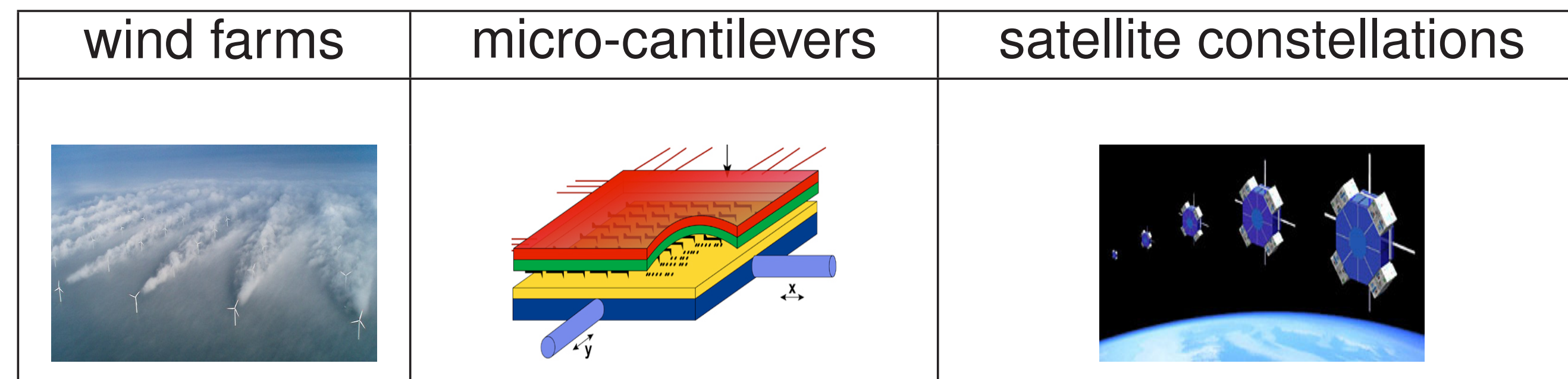
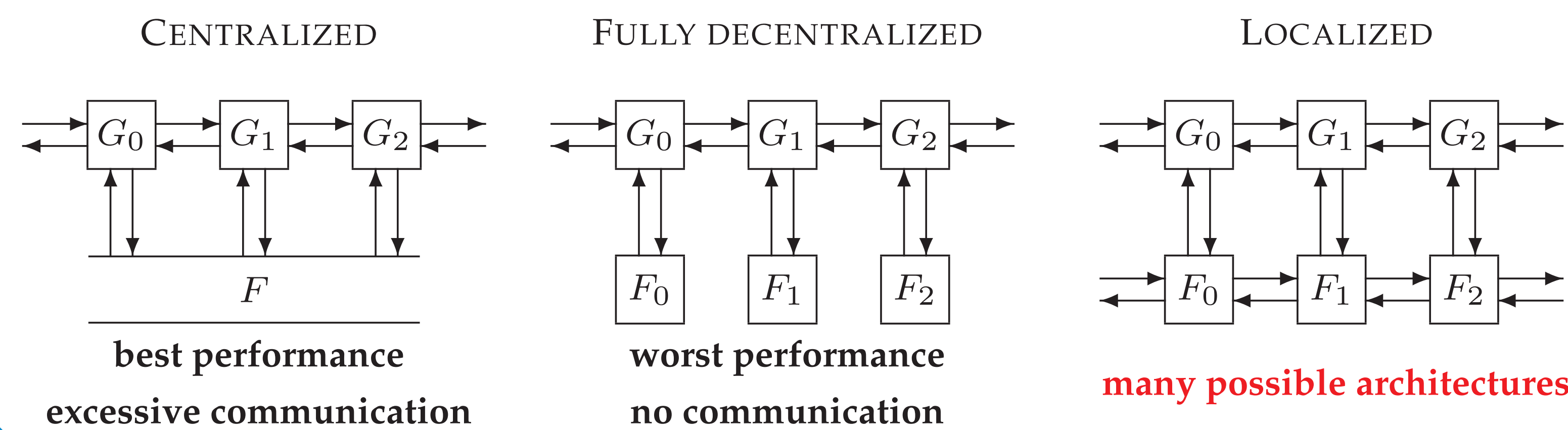


## LARGE DYNAMIC NETWORKS

- Ubiquitous in modern technology



- Controller architectures



## SPARSITY-PROMOTING OPTIMAL CONTROL

$$\begin{aligned} \dot{x} &= Ax + B_1 d + B_2 u \\ z &= \begin{bmatrix} Q^{1/2} x \\ R^{1/2} u \end{bmatrix} \quad u = -Fx \end{aligned}$$

OBJECTIVE:

Strike a balance between **variance amplification**  $d \rightarrow z$  and **sparsity of  $F$**

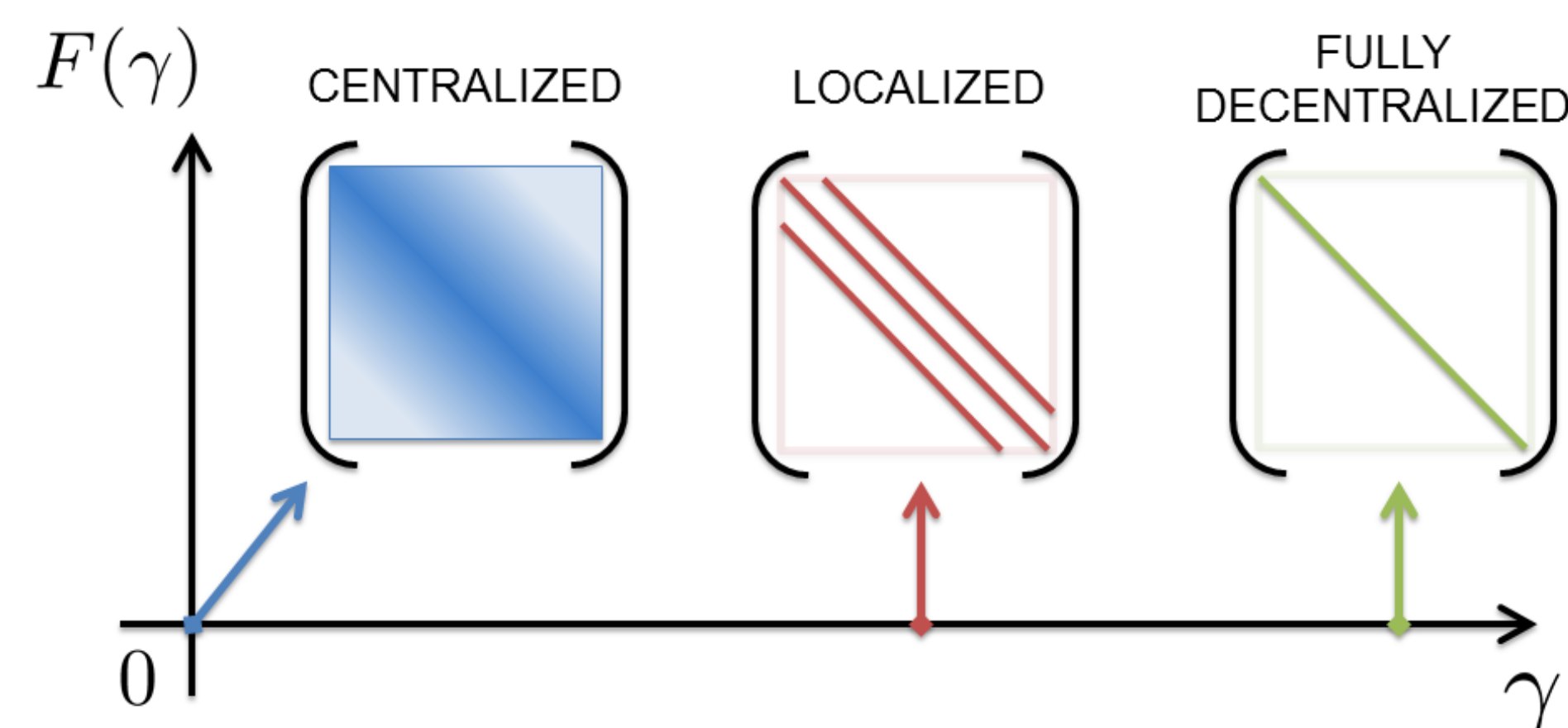
$$\text{minimize } J(F) + \gamma \text{card}(F)$$

DIFFICULTIES:

- Non-convex  $J = \text{trace} \left( \int_0^\infty e^{(A-B_2F)^T t} (Q + F^T R F) e^{(A-B_2F)t} dt B_1 B_1^T \right)$
- Non-convex & discontinuous **card** = number of nonzero elements

APPROACHES:

- Identify convex problems/relaxations
- Utilize powerful distributed optimization algorithms
- Identify parameterized family of feedback gains



## ADMM

- Step 1: introduce additional variable/constraint**

$$\begin{aligned} &\text{minimize } J(F) + \gamma g(G) \\ &\text{subject to } F - G = 0 \end{aligned}$$

- Step 2: introduce augmented Lagrangian**

$$\mathcal{L}_\rho = J(F) + \gamma g(G) + \text{trace}(\Lambda^T(F - G)) + \frac{\rho}{2} \|F - G\|_F^2$$

- Step 3: use ADMM for augmented Lagrangian minimization**

$$F^{k+1} := \arg \min_F \mathcal{L}_\rho(F, G^k, \Lambda^k)$$

$$G^{k+1} := \arg \min_G \mathcal{L}_\rho(F^{k+1}, G, \Lambda^k)$$

$$\Lambda^{k+1} := \Lambda^k + \rho(F^{k+1} - G^{k+1})$$

- Step 4: Polishing** – structured optimal design  $F \in \mathcal{S}$

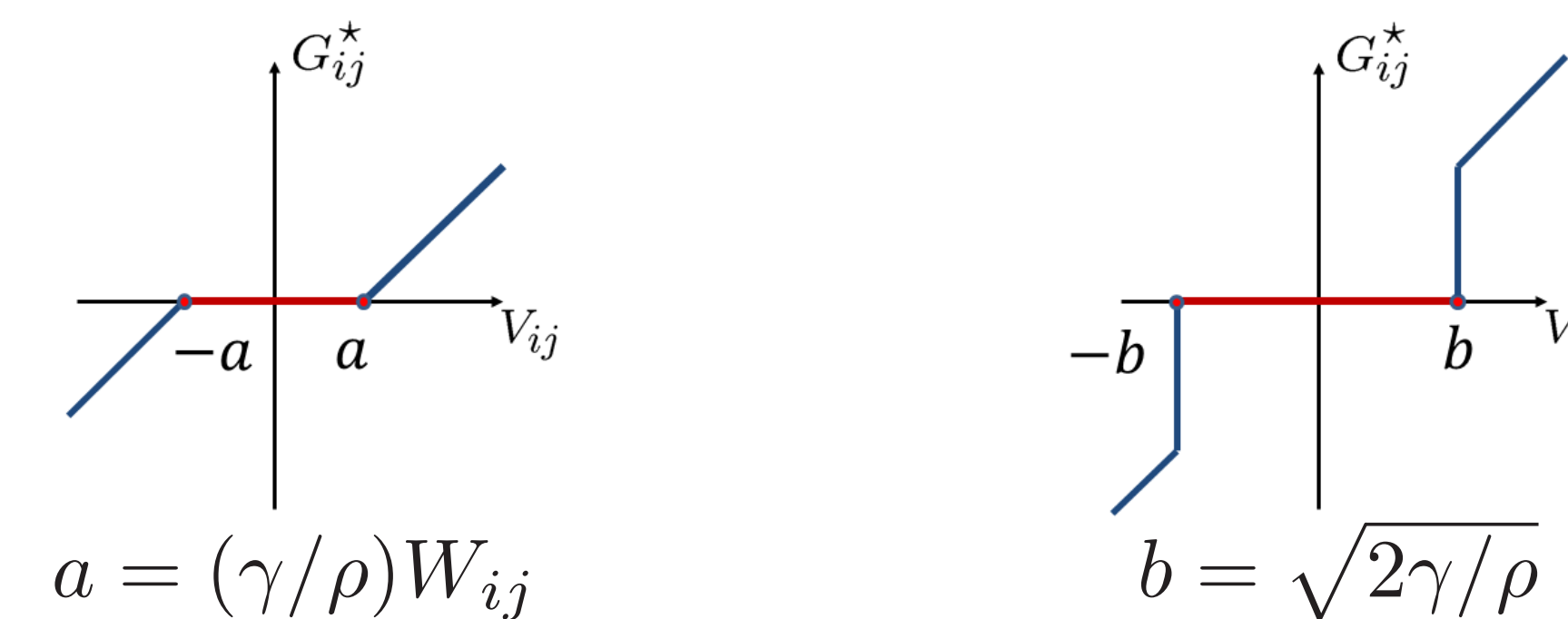
## ALGORITHM DETAILS

Sparsity-promoting penalty functions

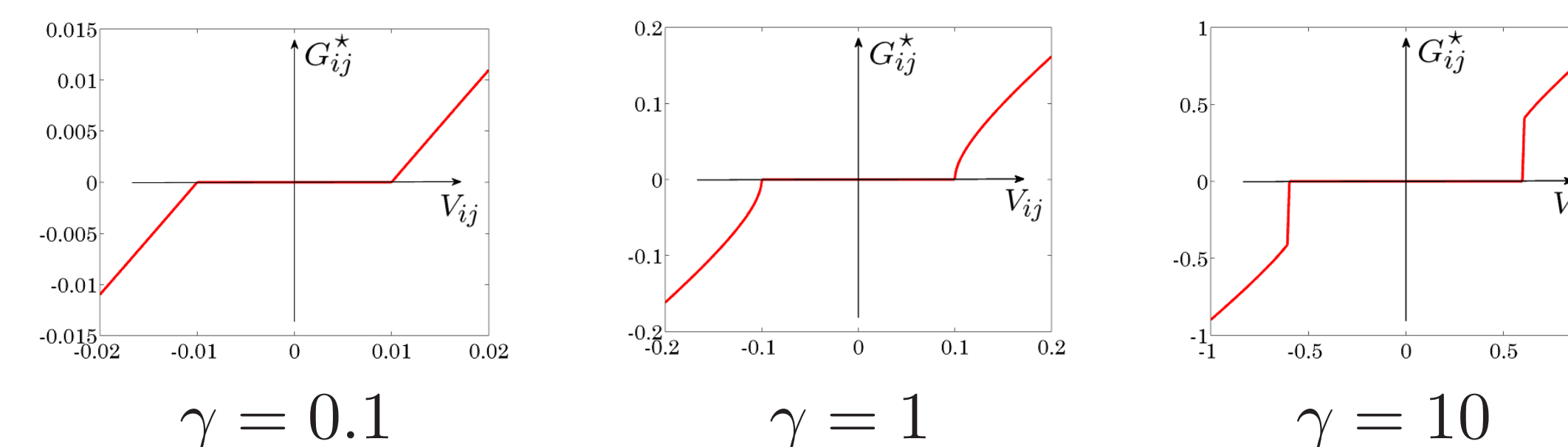
|                              |   |                                  |
|------------------------------|---|----------------------------------|
| weighted $\ell_1$            | sum-of-logs   | cardinality                      |
| $\sum_{i,j} W_{ij}  F_{ij} $ | $\sum_{i,j} \log \left( 1 + \frac{ F_{ij} }{\varepsilon} \right)$ | $\sum_{i,j} \text{card}(F_{ij})$ |

- G-minimization step: Elementwise analytical solution**

weighted  $\ell_1$ : **shrinkage**      cardinality: **truncation**



sum-of-log with  $\{\rho = 100, \varepsilon = 0.1\}$



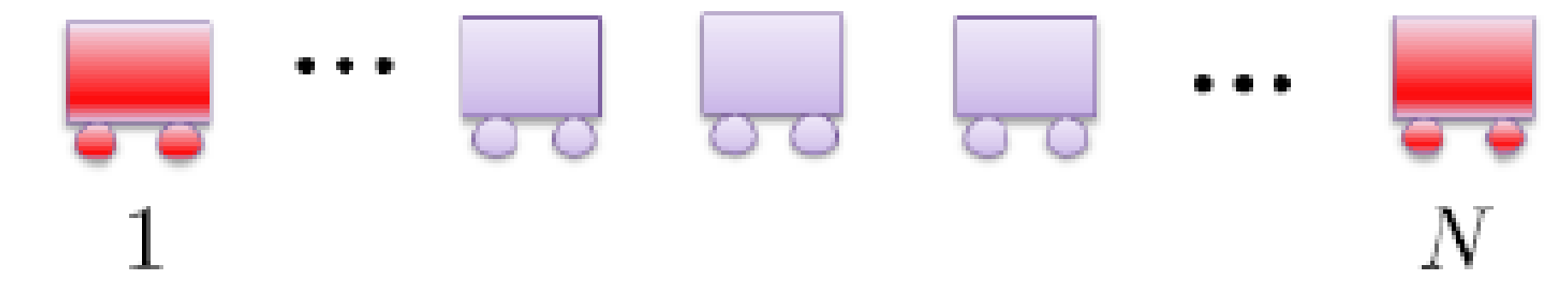
- F-minimization step: Necessary conditions for optimality**

$$\begin{aligned} (A - B_2 F) L + L (A - B_2 F)^T &= -B_1 B_1^T \\ (A - B_2 F)^T P + P (A - B_2 F) &= -(Q + F^T R F) \\ F L + \rho (2R)^{-1} F - R^{-1} B_2^T P L &= \rho (2R)^{-1} U \end{aligned}$$

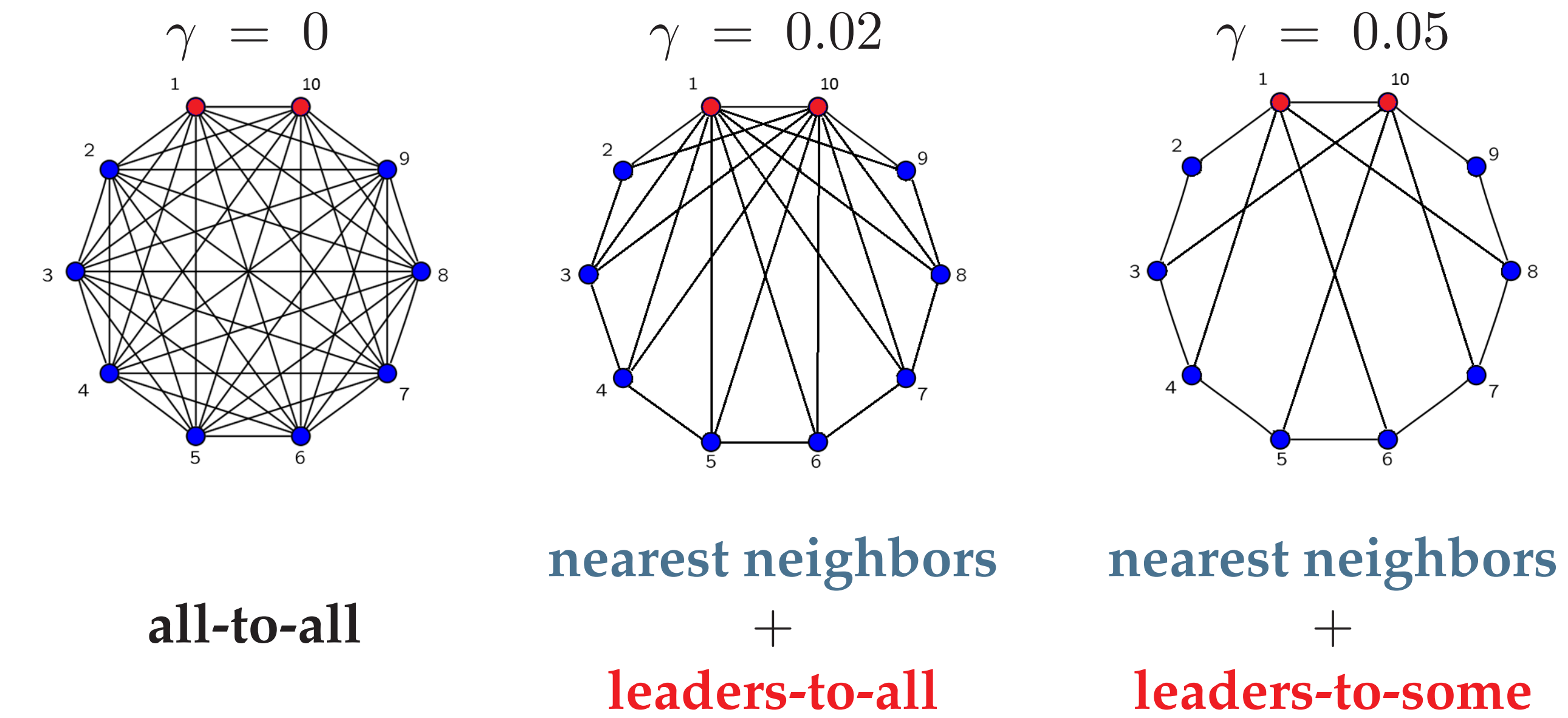
Descent scheme:  $F_0 \rightarrow \{L_1, P_1\} \rightarrow F_1 \rightarrow \{L_2, P_2\} \dots$

## EXAMPLES

- Formation of vehicles

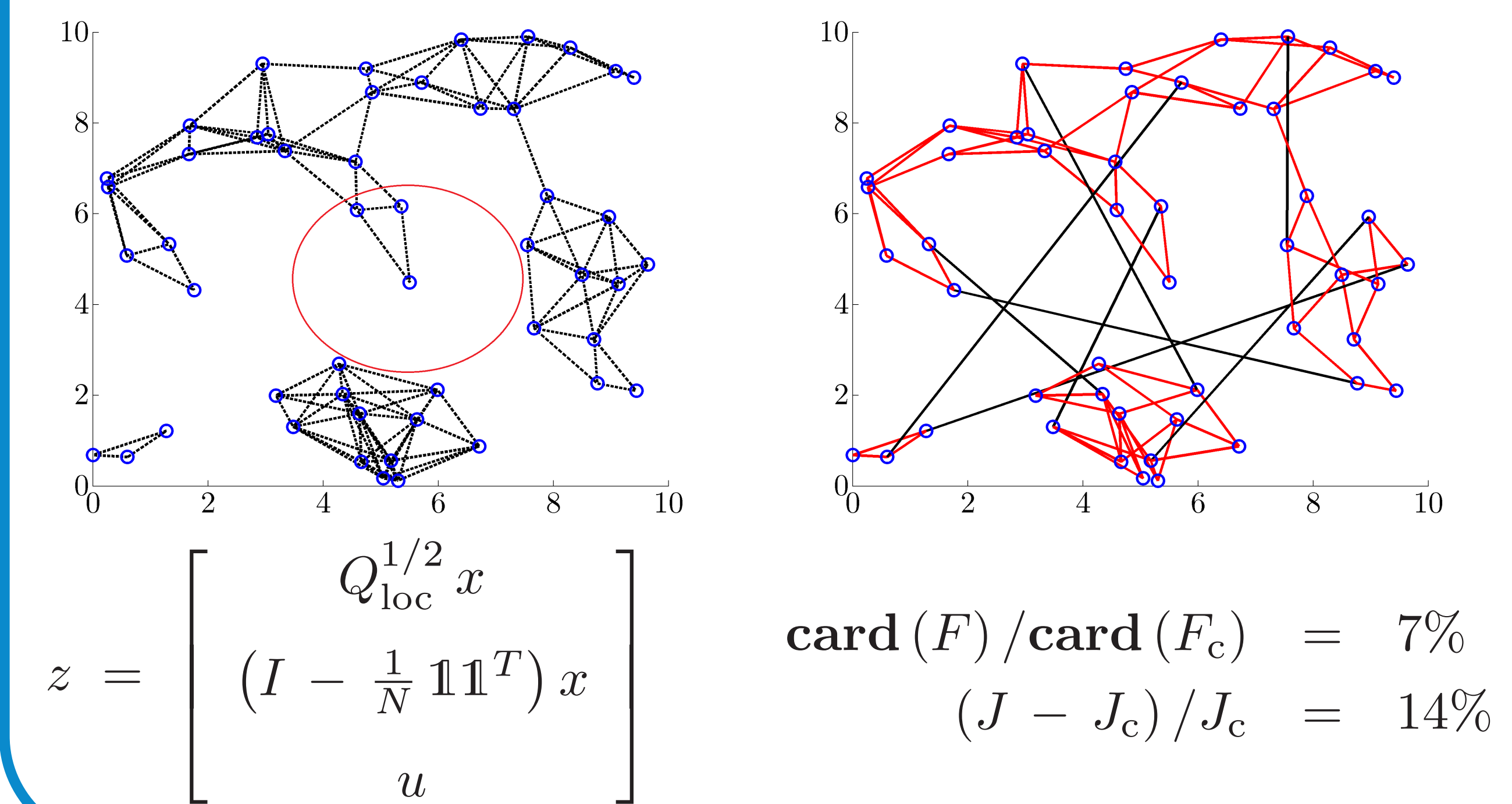


IDENTIFIED COMMUNICATION ARCHITECTURES



- Sensor network

local performance graph:      identified communication graph:



## SOFTWARE

[www.ece.umn.edu/~mihailo/software/lqrsp/](http://www.ece.umn.edu/~mihailo/software/lqrsp/)  
`>> solpath = lqrsp(A, B1, B2, Q, R, options);`

## PUBLICATIONS

- F. Lin, M. Fardad, and M. R. Jovanović, "Design of optimal sparse feedback gains via the alternating direction method of multipliers", *IEEE Trans. Automat. Control*, submitted, 2012; also arXiv:1111.6188v2.
- F. Lin, M. Fardad, and M. R. Jovanović, "Optimal control of vehicular formations with nearest neighbor interactions", *IEEE Trans. Automat. Control*, doi:10.1109/TAC.2011.2181790, 2012.
- F. Lin, M. Fardad, and M. R. Jovanović, "Augmented Lagrangian approach to design of structured optimal state feedback gains", *IEEE Trans. Automat. Control*, vol. 56, no. 12, pp. 2923-2929, 2011.