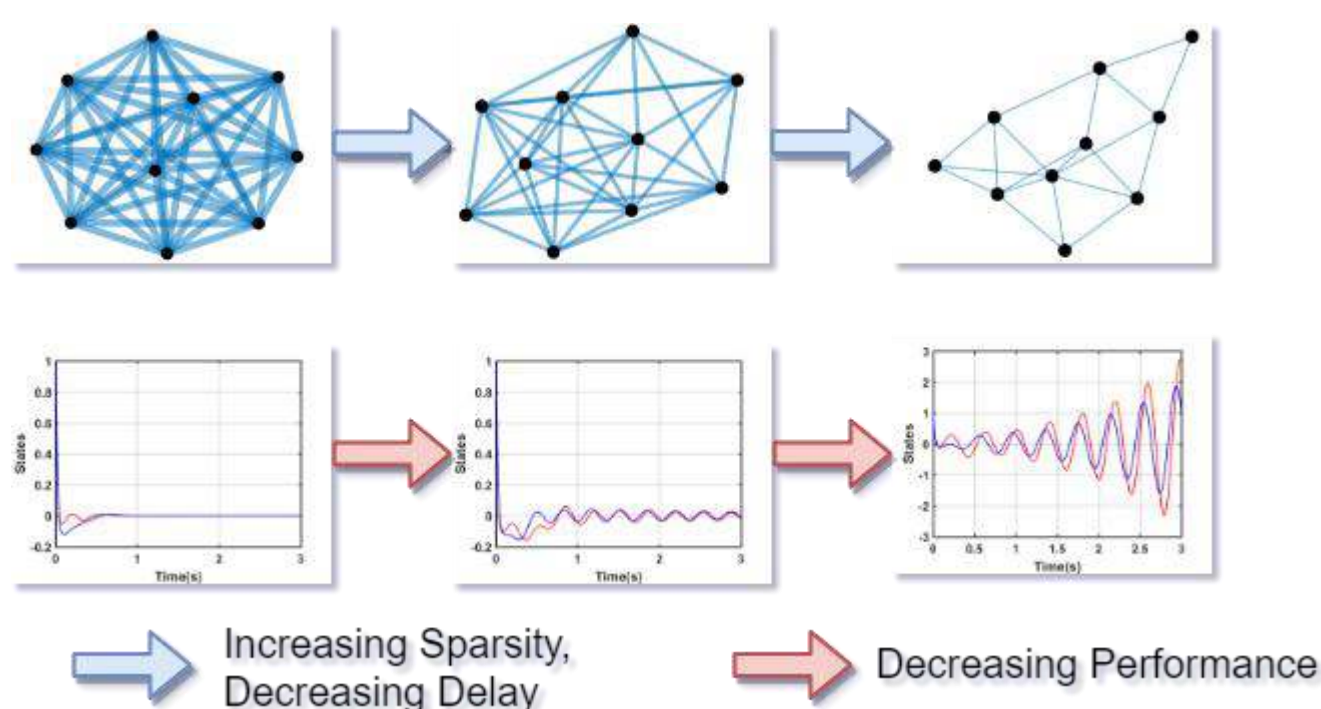


Fig: A Simplified Cyber-Physical System

- Integration of computation, networking and physical process.
- τ is the feedback delay. Control requires **all-to-all communication**.
- Result – Dense Communication network, Higher Infrastructure costs.
- Solution – ‘**Sparsify**’ the communication network. Gain matrix $K_{ij} \rightarrow 0$.
- Limited Bandwidth – leads to presence of feedback delays.
- Delay is a function of sparsity.
- Trade-off**: Increasing sparsity - degrades performance - decreases delay - improves performance

Problem Statement



- Closed loop delayed System:

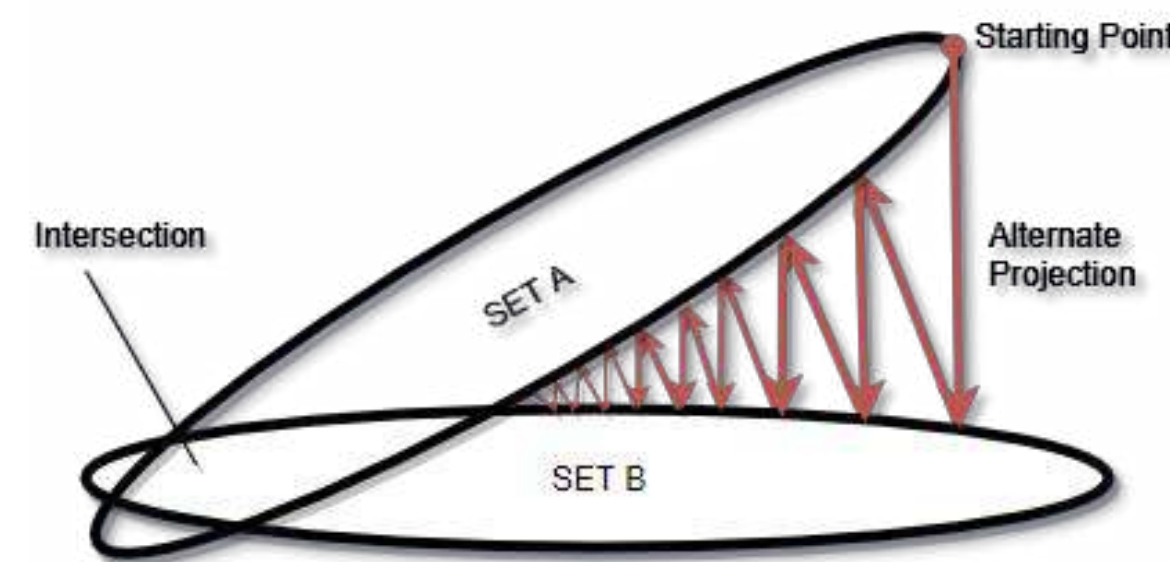
$$\dot{x}(t) = Ax(t) + BKx(t - \tau) + B_w w(t) \quad (1)$$

$$z(t) = [Q^{\frac{1}{2}} \quad R^{\frac{1}{2}}K]^T x(t) = C x(t).$$
- Let K stabilize the delayed system for τ .
- AIM – three objectives :
Delay Stable, Optimal H2 norm, Sparsify K.
- H2 Problem :
Find proper, rational K to internally stabilize (1) and minimize H2 norm of transfer function from disturbance $w(t)$ to output $z(t)$ for $\tau = 0$.

Algorithms Developed

- $J(F)$ is the H2 norm of F with delay τ . $g(K)$ is the weighted l_1 norm of K .

$$\text{minimize } \underbrace{J(F)}_{H_2} + \gamma \underbrace{g(K)}_{\text{Sparsity}}$$
subject to $F = K$
 $C_{BML_1}(K, \tau) > 0, C_{LMH_2}(\tau) > 0 \} \rightarrow$ Delay Stability
- Alternating Directions Method of Multipliers (ADMM) steps:



- **Set A** : K that satisfy delayed H2 optimality problem.
- **Set B** : Sparse K that are stabilizing for (1).

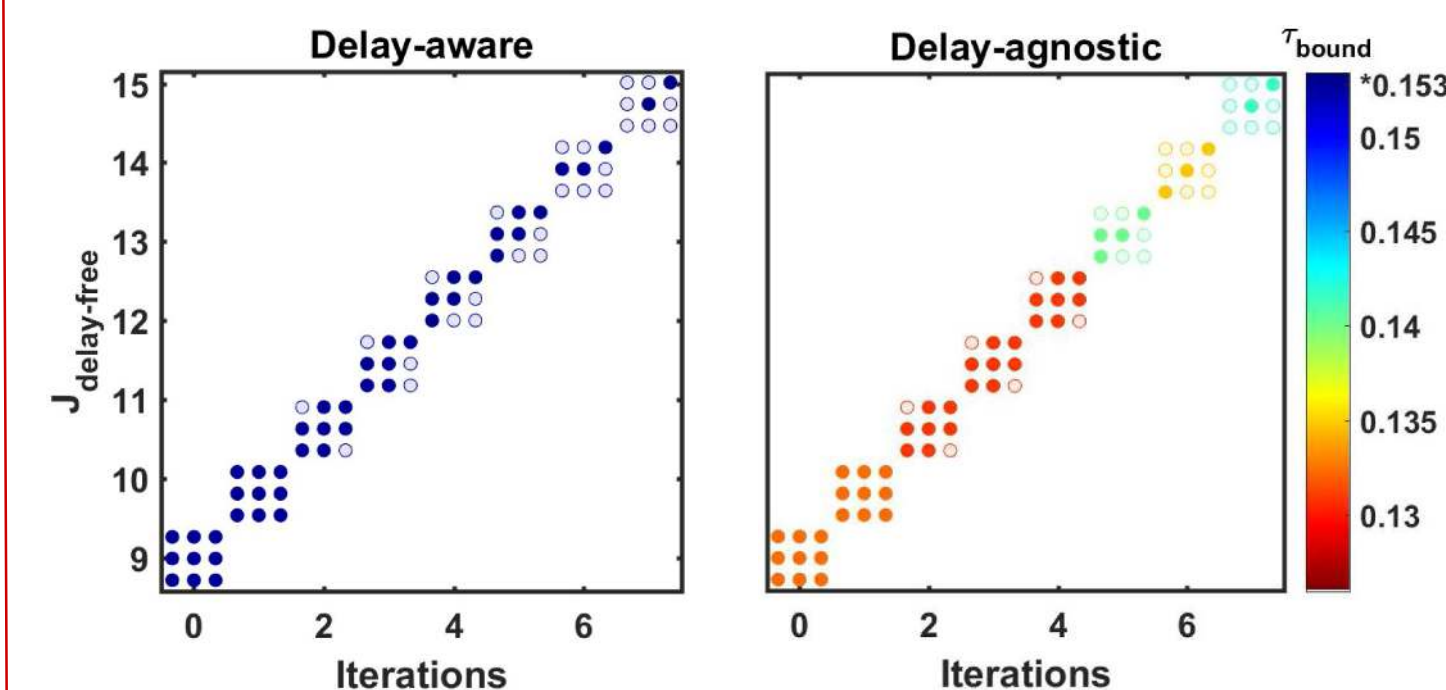
- Alternatively solve each objective while maintaining stability with respect to delay.
- Delay Vs Sparsity** – As sparsity increases, delay decreases.
- τ dependency is on $\text{card}(K)$. Cannot be explicitly included in ADMM.

- As sparsity increases, τ decreases $\Rightarrow K$ needs to be stabilizing for a smaller delay \Rightarrow **set of stabilizing solution expands.**
- For **optimal H2 norm** - need to include τ .
- Delayed performance is bad, especially, in case of large τ value.
- Delayed Systems are of infinite dimensions - **Discretize the interval $[0, \tau]$** and create an augmented vector.
- Augmented vector– **Includes past states.**
- Define $J_\tau(K)$ on the augmented vector.
- Positive and finite iff the system is stable with the feedback delay.
- \mathcal{A} – new state matrix is defined to **imitate the effect of delayed states.**

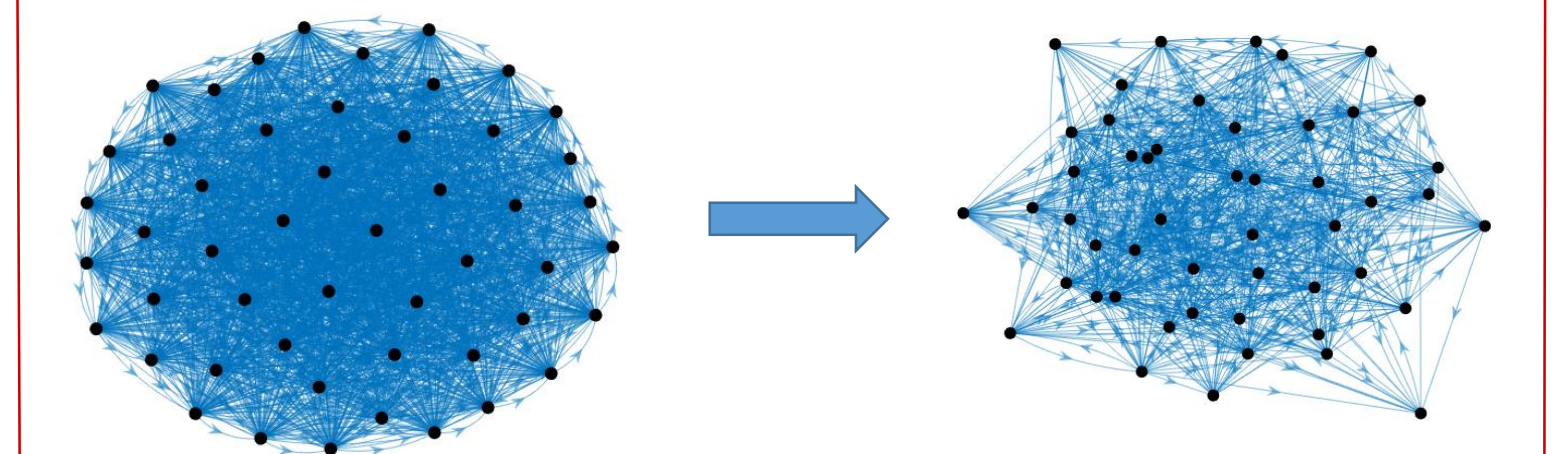
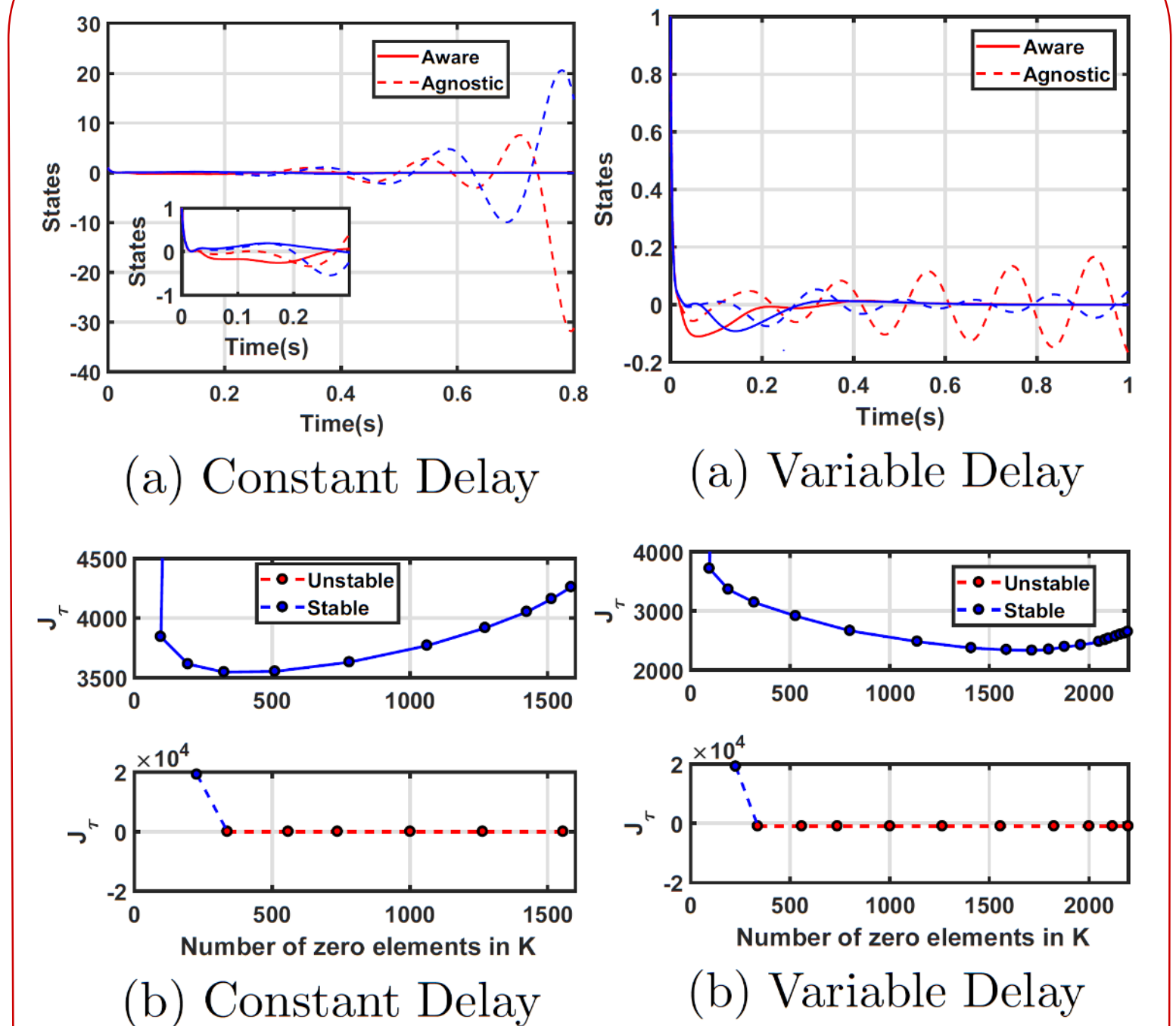
$$\eta = \begin{pmatrix} \phi(-\tau_N = -\tau) \\ \phi(-\tau_{N-1}) \\ \vdots \\ \phi(-\tau_2) \\ \phi(0) = x(t) \end{pmatrix} \Rightarrow \mathcal{A}\eta = \mathcal{A} \begin{bmatrix} \Phi(-\tau) \\ \vdots \\ x(t) \end{bmatrix}$$

Simulation Results

- Comparison with Delay-agnostic Algorithms.
- Case – 1 : K is a 3x3 matrix. Observe the consistent maintenance of delay-bounds in our delay-aware algorithm. This does not include optimality with respect to the delay.



- Case – 2 : K is a 50x50 matrix consisting of 2500 links. We include delay in H2 norm for this case.



- The number of non-zero elements of K is decreased from 2500 to 924.q

Conclusion

- We design algorithms with convergence guarantees to find sparse optimal controllers that are delay stable.
- As part of future work, we expand the idea to real world multi-user system where each user is dynamically decoupled from each other and share a common cloud-based communication network.

References

- N Negi and A Chakraborty. ACC 2018. “Sparse Optimal Control of LTI Systems under Sparsity-Dependent Delays“.