

# Pooling or Sampling: Collective Dynamics for Electrical Flow Estimation

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joint work with L. Becchetti<sup>2</sup> and V. Bonifaci<sup>3</sup>



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# Electrical Flow in CS and Biology

Computation of currents and voltages in resistive **electrical network** is a crucial primitive in many **optimization algorithms**

- **Maximum flow**
  - Christiano, Kelner, Madry, Spielman and Teng, STOC'11
  - Lee, Rao and Srivastava, STOC'13
- **Network sparsification**
  - Spielman and Srivastava, SIAM J. of Comp. 2011
- **Generating spanning trees**
  - Kelner and Madry, FOCS'09

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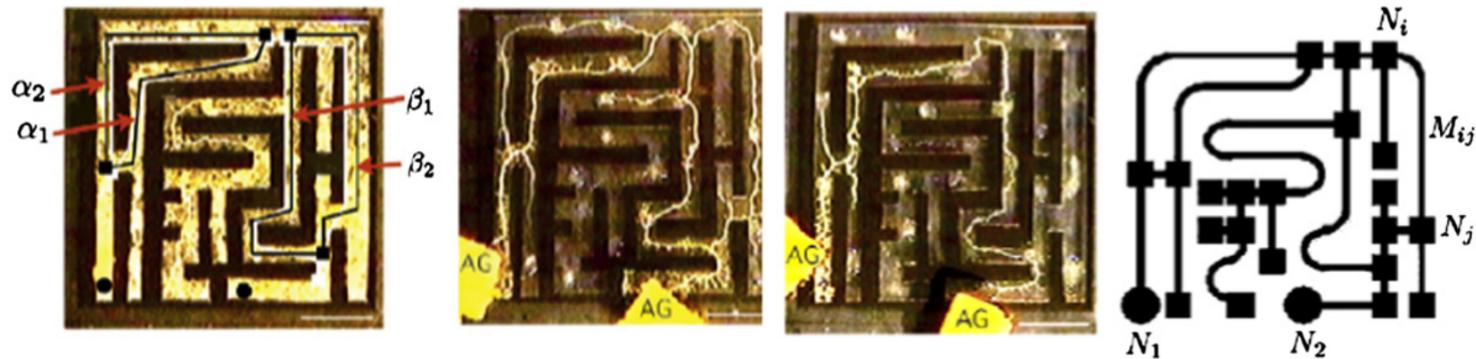
and as model of **biological computation**

- **Physarum polycephalum** implicitly **solving electrical flow** while forming **food-transportation networks**
- **Ants**



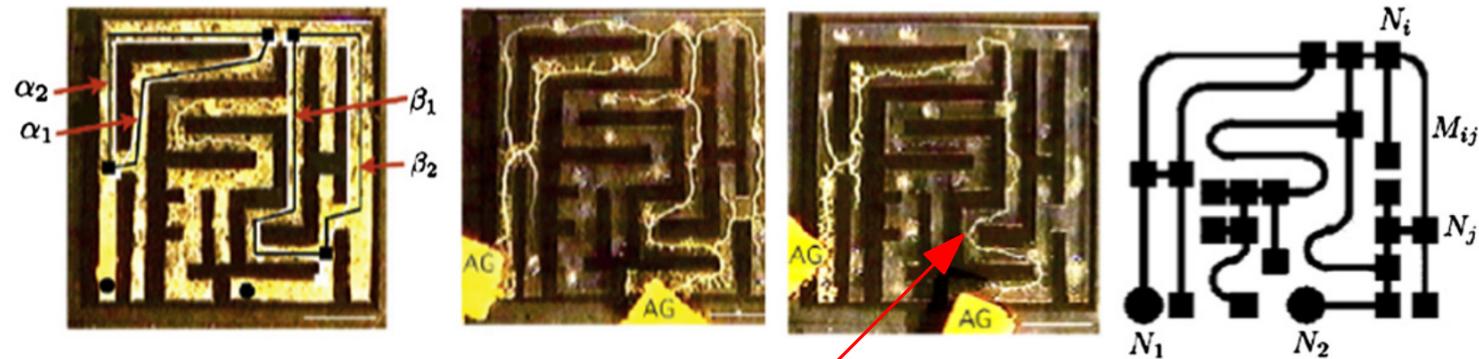
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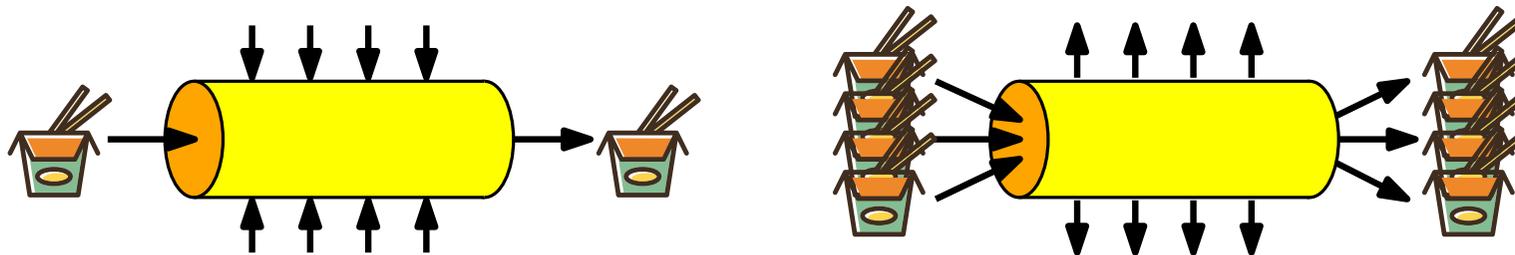


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*Physarum polycephalum* builds *tubes* to transport food.  
Amount of food flowing in tube determines growth or deterioration.



# Physarum Dynamics as an Algorithm

Dynamics:  $\dot{x}_e = |q_e| - x_e$

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$$x(t+1) - x(t) = h(|q(t)| - x(t))$$

Becchetti, Bonifaci, Dirnberger, Karrenbauer and Mehlhorn

**ICALP'13:** Discretized physarum computes  $(1 + \epsilon)$ -apx.

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Many sequels in TCS:

Bonifaci, Mehlhorn and Varma **SODA'12**, Bonifaci **IPL'13**,

Straszak and Vishnoi **ITCS'16**, Straszak and Vishnoi

**SODA'16**, Becker et al. **ESA'17**, ...

# How to Compute with Electrical Networks

Physarum have to compute the *pressures*,  
i.e. solve Kirchhoff's equations on nodes:

$$Lp = b$$

- edge weights  $x_e/\ell_e$  ( $\ell_e$  length of segment)
- $A$  weighted adjacency matrix
- $D$  diagonal matrix of nodes' volumes
- $L = D - A$  non-normalized graph laplacian
- $b$  is  $+1$  on source,  $-1$  on sink,  $0$  elsewhere

Previous approaches: **centralized computation**

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Previous approaches: **centralized computation**

*Biologically*, computation is achieved through a  
“microscopic” *local* process: what is it?

# Randomized Token Diffusion Process

**Process** [Ma, Johansson, Tero, Nakagaki and Sumpter, '13]

- At the beginning of each step,  $K$  new tokens *appear* at the source
- Each token independently performs a *weighted random walk* at each step
- Each token that hits the sink *disappears*

**Estimator**

$$V_K^{(t)} = \frac{Z_K^{(t)}(u)}{K \cdot \text{vol}(u)} \text{ where } Z_K^{(t)}(u) \text{ number of tokens on } u$$

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Cfr. Doyle and Snell, '84 & Tetali, '91:

Times random walk *transits on an edge before hitting sink*

# Analysis of Random Process

**Theorem.** In expectation the process converges to a *valid potential* with rate

$$1 - \frac{1 - \sqrt{1 - (\theta/\text{vol}_{\max})^2}}{2(n-1)} \sum_i \frac{w_{in}}{w_{in} + \text{vol}_{\max} - \sqrt{\text{vol}_{\max}^2 - \theta^2}}$$

where  $\theta$  is the edge expansion of the graph with sink removed.

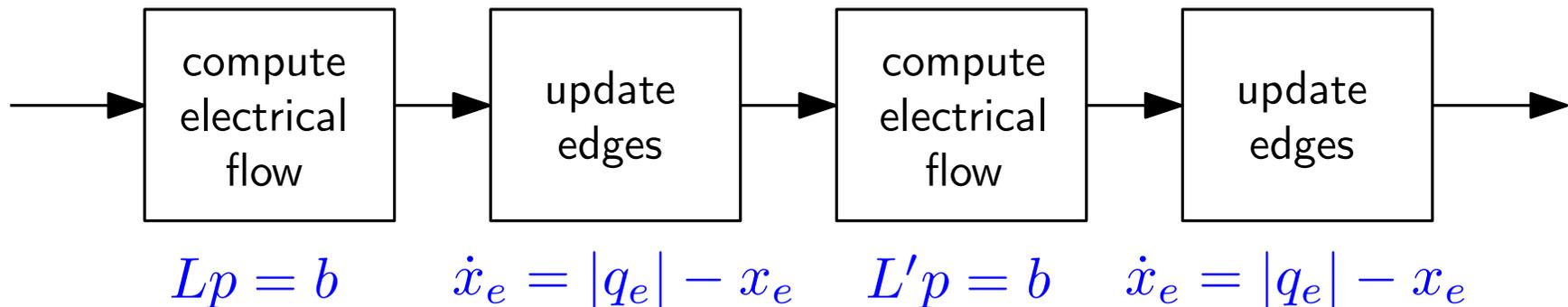
**Proof.**

- Bound w.r.t. spectral radius of  $P$  with row and column of sink set to zero, then
- relate bound to eigenvalue of non-norm. laplacian of graph with sink removed, finally
- use known relation to edge expansion.

# Open Problem: Analysis of *Distributed* Physarum?

Physarum dynamics et sim.:

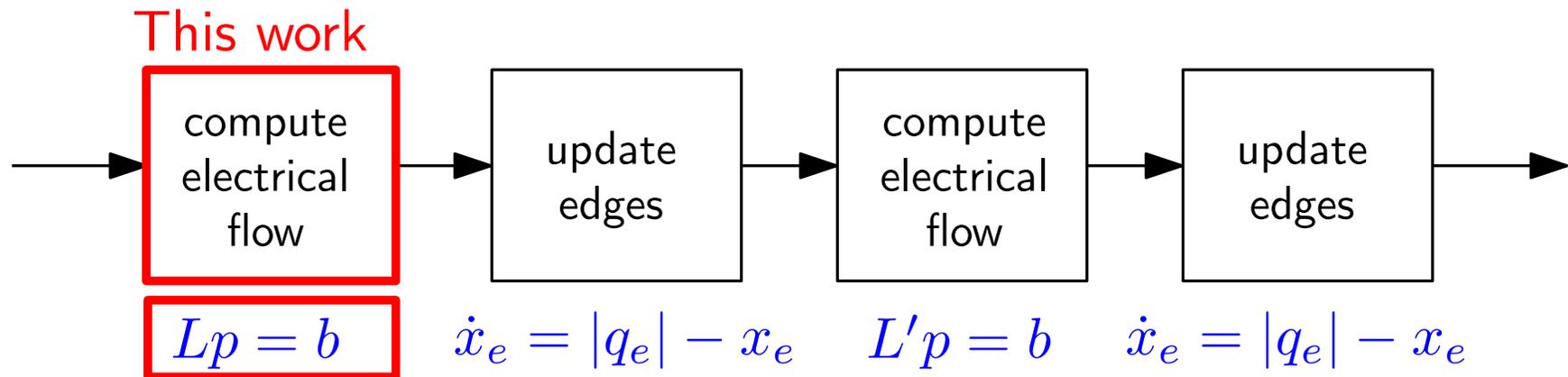
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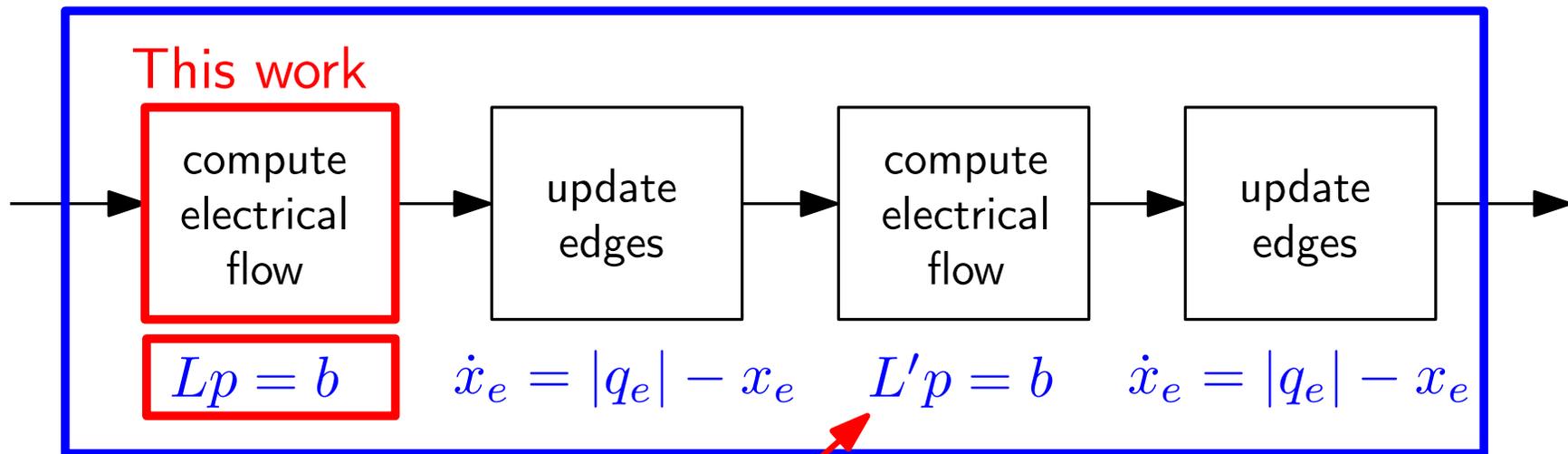
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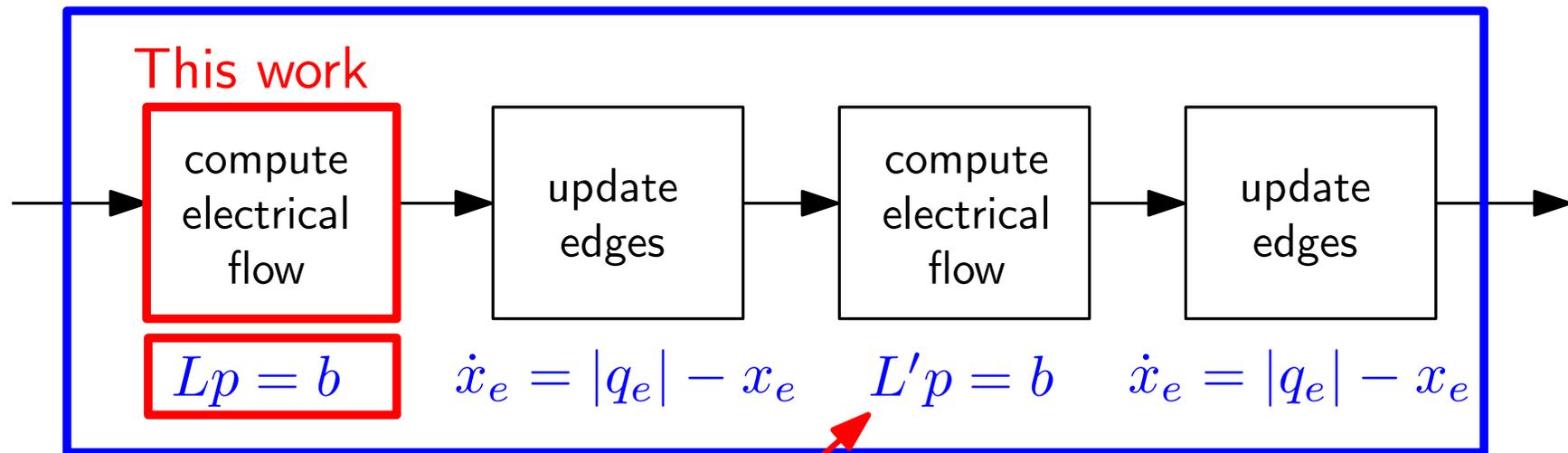


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# Open Problem: Analysis of *Distributed* Physarum?

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Compute *electrical flow*, then *update edge-weights*



- Spectral structure of  $L'$ ?
- **Global convergence time?**

Thank You!

# Distributed Jacobi's Method

**Thm.** Let

$$\tilde{p}(t+1) = P\tilde{p}(t) + D^{-1}b$$

where  $P = D^{-1}A$  is the transition matrix of graph. The system converges to a *valid potential* with rate

$$\|e_{\perp}(t)\| \leq \sqrt{\frac{\text{vol}_{\max}}{\text{vol}_{\min}}} \sqrt{2\phi}^t \|e_{\perp}(0)\|$$

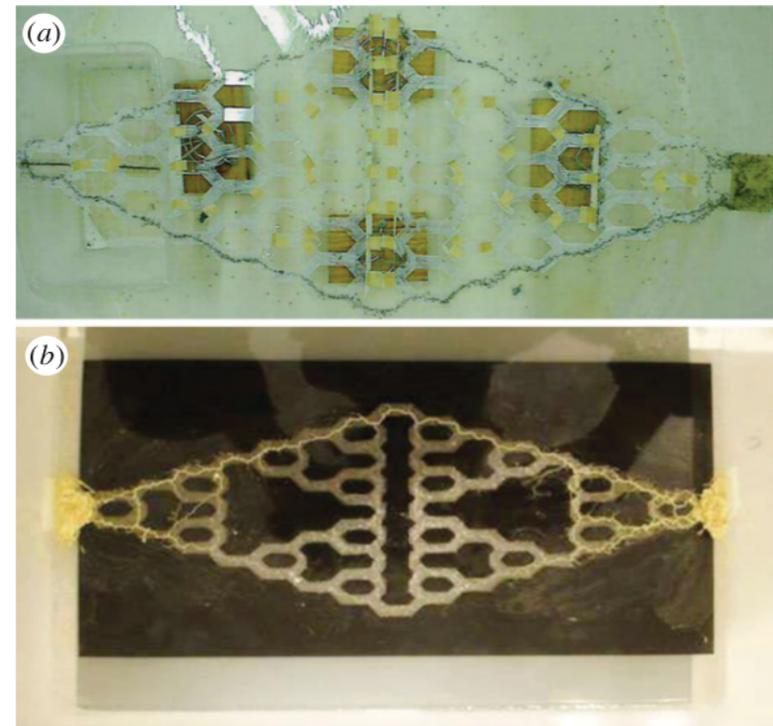
where  $e_{\perp}(t)$  is the component of the error orthogonal to  $\mathbf{1}$  and  $\phi$  is the *graph conductance*.

**Proof.** New analysis of Jacobi's iterative method exploiting structure of laplacian.

# The Slime Mold *Physarum Polycephalum*

electric network	<i>Physarum</i>	ant trails
length in space	length in space	length in space
potential/voltage	amount of nutrient	number of ants
current	flow of nutrient	flow of ants
conductivity	thickness of tube	pheromone concentration
capacitance	transport efficiency	total pheromone density
reinforcement intensity	tube expansion rate	pheromone drop rate
conductivity decrease rate	tube decay rate	evaporation rate

Ma, Johansson, Tero, Nakagaki and Sumpter, J. of the Royal Society Interface '13



# Message Complexity and Stochastic Accuracy

**Lemma.** As  $t \rightarrow \infty$ , the *expected message complexity* per round of Token Diffusion Algorithm is  $O(K n \text{vol}_{\max} \cdot E)$ , where  $E = p^\top L p$  is the *energy* of the electrical flow.

**Lemma.** For any  $K$ ,  $0 < \epsilon, \delta < 1$ ,  $t$  and  $u$ , such that  $p_u^{(t)} \geq \frac{3}{\epsilon^2 K \text{vol}(u)} \ln \frac{2}{\delta}$ , the estimator provides an  $(\epsilon, \delta)$ -approximation\* of  $p_u^{(t)}$ .

**Vice versa.**  $(\epsilon, \delta)$ -approximation of the potentials  $p_u^{(t)}$  greater than  $p_\star^{(t)}$  is achieved by setting  $K \geq \frac{3}{\epsilon^2 p_\star^{(t)} \text{vol}(u)} \ln \frac{2}{\delta}$ .

**Proofs.** Chernoff bound requires  $Y > \frac{3 \ln \frac{1}{\delta}}{\epsilon^2}$ .

\*  $X$  gives  $(\epsilon, \delta)$ -approximation of  $Y$  if  $\mathbf{P}(|X - Y| > \epsilon Y) \leq \delta$ .