

2017 - NS - 196-AOC - Distributed Algorithms for Solving Linear Equations - Xuan Wang

# CERIAS

The Center for Education and Research in Information Assurance and Security

## **Distributed Algorithms for Solving Linear Equations**

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### Introduction

The goal of our research is to propose distributed algorithms for solving linear equations by *multi-agent networks*. In the network, each agent only knows part of these linear equations and is able to communicate with its nearby neighbors. The key idea behind the algorithms is a so-called ``*agreement principle*'', in which each agent limits the update of its state to satisfy its own equation meanwhile trying to reach a consensus with its nearby neighbors' states.

#### Motivation

- Requirement for data security: Each agent knows part of equation and the sub-equation is not allowed to be shared through the network.
- Requirement on calculation Efficiency when the size of equation are extremely large.
- Requirement for real time data processing when the data acquisition units are physically distant from each other.

#### The Algorithms

> When the equation has **at least one solution**, to get it:

Initial  $x_i(t)$  as an arbitrary solution of  $A_i x = b_i$ , then follow the update:

$$x_{i}(t+1) = x_{i}(t) + P_{i}\left(\frac{1}{d_{i}(t)}\sum_{j\in\mathcal{N}_{i}(t)}x_{j}(t) - x_{i}(t)\right)$$

Where  $P_i$  is the orthogonal projection matrix to the kernel of  $A_i$ , that is  $P_i = I - A_{i'} (A_i A_{i'})^{-1} A_i$ .

Furthermore, when equation has multiple solutions, to get a solution with minimum l<sub>2</sub>norm. By a special initialization:

#### The Problem

To solve a linear equation Ax = b by a network of *m* agents, each agent *i* knows a sub-equation  $A_ix = b_i$ 

$$\begin{bmatrix} A & b \end{bmatrix} = \begin{vmatrix} A_1 & b_1 \\ A_2 & b_2 \\ \vdots & \vdots \\ A_m & b_m \end{vmatrix}, \quad A \in \mathbb{R}^{\overline{n} \times n}$$

Each agent controls a state vector  $x_i(t) \in \mathbb{R}^n$ , which could be looked as an estimate of agent *i* to the solution of Ax = b. At time *t*, each agent only shares its state  $x_i(t)$  to its current neighbors denoted by set  $\mathcal{N}_i(t)$ .

#### **Application Examples**

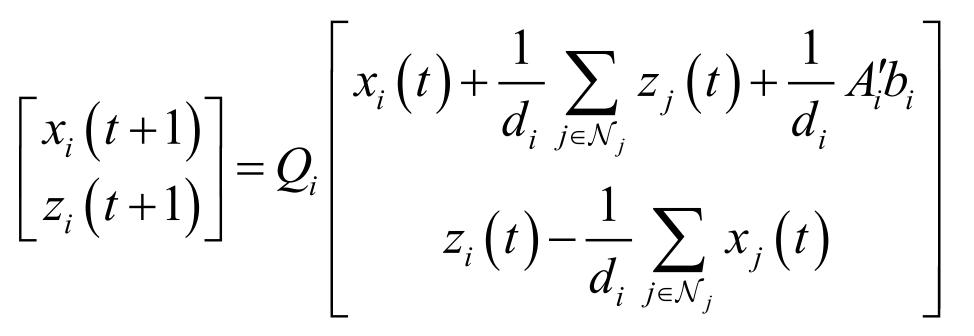
- The mode estimation of voltage oscillation In power networks
  - The phase sensors to monitor the power network is privately owned by different companies.
  - The companies don't want to share their information with their competitors.

 $x_{i}(0) = \left(M_{i}'M_{i}\right)^{-1}M_{i}'\begin{bmatrix}b_{i}\\b\end{bmatrix}$  $M_{i} = \begin{bmatrix}A_{i}\\P\end{bmatrix}$ 

where

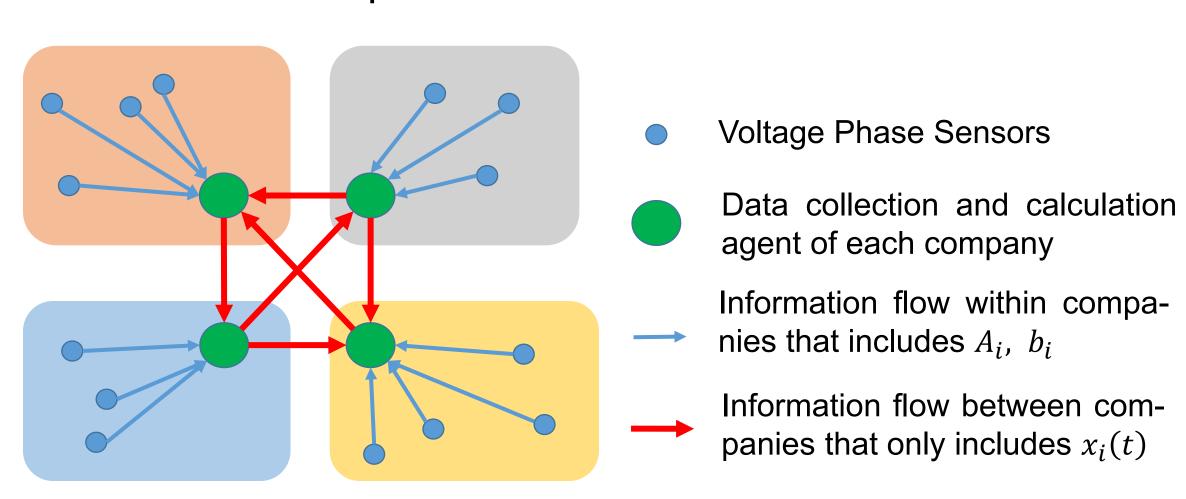
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When the equation does not have solution, to get its least square solution, by any initial point and the update:

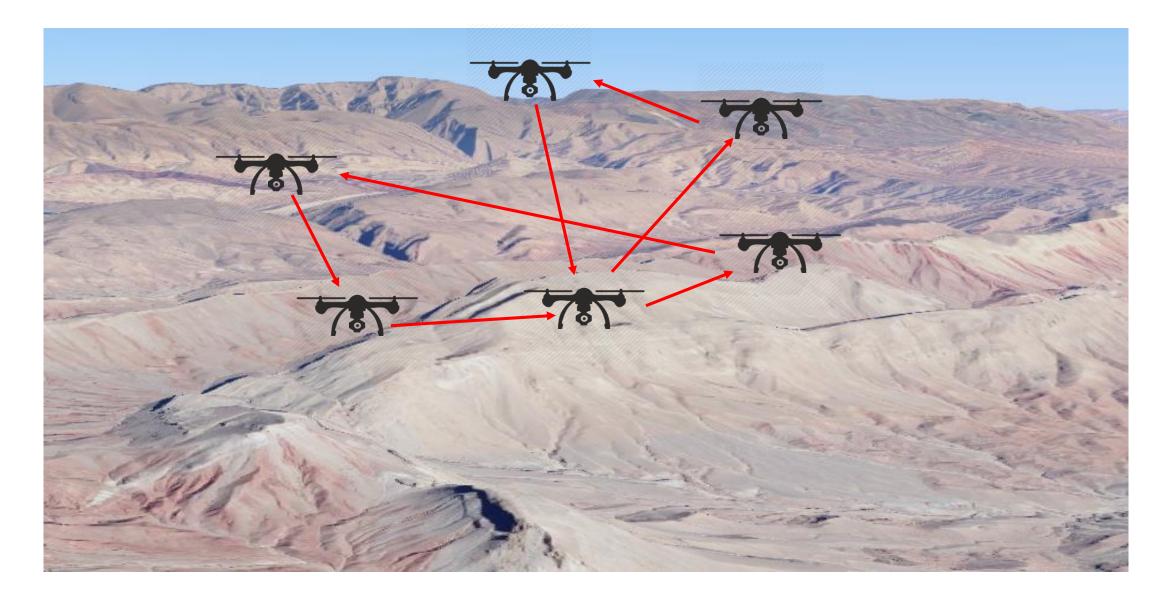


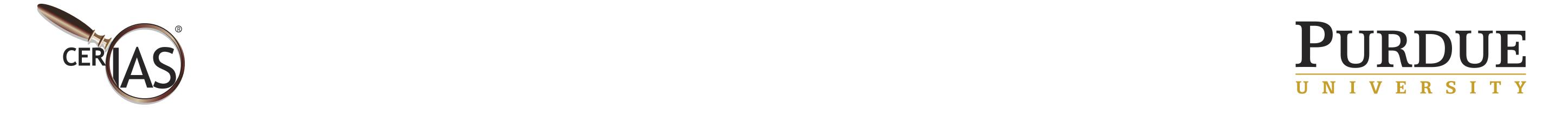
Where  $z_i(t) \in \mathbb{R}^n$  is an extra state hold by each agent and

$$Q_{i} = \begin{bmatrix} I_{n} + \frac{1}{d_{i}}A_{i}'A_{i} & I_{n} \\ -I_{n} & I_{n} \end{bmatrix}^{-1}$$



Large scaled distributed terrain-matching by UAVs





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