

WELCOME

GOSSIP ALGORITHMS FOR DISTRIBUTED COMPUTING

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Presentation at the Weekly Group Meet,
Signal Processing for Communications Lab,
Indian Institute of Science,
29 September, 2012

- **Introduction**

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- **An Overview on the Field of Interest**
 - The decentralized or distributed computation,
a short historical perspective.
 - Applications
 - On the amount of literature
 - The elements of gossip algorithms
 - General analysis of gossip algorithms
 - The issues related to wireless
 - Few example works
 - The variant of consensus

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- **The current work from our lab**
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- **Conclusions**

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(ppt by Bharath, on work by D.Shah etal, JSTSP'11.)
- The work presents some of the recent developments in a broad class of problems.
- As we go along, we shall see where the work belongs to, in a much broader sense.

A HISTORICAL VIEW

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- Dates back to more than 60 years old references..!
- The authoritative (& popularly cited) references of original work are
 - The team decision problem — Marschak et al. [1955]
 - The dissertation by Tsitsiklis at MIT [1984].
- An example: The hats problem (a well-known puzzle).

FEW EXCERPTS FROM TSITSIKLIS DISSERTATION, 1984

QUESTIONS

“ Given a set of *processors* with partial information, is it possible that they make satisfactory decisions (in certain sense) without communicating among them??”

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RESULTS

“ Optimal design of a decentralized system even in the absence of any dynamics, is very hard computationally.”

“ The basic, static problems of decentralized decision making are algorithmically hard (NP-hard) even though the corresponding centralized problem is *trivial*.”

The techniques developed for his problem aimed for “*exchange of optimal decisions converging to consensus*”.

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- “*Probably it is due to the fact that it has become hard to think of sensing etc problems in electronic systems, without involving a networked platform*”
[intro to JSTSP,2011]
- “Gossip algorithms or distributed consensus algorithms”
— current active variants of *decentralized computation*.

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Span over a wide range of fields..

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- **Wireless Networks:** Collaborative in-network processing (eg: spectrum-sensing in cognitive radio),
 - Reducing the bandwidth, energy, latency thereby increasing the lifetime of the system.
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- **Computing:** Load balancing in distributed multi processors.
- **Automation and control:** Multi-agent coordination for vehicle formation, alignment, cooperative control.

A QUANTITATIVE VIEW ON THE LITERATURE

- “Distributed computation” has been a popular problem and there are over 10 thousand papers written on the subject.
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- A special issue only on Gossip algorithms is issued by IEEE JSTSP in August,2011.
- The paper discussed in the last meet is one of the papers appeared in this issue.
- The terms of “distributed computation”, “Gossip”, “Consensus” have become more generic.
- This presentation is an attempt to give an overall flavor..

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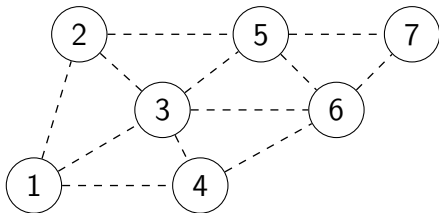
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- Readily extendable to a larger class of problems called *pair-wise* computable functions.

CONTD.] GOSSIP ALGORITHMS..

- “Who should communicate to whom and what, how often?”



TYPICAL ANALYSIS FOR ALGORITHMS:

- A proposal of communication protocol to converge to an optimal system state.
- Proof of convergence
- The convergence rate (mixing-time, second-eigenvalue..) computation
- Optimizing the convergence rate

AN ILLUSTRATION OF CONVERGENCE:

- We consider,
 - ① Average consensus
 - ② A static network with N nodes
 - ③ At each time instant, an independently chosen subset of nodes exchange data and compute a linear average at each node.

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- As only a partial information is conveyed throughout the network, it is required to perform an appropriate sequence of such communications (at least equal to the diameter of network).
- For example, two of such operations are required if just three nodes are present in a line topology.

- If we consider after an arbitrary n consecutive operations,

$$x(n+1) = W(n).W(n-1)...W(2).W(1).x(1) \quad (1)$$

$$= \left(\prod_{i=1}^n W(i) \right) x(1) \quad (2)$$

- When the appropriate randomization is employed, $W(n)$ become i.i.d. for each n . (example: randomized gossip, broadcast gossip etc.)

- As we are dealing with randomized algorithm, we are interested in the average performance of the system. Hence,

$$\mathbb{E}x(n+1) = \mathbb{E} \left(\prod_{i=1}^n W(i) \right) x(1) \quad (3)$$

$$= \left(\prod_{i=1}^n \mathbb{E}W(i) \right) x(1) \quad (4)$$

$$= (\mathbb{E}W)^n x(1) \quad (5)$$

- Hence the convergence of the system desires to have:

$$\lim_{n \rightarrow \infty} x(n) = \underline{1}x_{avg} \iff \lim_{n \rightarrow \infty} \mathbb{E}W^n = \frac{1}{N} \underline{1} \cdot \underline{1}^T \quad (6)$$

THEOREM

If $P_{N \times N}$ is a doubly stochastic matrix,

$$\lim_{n \rightarrow \infty} P^n = \frac{1}{N} \mathbf{1} \mathbf{1}^T \iff |\lambda_2(P)| < 1$$

Where $\lambda_i(A)$ denotes the i^{th} largest eigenvalue of the matrix A , in magnitude.

Proof:

Consider the difference,

$$P^n - \frac{1}{N} \mathbf{1} \mathbf{1}^T = P^n - P \frac{1}{N} \mathbf{1} \mathbf{1}^T, \quad \mathbf{1} \text{ is eigenvector} \quad (7)$$

$$= P^n - P^n \frac{1}{N} \mathbf{1} \mathbf{1}^T \quad (8)$$

$$= P^n \left(I - \frac{1}{N} \mathbf{1} \mathbf{1}^T \right) \quad (9)$$

$$= P^n \left(I - \frac{1}{N} \underline{1} \cdot \underline{1}^T \right)^n, \quad \text{a projection matrix} \quad (10)$$

$$= \left(P - P \frac{1}{N} \underline{1} \cdot \underline{1}^T \right)^n \quad (11)$$

$$= \left(P - \frac{1}{N} \underline{1} \cdot \underline{1}^T \right)^n \quad (12)$$

Hence, LHS of theorem is true iff $|\lambda_1(P - \frac{1}{N} \underline{1} \cdot \underline{1}^T)| < 1$.
Using simultaneous diagonalizability of P and $\frac{1}{N} \underline{1} \cdot \underline{1}^T$,

$$\lambda_1 \left(P - \frac{1}{N} \underline{1} \cdot \underline{1}^T \right) = \lambda_2(P) \quad \square$$

The immediate questions are

- ① The speed of convergence [second eigenvalue(stoch. matx. convergence in its power) or mixing parameter(as a Markov chain)].
- ② Optimize the speed of convergence for W .

GOSSIP ALGORITHMS FOR WIRELESS:

Issues:

- Synchronized and asynchronous setups, while latter is more relevant.
- Time varying connectivity (time varying topology)
- Time varying values at nodes (advanced & is discussed by D.Shah et al., We saw it in last ppt)
- Noise & uncertainties in elements

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Comments:

- A perfectly-synchronized setup with static-topology itself is complex in most cases.
- Like most wireless network problems, design of optimal protocols for fast convergence is an open problem.

ILLUSTRATING FEW EXAMPLES

1) Boyd S., et al., “Randomized Gossip Algorithms”, *IEEE Transactions on Information Theory*, June, 2006.

- 1 A gossip protocol — communication between a random pair of nodes
- 2 Synchronous model and asynchronous model. (Poisson timers at nodes)
- 3 Convergence Analysis
- 4 Rate of convergence by second eigenvalue
- 5 Optimizing the rate of convergence as an SDP

2) Aysal T. C., et al., “Broadcast Gossip Algorithms for Consensus”, *IEEE Transactions on Signal Processing*, July, 2009.

- 1 An asynchronous gossip protocol — A randomly selected node broadcasts its data to its neighbors. (Poisson timers at nodes)
- 2 Analysis of convergence in mean, mean-square sense.
- 3 Optimal mixing parameter computation.
- 4 Performance analysis — bound for mean square error and communication cost for achieving consensus.

3) Rajagopalan S., et al., "Distributed Averaging in Dynamic networks", *IEEE Journal of Selected Topics in Signal Processing*, August, 2011.

- 1 Seeking insights on effects in gossip based computations, when the network topology changes and the values themselves change
- 2 Bounded additive variations and Bounded multiplicative variations
- 3 Bharath has already discussed it in detail

THE PROBLEM OF CONSENSUS

- A special case of gossip algorithms, where the optimal state of each node is same.
- The version focused by most of the gossip algorithms at their initial level.

AN INTERESTING VARIANT OF CONSENSUS

- A higher layer problem so far.
- The physical layer was assumed to be well-serving, to address the problems related to higher layer precisely.
- All PHY issues such as communication noise and fading were ignored, given a link is present.

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- In the same lines of gossip algorithms, the question in its simplest form is posed asking nodes with discrete *binary* data values to achieve majority consensus among the nodes.
- The problem of “Binary Consensus in Fading environment”: my current focus. (A future presentation)

Thank You.