

# Power Distribution Systems: Optimization, Control and Communication

Baosen Zhang

UC Berkeley

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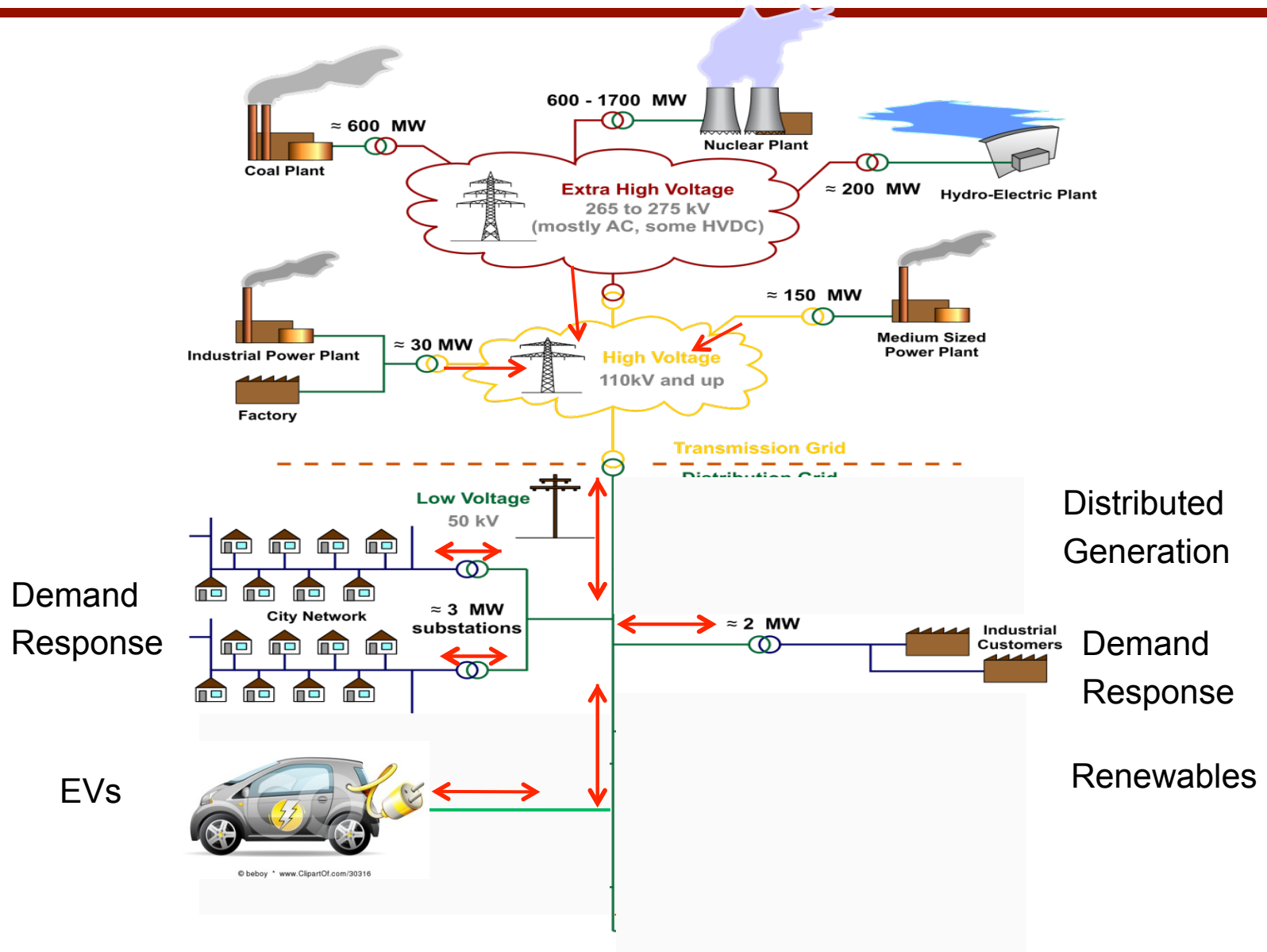
Joint work with David Tse(UCB), Albert Lam (HKBU), Javad Lavaei (Columbia),  
Alejandro Dominguez-Garcia (UIUC).

# Optimization problems in power networks

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- Optimization used for resource allocation
- Increasing complexity:
  - Optimal Power Flow (OPF)
  - Unit Commitment
  - Security Constraint Unit Commitment
- All are done at the level of **transmission** networks
- Smart grid: Optimization in **distribution** networks

# Power flow in the smart grid



# Power Flow

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- Focus on the basic problem: Optimal Power Flow (OPF)
- Review of AC power flow

- Network with  $n$  buses

$$\begin{array}{c}
 i \quad \text{-----} \quad k \\
 \bullet \quad \quad \quad \bullet \\
 y_{ik} = g_{ik} - jb_{ik} = 1/z_{ik}
 \end{array}$$

*Bus matrix:*  $Y \in \mathbb{R}^{n \times n}$

*Complex Voltage:*  $v \in \mathbb{R}^n$

- Power Flow

$$p + jq = \text{diag}(v \underbrace{v^T H Y^T H}_{\text{Quadratic}})$$

# Optimal power flow

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- OPF: optimize over complex voltages

$$\text{minimize } \sum f_i(P_i)$$

subject to Voltage Constraints

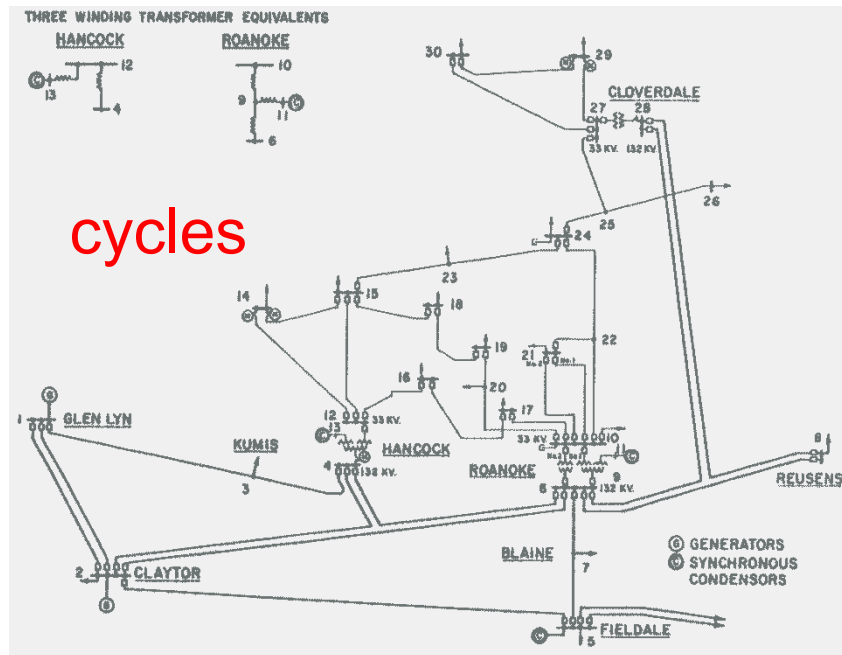
Bus Power Constraints

Network Constraints

- **Non-convex**
- **DC power flow** often used for transmission network
  - Lossless network, ignore reactive power
- Not satisfactory for distribution network
  - Higher losses
  - Reactive power coupled with active power flow

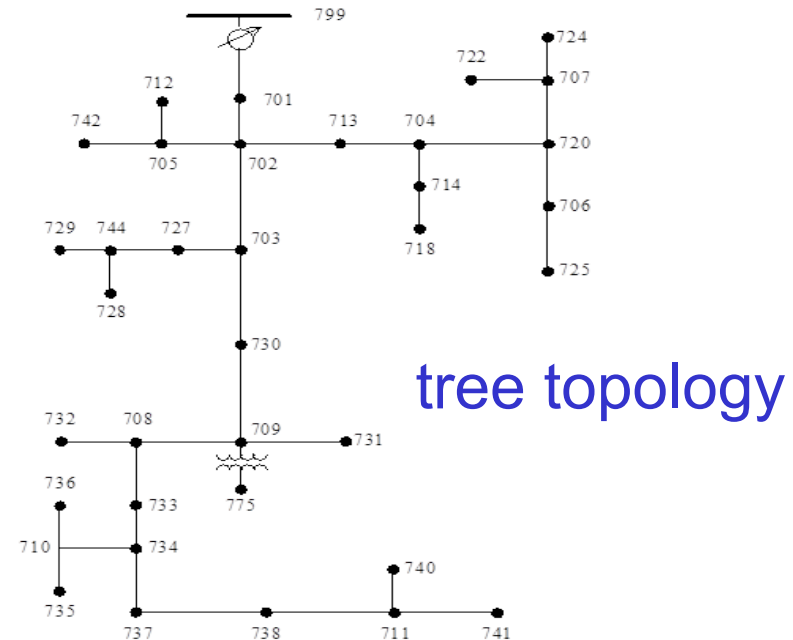
# Transmission vs. distribution

- Transmission Network



- OPF is non-convex, hard

- Distribution Network



- OPF still non-convex
- We show:
  - convex relaxation tight
  - decentralized solution

# OPF on trees: take 1

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**Theorem 1** (Z & Tse., 2011):

Convex rank relaxation for OPF is **tight** if:

- 1) the network is a **tree**
- 2) no two connected buses have tight bus power lower bounds.

(See also: Sojoudi & Lavaei 11, Steven Low's group)

# Proof approach

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- Focus on the underlying **injection region** and investigate its convexity.
- Used a matrix-fitting lemma from algebraic graph theory.

Drawbacks:

- Role of tree topology unclear.
- Restriction on bus power lower bounds unsatisfactory.



# OPF on trees: take 2

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**Theorem 2** (Lavaei, Tse. & Z 13):

Convex relaxation for OPF is **tight** if

- 1) the network is a **tree**
- 2) angle differences along lines are “reasonable”

More importantly:

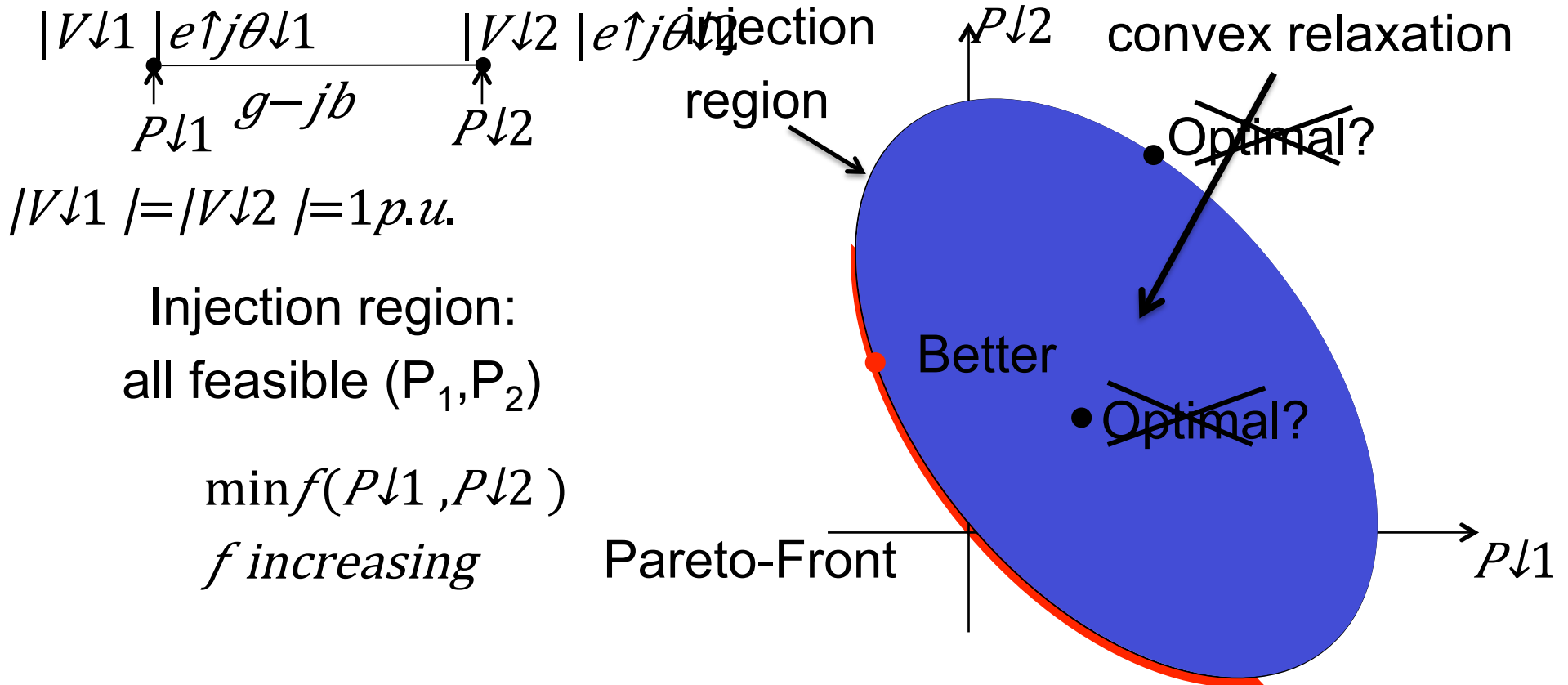
Proof is entirely geometric and from first principles.

# Outline of talk

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- Results on optimal power flow on trees.
- A **geometric** understanding.
- Application to the **voltage regulation** problem in distribution networks with renewables.
- An optimal decentralized algorithm for solving this problem.
- What happens when there is no communication?

# Example: Two Bus Network

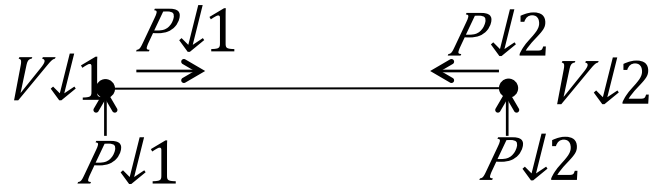


Pareto-Front = Pareto-Front of its Convex Hull

=> convex relaxation is tight.

# Add constraints

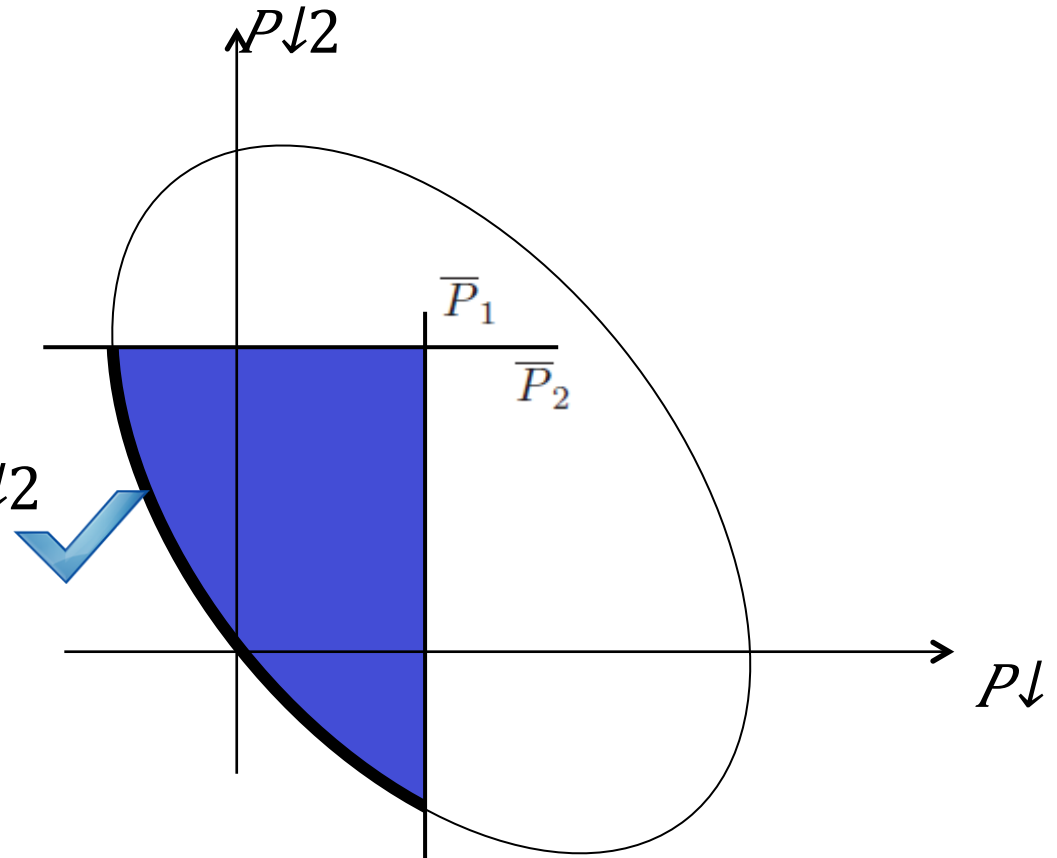
- Two bus network



- Active power upper bounds

$$P1 \leq \bar{P}_1, P2 \leq \bar{P}_2$$

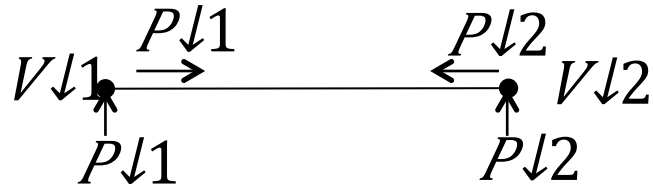
- Reactive power upper bounds



- Power lower bounds

# Add constraints

- Two bus network



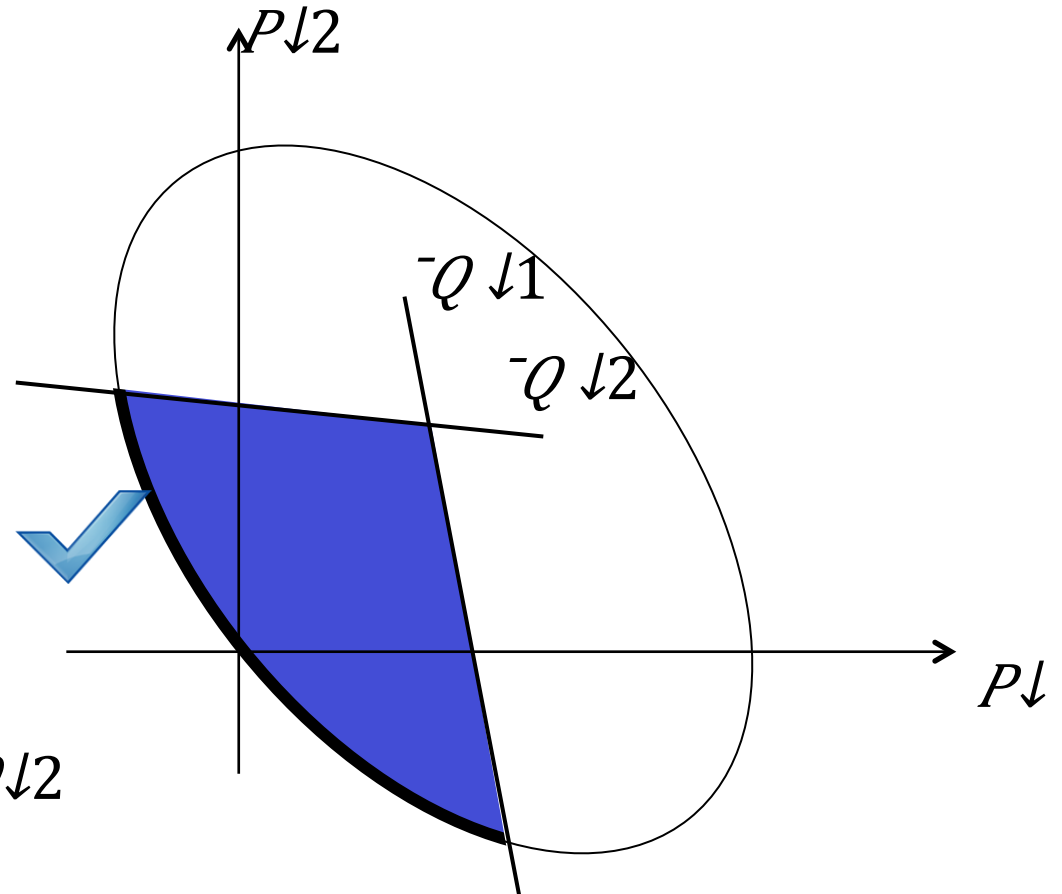
- Active power upper bound



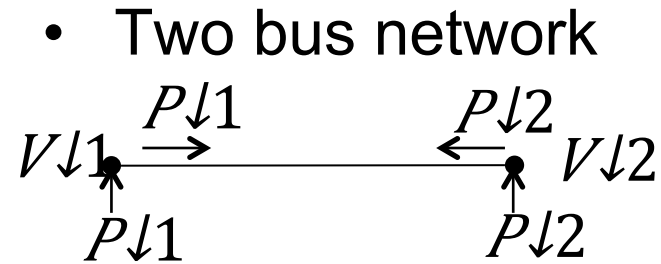
- Reactive power upper bounds

$$Q_1 \leq -Q_1, Q_2 \leq -Q_2$$

- Power lower bounds



# Add constraints



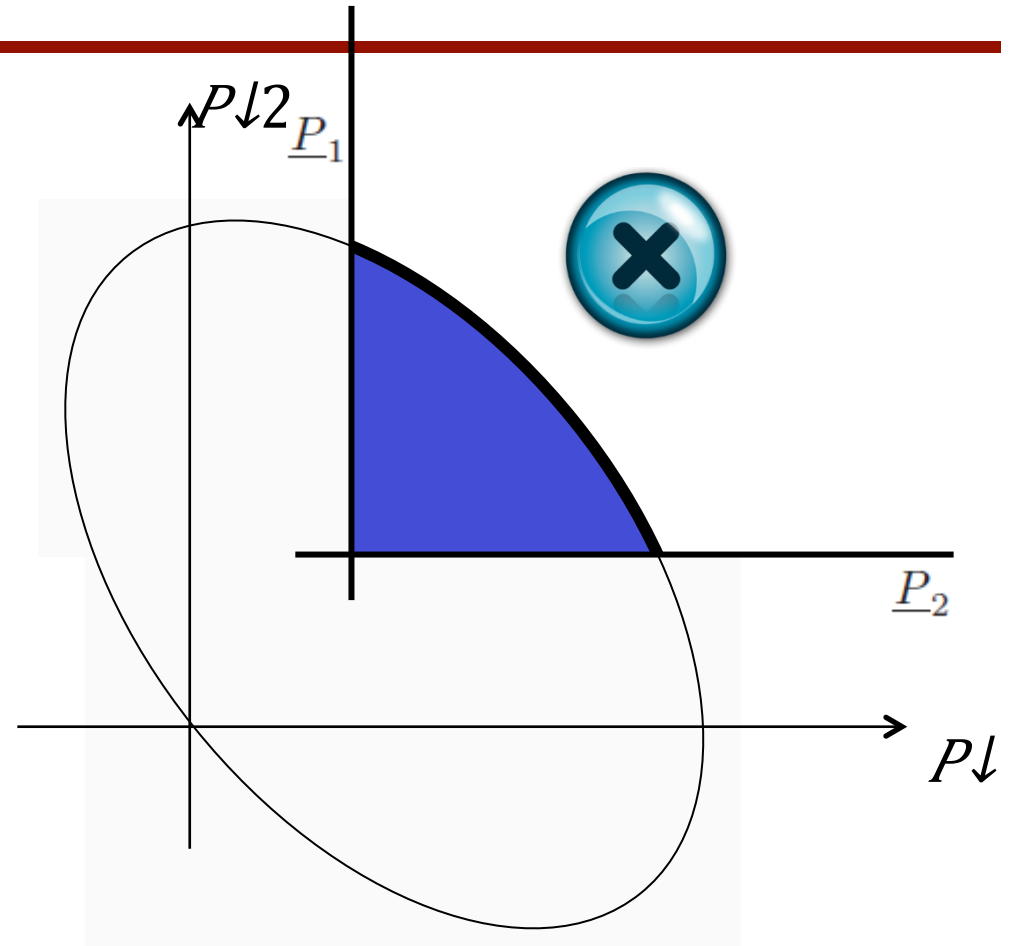
- Loss



- Power upper bounds



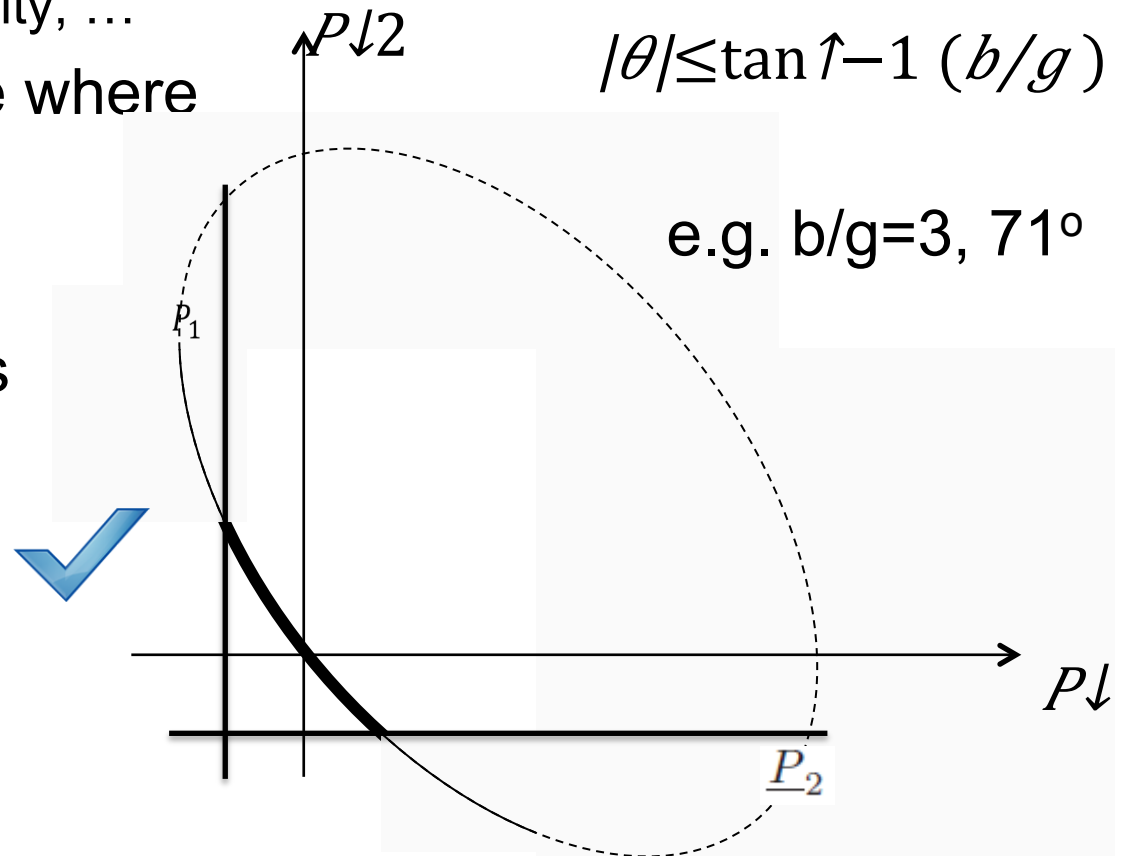
- Power lower bounds



This situation is avoided by adding **angle** constraints

# Angle Constraints

- Angle difference is often constrained in practice
  - Thermal limits, stability, ...
- Only a partial ellipse where all points are Pareto optimal.
- Power lower bounds

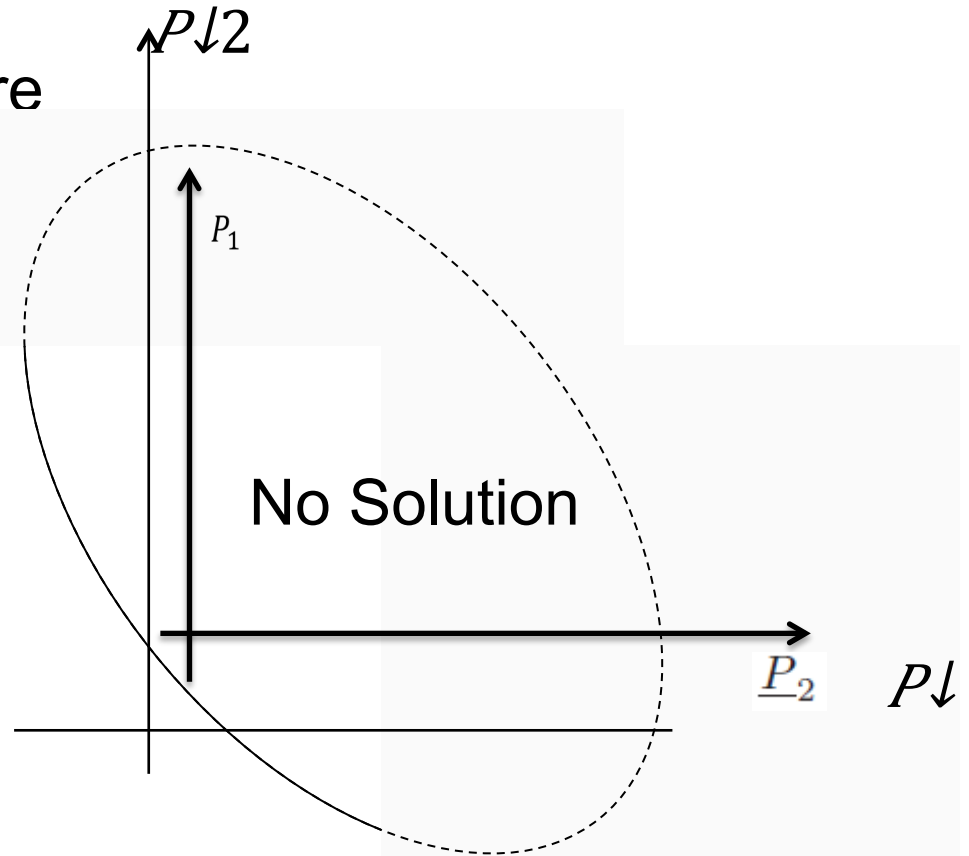


# Angle constraints

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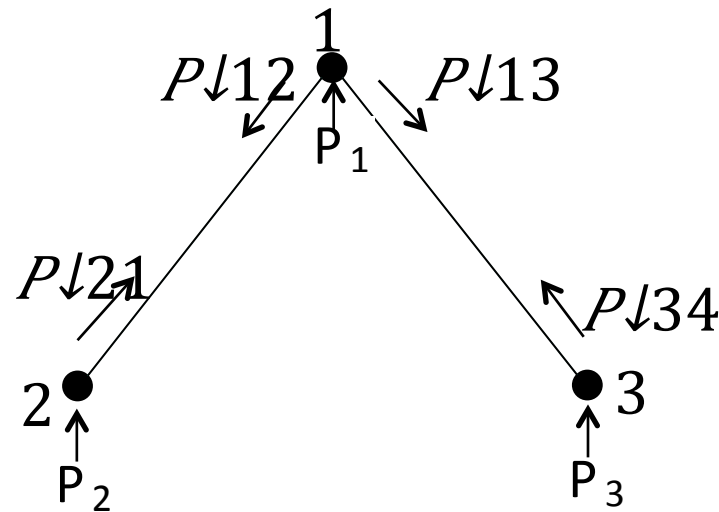


Or Problem is Infeasible





# Injection Region of Tree Networks



$$P_{\downarrow 1} = P_{\downarrow 12} + P_{\downarrow 13}$$

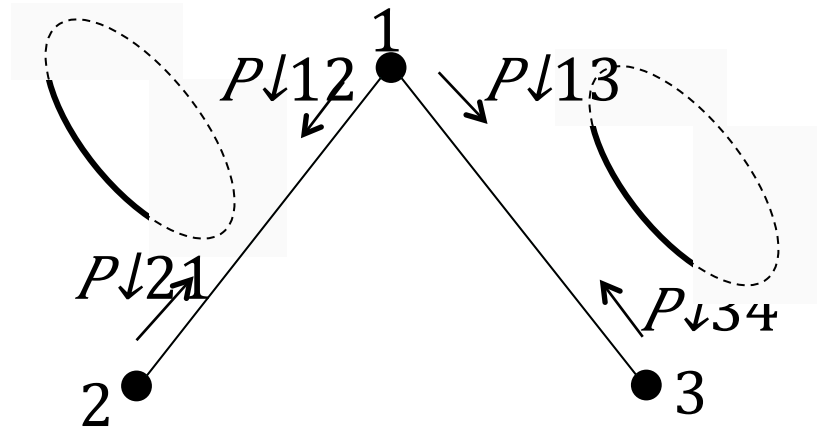
$$P_{\downarrow 2} = P_{\downarrow 21}$$

$$P_{\downarrow 3} = P_{\downarrow 34}$$

- Injections are sums of line flows
- **Injection region** =  
monotone linear transformation of the **flow region**
- Pareto front of **injection region** is preserved under convexification if same property holds for **flow region**.
- Does it?

# Flow Region

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- One partial ellipse per line
- Trees: line flows are **decoupled**

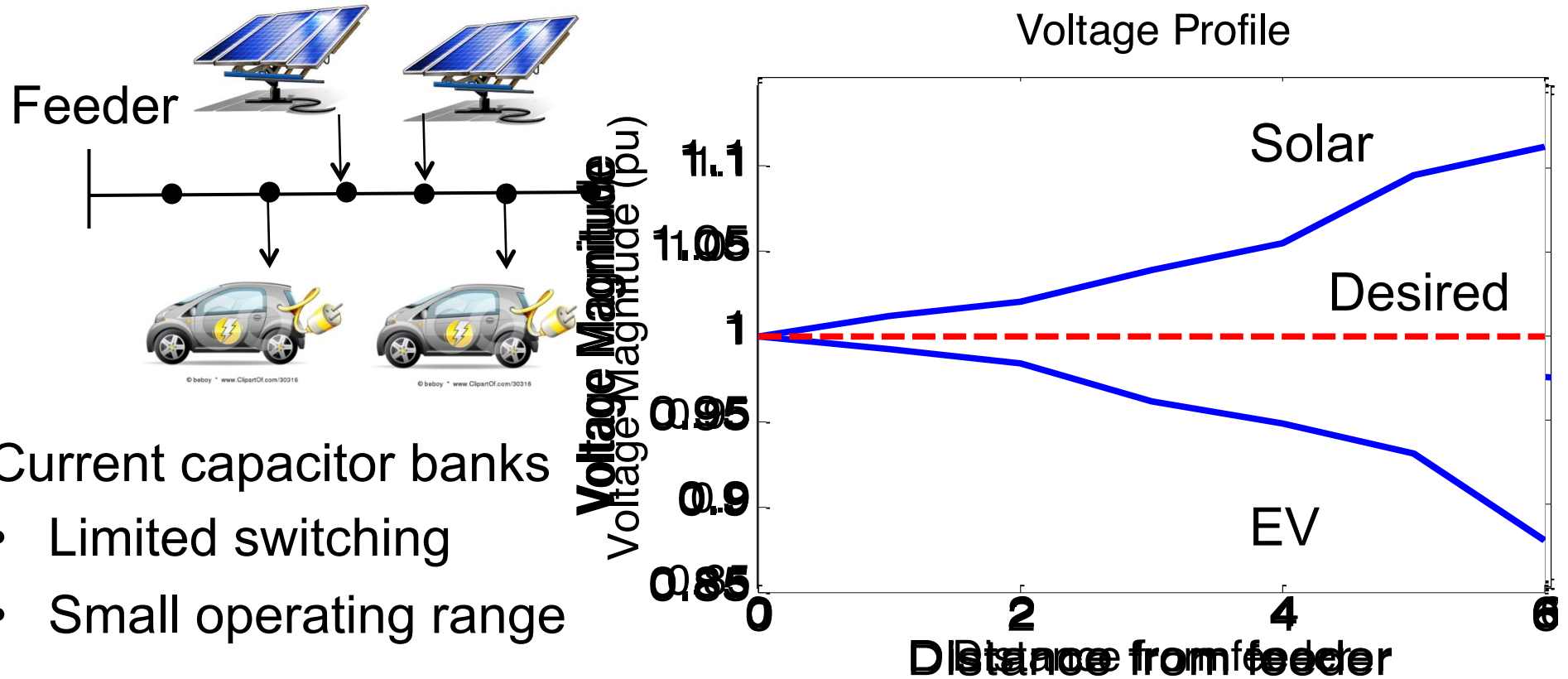
Flow region = Product of  $n-1$  ellipses

Pareto-Front of  
Flow Region = Pareto-Front of  
its Convex Hull

# Application: Voltage regulation

(Lam, Z, Dominguez-Garcia & Tse., 12)

Voltage Profile



Current capacitor banks

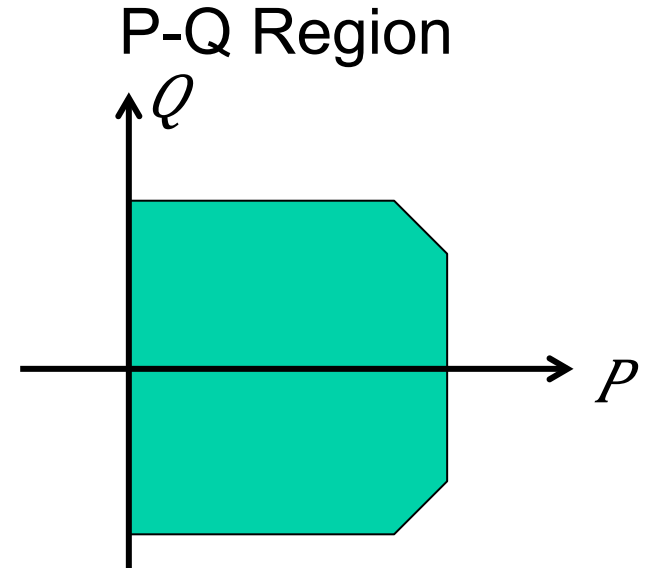
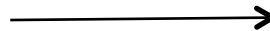
- Limited switching
- Small operating range

Use power electronics to regulate voltage via **reactive power**.

# Power Electronics

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Solar Inverter

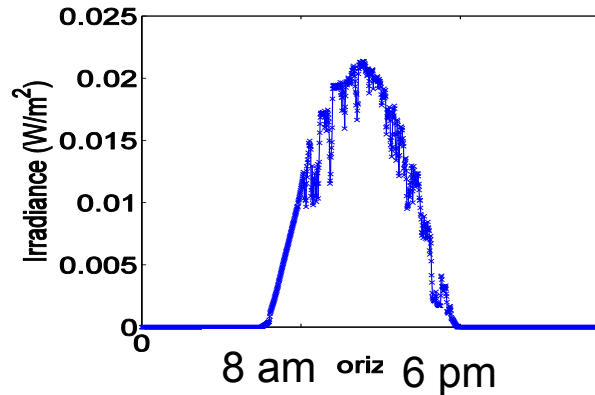
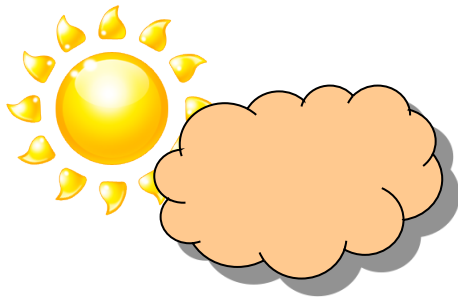


- The reactive powers can be used to regulate voltages

# Random Solar Injections

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- Solar Injections are random



Random

$$\text{Net Load} = \text{Load} - \text{Solar}$$

$$\in [\text{Load} - \text{Solar}, \text{Load}]$$

Random Bus Active  
Power Constraints

# Voltage Regulation Problem

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- Can formulate as an loss minimization problem

minimize  $\sum P_i$  ← System Loss

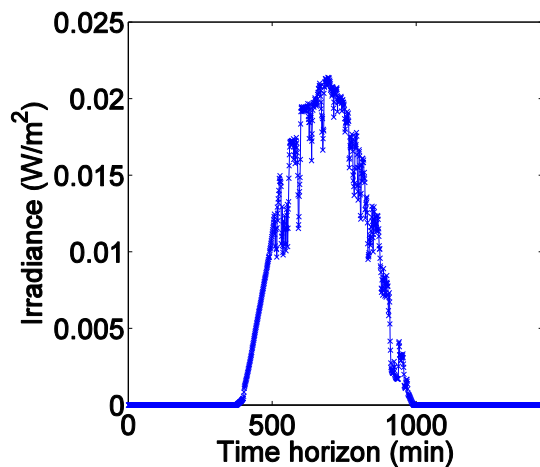
subject to  $|V_i| = 1$  ← Voltage Regulation

$\underline{P}_i \leq P_i \leq \bar{P}_i$  ← Active Power

$\underline{Q}_i \leq Q_i \leq \bar{Q}_i$  ← Reactive Power

Control

Network Constraints



How to do relaxation algebraically?

# Algebraic Representation

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- Everything is linear in  $VV^H$

$$\text{maximize } \sum P_i$$

$$\text{subject to } |V|_i^2 = 1, \forall i$$

$$\underline{P}_i \leq P_i \leq \overline{P}_i$$

$$\underline{Q}_i \leq Q_i \leq \overline{Q}_i$$

$$\mathbf{p} + j\mathbf{q} = \text{diag}(\mathbf{v}\mathbf{v}^H \mathbf{Y}^H)$$

- Replace  $VV^H$  by  $W$

$$W \succeq \mathbf{0}$$

$$\text{rank}(W) = 1$$

Convex rank relaxation

- SDP

# Decentralized Algorithm

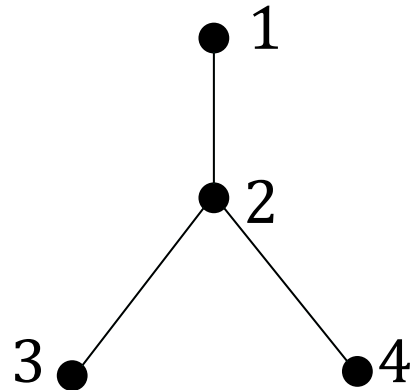
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- Convex relaxation gives an SDP, does not scale
- No infrastructure to transfer all data to a central node
- We exploit the tree structure to derive a **decentralized, asynchronous** algorithm
- Communication along physical topology.

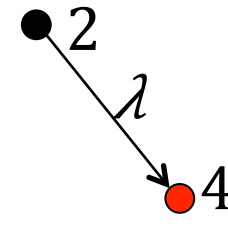
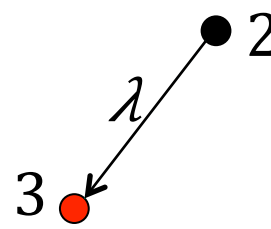
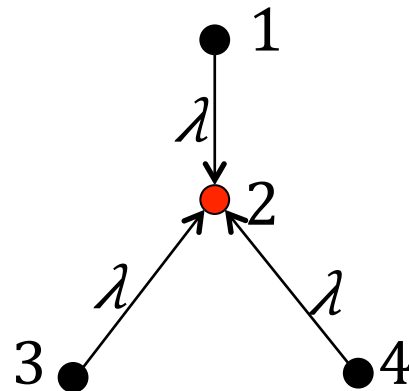
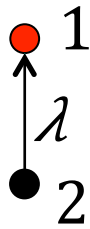


# Example

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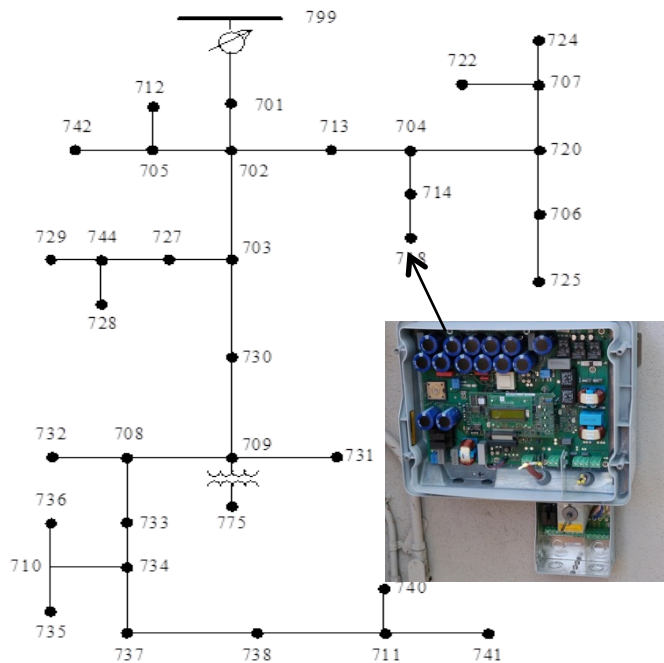
- Each node solves its sub-problem with
  - Its bus power constraints
  - Lagrangian multipliers for its neighbors (line flow constraints)



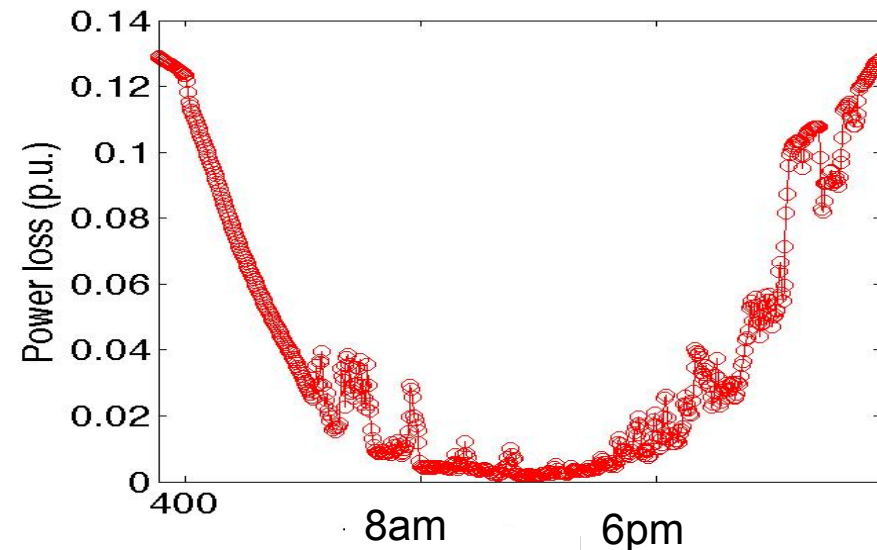
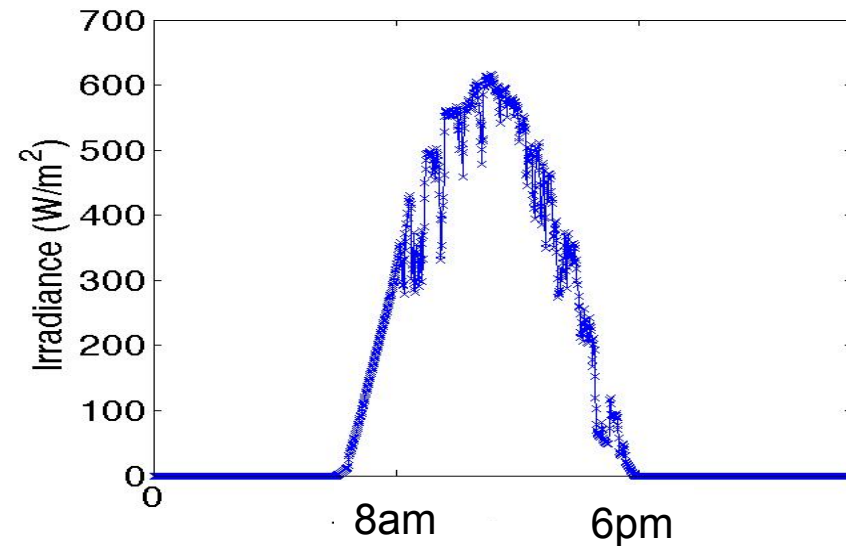
- Update Lagrangian multipliers
- Robust to asynchrony

# Simulations

- 34 Bus Network



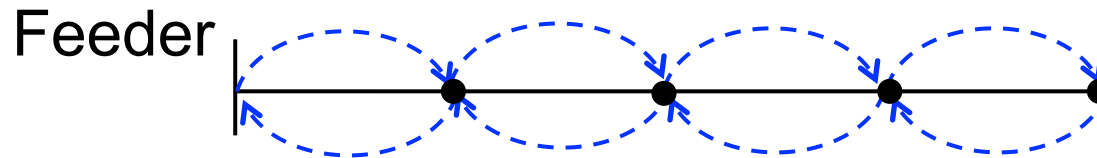
- 2.4 KV
- Bus~ 10 households
- Nominal Loads
- Fixed capacitor banks



# Cyber-physical System

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- An optimal distributed algorithm for voltage regulation



- PG&E: Communication links fails about 50% of the time (dies completely)
- What happens if communication is not complete?
- Can we maintain voltage stability?

# Local Control Scheme

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- No real-time communications at all (today's system)
- Solar and EV penetrations are increasing
- How do maintain **voltage stability**?
- Each bus senses its voltage, adjusts its reactive power (Active power not changed)
- Are **local actions** enough?

# Iterative Algorithm

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At time  $t$ ,

- If  $|V_{i,j}[t]| < 1$ , Increase  $Q_{i,j}$
- If  $|V_{i,j}[t]| > 1$ , Decrease  $Q_{i,j}$

Update algorithm

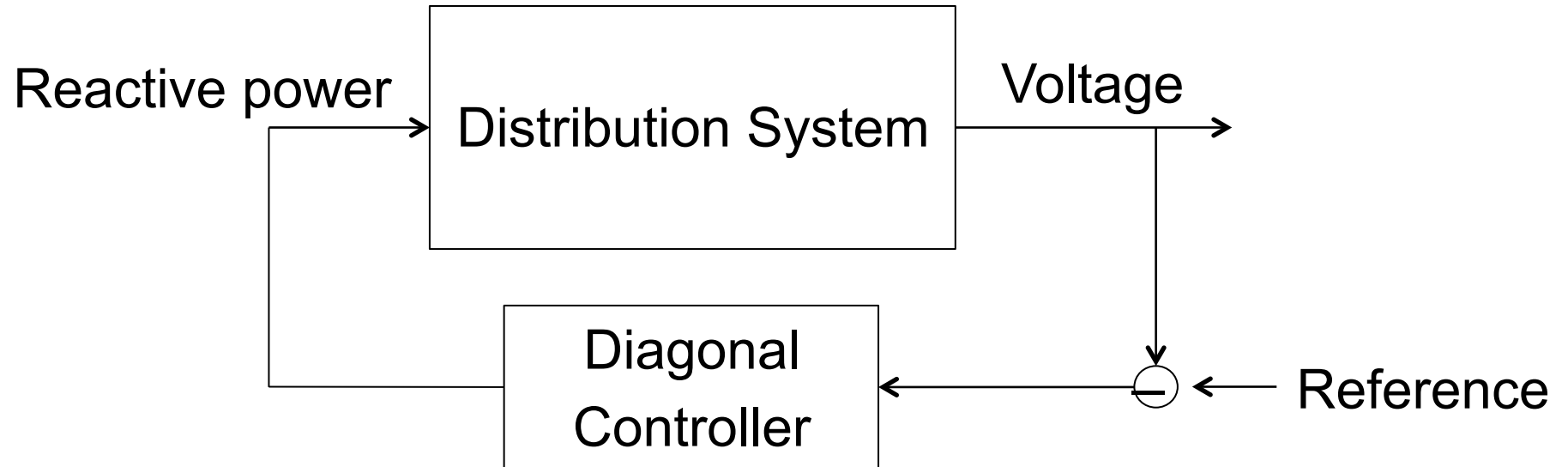
$$Q_{i,j}[t+1] = Q_{i,j}[t] + \overset{\text{Gain}}{d_{i,j}} (1 - |V_{i,j}[t]|)$$

Question: Does this algorithm ever terminate?

We show sufficient and necessary conditions

# Dynamical System

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- Linearize the system

$$v=Aq$$

- Matrix  $A$  depends on active powers

# Linear System

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- Given active powers,

$$v=Aq$$

- Does there exist a **diagonal controller** to stabilize the system?
- Does there exist  $D$ , diagonal, such at

$$DA+A^T D > 0$$

- Given  $p$ , easy to check

# No Communication

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- Active powers lies in a region

$$P_{\downarrow i} \leq P_{\downarrow i} \leq \bar{P}_{\downarrow i}$$

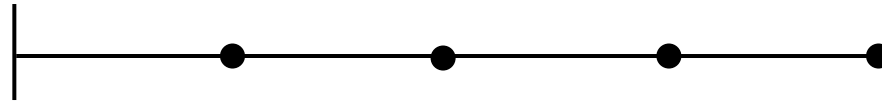
- One diagonal controller needs to work for **all active powers**
- Theorem: It is sufficient to design a controller with respect to  $(P_{\downarrow 1} , P_{\downarrow 2} , \dots, P_{\downarrow n} )$
- Proof not trivial, careful analysis of the system matrix



# Stability Regions

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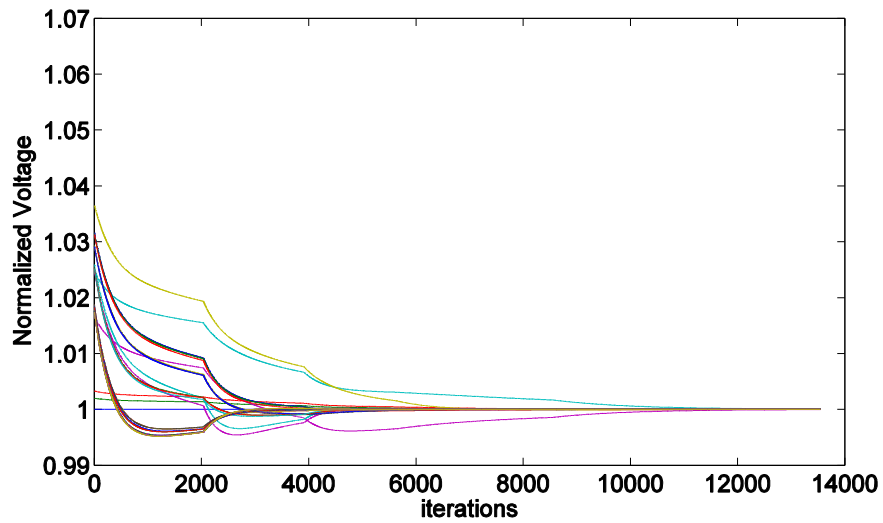
- Stability region: set of all active powers that can be stabilized by some diagonal controller
- Line networks are hardest to control



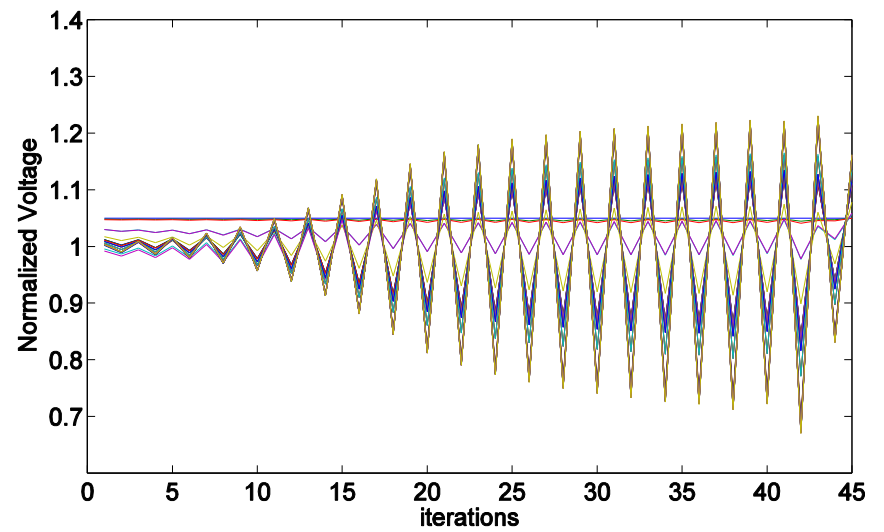
- Size of stability region depends on the depth of the network
- As length go to  $\infty$ , stability region goes to a point

# Simulation

- IEEE 34-bus test feeder



Normal Load



5x the Load

# Summary

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- Geometrical view of power flow
- Optimization problems on tree networks can be convexified
- Applied to design an optimal distributed algorithm for voltage regulation.
- Communication important for long networks