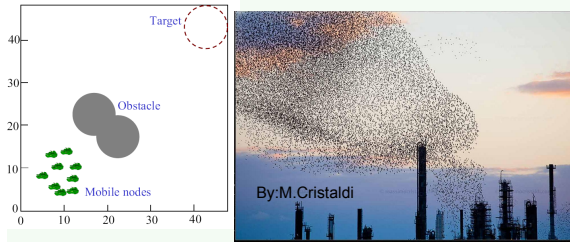


Introduction and Motivation



Distributed algorithms

- Group of agents with simple/complex abilities
- Agents sense their **local** neighborhood
- Communicate with neighbors and process the information
 - Perform a **local** action
 - Emergence of a **global** behavior.

Effectiveness of these algorithms depends on:

- The speed of convergence
- Robustness to agent/connection failures
- Energy/ communication efficiency

Graph theoretic abstraction of network

- Group topology affects group performance critically
- Graphs as structural abstractions of neighborhoods/connectivity
- Agents' knowledge of connectivity effects their dynamics
- Structural properties of graphs characterized by relevant matrices

Important graph-related matrices

- Graph Laplacian: $L=D-A$
- Natural Random walk matrix $P=D^{-1}A$
- Spectrum provides important structural information

Objective

Design problem: Find graph topologies with favorable tradeoff between performance **improvement (benefit)** of collaborative behaviors vs. **costs** of collaboration

Performance measures

- The speed of convergence of many distributed algorithms is determined by Second Largest Eigenvalue (SLEM) of P
- The number of spanning trees of a graph is a measure of robustness to losses in many applications

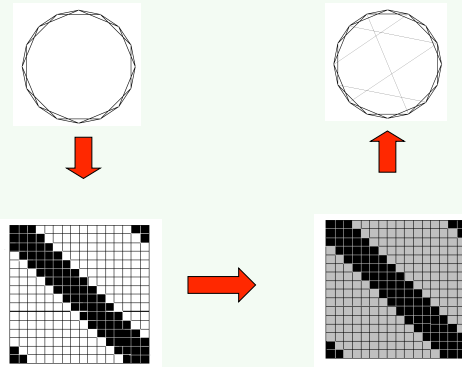
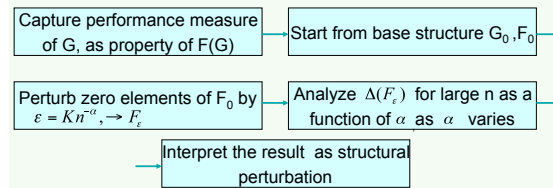
Problem 1 Characterization of Small World networks as **efficient topologies** (Asymptotic)

Problem 2 Optimal performance enhancement by adding few links




Problem statement and Analysis

Problem 1 Characterization of Small World graphs

Φ -model: Adding small number of new edges into a regular lattice $G_0=C(n,k)$, (n =number of nodes, $2k$ =number of initial neighbors of each node)



Results for spectral gap gain

	Initial S.G.	SW onset	SW
	$O(n^{-2})$	$\epsilon = O(n^{-3})$	$\epsilon = O(n^{-2})$
	$O(n^{-1})$	$\epsilon = O(n^{-2})$	$\epsilon = O(\frac{1}{n \log n})$
	$O((\log n)^{-1})$	$\epsilon = O(\frac{1}{n \log n})$	$\epsilon = O(\frac{1}{n \log \log n})$

Problem 2: Robust network design

Given a base graph topology, add k edges that result in maximum possible number of spanning trees

Matrix-tree theorem

$$\tau(G) = \frac{1}{n} \prod_{j=2}^n \lambda_j(L) = \frac{1}{n} \det \left(L + \frac{1}{n} J \right)$$

Optimization problem

Let f_l denote the incidence vector for edge l

Maximize $\tau \left(L_0 + \sum_{i=1}^m x_i f_i f_i^T \right)$ or equivalently

$$\log \det \left(L_0 + \frac{1}{n} J + \sum_{i=1}^m x_i f_i f_i^T \right)$$

Subject to:

$$1^T x = k$$

$$x \in \{0, 1\}^m$$

We relax the problem to find heuristics for design of optimal topologies

Result:

- The optimal graph is determined as a compromise between symmetrizing the graph and minimizing a notion of distance, **effective resistance distance**.
- Effective resistance distance between two nodes i and j : If we consider the graph as a resistive network with unit resistance on the edges, this is the effective resistance between i and j when a unit potential difference is applied between the two nodes
- The small world effect holds for spanning trees. Asymptotically, union of uniformly generated random spanning trees leads to construction of **expander graphs**.

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