



# Distributed Cooperative Charging for Plug-in Electric Vehicles: A Consensus+Innovations Approach

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# Outline

- **Motivation and Background**
- Approach and Methodology
- Problem Formulation
  - Plug-in Electric Vehicles' Cooperative Charging (PEV-CC)
- Performance analysis
- Summary

# Motivation

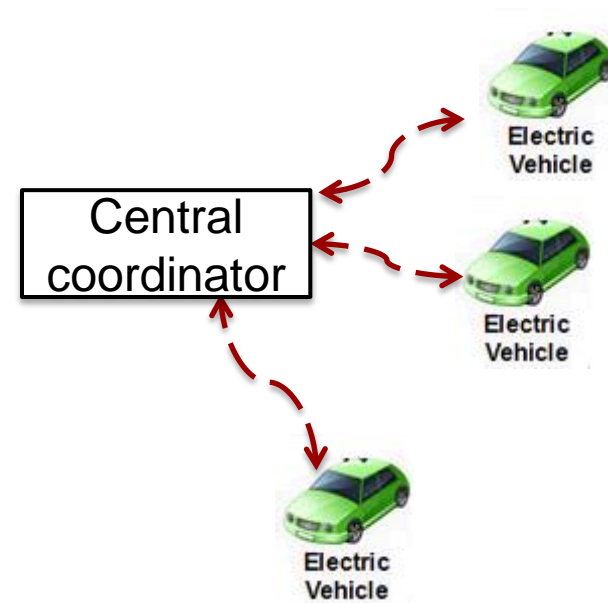
- Towards a smarter grid
  - Abundance of local computation/communication
  - Emergence of affordable small scale energy resources
  - Increased integration of power electronics



Source: [http://solutions.3m.com/wps/portal/3M/en\\_EU/SmartGrid/EU-Smart-Grid/](http://solutions.3m.com/wps/portal/3M/en_EU/SmartGrid/EU-Smart-Grid/)

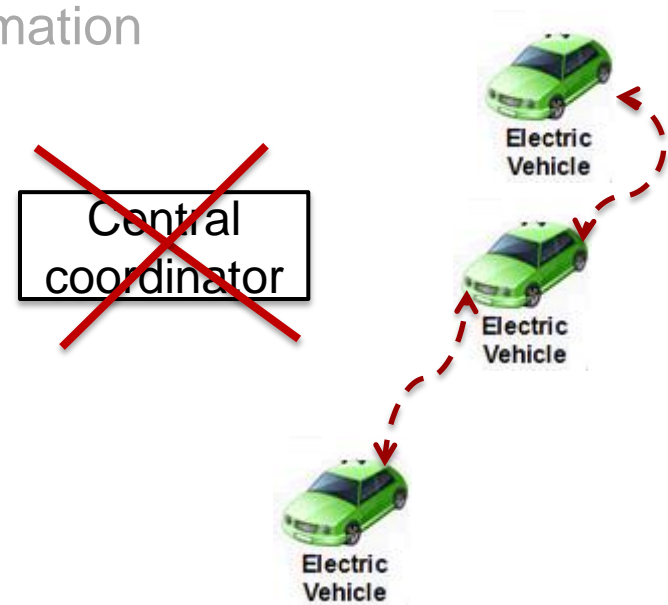
# Motivation

- Traditional/Centralized
  - Requires complete set of information



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- Traditional/Centralized
  - Requires complete set of information
- Distributed
  - No need for complete set of information



# Distributed approach



- Distributed control approaches:
  - No need for central coordinator
  - Distributes the computation among entities.
  - Each entity exchanges limited information with a few other entities

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- Distributed control approaches:
  - No need for central coordinator
  - Distributes the computation among entities.
  - Each entity exchanges limited information with a few other entities
- Advantages:
  - Suitable for smart grid
    - No need for complete set of information
    - A promising solution approach for coordinating geographically scattered resources like PEVs.
    - Preserves data confidentiality

# Existing work

- Non-cooperative agents:
  - Game theory approach
- Cooperative agents:
  - Decomposition based methods, e.g., ADMM
    - Requires information exchange with a central entity
  - Consensus based methods:
    - Enforcing agreement on a common variable
    - Requires a leader node to have access to total load information



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Our assumption: Cooperative agents

Our approach: Fully distributed consensus-based approach

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# Problem

- Optimal dispatch problem

$$\begin{aligned} \min_{\mathbf{Y}} \quad & \mathbf{F}(\mathbf{Y}) \\ \text{s.t.} \quad & \mathbf{M}(\mathbf{Y}) = 0 \\ & \mathbf{N}(\mathbf{Y}) \leq 0 \end{aligned}$$

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- Lagrangian Function

$$L = \mathbf{F}(\mathbf{Y}) + \lambda^T \mathbf{M}(\mathbf{Y}) + \mu^T \mathbf{N}(\mathbf{Y})$$

Dual variables

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Vector of primal variables

- Lagrangian Function

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- Optimality conditions

$$\frac{\partial L}{\partial x} = g(x) \quad \text{and Complementary Slackness Condition}$$

Vector of primal and dual variables

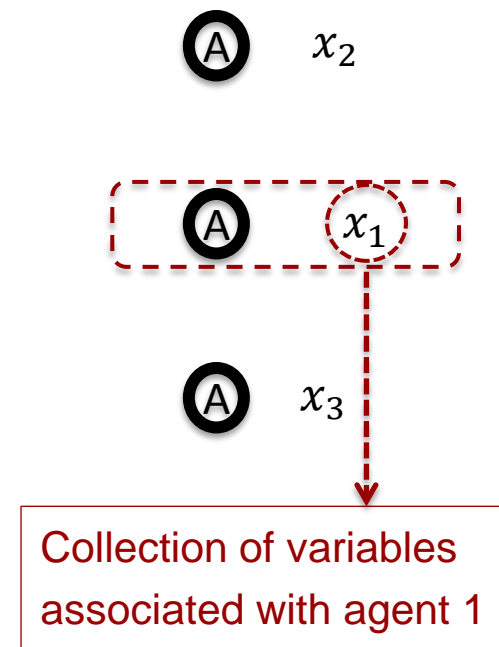
where  $x^T = [\mathbf{Y}, \lambda, \mu]^T$

# Consensus+Innovations based method

- Iterative procedure
  - Solve first order optimality conditions
  - These conditions involve local and global information

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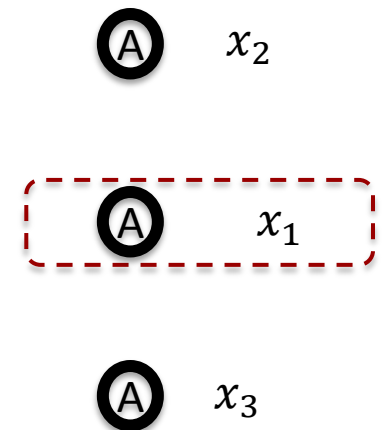
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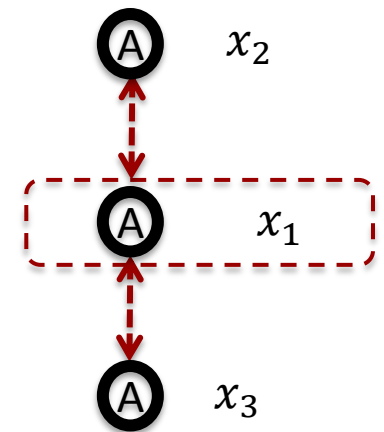
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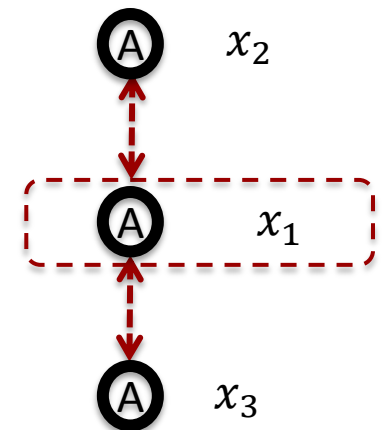
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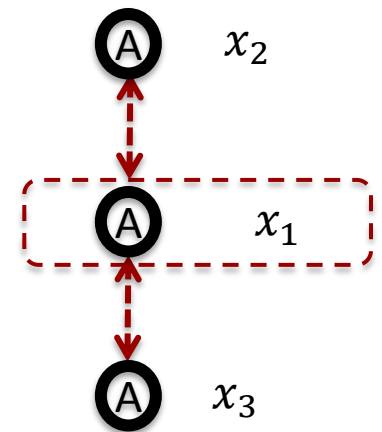
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- Update local variables

$$x_i(k+1) = x_i(k) + C \sum_{j \in \Omega_i} x_i(k) - x_j(k) + \Phi^T \cdot g_i(x_j(k))$$



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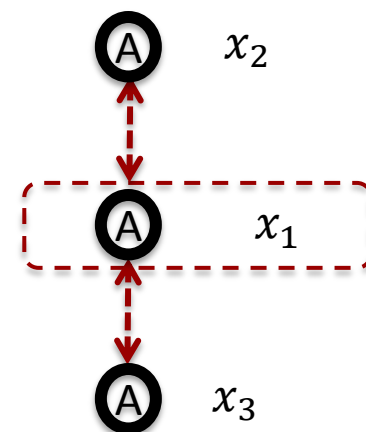
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The term  $\sum_{j \in \Omega_i} x_i(k) - x_j(k)$  is circled in red in the original image. A red dashed arrow points from this circled term to a red-bordered box containing the text "Consensus term".

Consensus term

$$j \in \Omega_i$$



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Power constraint

←--  $\underline{X}_v \leq X_v \leq \bar{X}_v$

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Power constraint

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Energy constraint

←---  $AX_v \leq b_v$

# PEV-CC formulation

$$\text{minimize}_{\mathbf{x}_v, \mathbf{L}} \quad \mathbf{L}^\top \cdot c_1 \cdot \mathbf{L} + c_2^\top \cdot \mathbf{L}$$

s.t.

$$\mathbf{L} = \sum_{v \in V} \mathbf{x}_v$$

$$A \cdot \mathbf{x}_v \leq b_v, \quad \forall v \in \{1, \dots, V\}$$

$$\underline{\mathbf{x}}_v \leq \mathbf{x}_v \leq \overline{\mathbf{x}}_v, \quad \forall v \in \{1, \dots, V\}$$

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# PEV-CC formulation

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s.t.

$\lambda$

$$\mathbf{L} = \sum_{v \in V} \mathbf{x}_v$$

$\rightarrow$  Global Constraint

$\mu_v$

$$\left. \begin{aligned} A \cdot \mathbf{x}_v &\leq b_v, & \forall v \in \{1, \dots, V\} \\ \underline{x}_v &\leq \mathbf{x}_v \leq \bar{x}_v, & \forall v \in \{1, \dots, V\} \end{aligned} \right\}$$

$\rightarrow$  Local Constraints

# Optimality conditions



$$\frac{\partial \mathcal{L}}{\partial \mathbf{L}} = 2c_1 \mathbf{L} + c_2 \lambda = 0$$

$$\frac{\partial \mathcal{L}}{\partial \mathbf{x}_v} = \lambda A^\top \cdot \mu_v + \mu_{v,+} - \mu_{v,-} = 0$$

$$\frac{\partial \mathcal{L}}{\partial \lambda} = \mathbf{L} + \sum_{v \in V} \mathbf{x}_v = 0$$

$$\frac{\partial \mathcal{L}}{\partial \mu_v} = A \cdot \mathbf{x}_v - b_v \leq 0$$

$$\frac{\partial \mathcal{L}}{\partial \mu_{v,+}} = \mathbf{x}_v - \bar{x}_v \leq 0$$

$$\frac{\partial \mathcal{L}}{\partial \mu_{v,-}} = -\mathbf{x}_v + \underline{x}_v \leq 0$$

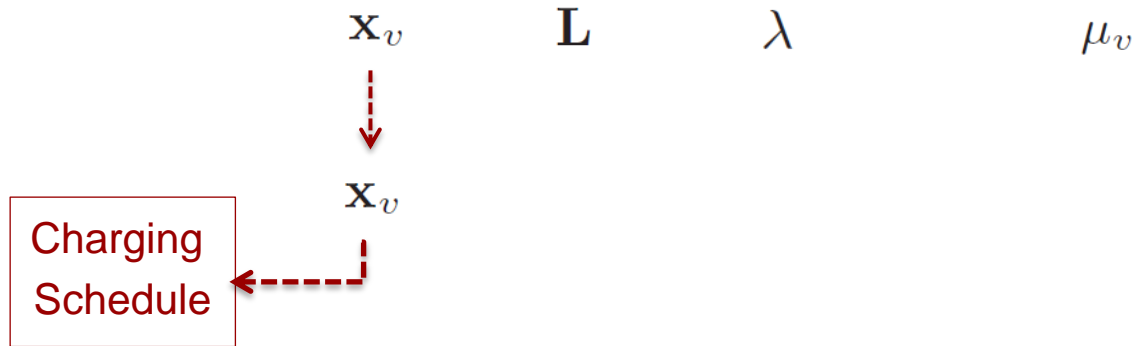
These equations involve both local and global information



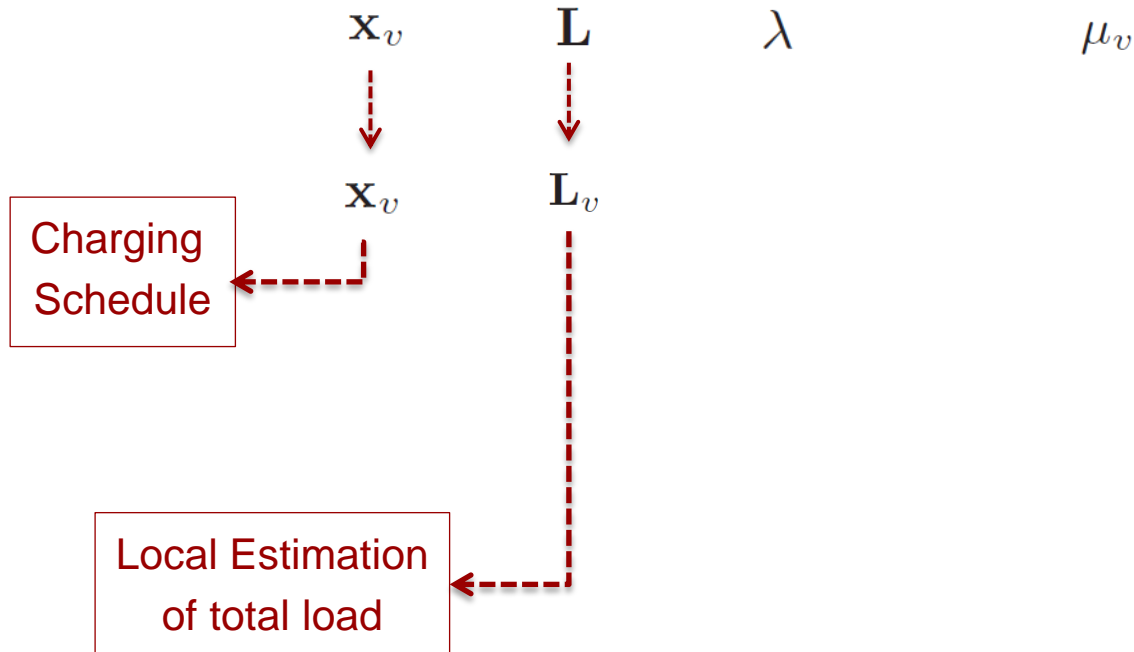
# Distributed variables

 $\mathbf{x}_v$  $\mathbf{L}$  $\lambda$  $\mu_v$

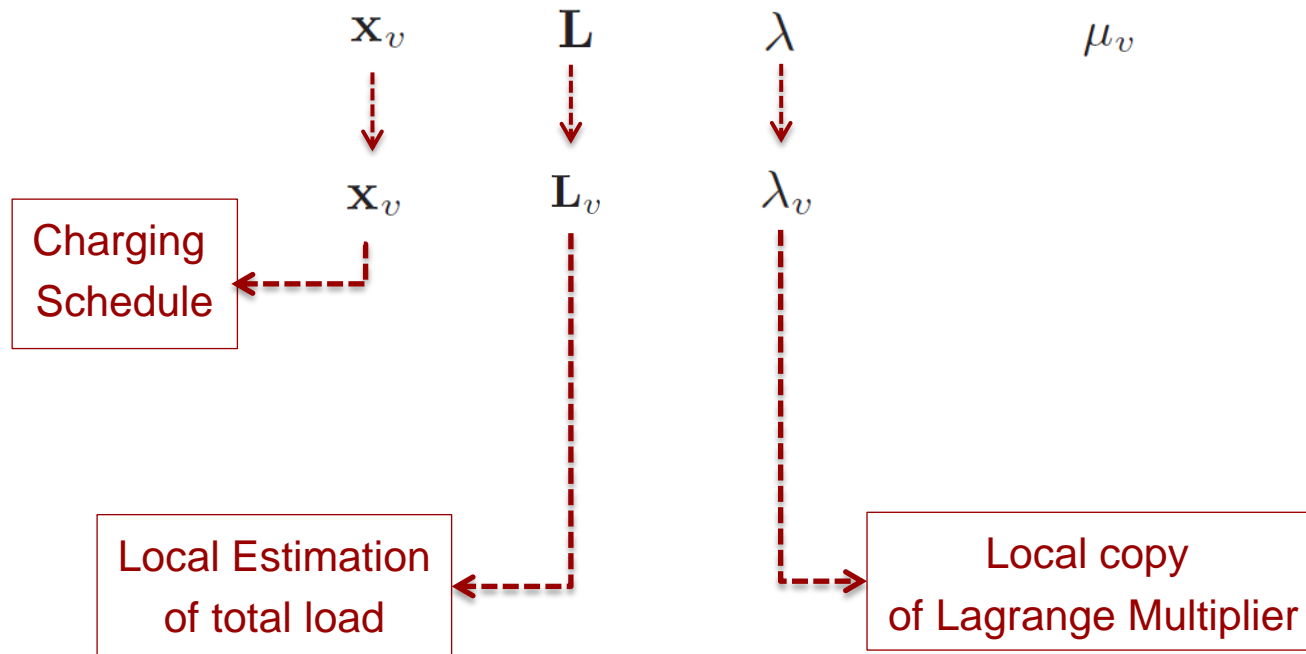
# Distributed variables



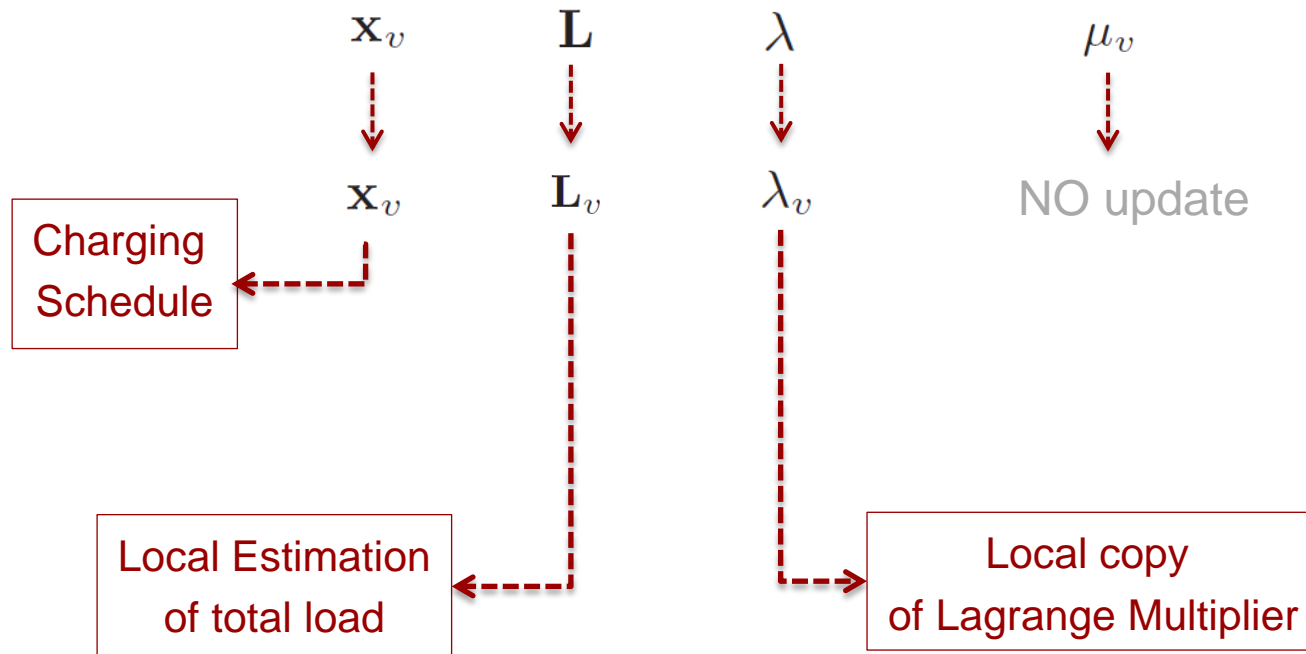
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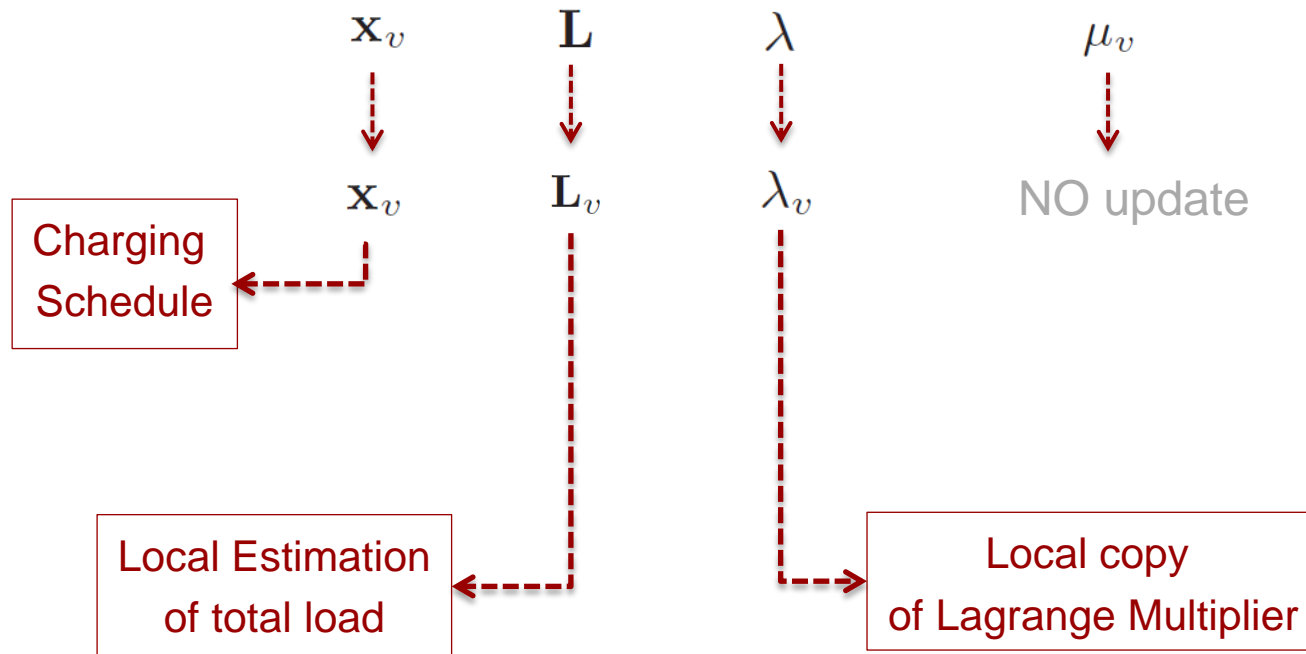
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$$\begin{array}{l} L_v \longrightarrow L^* \\ \lambda_v \longrightarrow \lambda^* \end{array}$$

# Distributed updates



The proposed update for  $\lambda_v$  :

$$\lambda_v(k+1) = \mathbb{P} \left[ \begin{array}{l} \lambda_v(k) - \beta_k \overbrace{\left( \sum_{w \in \Omega_v} (\lambda_v(k) - \lambda_w(k)) \right)}^{\text{neighborhood consensus}} \\ - \underbrace{\alpha_k \left( \frac{\mathbf{L}_v(k)}{V} - \mathbf{x}_v(k) \right)}_{\text{local innovation}} \end{array} \right]_{[c_2, \infty)} .$$

The proposed update for  $\mathbf{L}_v$  :

$$\mathbf{L}_v(k+1) = \frac{\lambda_v(k) - c_2}{2c_1}$$

# Distributed updates



The proposed update for  $\mathbf{x}_v$  :

Step1: Variable update

$$\mathbf{x}_v(k+1) = \mathbb{P}\left[\mathbf{x}_v(k) + \delta_k \left( \frac{\mathbf{L}_v(k)}{V} - \mathbf{x}_v(k) \right) - \eta_k (\lambda_v(k))\right]_{\mathcal{F}}$$



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Step2: Feasibility

$$\begin{aligned} A \cdot \mathbf{x}_v &\leq b_v \\ \underline{x}_v &\leq \mathbf{x}_v \leq \bar{x}_v \end{aligned}$$

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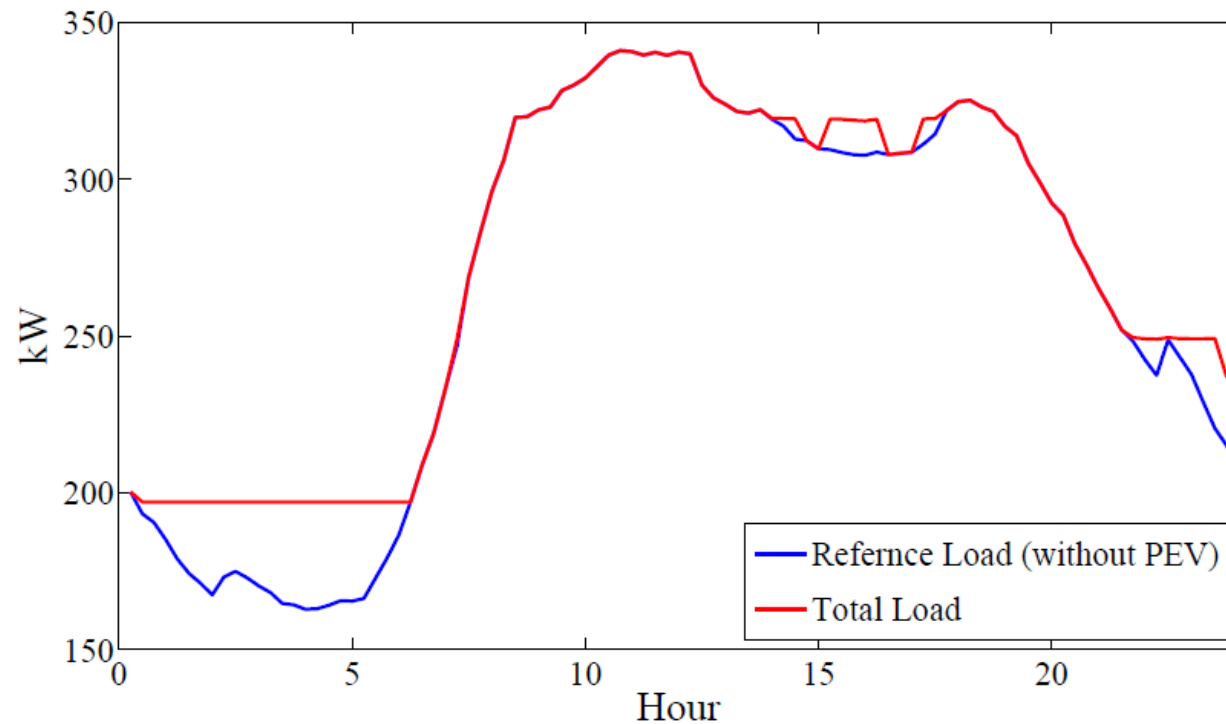
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# Test system

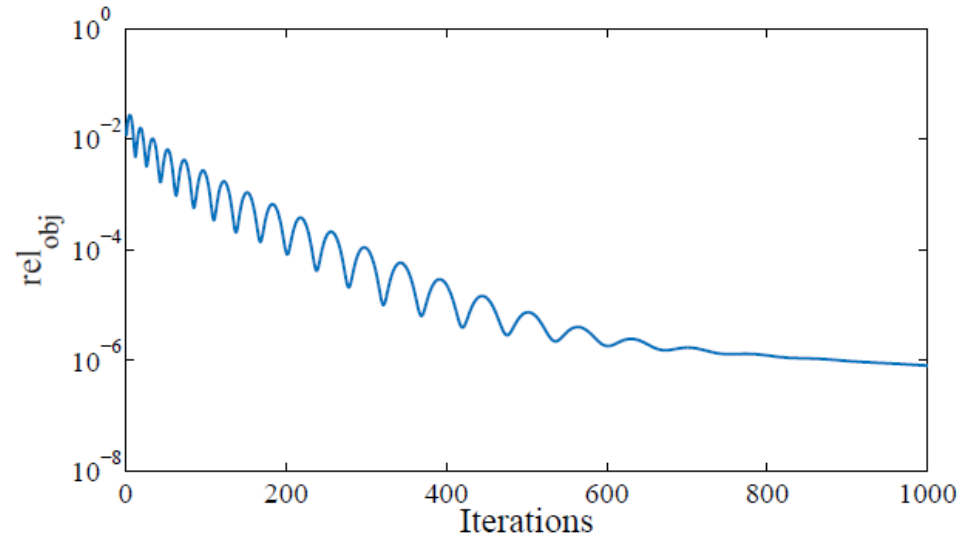
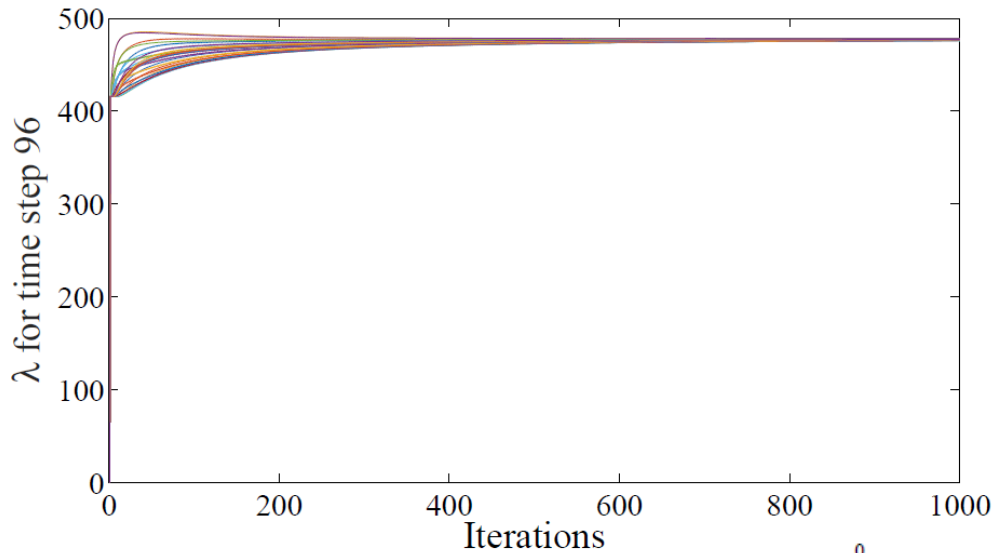
- Fleet of 25 EVs
  - Communication topology: ring graph
- Convergence Measures

$$rel_{obj} = \frac{|f - f^*|}{f^*},$$

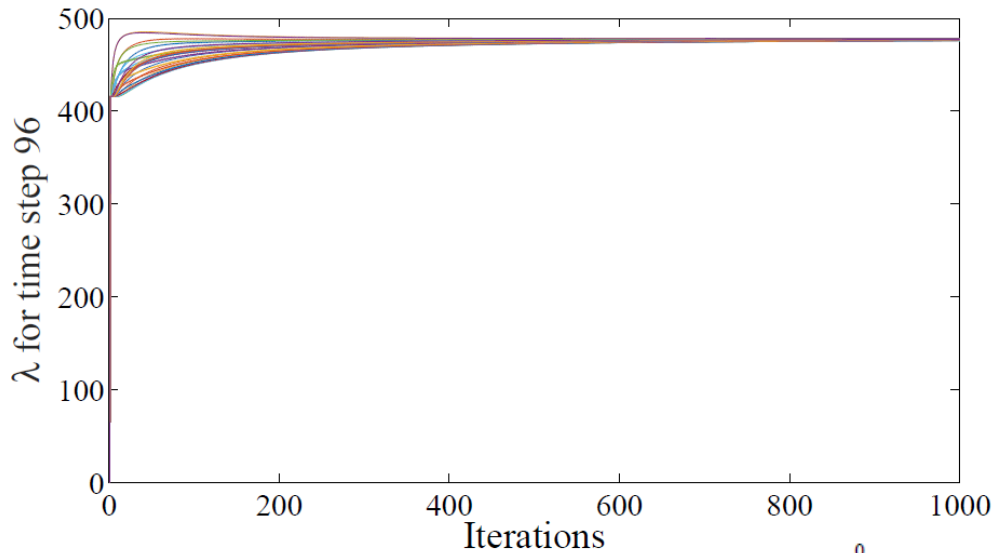
# Valley filling



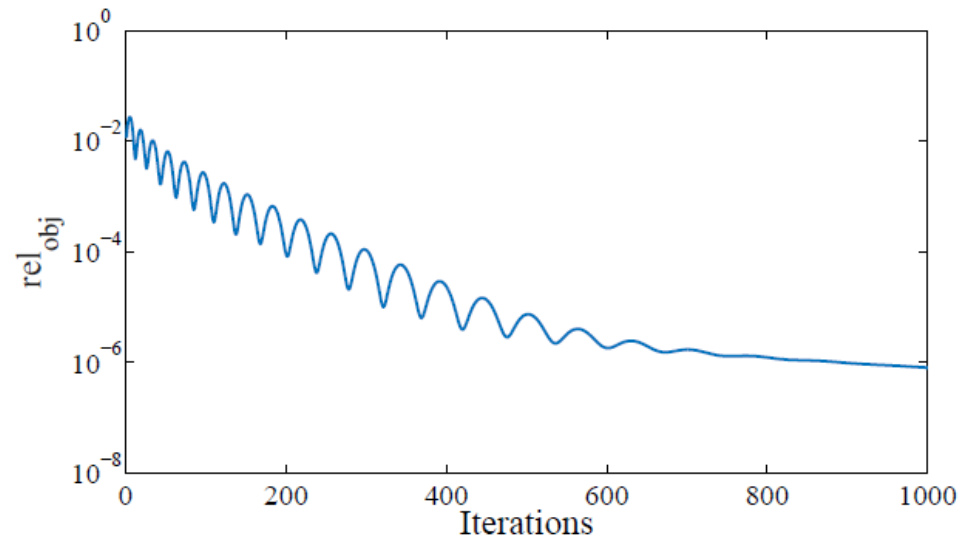
# Convergence analysis



# Convergence analysis



Feasible solution  
at each iteration



# Summary



- Proposed distributed solution:
  - Distributes the computation among distributed entities (PEVs).
  - Each PEV exchanges limited information with a few other PEVs.
  - Achieves feasible solution at each iteration.





# Questions