

Topology optimization for a trade off between energy cost and network lifetime in average consensus

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Summary of Contributions

New topology optimization method to:

- 1) Reduce the convergence time
- 2) Reduce the total energy consumption
- 3) Increase network lifetime

Based on multi-objective minimization.

Background

- Set of nodes and edges: \mathbf{V} and \mathbf{E} .

- Communication flow: $\mathbf{G} = (\mathbf{V}, \mathbf{E})$.

- Adjacency matrix: \mathbf{A} such that $[\mathbf{A}]_{ij}$ is equal to 1 if $(i, j) \in \mathbf{E}$ and 0 otherwise.

- Degree matrix: \mathbf{D} diagonal matrix, whose entries are given by $d_i = \sum_{j=1}^N \mathbf{A}_{ij}$.

- Laplacian matrix: $\mathbf{L} = \mathbf{D} - \mathbf{A}$.

System evolution

- Initial data at time instant $t = 0$: $\mathbf{x}(0)$, whose average is $\mathbf{x}_{avg} = \frac{\mathbf{1}\mathbf{1}^T \mathbf{x}(0)}{N}$.

General linear update of sensors at time instant t :

$$\dot{\mathbf{x}}(t) = -\mathbf{L}\mathbf{x}(t)$$

- Convergence time T_{conv} : time required by slowest mode to be reduced by a factor $\rho < 1$

- Expressed as $T_{conv} = -T_s \frac{\log(\rho)}{\lambda_2(\mathbf{L})}$, where T_s is the duration of a time slot unit and $\lambda_2(\mathbf{L})$ is the so-called algebraic connectivity.

Multi-objective function

Convergence time: T_{conv} .

Total energy consumption:

$$\mathcal{E}_{tot} = T_{conv} P_{iter} = T_{conv} \sum_{i \in \mathbf{V}} P_i = T_{conv} \sum_{(i,j) \in \mathbf{E}} p_{ij}$$

Network lifetime proportional to:

$$\mathcal{L} = \frac{\mathcal{E}_{node}}{T_{conv} P_{i_{max}}}$$

Numerical results

Table 1. Simulation Results with $a_{th} = 0.45$

Graph	$\mathcal{L}^{\beta=1}$	$\mathcal{E}_{tot}^{\beta=1}$	$\mathcal{L}^{\beta=0}$	$\mathcal{E}_{tot}^{\beta=0}$	$\mathcal{L}^{\beta=1/2}$	$\mathcal{E}_{tot}^{\beta=1/2}$
none	795	0.037	475	0.035	734	0.036
RGG	1000	0.038	600	0.027	750	0.027
SWG	1100	0.031	440	0.024	850	0.025

The proposed algorithm: Projection is divided in several consensus process

Optimization problem (non convex) **P1**:

$$\begin{aligned} & \text{minimize}_{\{\mathbf{A}\}} && \beta \frac{P_{i_{max}}(\mathbf{A})}{\lambda_2(\mathbf{L}(\mathbf{A}))} + (1 - \beta) \frac{P_{iter}(\mathbf{A})}{\lambda_2(\mathbf{L}(\mathbf{A}))} \\ & \text{s. t.} && \xi \leq \lambda_2(\mathbf{L}(\mathbf{A})) \\ & && \mathbf{A} = \mathbf{A}^T \\ & && [\mathbf{A}]_{ij} \in \{0, 1\} \end{aligned}$$

- ξ : arbitrary small positive constant to ensure resulting value of $\lambda_2(\mathbf{L}(\mathbf{A}))$ greater than zero.

- β : constant between 0 and 1 that controls trade off between total energy consumption and network lifetime.

Relaxed optimization problem (convex) **P2**:

$$\begin{aligned} & \text{min.}_{\{s, \mathbf{A}\}} && s \\ & \text{s. t.} && \beta P_1(\mathbf{A}) + (1 - \beta) P_{iter}(\mathbf{A}) - \mu \lambda_2(\mathbf{L}(\mathbf{A})) \leq s \\ & && \vdots \\ & && \beta P_N(\mathbf{A}) + (1 - \beta) P_{iter}(\mathbf{A}) - \mu \lambda_2(\mathbf{L}(\mathbf{A})) \leq s \\ & && \xi \leq \lambda_2(\mathbf{L}(\mathbf{A})) \\ & && \mathbf{A} = \mathbf{A}^T \\ & && 0 \leq [\mathbf{A}]_{ij} \leq 1 \end{aligned}$$

P2 is the relaxed epigraph form of the original problem **P1**.

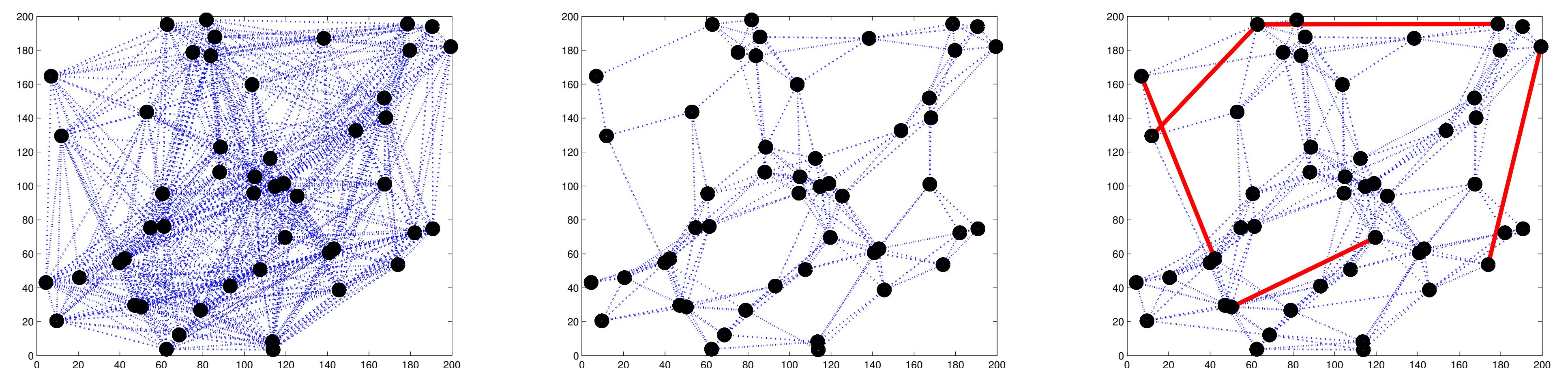
Dinkelbach's algorithm:

Set matrix \mathbf{A} as a feasible solution

WHILE $(\beta P_{i_{max}}(\mathbf{A}) + (1 - \beta) P_{iter}(\mathbf{A}) - \mu \lambda_2(\mathbf{L}(\mathbf{A})) > \epsilon)$

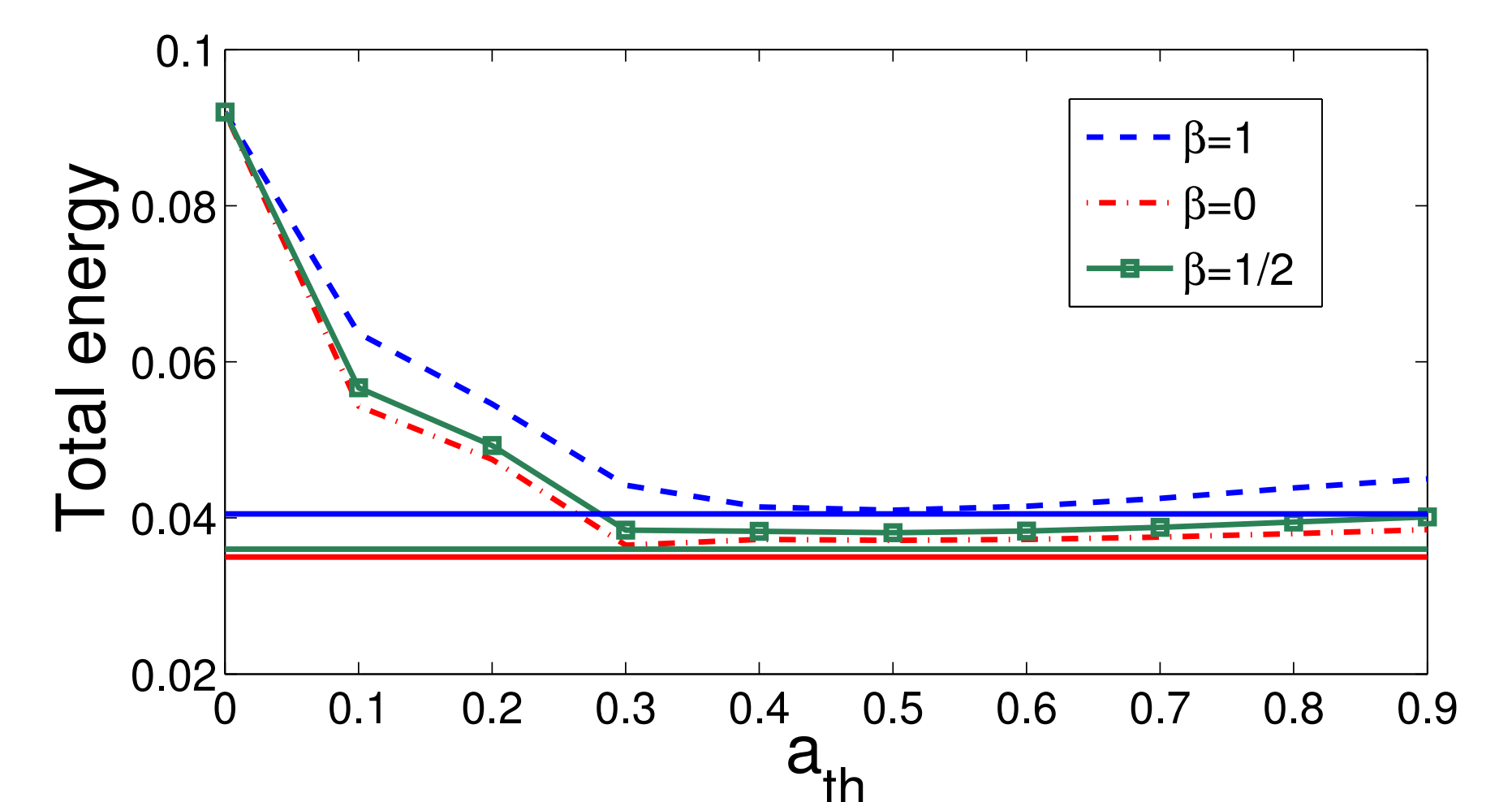
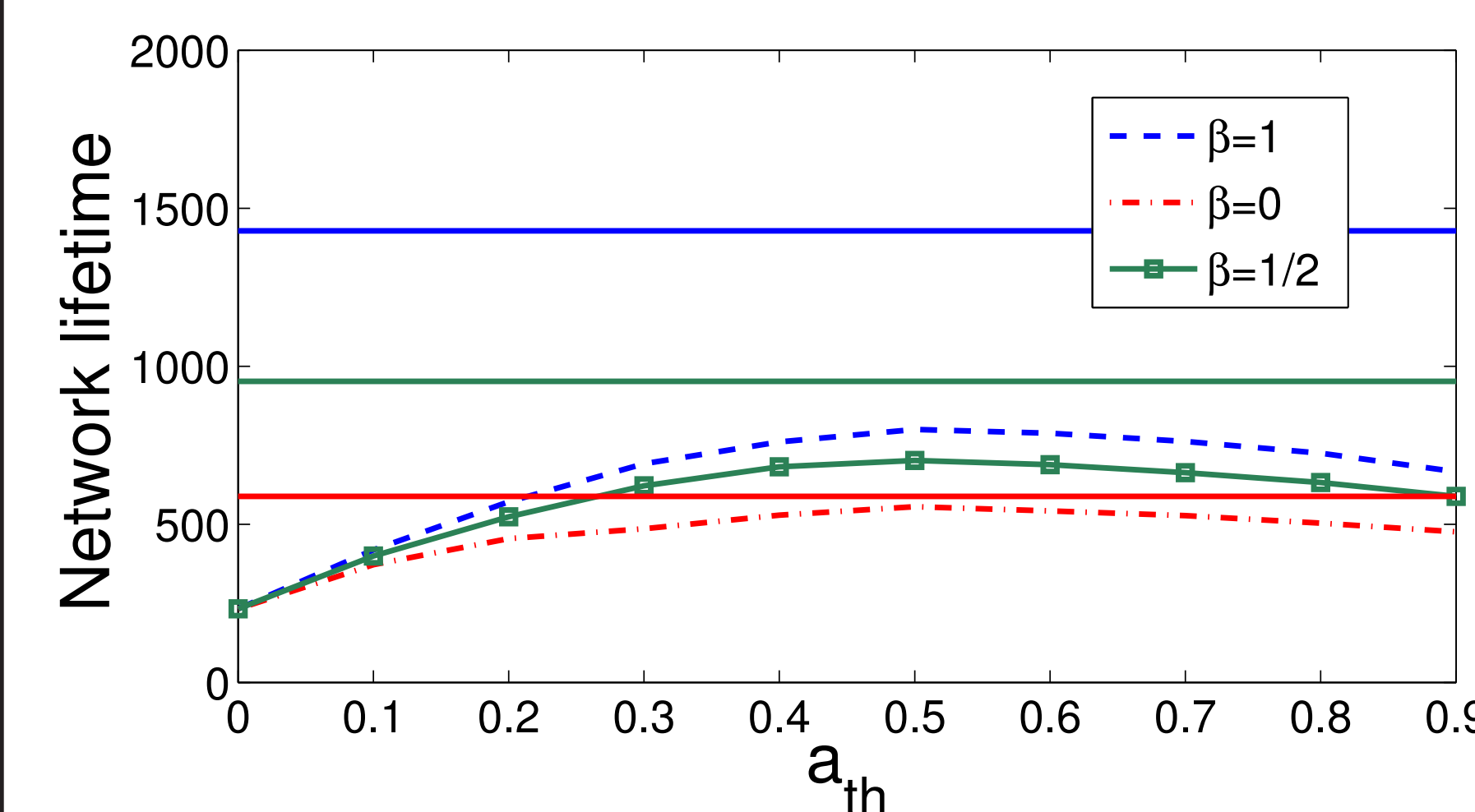
- set μ as $\frac{\beta P_{i_{max}}(\mathbf{A}) + (1 - \beta) P_{iter}(\mathbf{A})}{\lambda_2(\mathbf{L}(\mathbf{A}))}$
- Solve **P2** with the current μ

END WHILE



Results: Projecting to the original solution set

Simulation setup: $N = 50$ nodes. Path loss exponent $\gamma = 3$, a signal to noise threshold $\Phi = 10$, and background noise $N_0 = 10^{-10}$ mW., such that $p_{ij} = \Phi N_0 r_{ij}^\gamma$ mW. Finally, $\mathcal{E}_{node} = 1$ and $T_s = 1$ ms.



- a) $\beta = 1$: a lifetime maximum is obtained for threshold values around $a_{th} = 0.45$.
- b) $\beta = 0$: minimization of the total required energy.
- c) $\beta = 1/2$: Pareto optimal solutions.

Funding

This work was supported by the Spanish MEC Grants TEC2010-19545-C04-04 "COSIMA", CONSOLIDER-INGENIO 2010 CSD2008-00010 "COMONSENS", the European STREP Project "HYDROBIONETS" Grant no. 287613 within the FP 7 and by a Telefonica Chair.