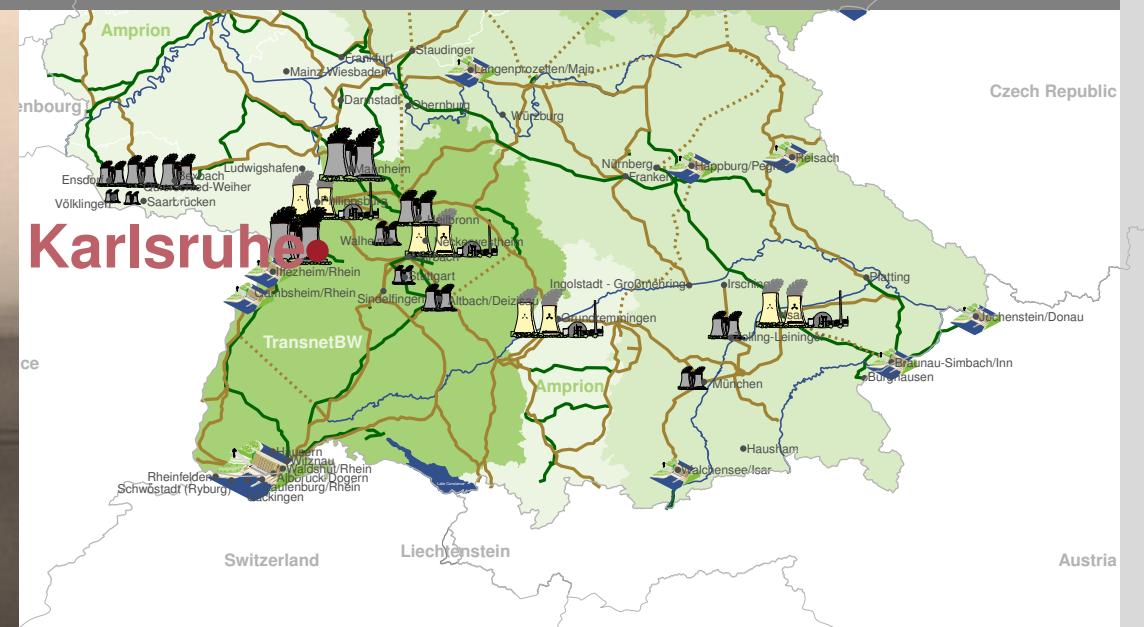


# Placement and Planning Problems in Power Grids

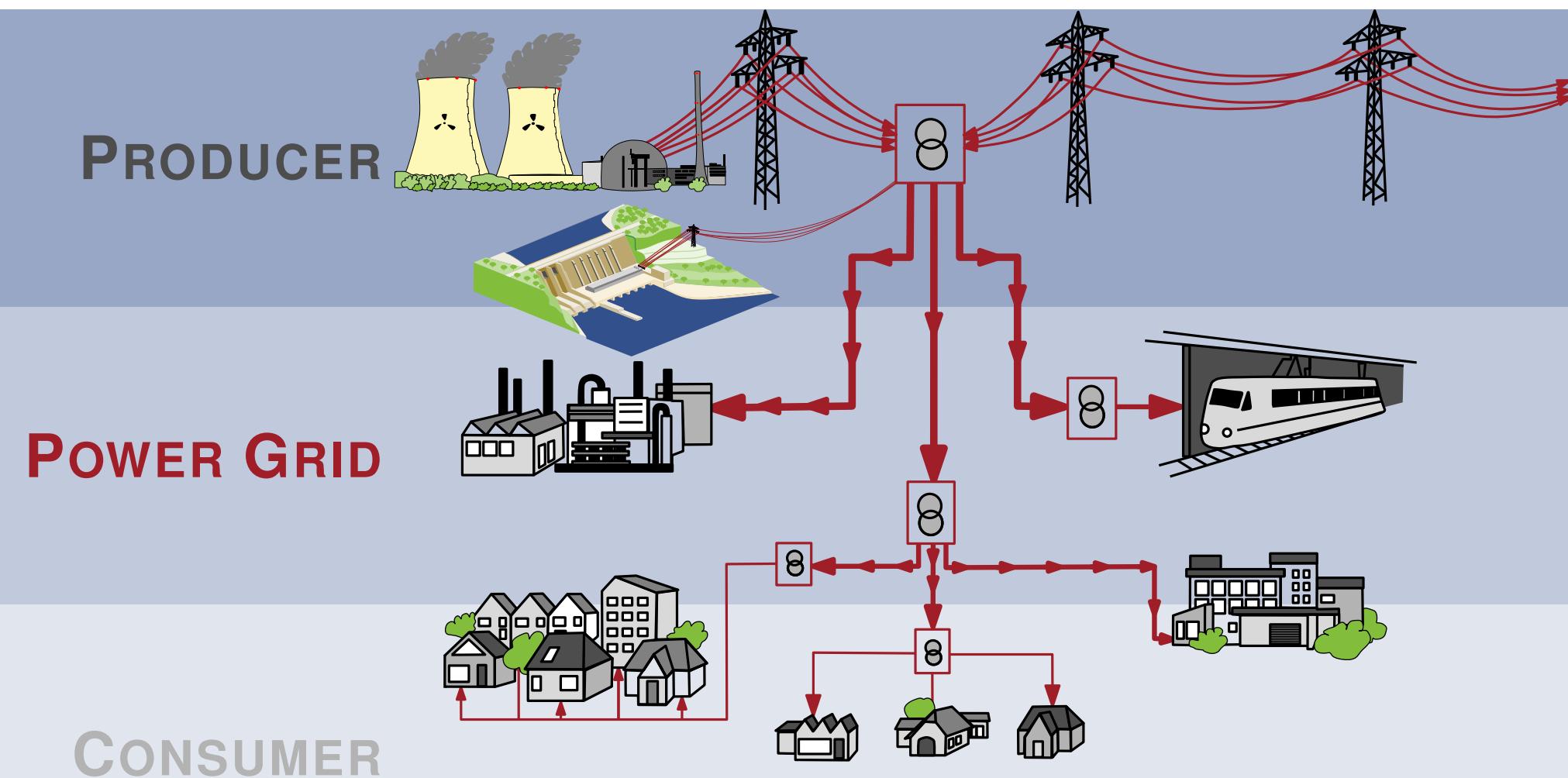
FRICO · 16. August 2017

Franziska Wegner

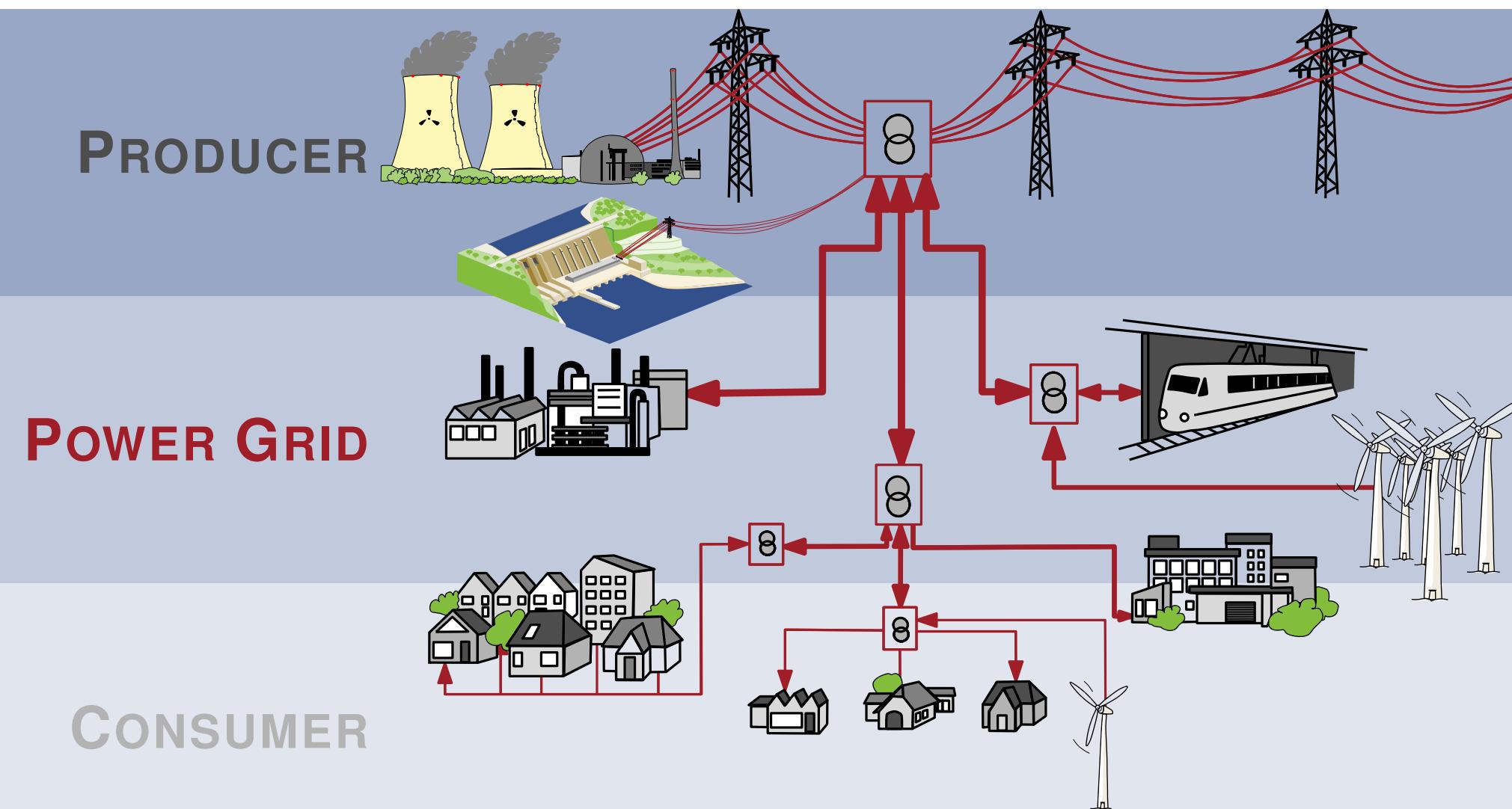
INSTITUTE OF THEORETICAL INFORMATICS · ALGORITHMIC GROUP



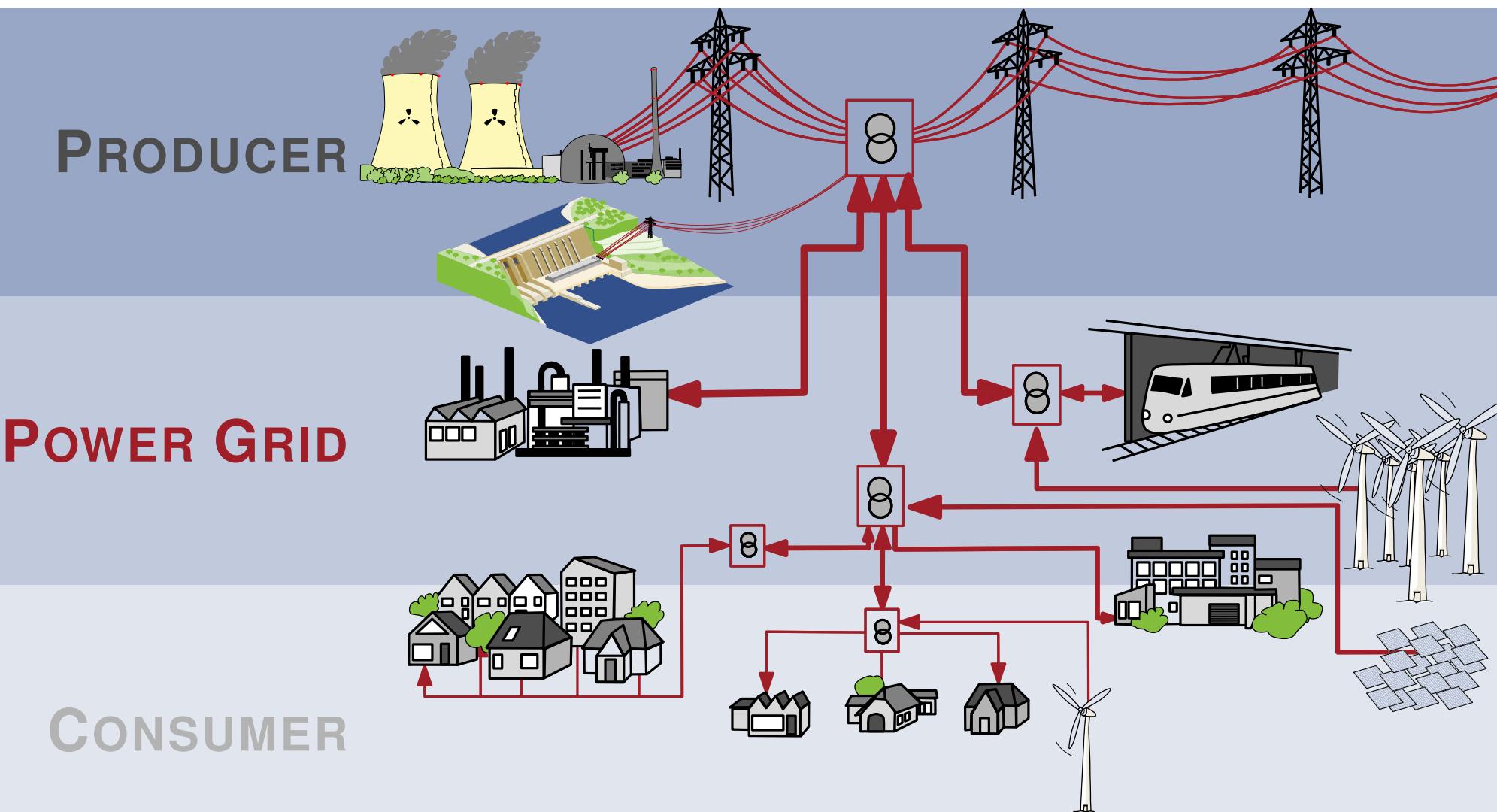
# Recent Development in Power Grids



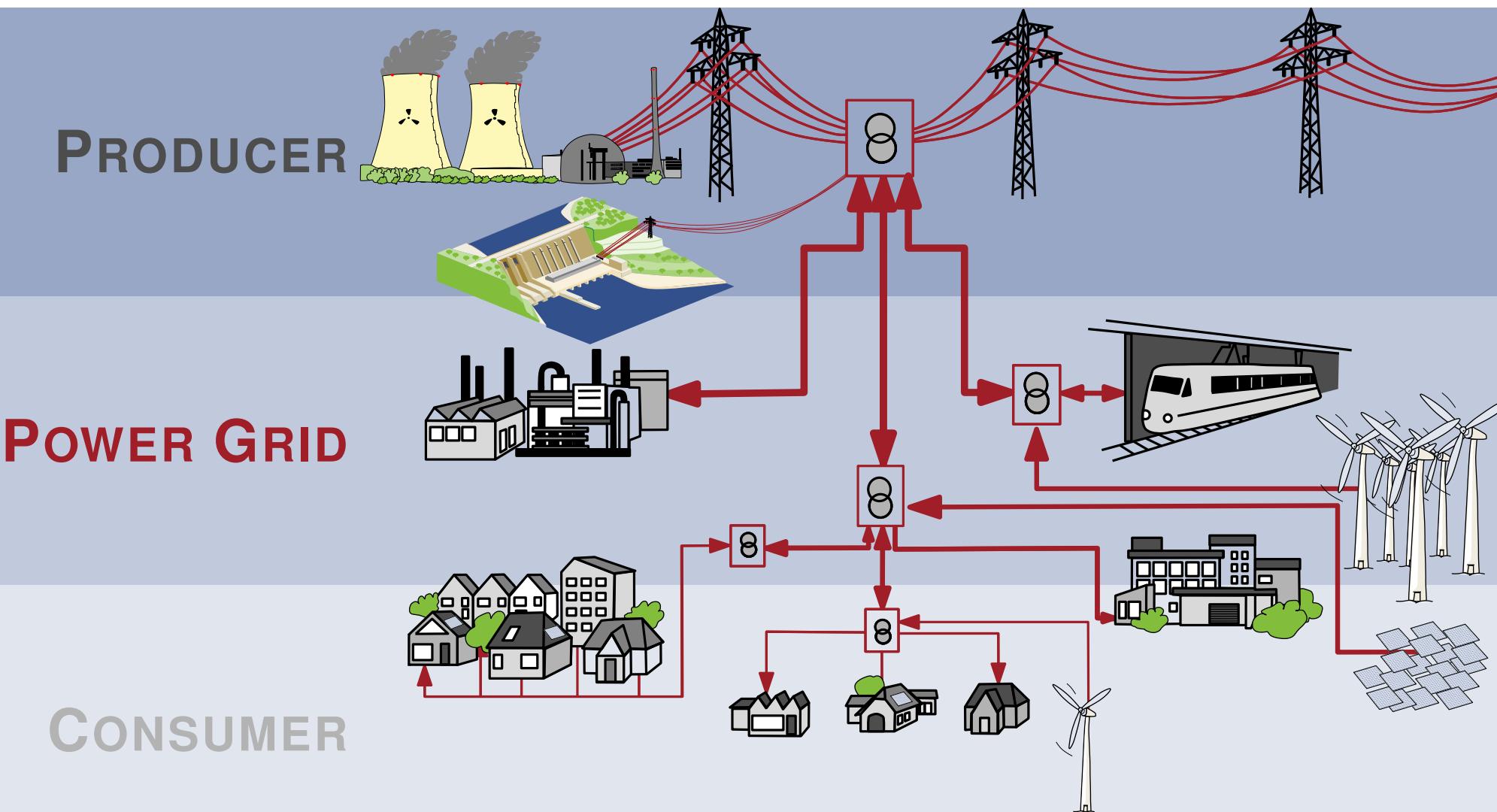
# Recent Development in Power Grids



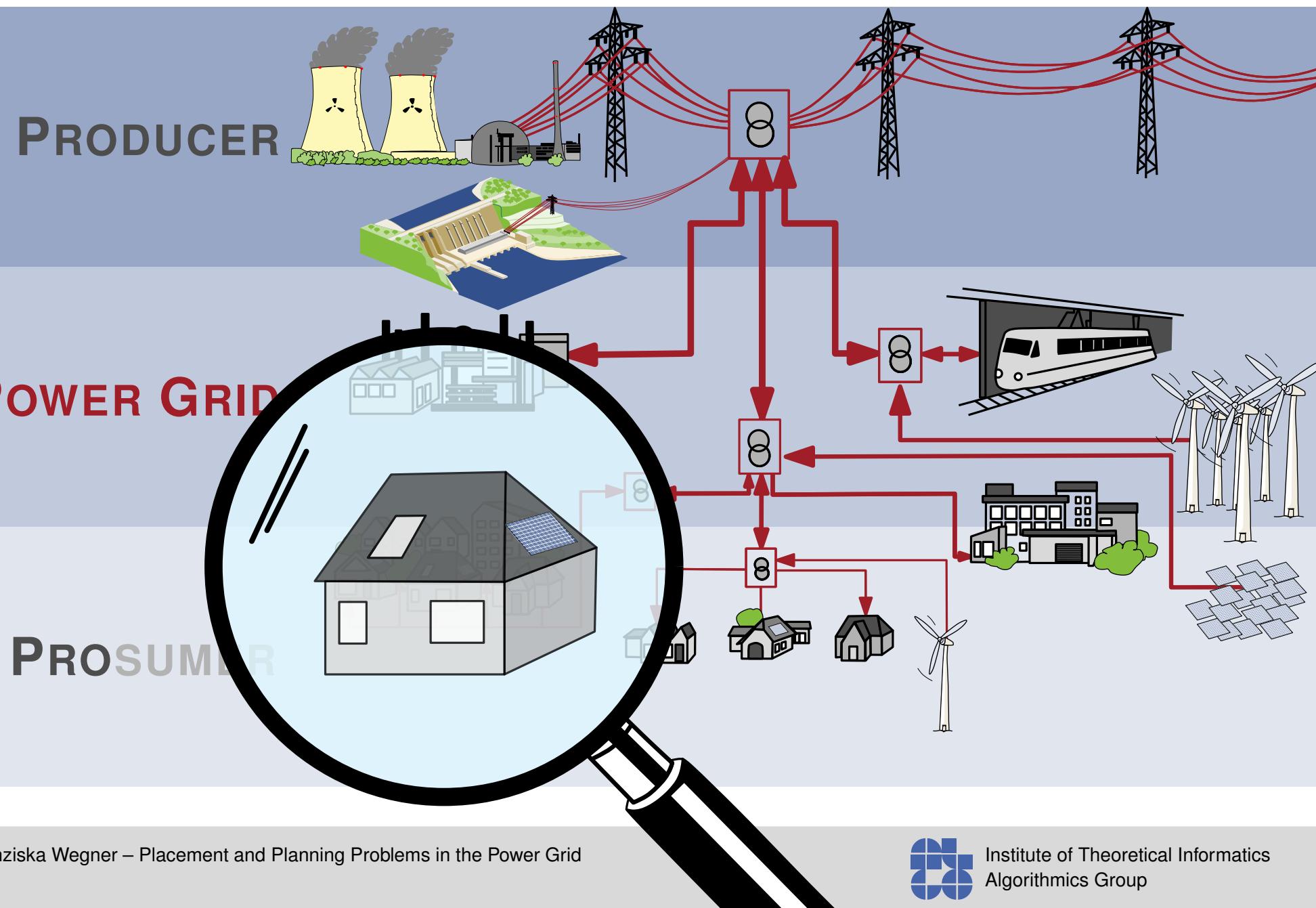
# Recent Development in Power Grids



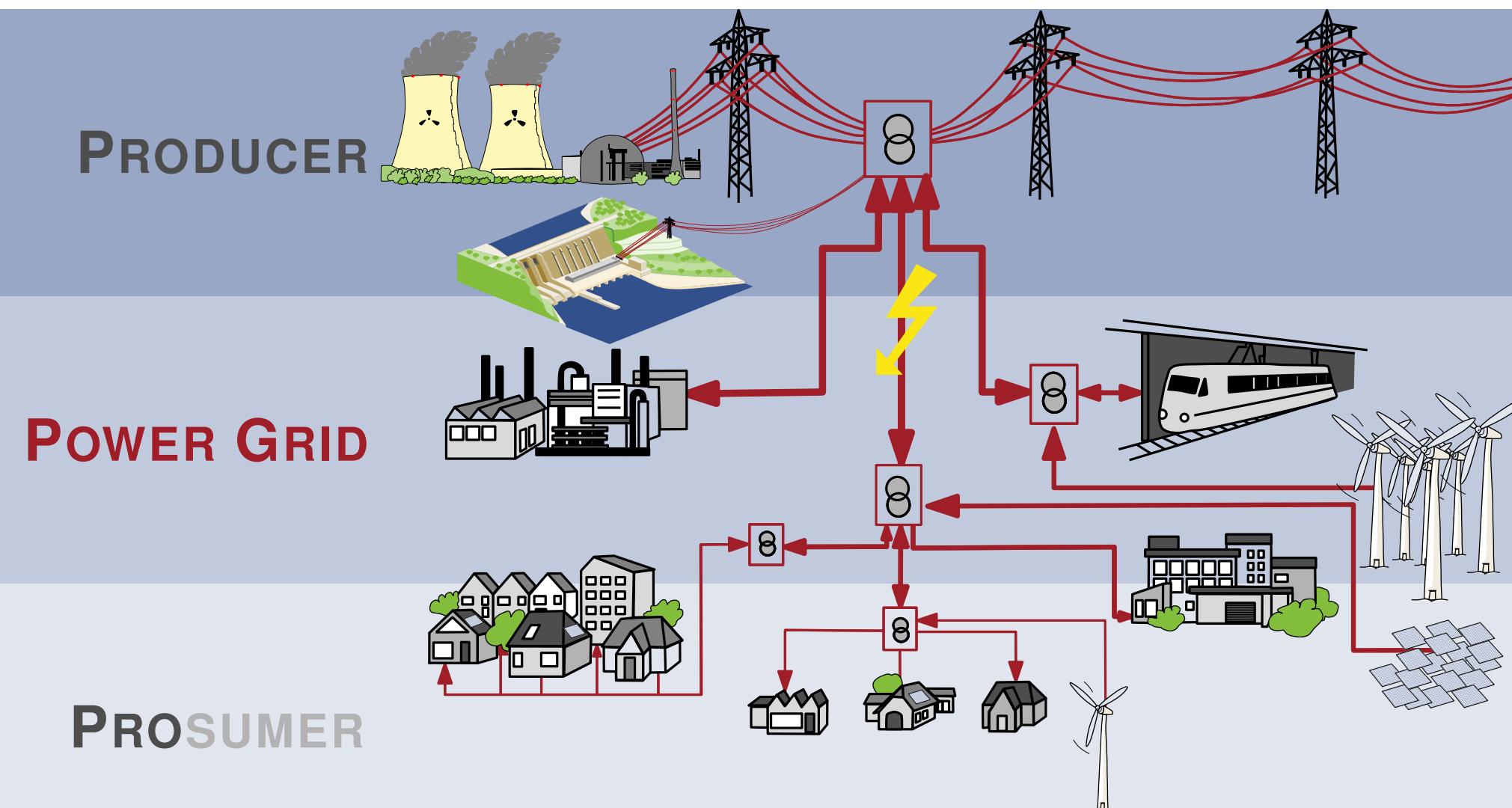
# Recent Development in Power Grids



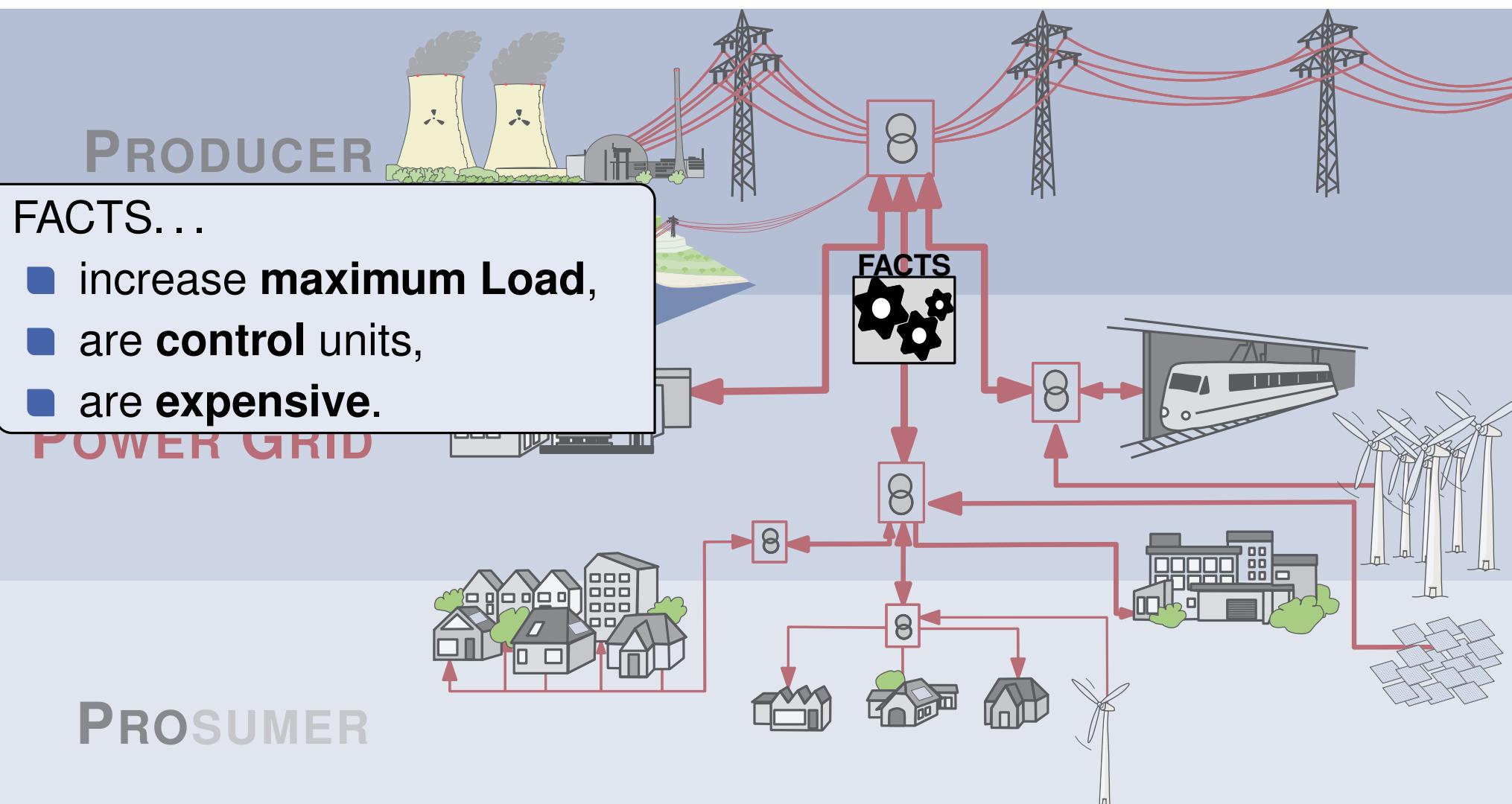
# Recent Development in Power Grids



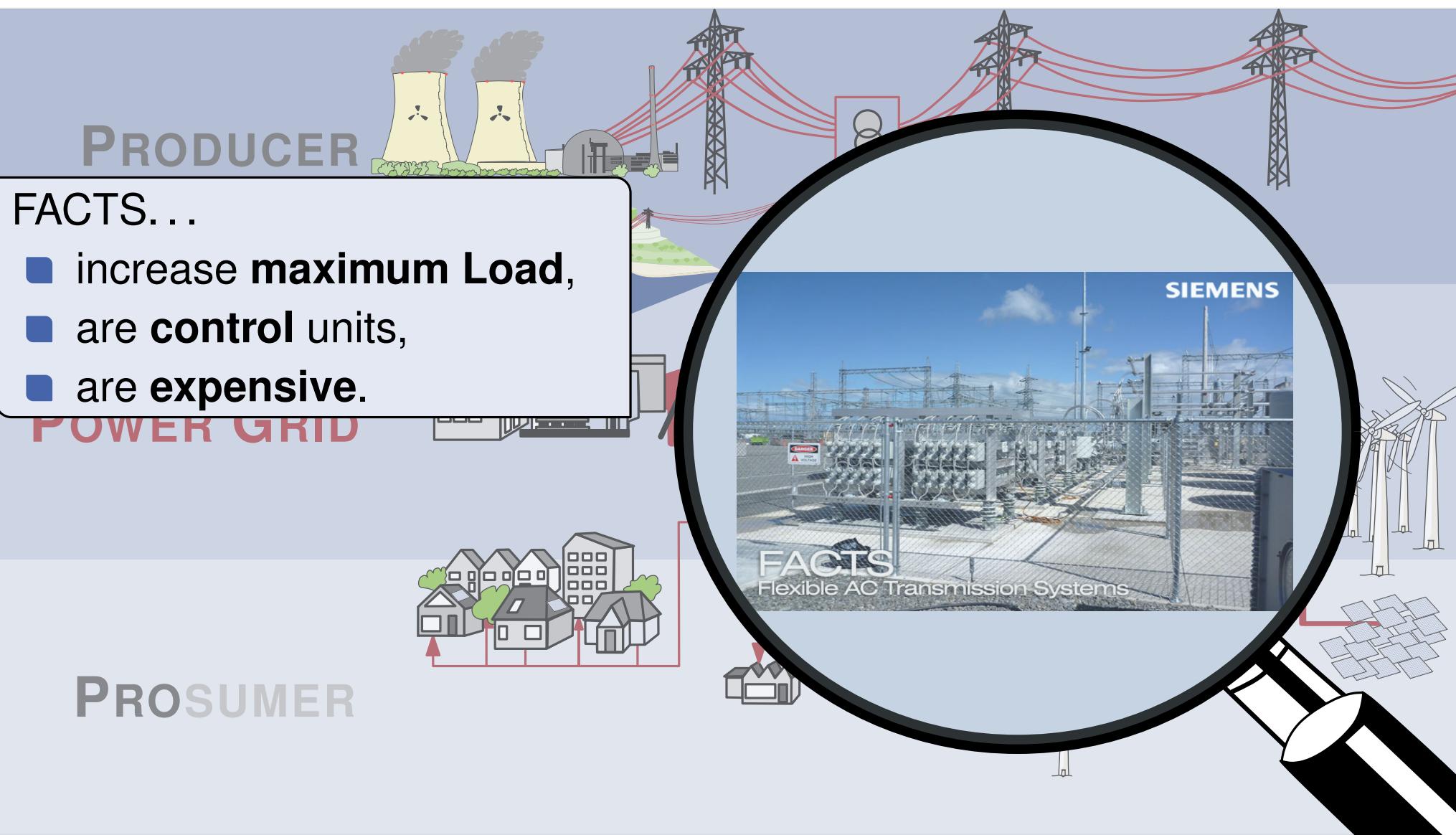
# Recent Development in Power Grids



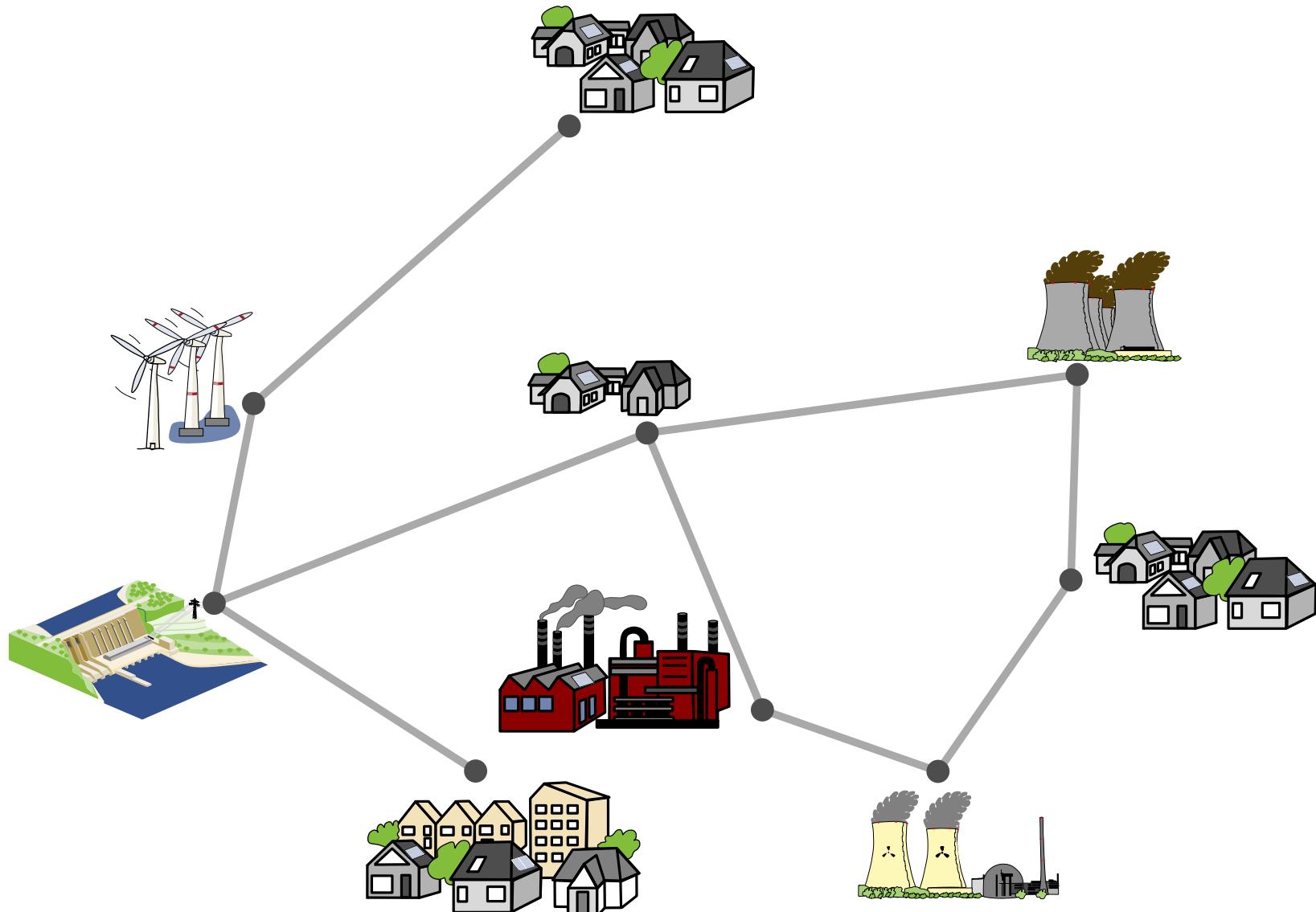
# Recent Development in Power Grids



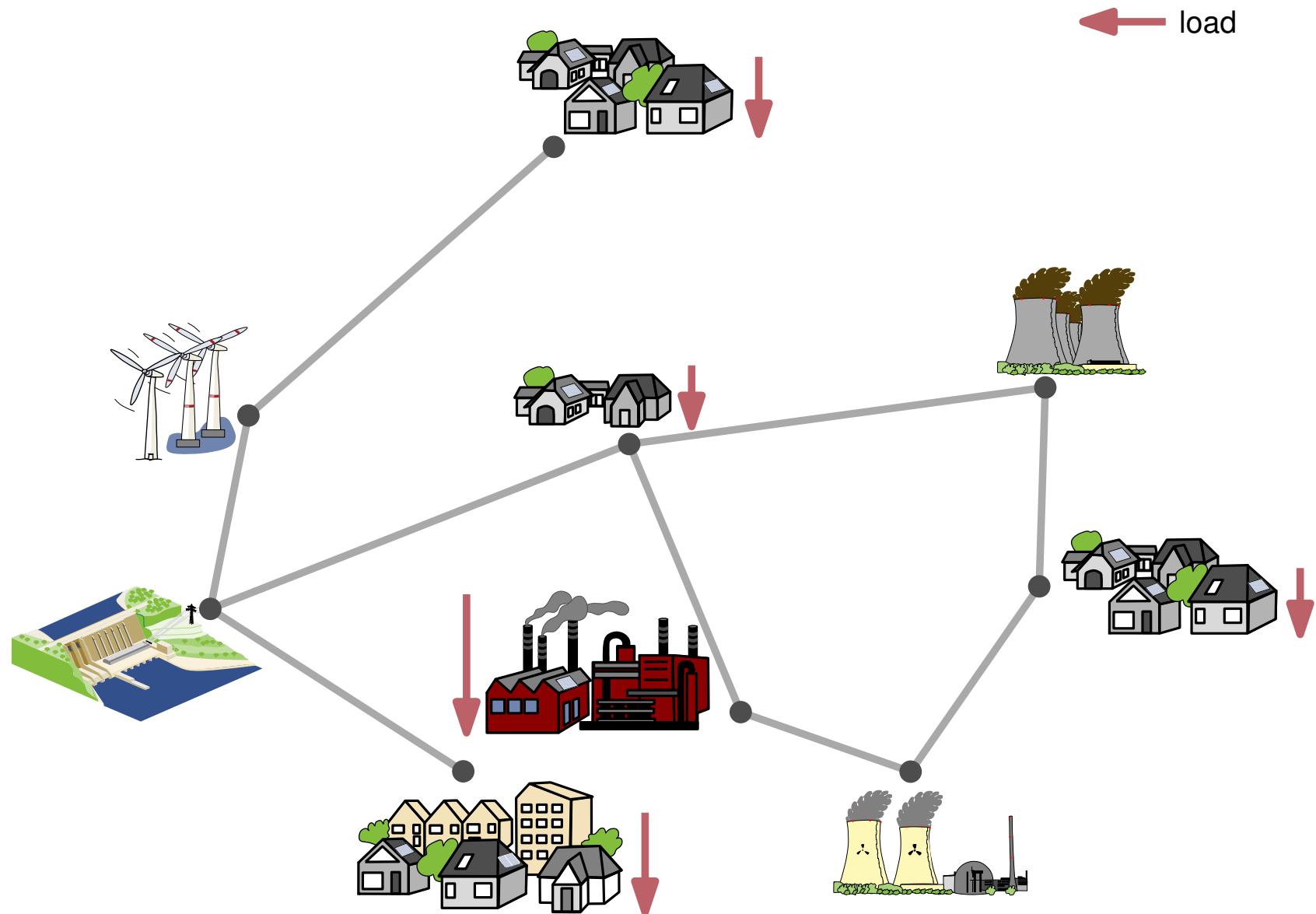
# Recent Development in Power Grids



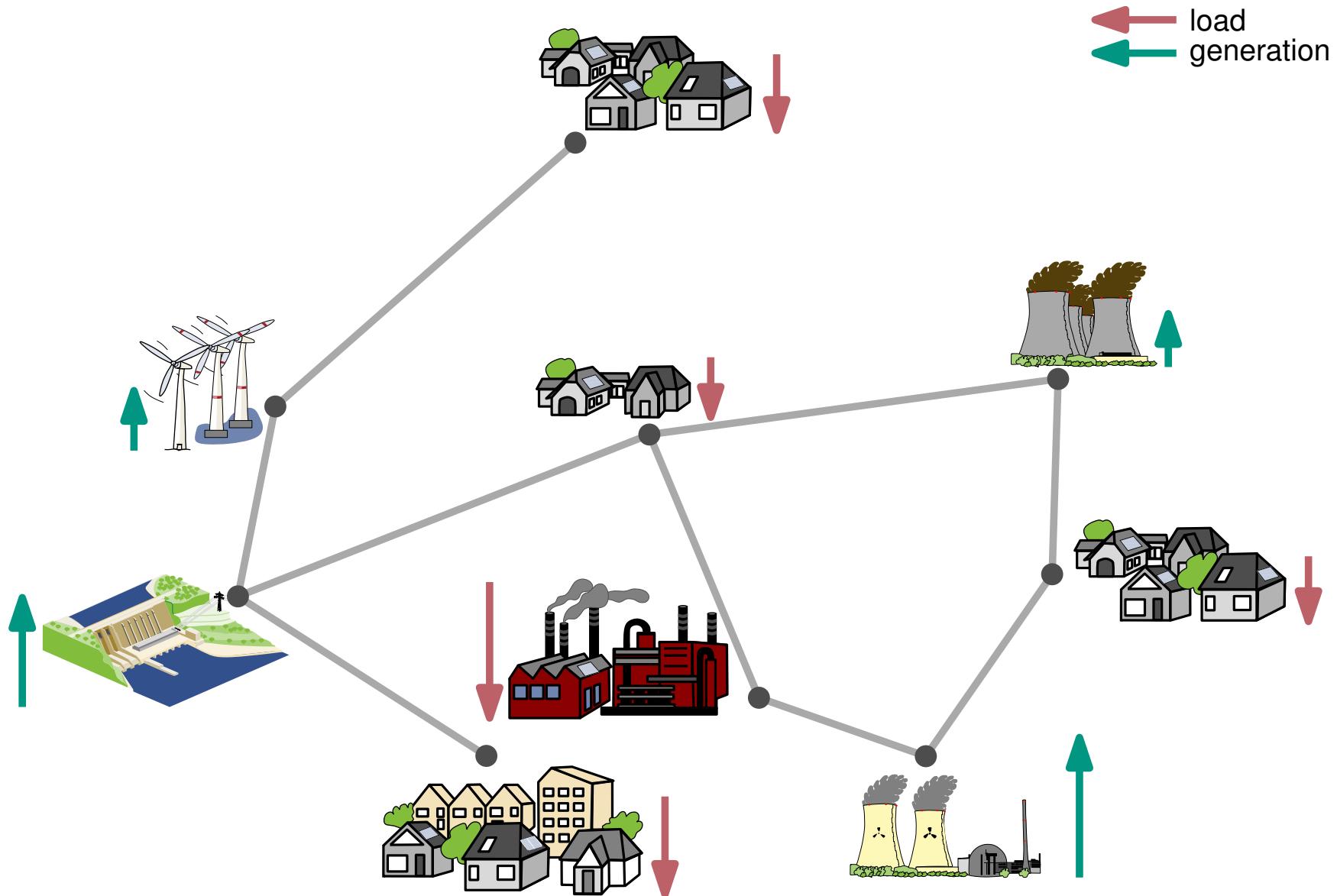
# AC Conservation of Flow in Power Grids



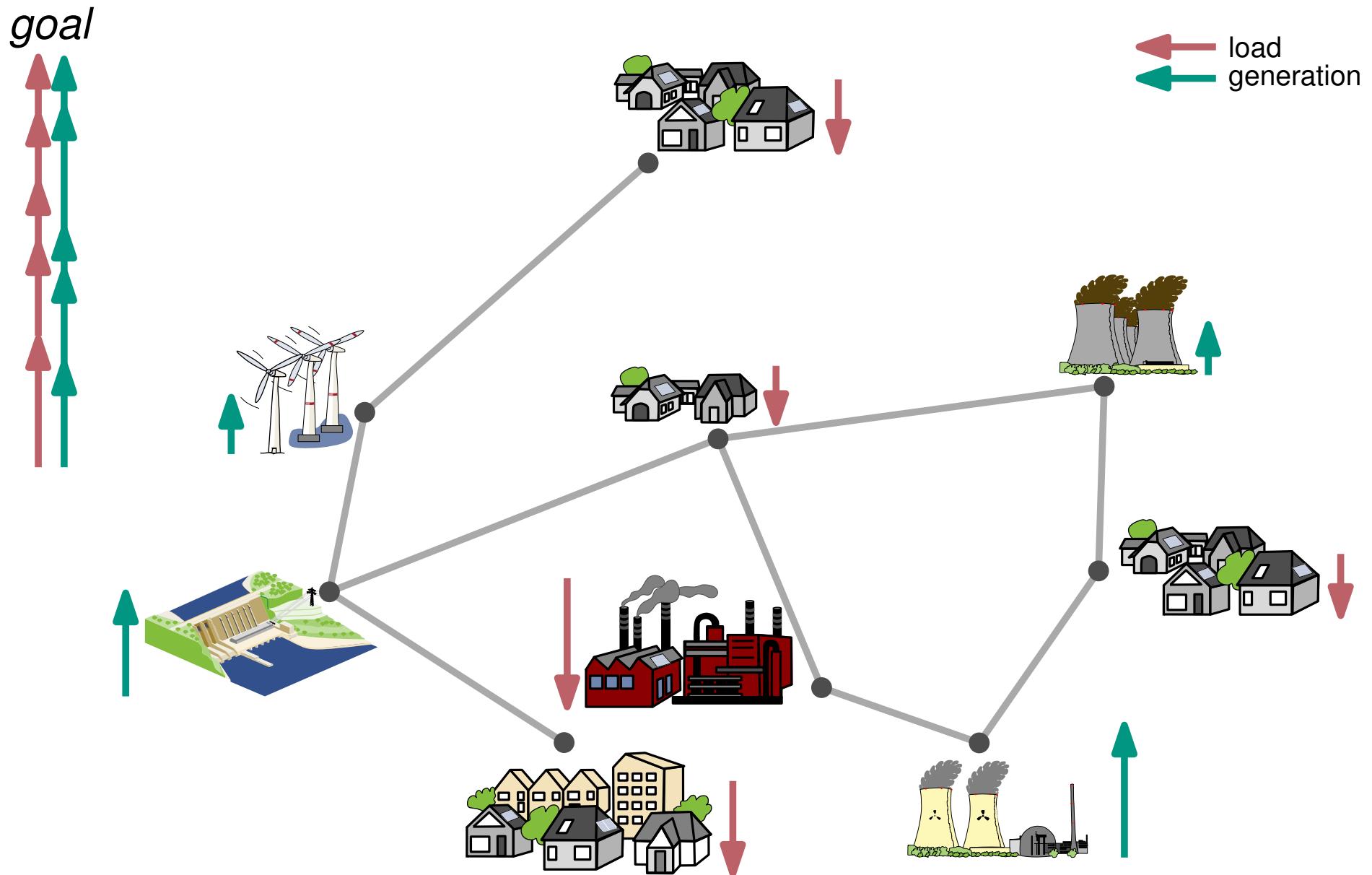
# AC Conservation of Flow in Power Grids



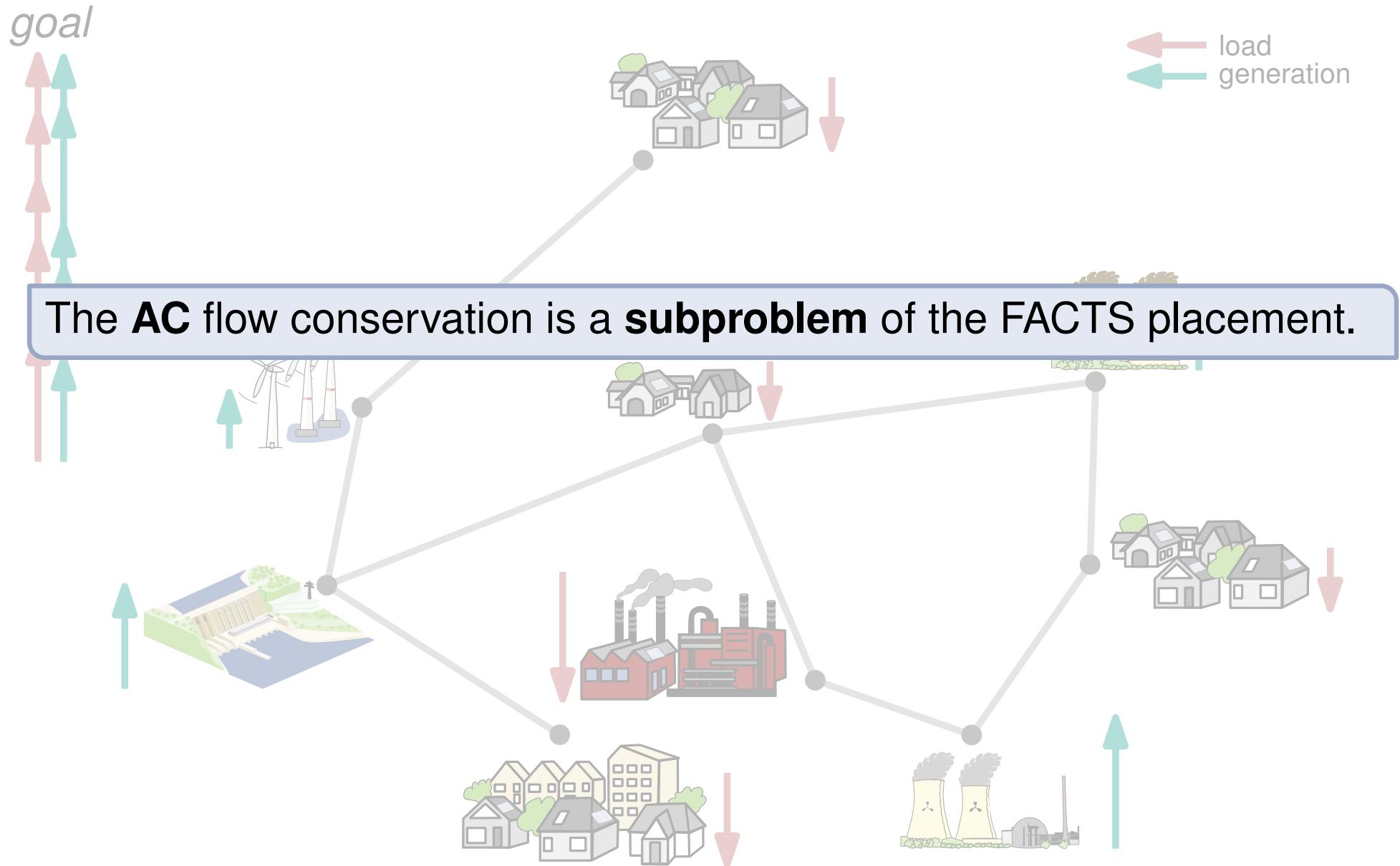
# AC Conservation of Flow in Power Grids



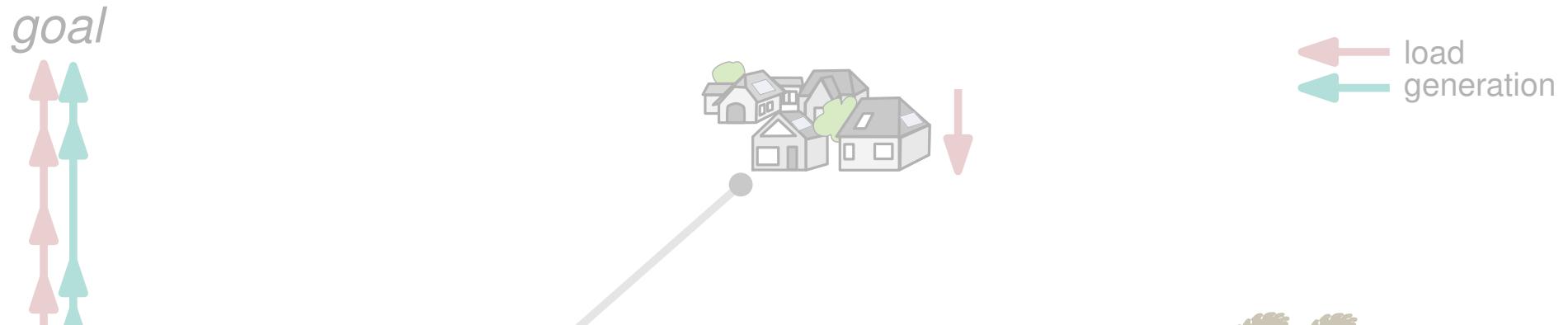
# AC Conservation of Flow in Power Grids



# AC Conservation of Flow in Power Grids



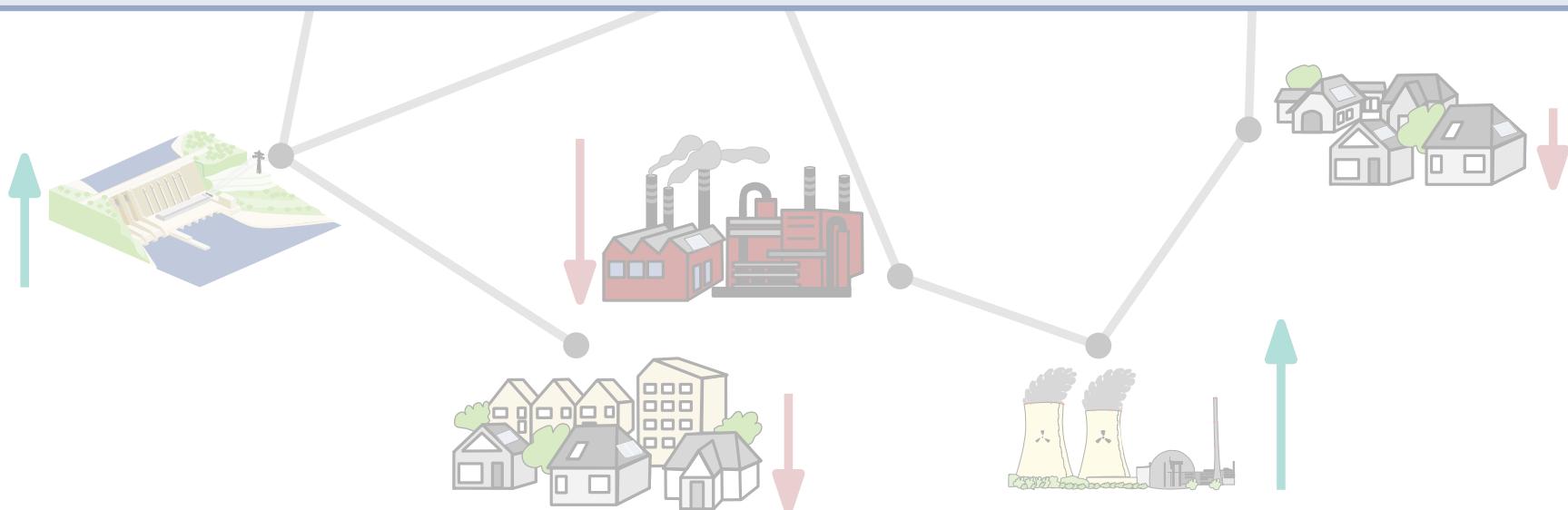
# AC Conservation of Flow in Power Grids



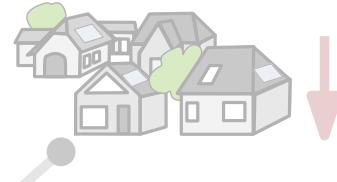
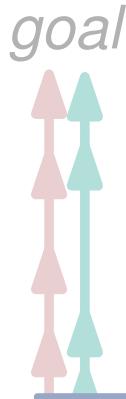
The **AC** flow conservation is a **subproblem** of the FACTS placement.

**AC** flow conservation is already **NP-hard** on **trees**.

[Lehmann et al., 2015]



# AC Conservation of Flow in Power Grids



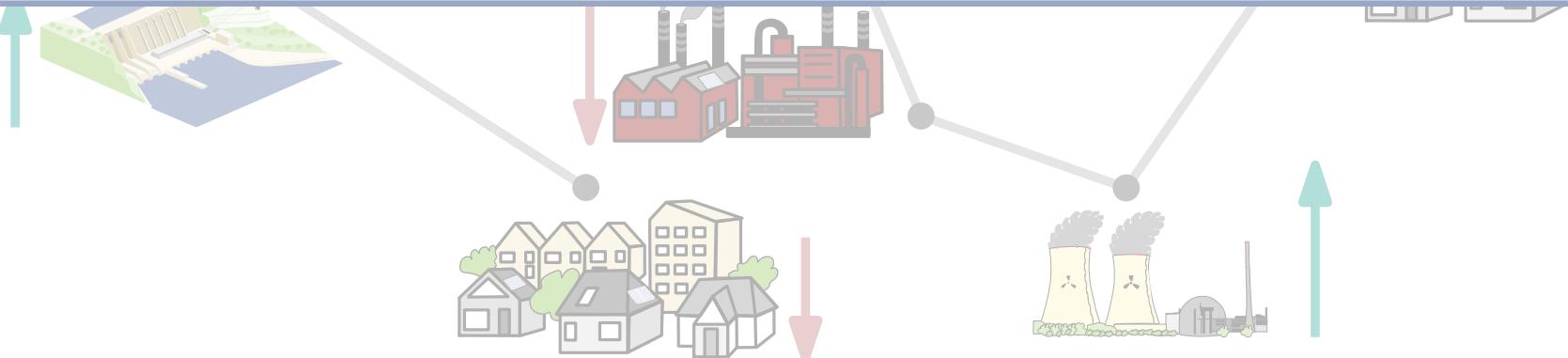
load  
generation

The **AC** flow conservation is a **subproblem** of the FACTS placement.

**AC** flow conservation is already **NP-hard** on **trees**.

[Lehmann et al., 2015]

- Power grids are non-trivial.
- **Linearized AC** flow conservation is **easy** to solve.





## FACTS placement in the power grid

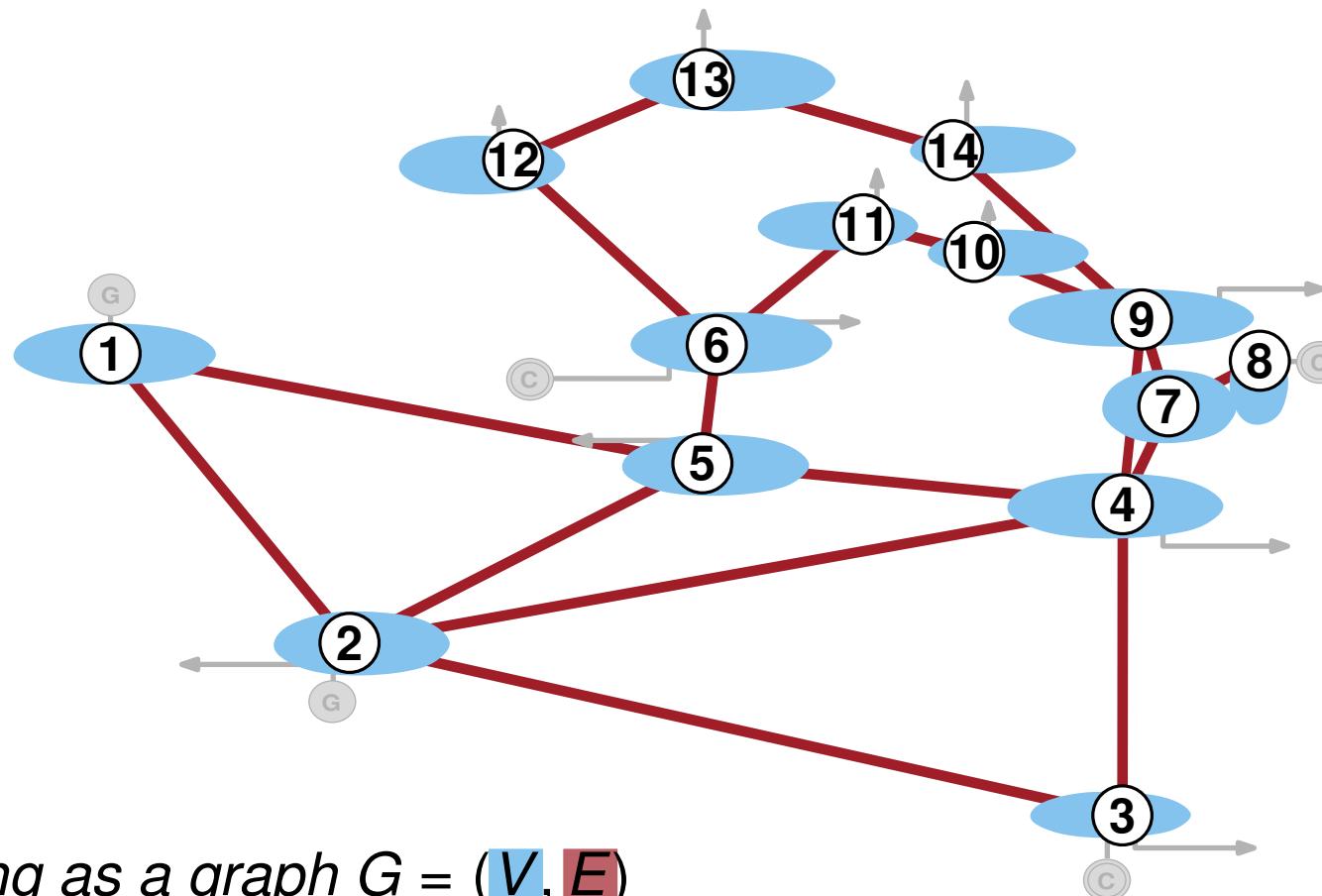
Left Figure:

<sup>2</sup> [http://www.lichtenwald-mentaltraining.de/files/bild\\_licht\\_im\\_wald.jpg](http://www.lichtenwald-mentaltraining.de/files/bild_licht_im_wald.jpg)

# Optimal ideal FACTS Flow (OiFF)

*given*  $V$  set of grid vertices

## $E$ set of branches



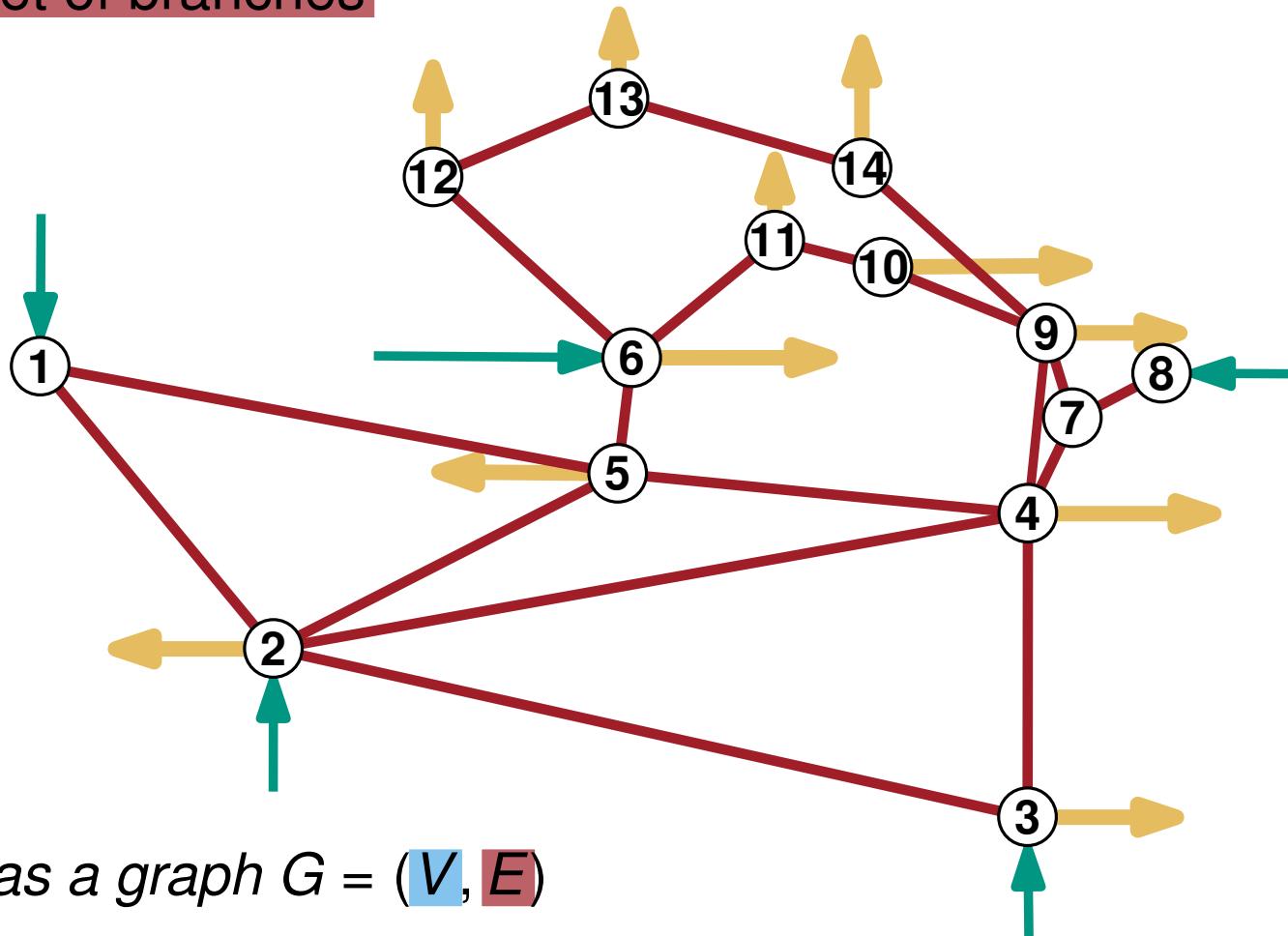
## *Modeling as a graph $G = (V, E)$*

# Optimal ideal FACTS Flow (OiFF)

given  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches



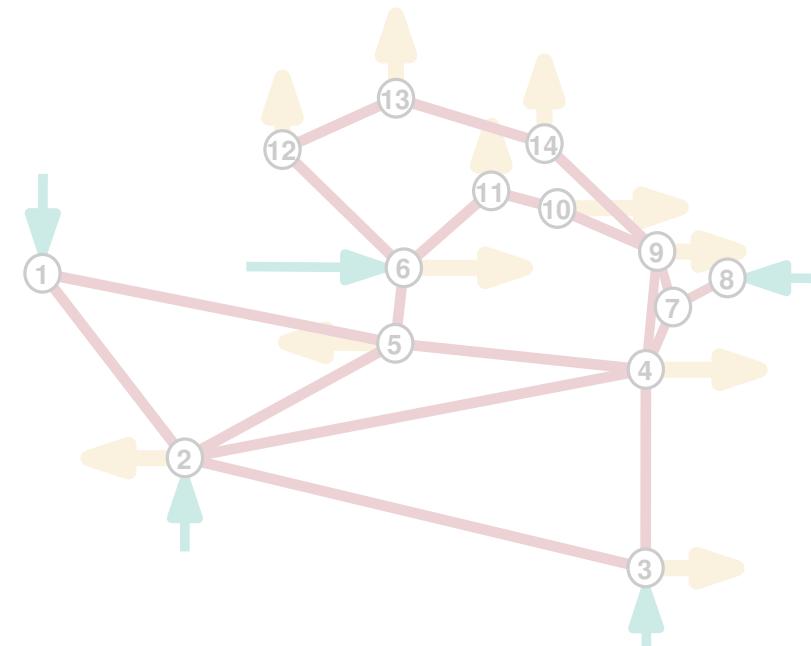
# Optimal ideal FACTS Flow (OiFF)

given  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches

Input



# Optimal ideal FACTS Flow (OiFF)

*given*  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

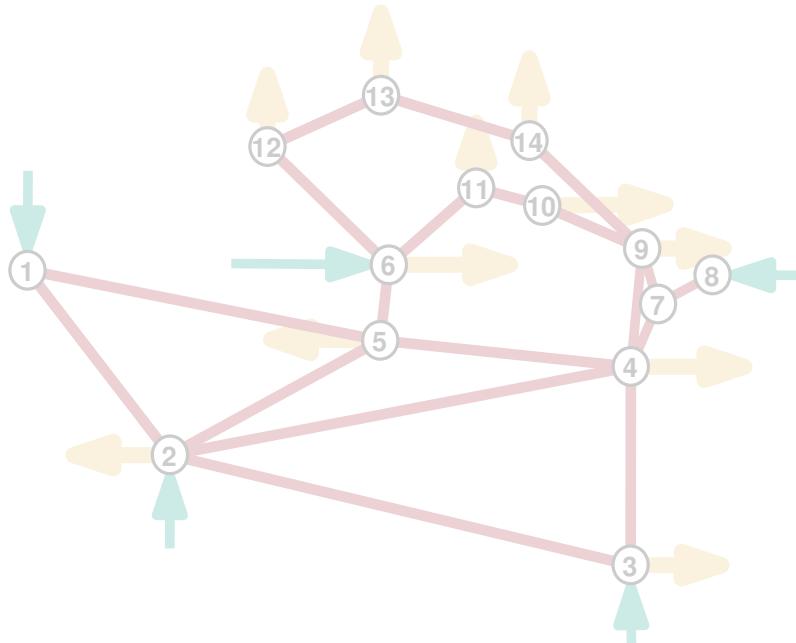
$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches

*find* for every generator: the **power generation**

for every branch: **FACTS placement**

variables



# Optimal ideal FACTS Flow (OiFF)

*given*  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

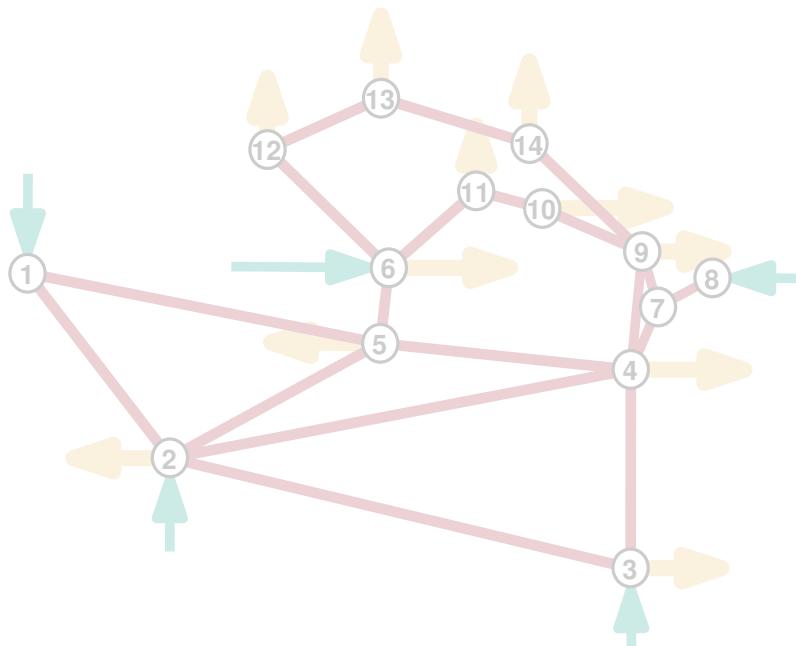
$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches

*find* for every generator: the **power generation**

for every branch: **FACTS placement**

*minimize* **costs**



# Optimal ideal FACTS Flow (OiFF)

given  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches (with **impedances**, **susceptance**,

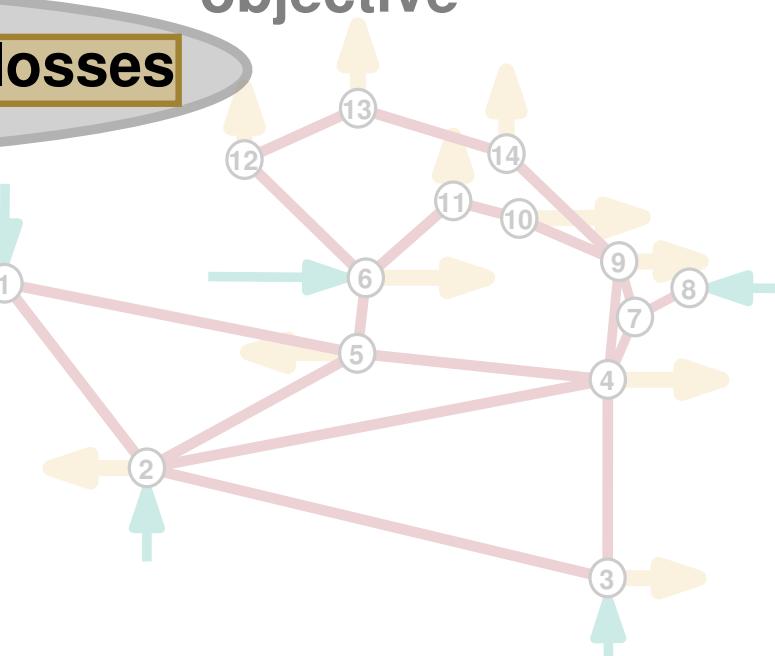
**capacities**)

find for every generator: the **power generation**

for every branch: **FACTS placement**

minimize **total generation costs** and **branch losses**

objective



# Optimal ideal FACTS Flow (OiFF)

*given*  $V$  set of grid vertices,  $V_L \subseteq V$  set of consumers

$V_G \subseteq V$  set of generators (with **capacities** and **costs**)

$E$  set of branches (with **impedances**, **susceptance**,

**capacities**)

*find* for every generator: the **power generation**

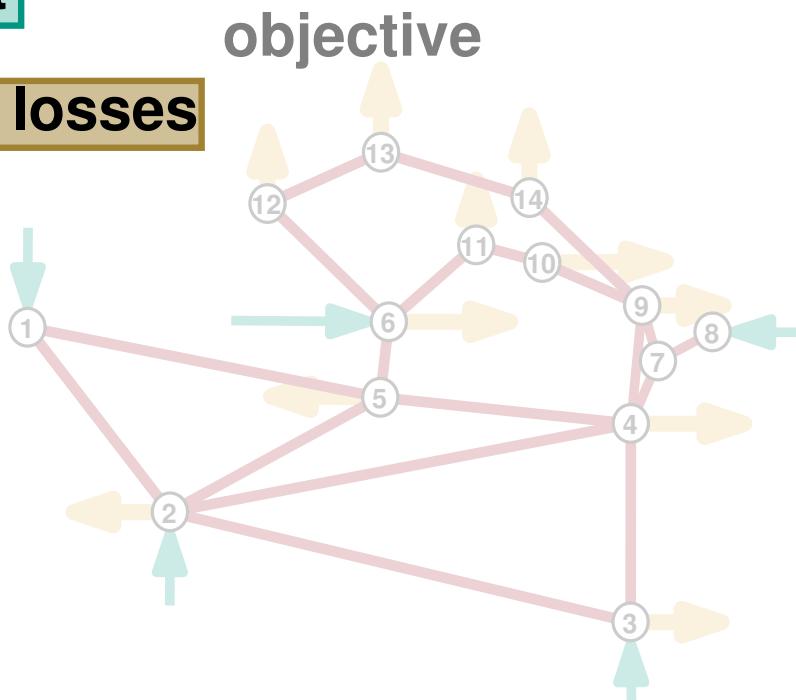
for every branch: **FACTS placement**

*minimize* **total generation costs** and **branch losses**

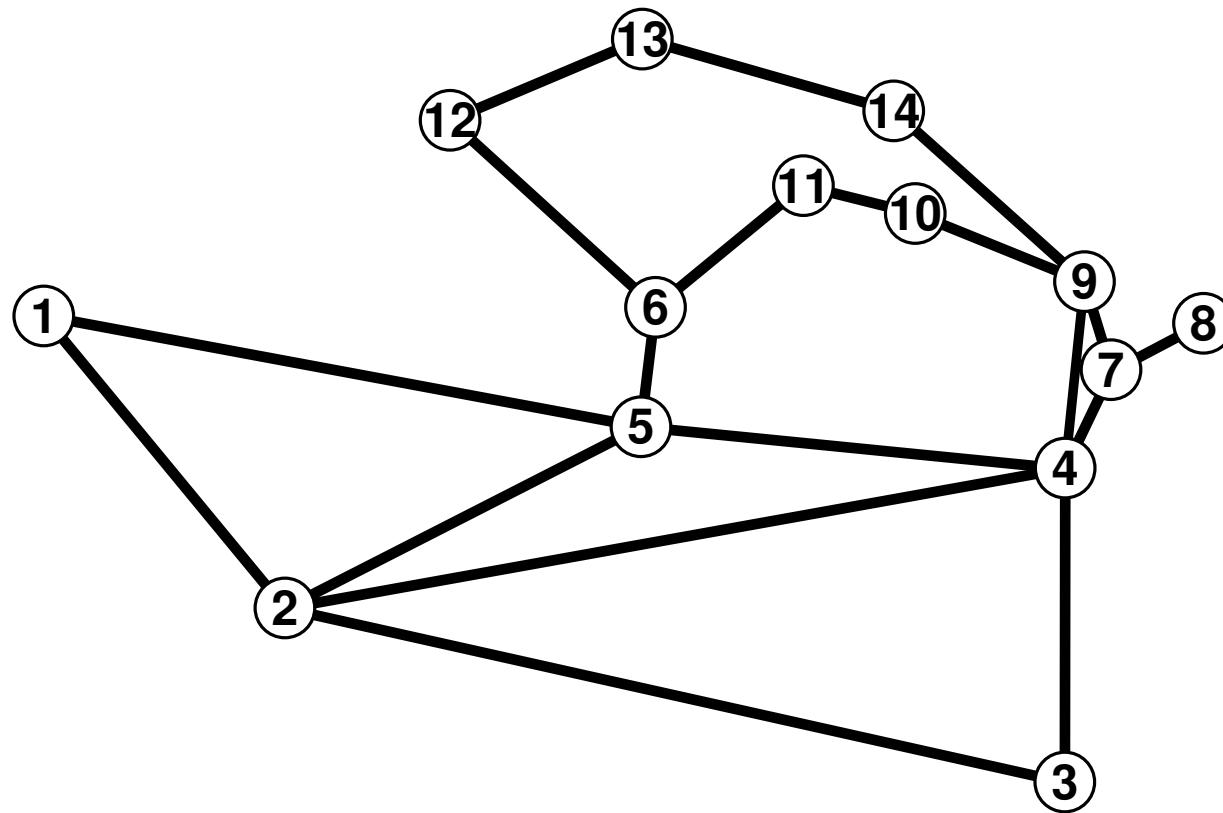
*subject to* branch **capacities**

generator **capacities**

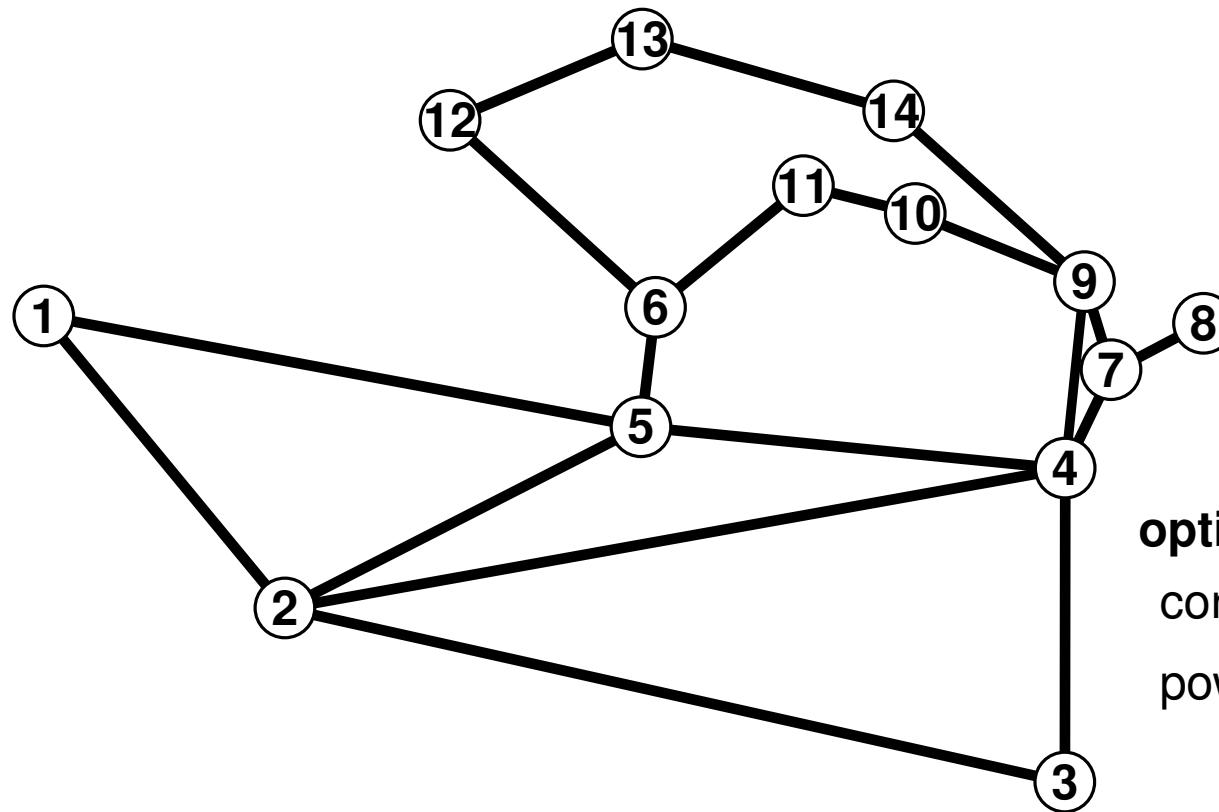
**power flow constraints**



# Modeling the Power Flow Constraint

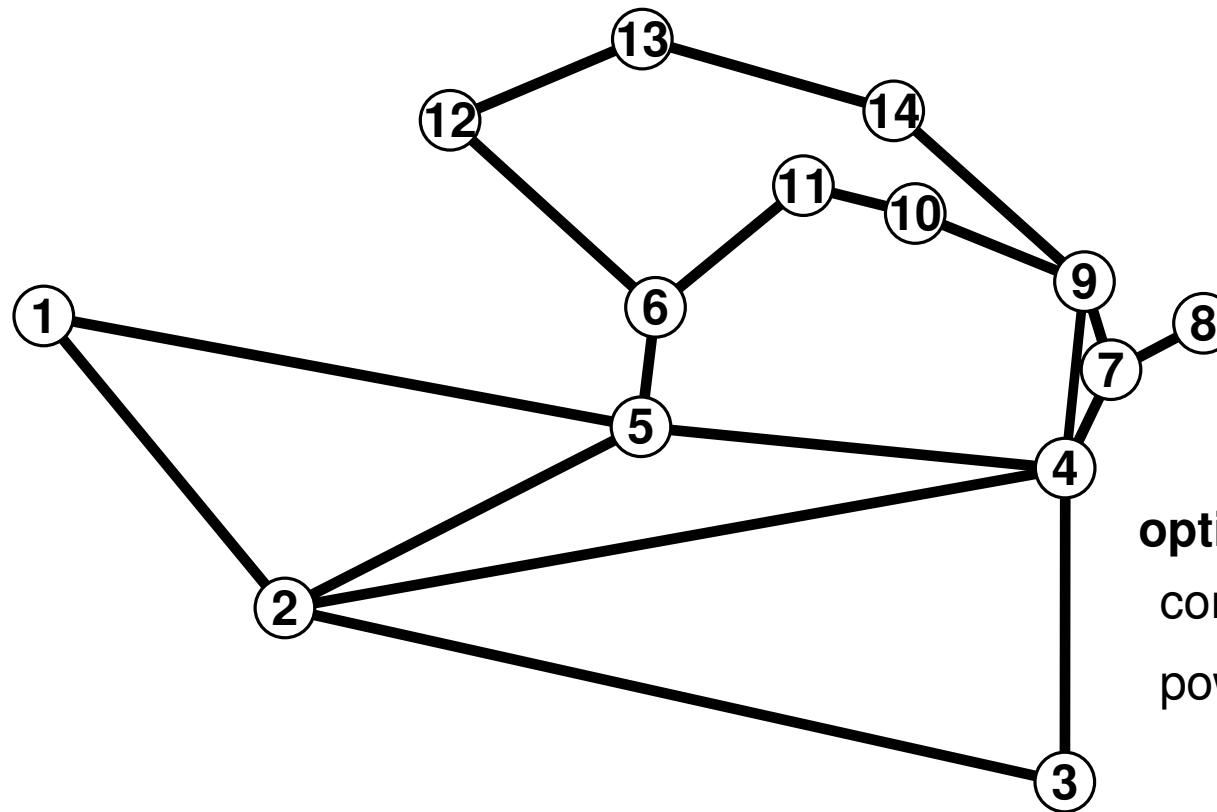


# Modeling the Power Flow Constraint



**optimize subject to:**  
conservation of flow  
power flow constraints

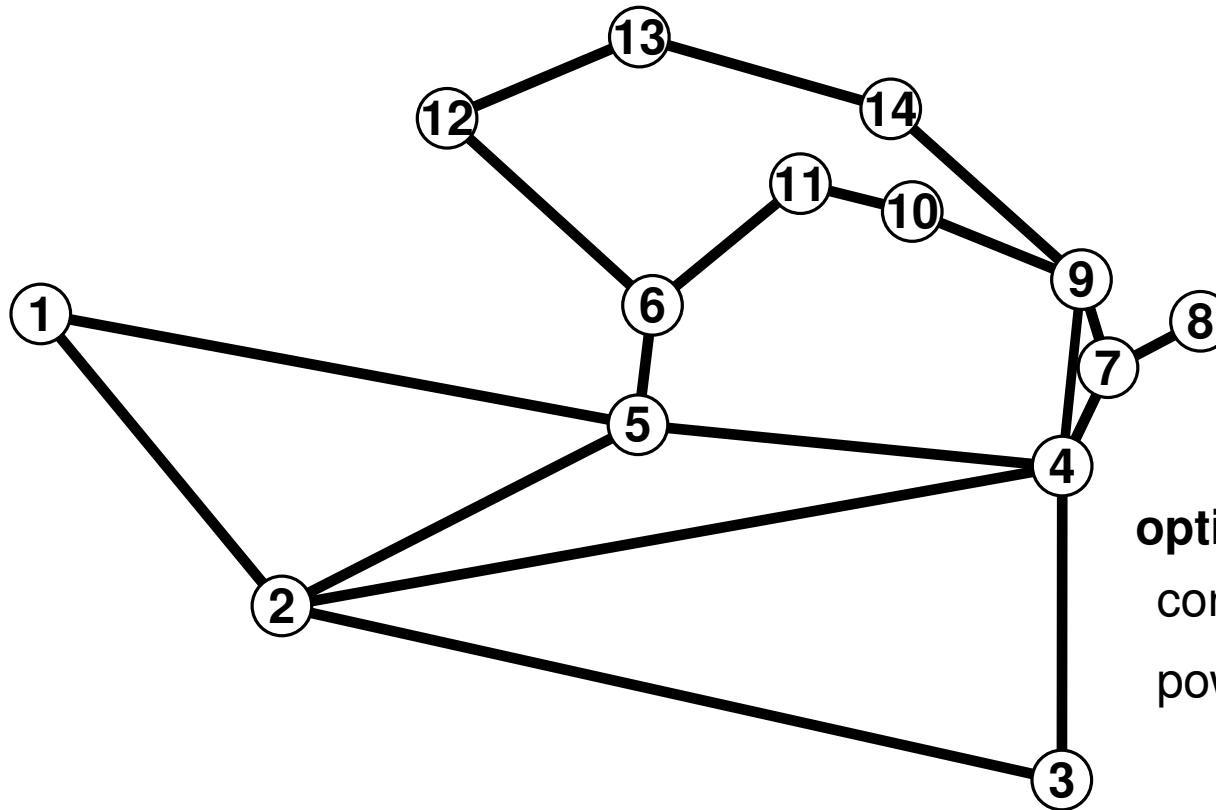
# Modeling the Power Flow Constraint



**optimize subject to:**  
conservation of flow  
power flow constraints

**minimize** costs

# Modeling the Power Flow Constraint



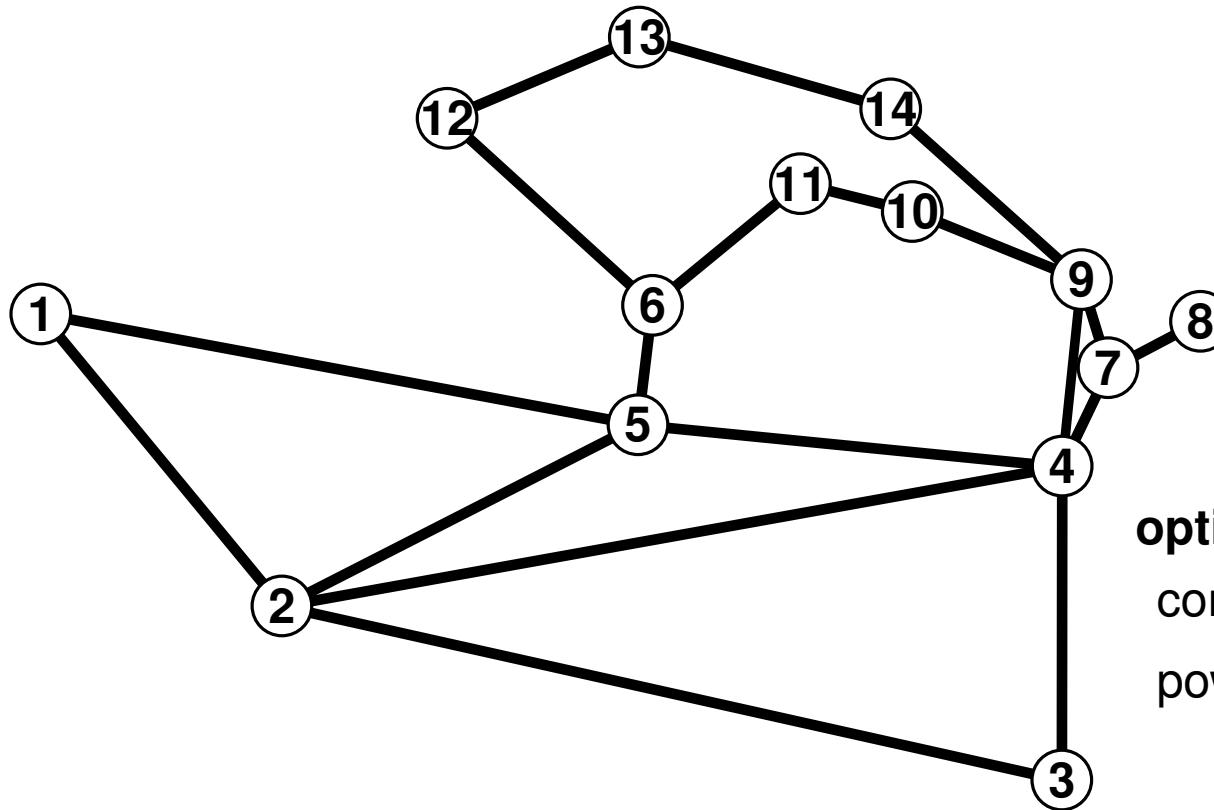
minimize costs

optimize subject to:  
conservation of flow  
power flow constraints



physical model

# Modeling the Power Flow Constraint



**optimize subject to:**  
conservation of flow  
power flow constraints

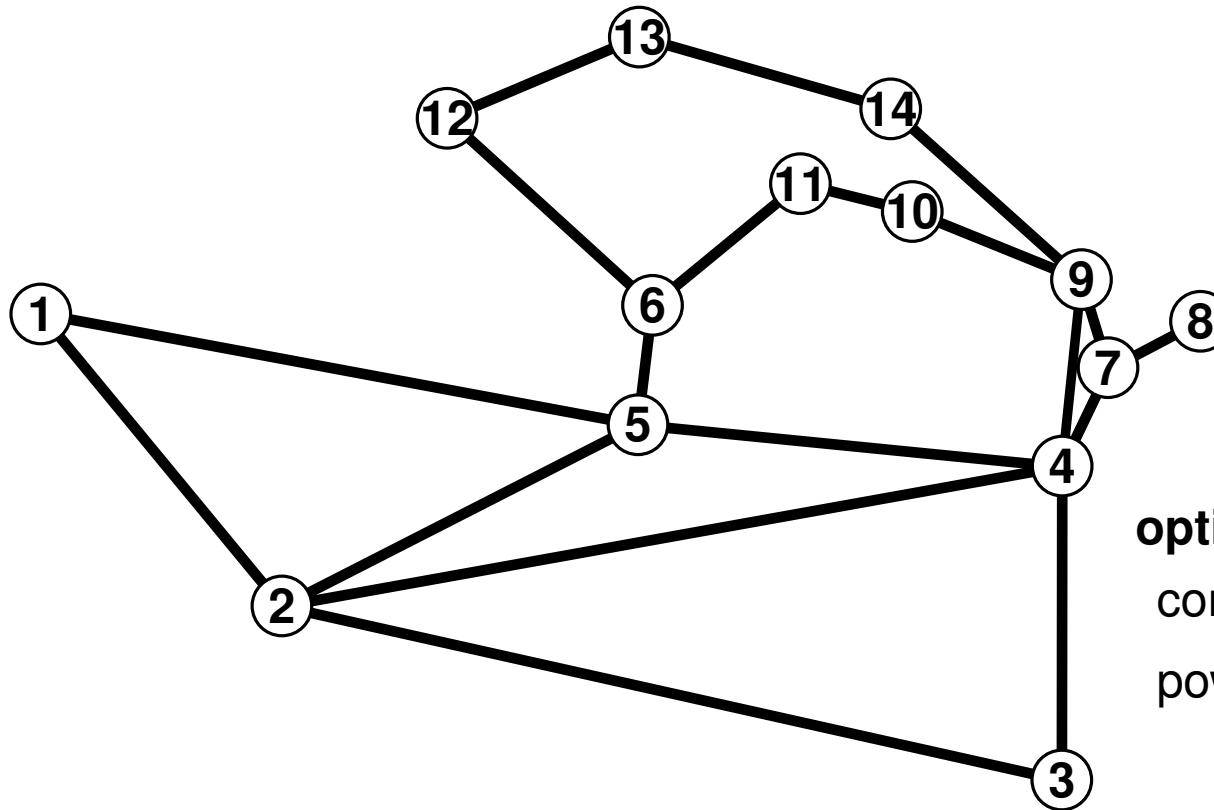


**flow model**

minimize costs

**physical model**

# Modeling the Power Flow Constraint



**optimize subject to:**  
conservation of flow  
power flow constraints



**flow model**

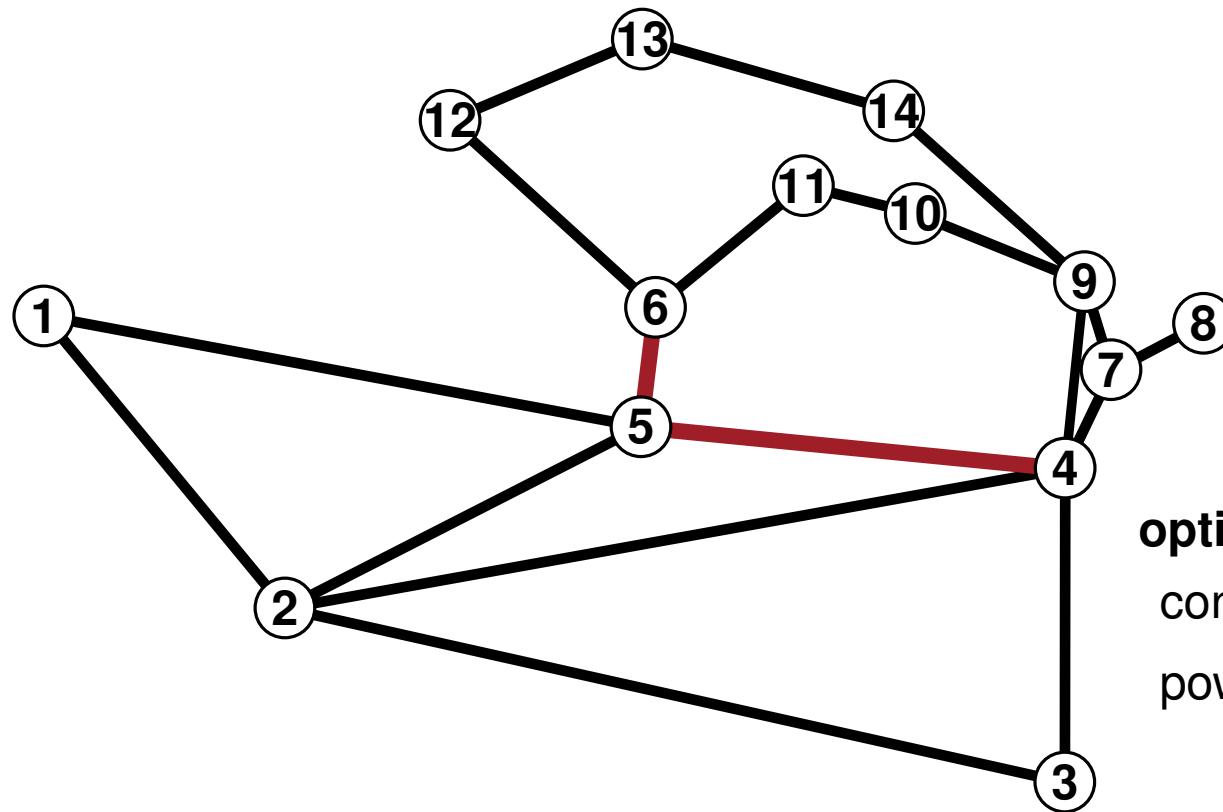
lower bound

minimize costs

**physical model**

upper bound

# Modeling the Power Flow Constraint

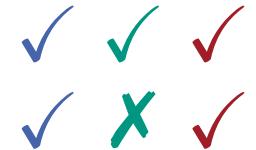


— ideal FACTS

**optimize subject to:**

conservation of flow

power flow constraints

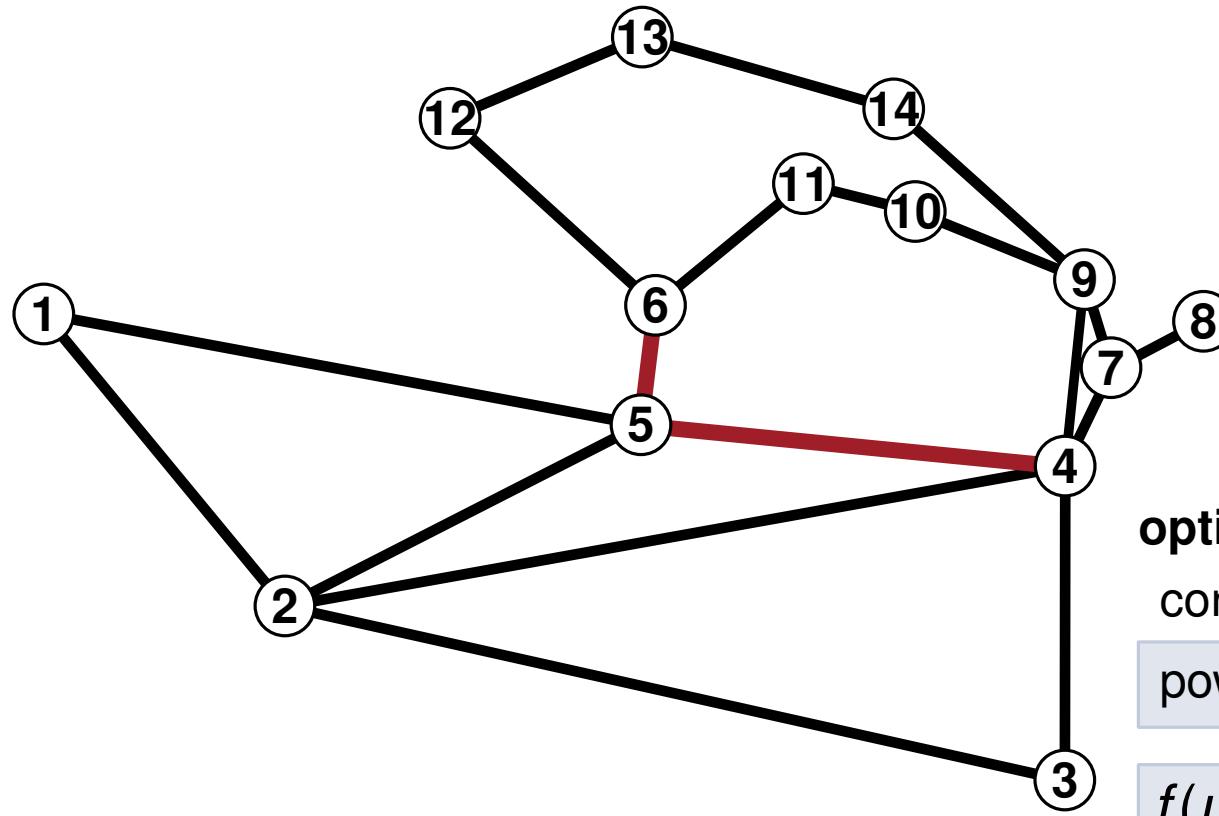


**flow model**  
lower bound

minimize costs  
**objective**

**physical model**  
upper bound

# Modeling the Power Flow Constraint



— ideal FACTS

optimize subject to:

conservation of flow



power flow constraints



$$f(u, v) = B(u, v) \cdot (\theta(u) - \theta(v))$$

flow model

lower bound

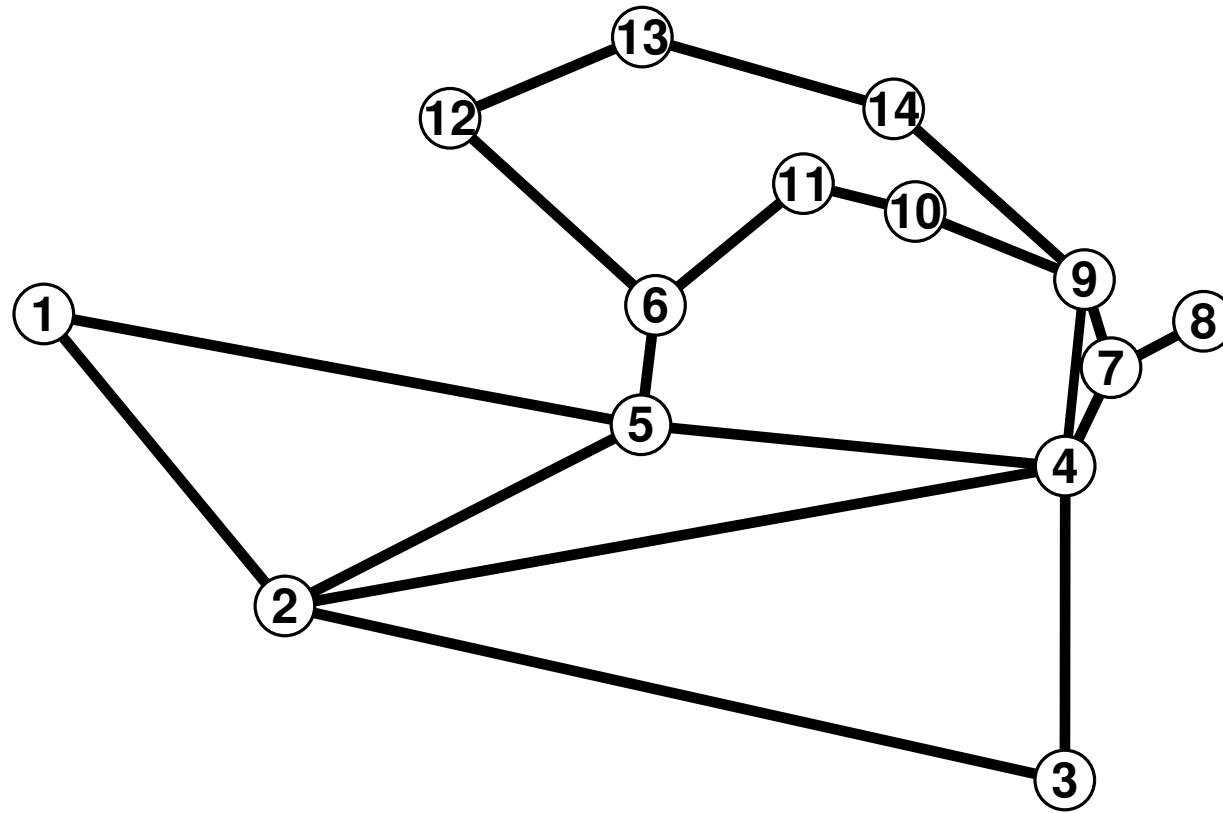
minimize costs

objective

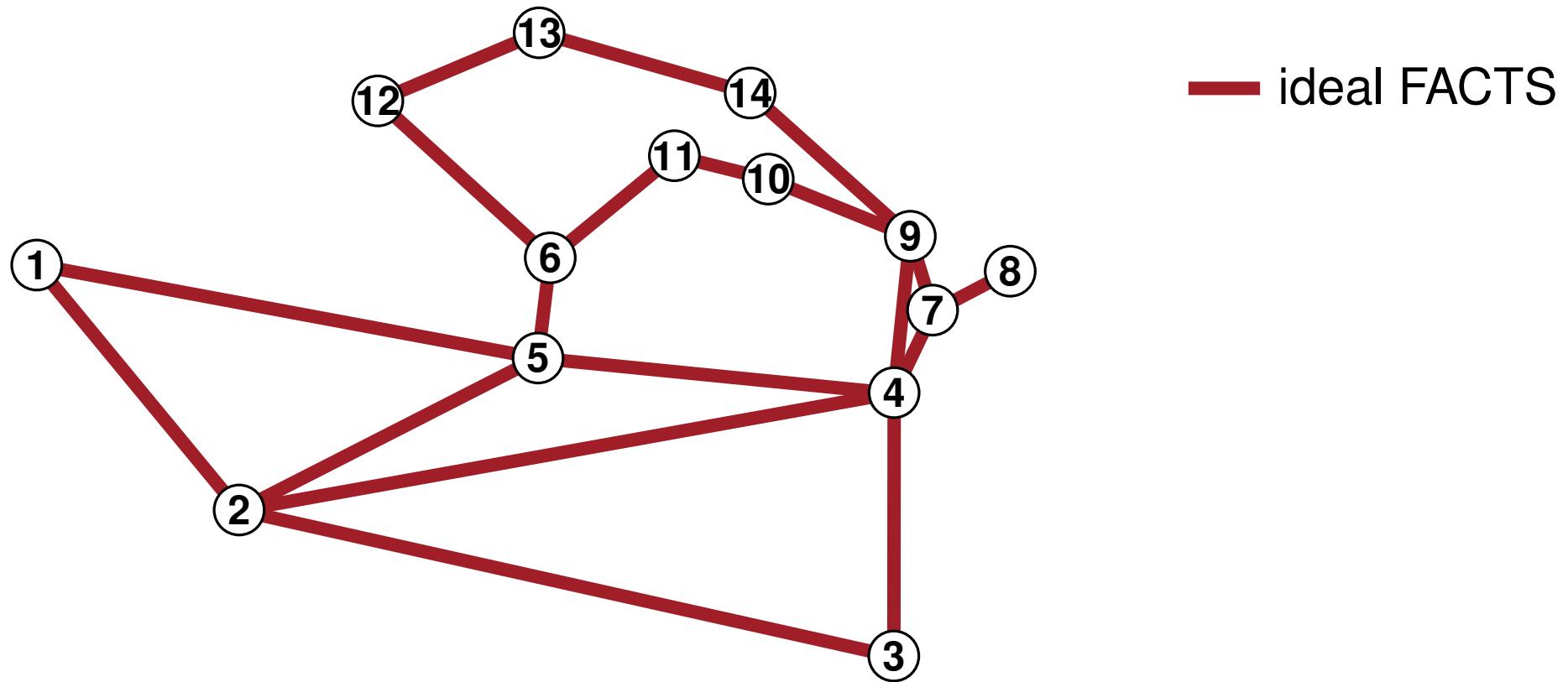
physical model

upper bound

# Global Optimal Power Flow



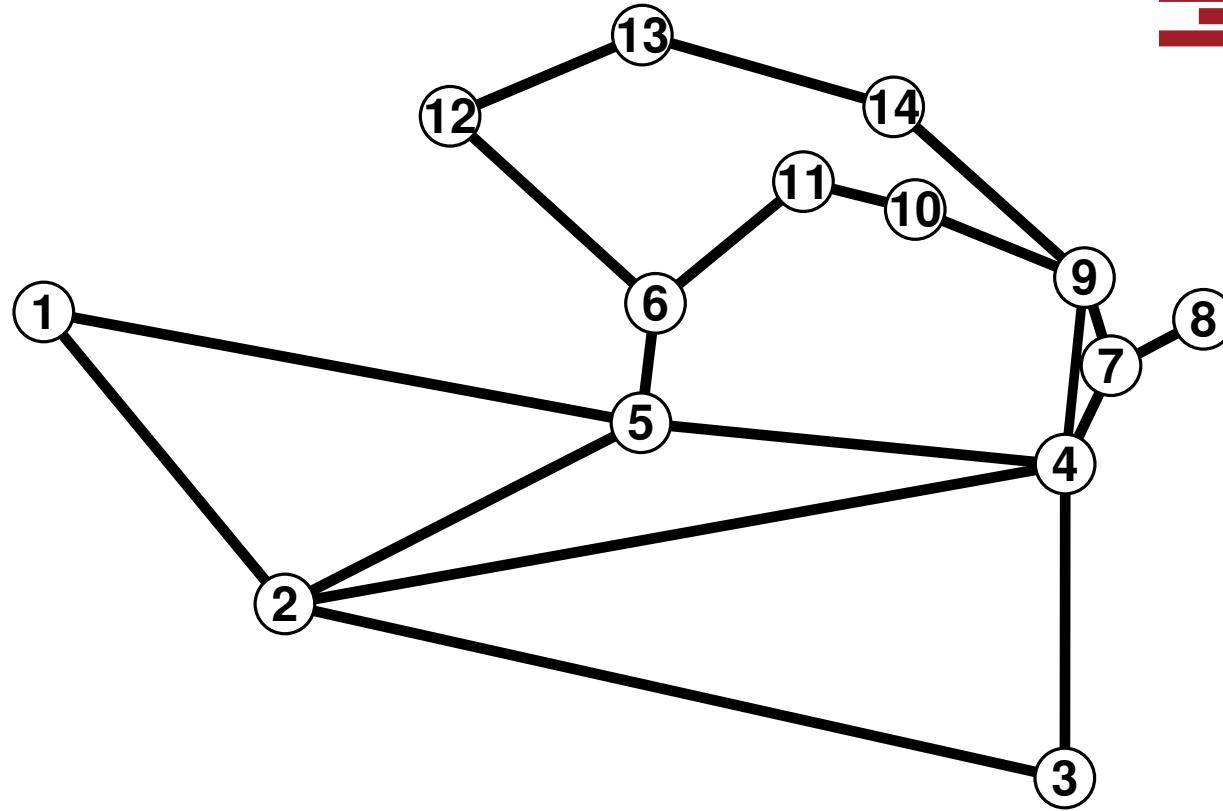
# Global Optimal Power Flow



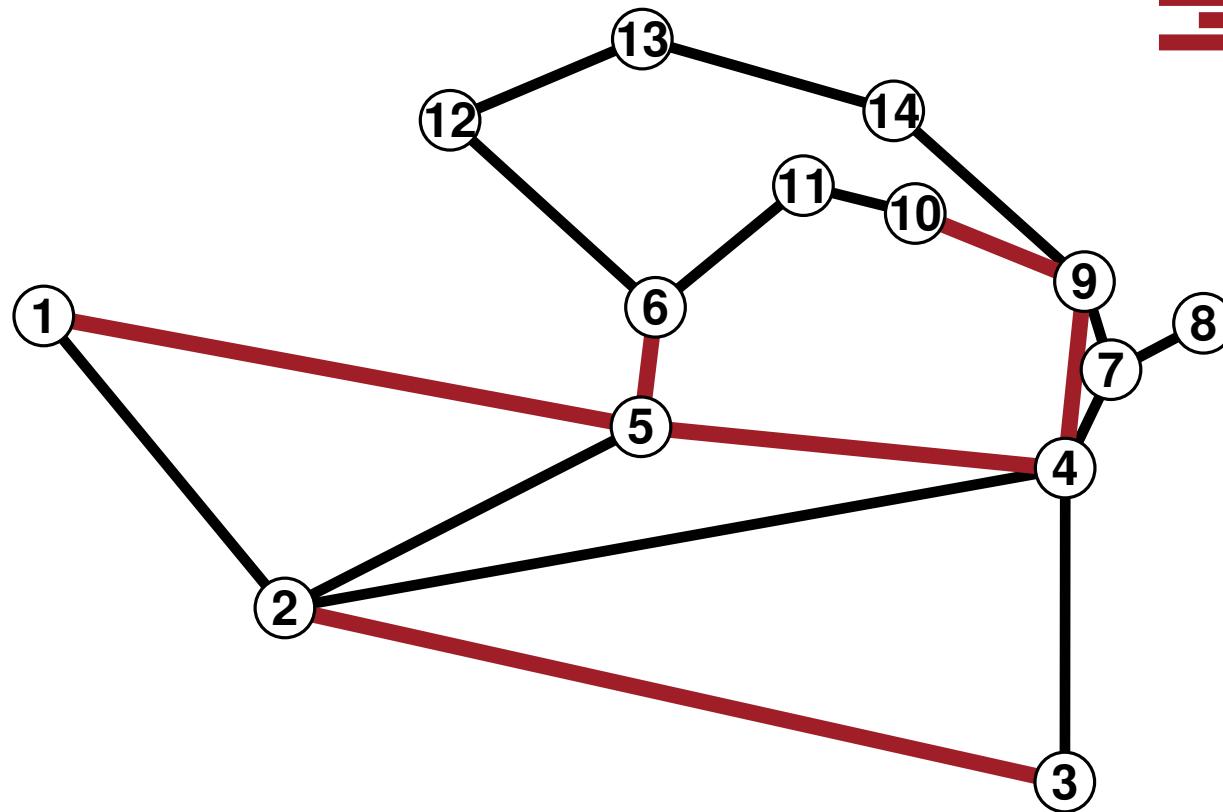
Can we become as good as the flow model  
with fewer FACTS?

# Feedback Forest Set

 *Feedback Forest Set*

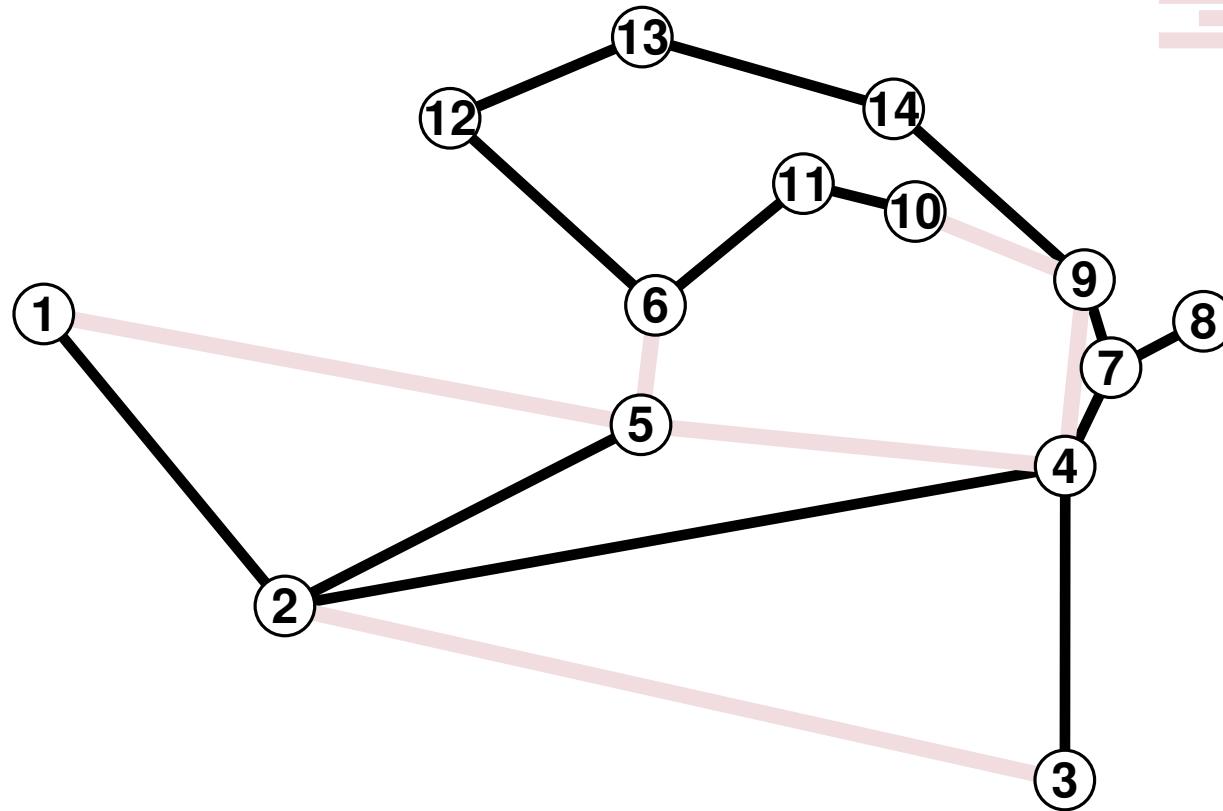


# Feedback Forest Set



 *Feedback Forest Set*  
↓  
A set of Trees (*Forests*) remains!

# Feedback Forest Set

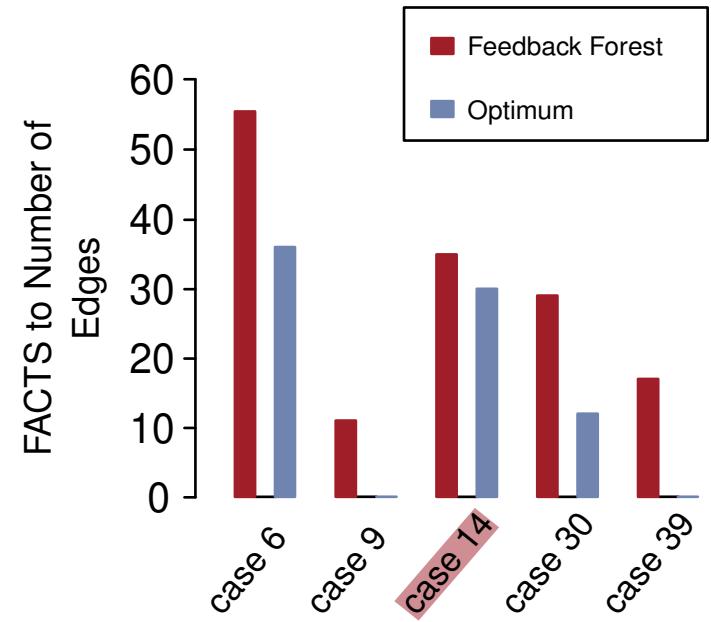
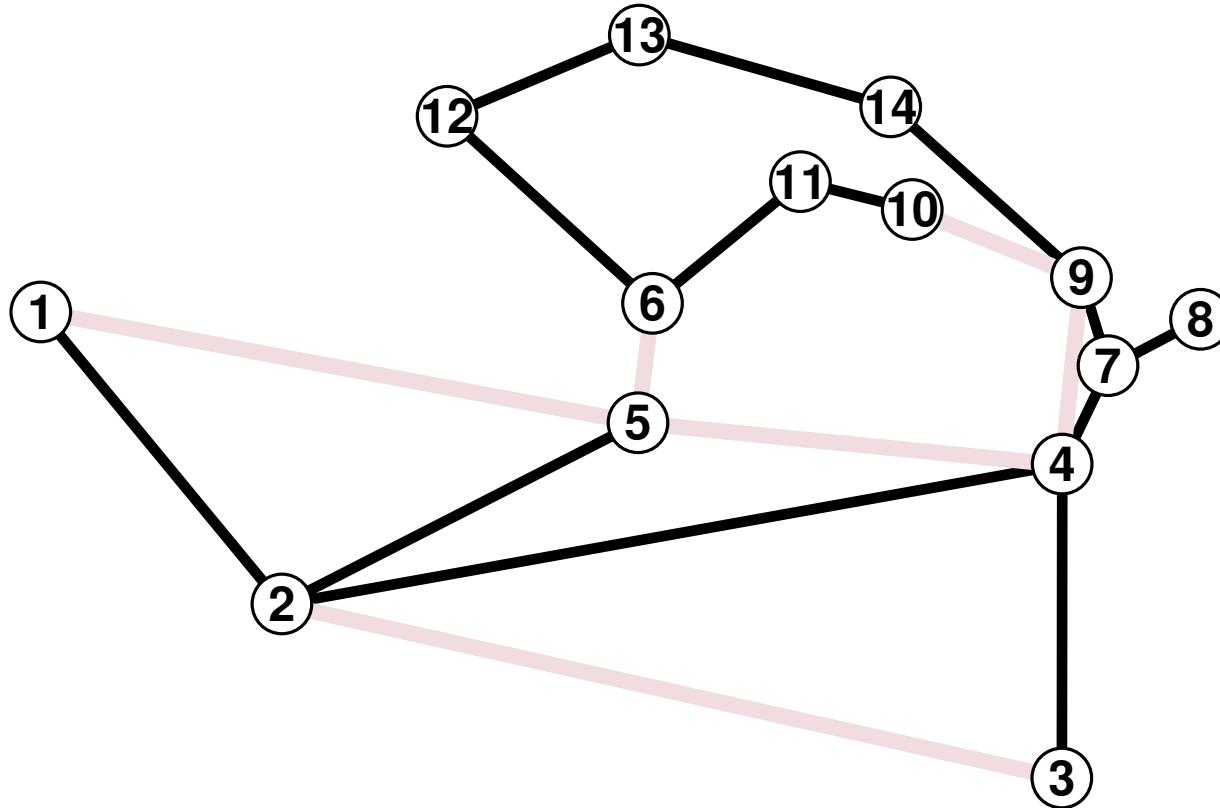


 *Feedback Forest Set*  
↓  
A set of Trees (*Forests*) remains!

## Theorem 1

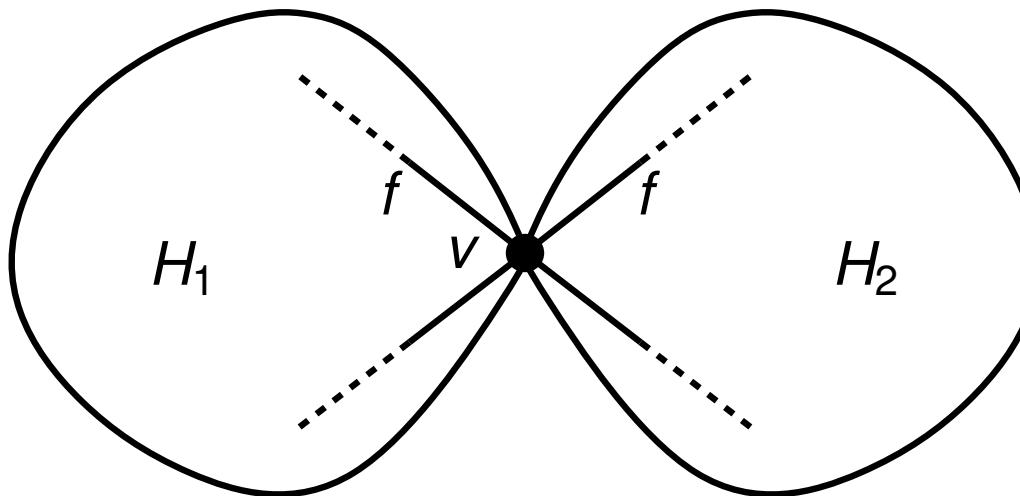
Physical subgrid forest  $\Rightarrow$  All graph theoretical flows are electrical flows

# Feedback Forest Set



Are less than 6 FACTS possible?

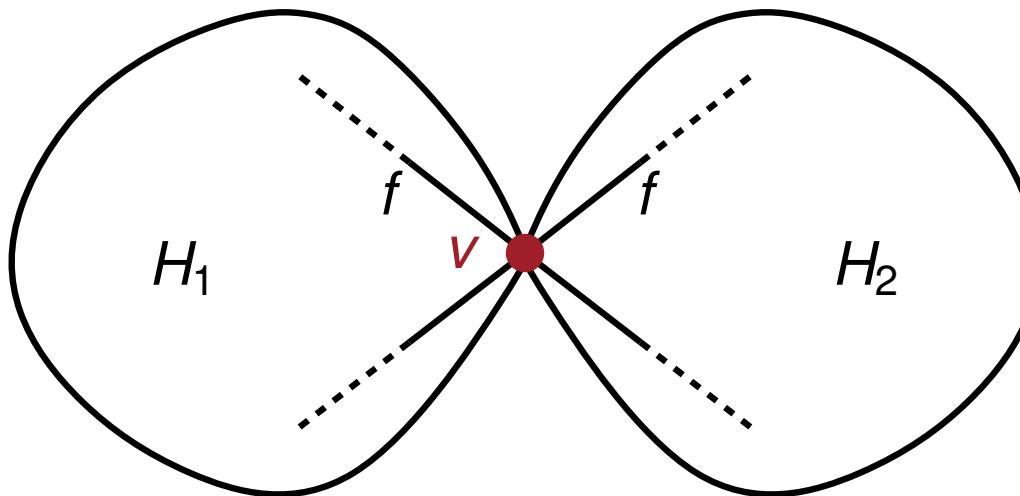
# Idea of Proof for the Feedback Forest Set



## Assumption

- network  $N(G)$  has an electrically feasible flow  $f$
- network  $N(G)$  has a cut vertex  $v$

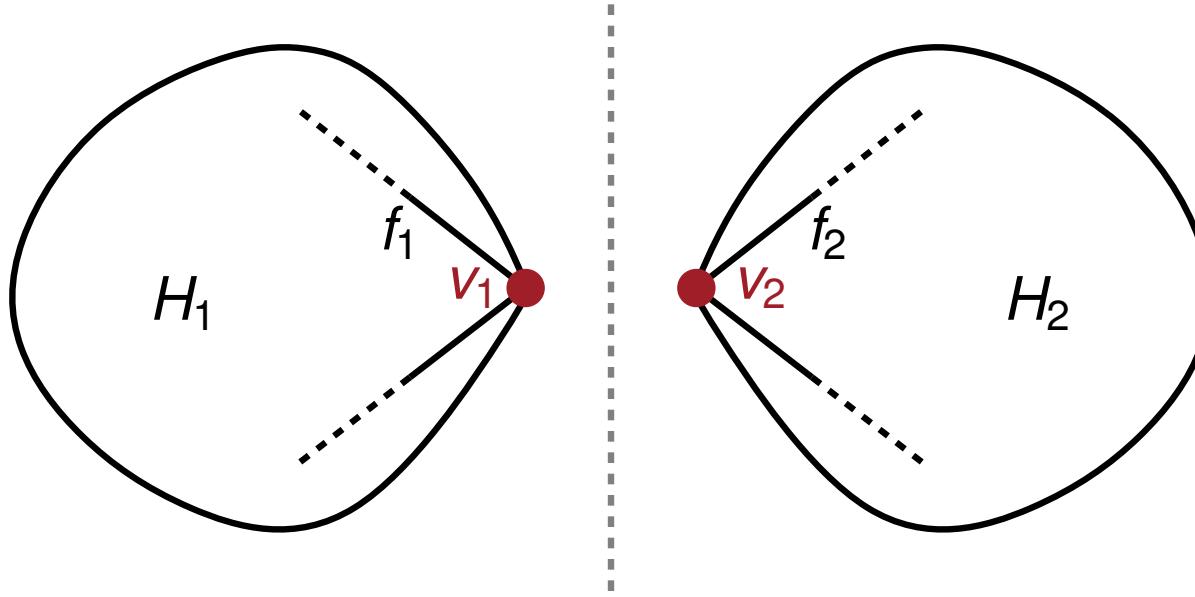
# Idea of Proof for the Feedback Forest Set



## Assumption

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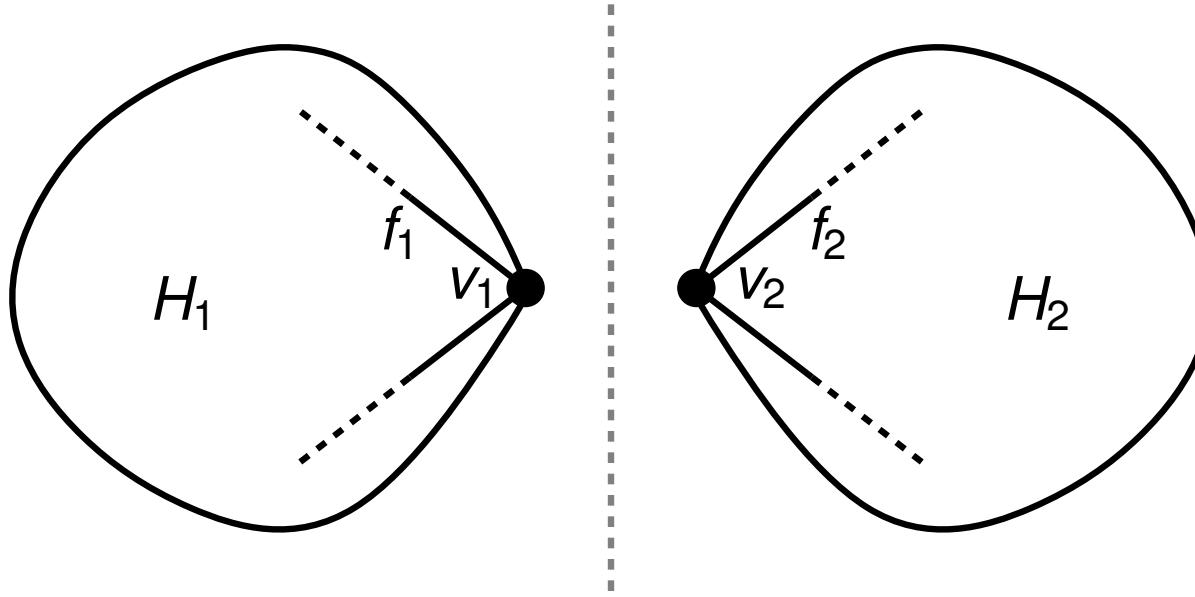
# Idea of Proof for the Feedback Forest Set



## Idea

- cut  $N(G)$  at the cut vertex  $v$  with  $v = v_1 = v_2$
  - function  $f_1$  in  $H_1$  und  $f_2$  in  $H_2$  remain unchanged
- ⇒ electrically feasible

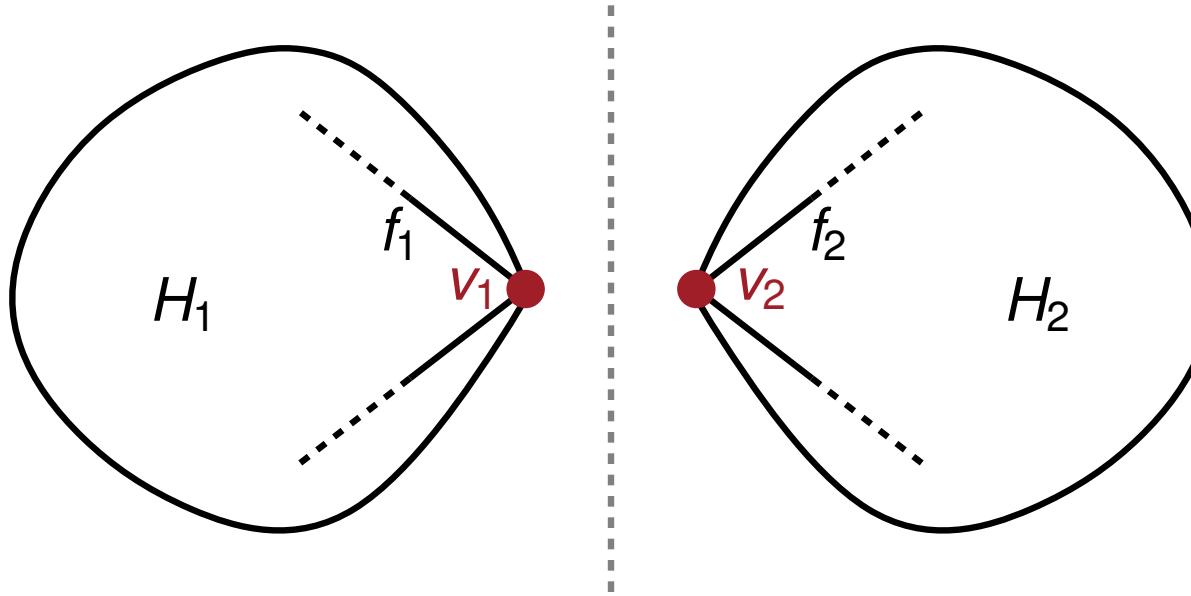
# Idea of Proof for the Feedback Forest Set



## Assumption

- function  $f_1$  in  $H_1$  and  $f_2$  in  $H_2$  are electrically feasible
- find all copies of  $v = \{v_1, v_2\}$

# Idea of Proof for the Feedback Forest Set

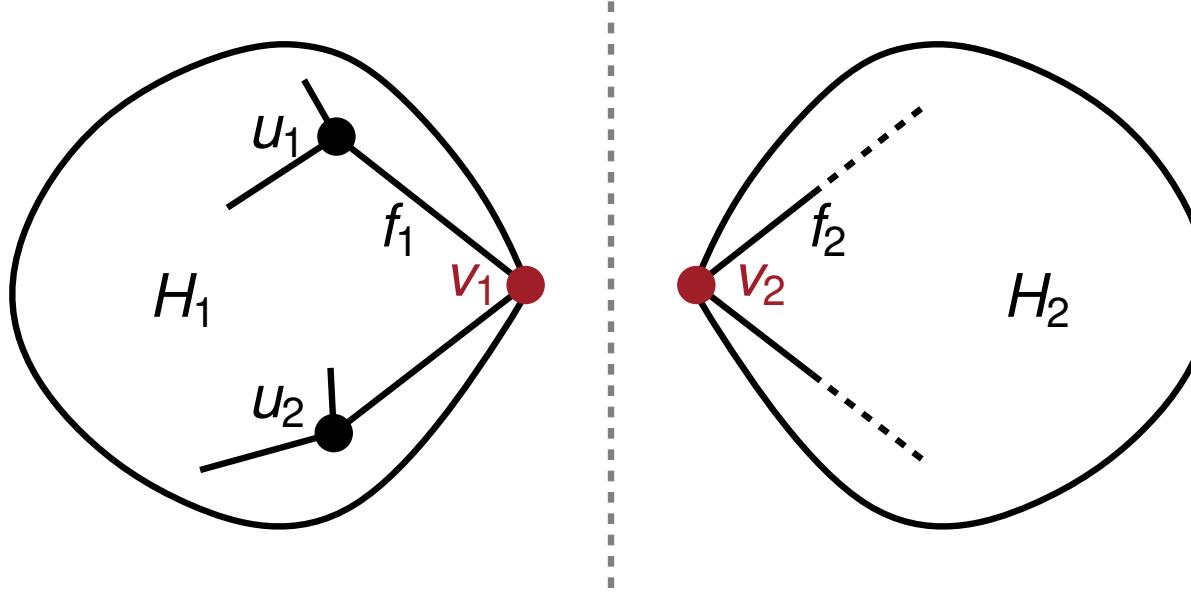


## Idea

$$\Theta'(v_1) = \Theta'(v_2) = c$$

$$\Delta_1 = \Theta(v_1) - c \quad \Delta_2 = \Theta(v_2) - c$$

# Idea of Proof for the Feedback Forest Set

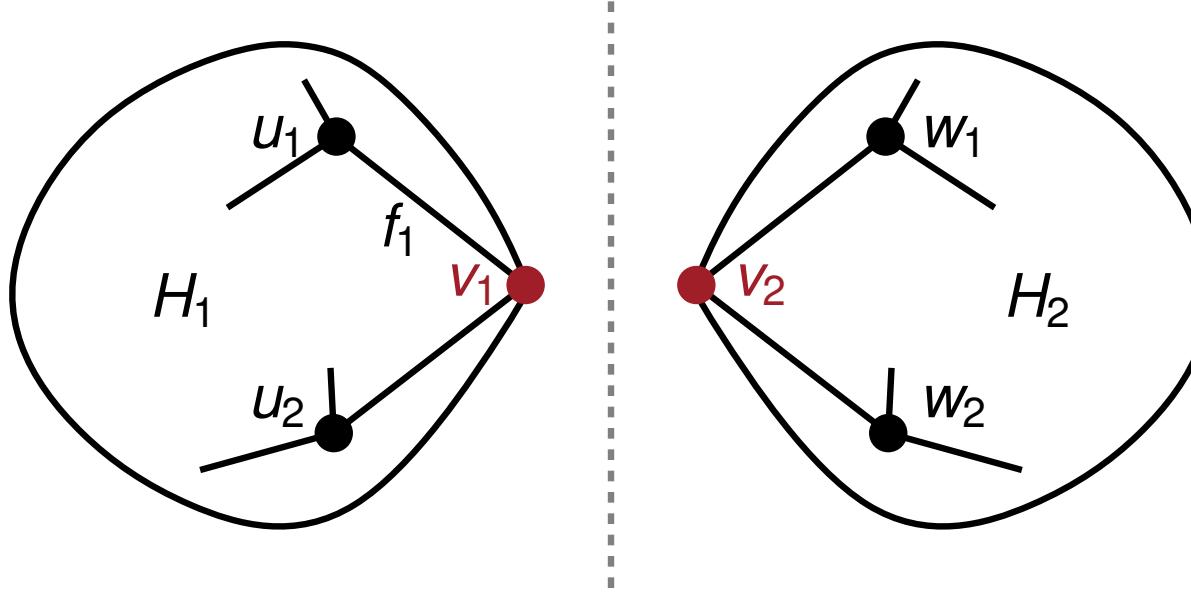


## Idea

$$\Theta'(u_1) = \Theta(u_1) - \Delta_1$$

$$\Theta'(u_2) = \Theta(u_2) - \Delta_1$$

# Idea of Proof for the Feedback Forest Set

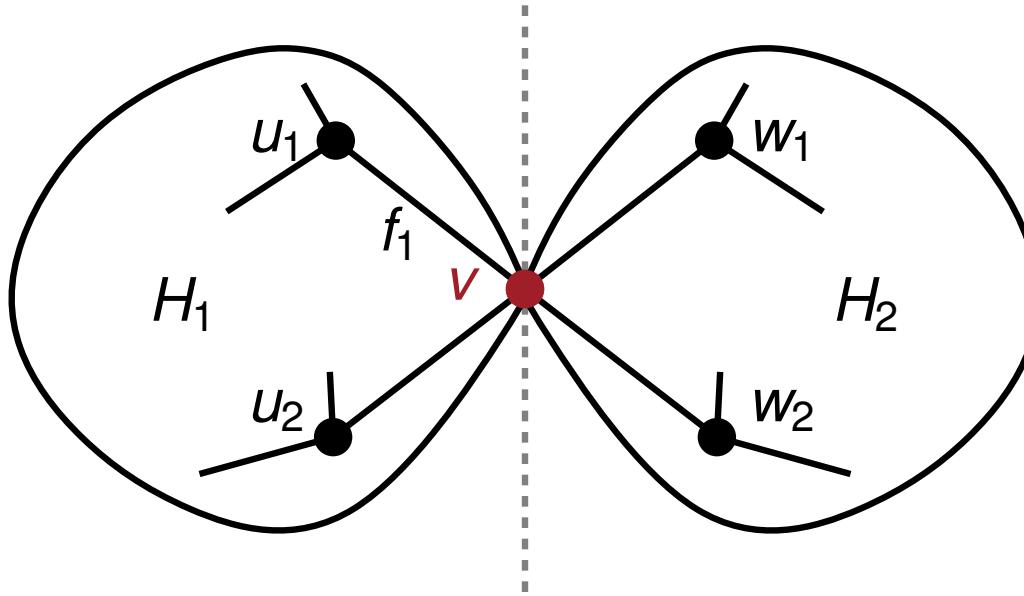


## Idea

$$\Theta'(w_1) = \Theta(w_1) - \Delta_2$$

$$\Theta'(w_2) = \Theta(w_2) - \Delta_2$$

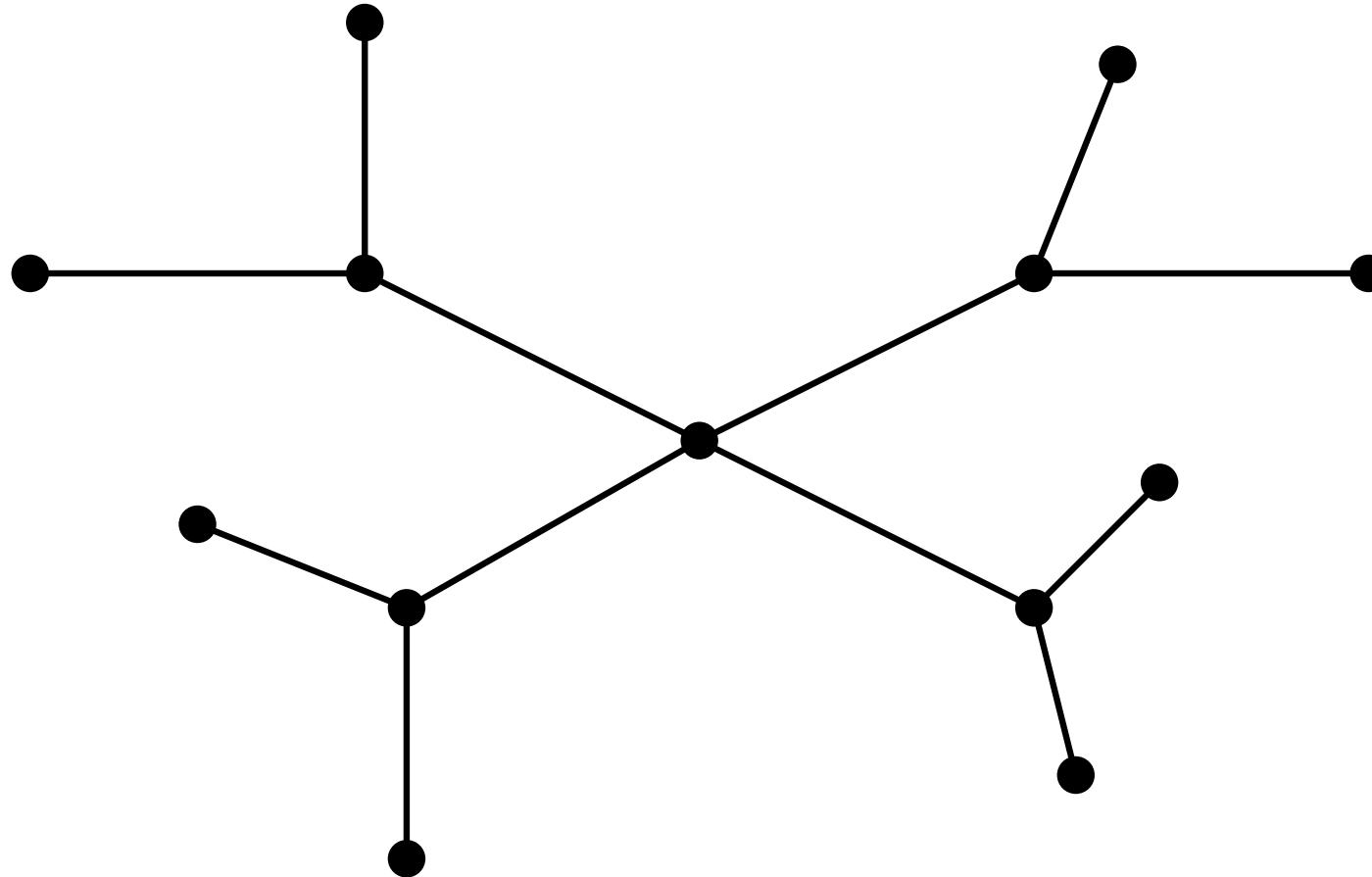
# Idea of Proof for the Feedback Forest Set



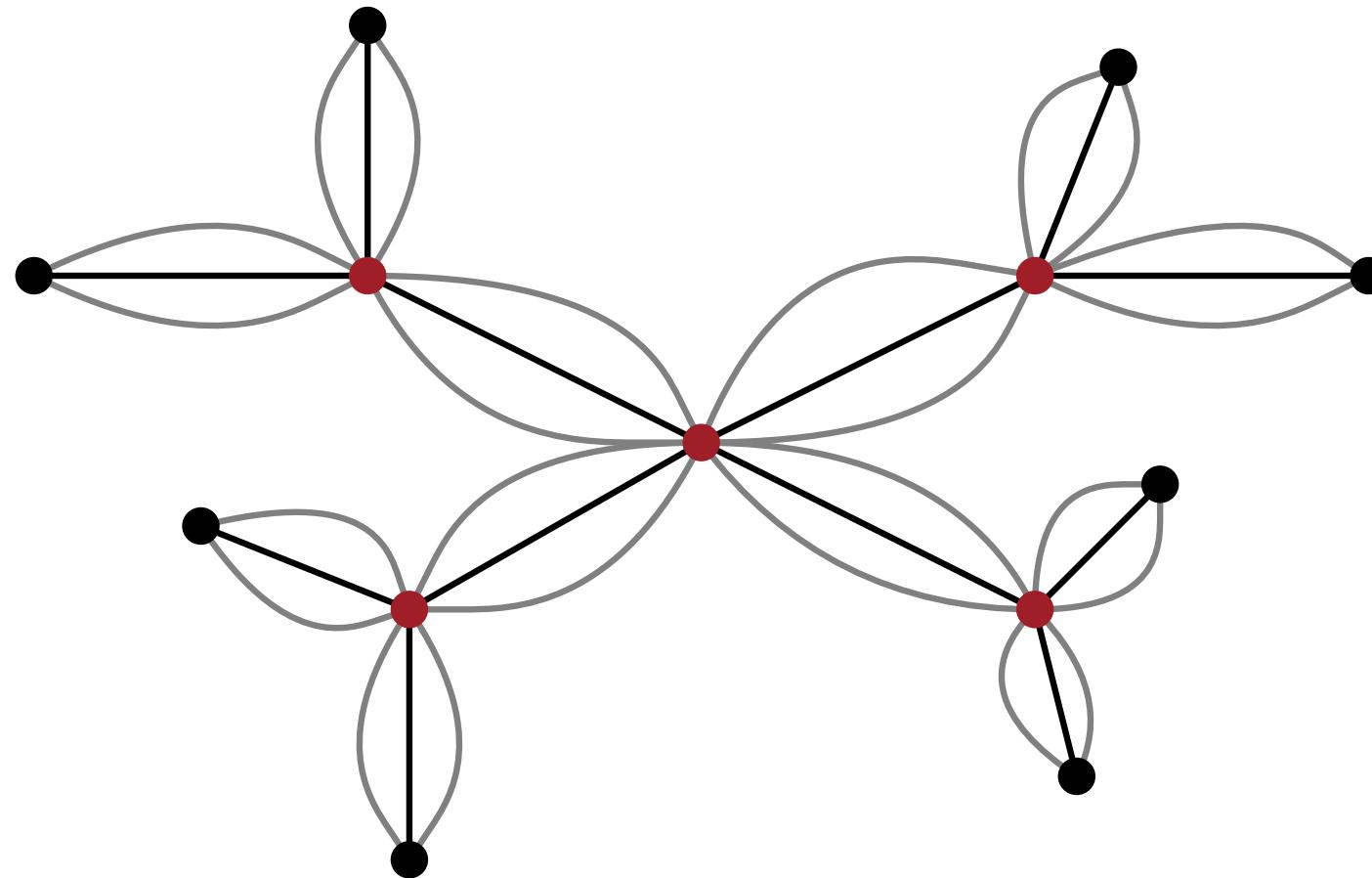
## Idea

$$\Theta(v) = \Theta'(v_1) = \Theta'(v_2)$$

# Idea of Proof for the Feedback Forest Set

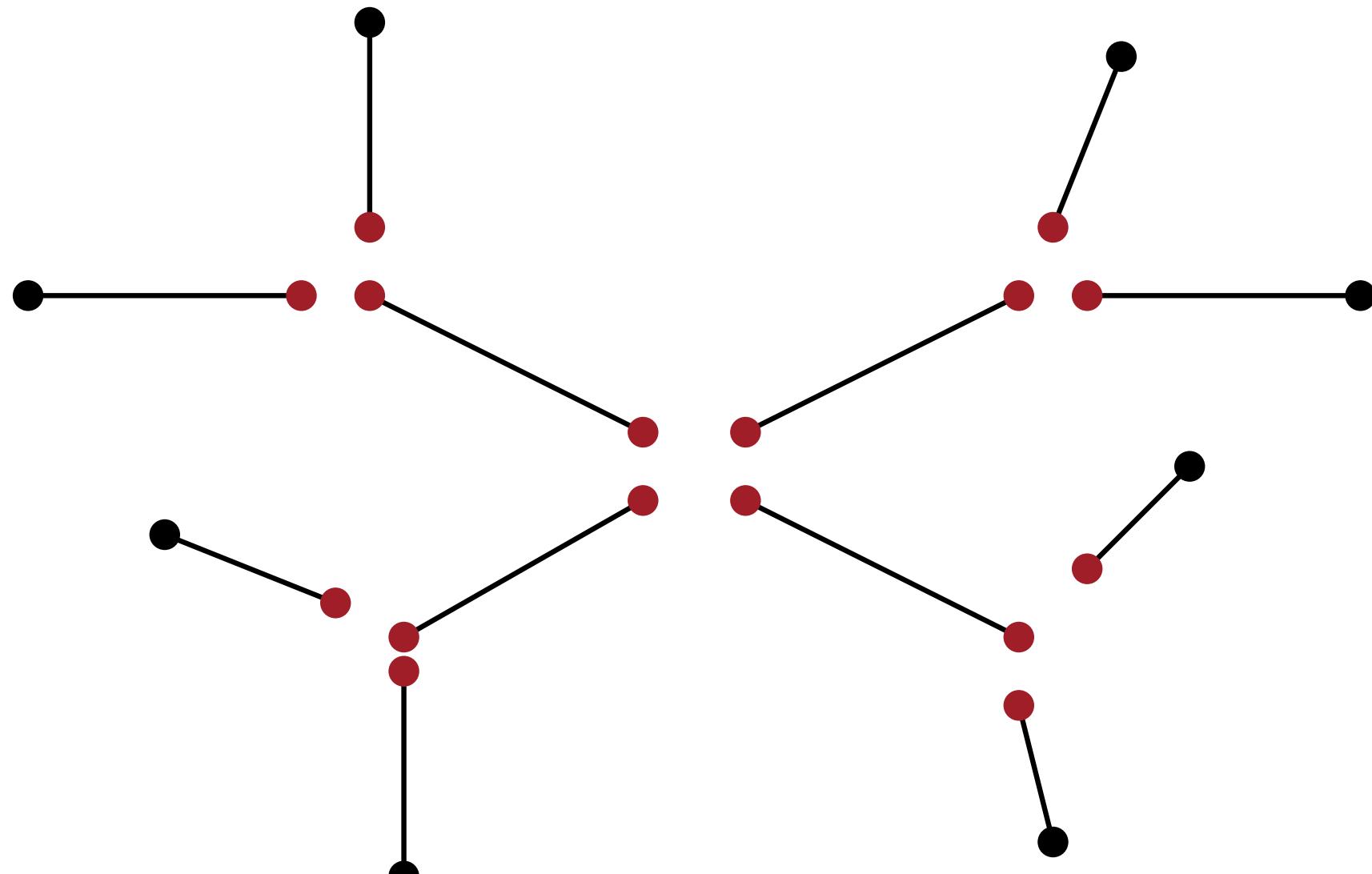


# Idea of Proof for the Feedback Forest Set



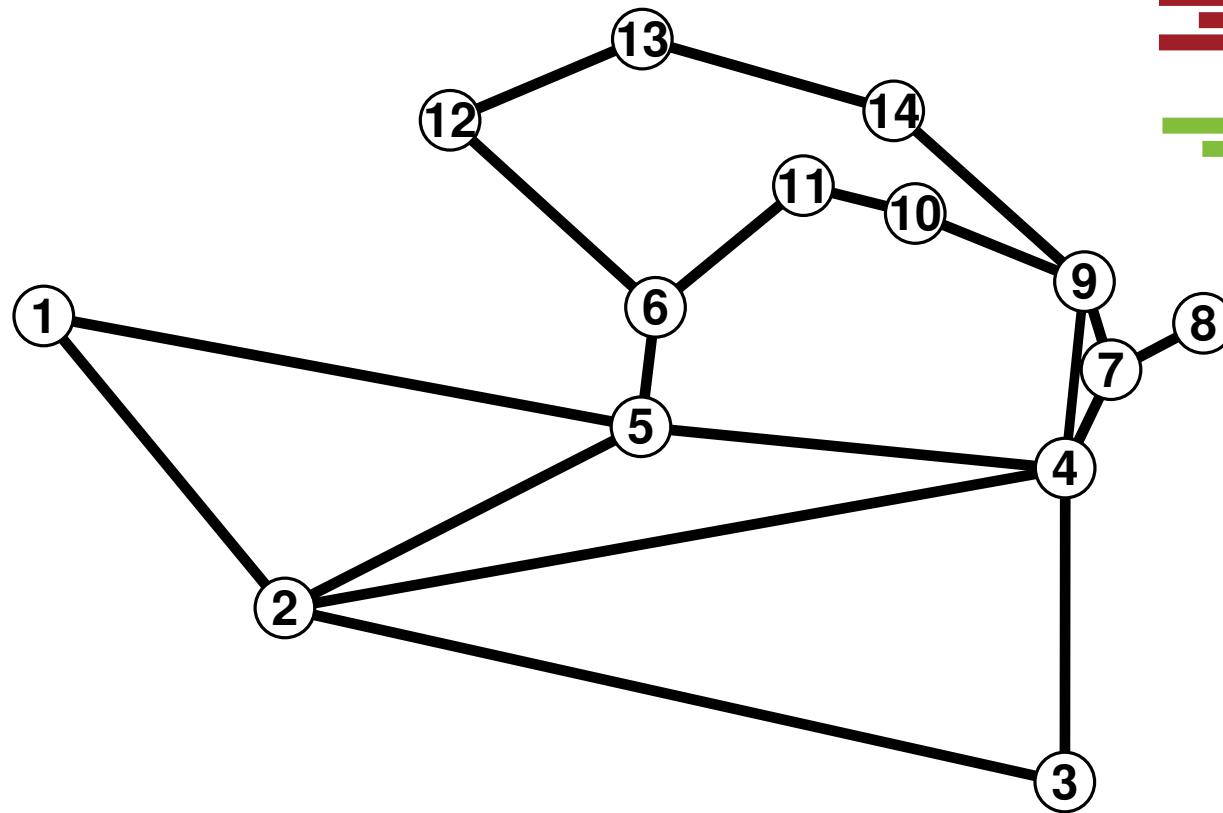
⇒ Trees can be cut into *edges*

# Idea of Proof for the Feedback Forest Set



⇒ For every edge there exists a solution!

# Feedback Cactus Set

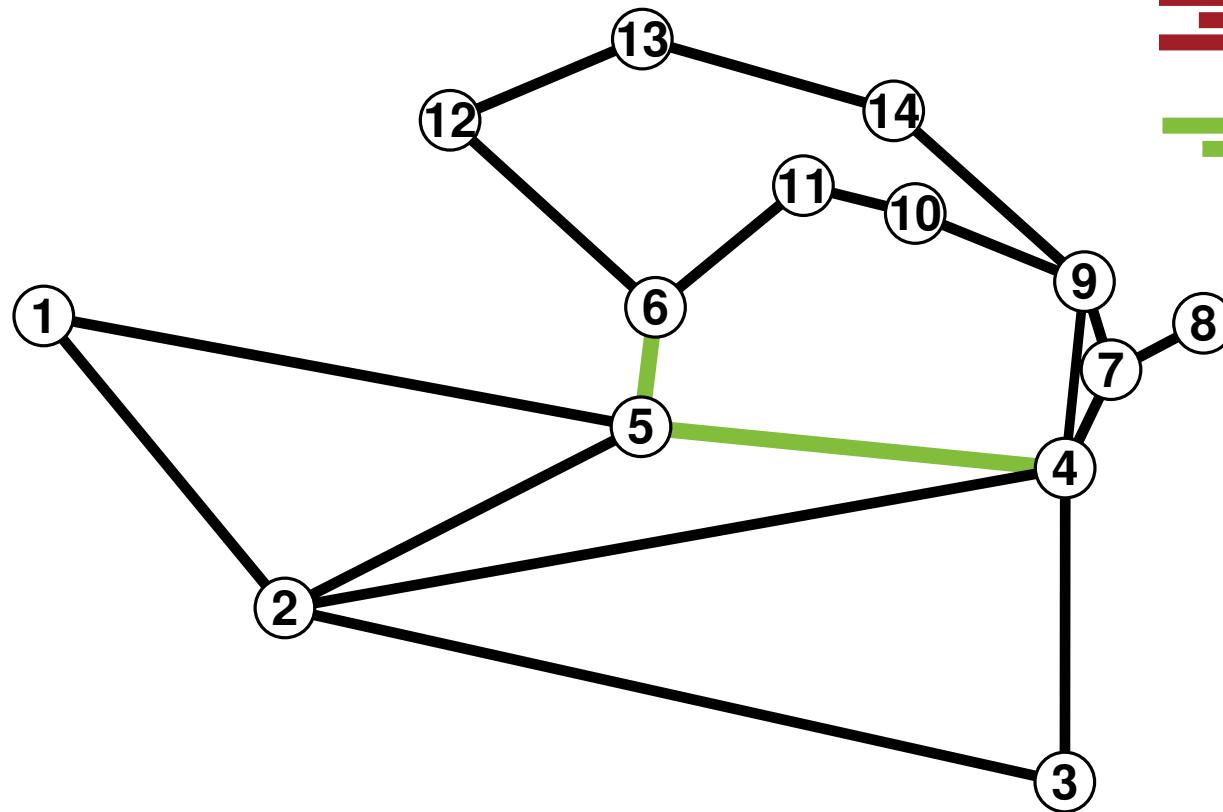


*Feedback Forest Set*



*Feedback Cactus Set*

# Feedback Cactus Set



Feedback Forest Set

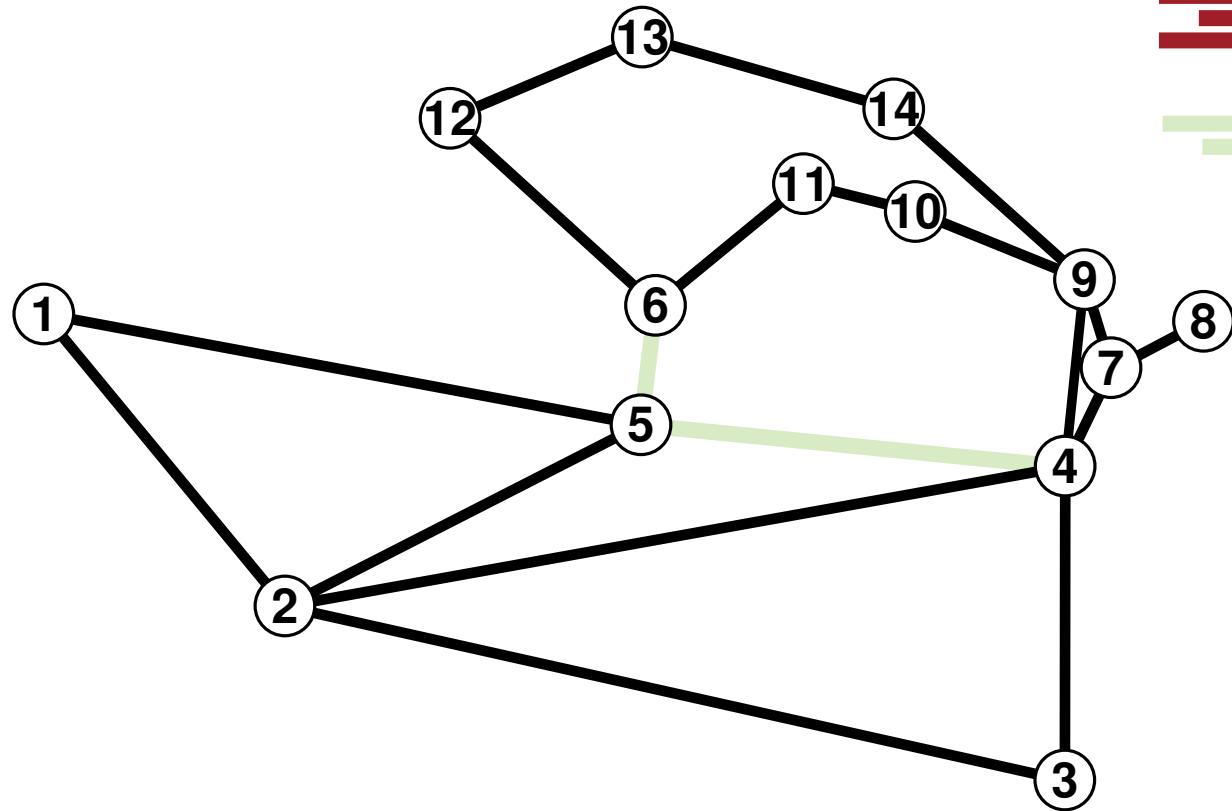


Feedback Cactus Set

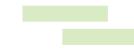


A set of *cacti*  
remains!

# Feedback Cactus Set



Feedback Forest Set



Feedback Cactus Set



A set of *cacti*  
remains!

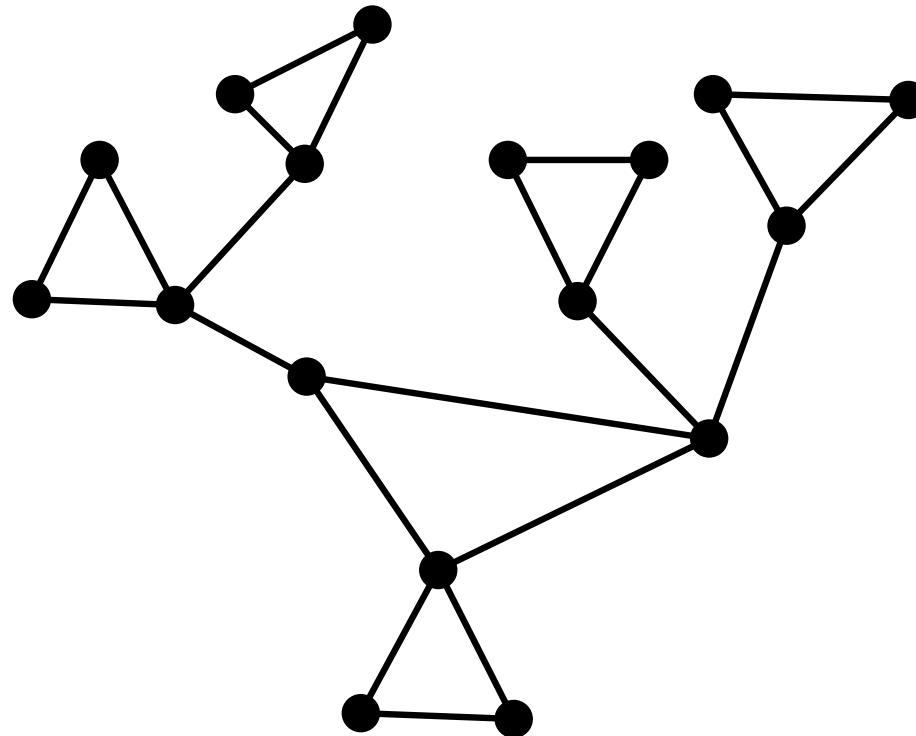
## Theorem 2

Physical subgrid **cactus**, line limits on cactus suitably bounded.

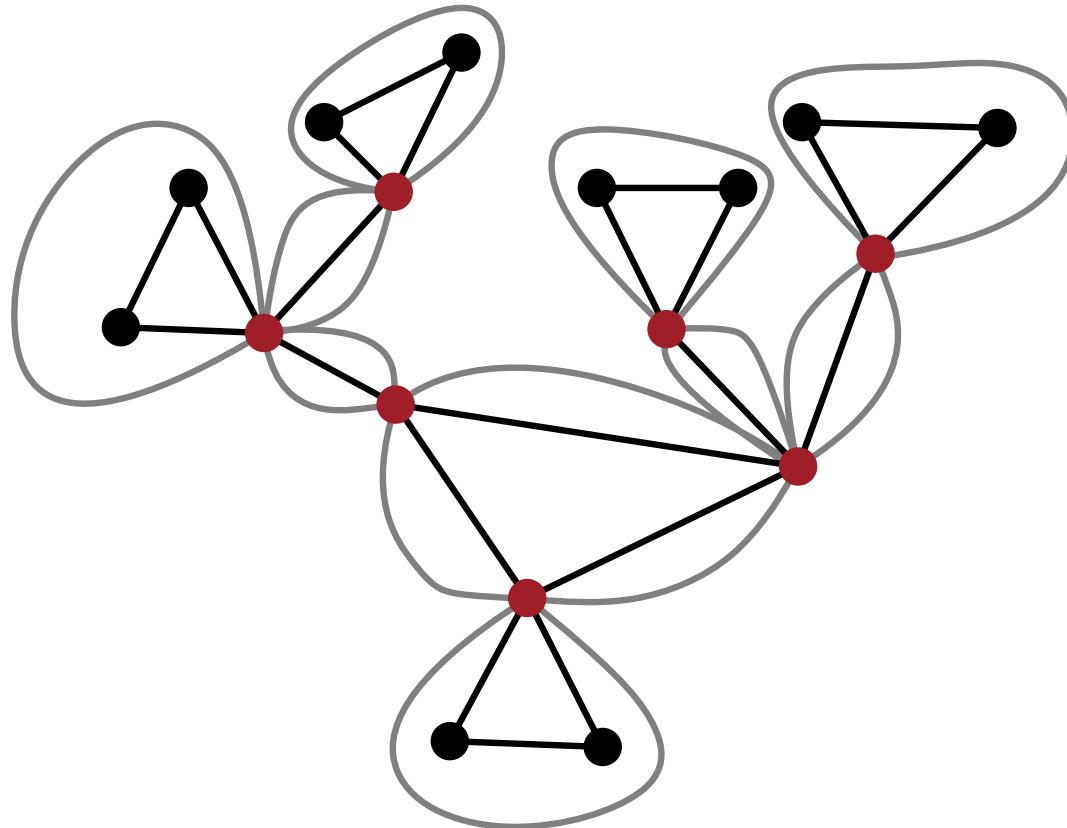


For every **graph theoretical flow** there is a **cost-equivalent flow** obeying voltage laws.

# Idea of Proof for the Feedback Cactus Set

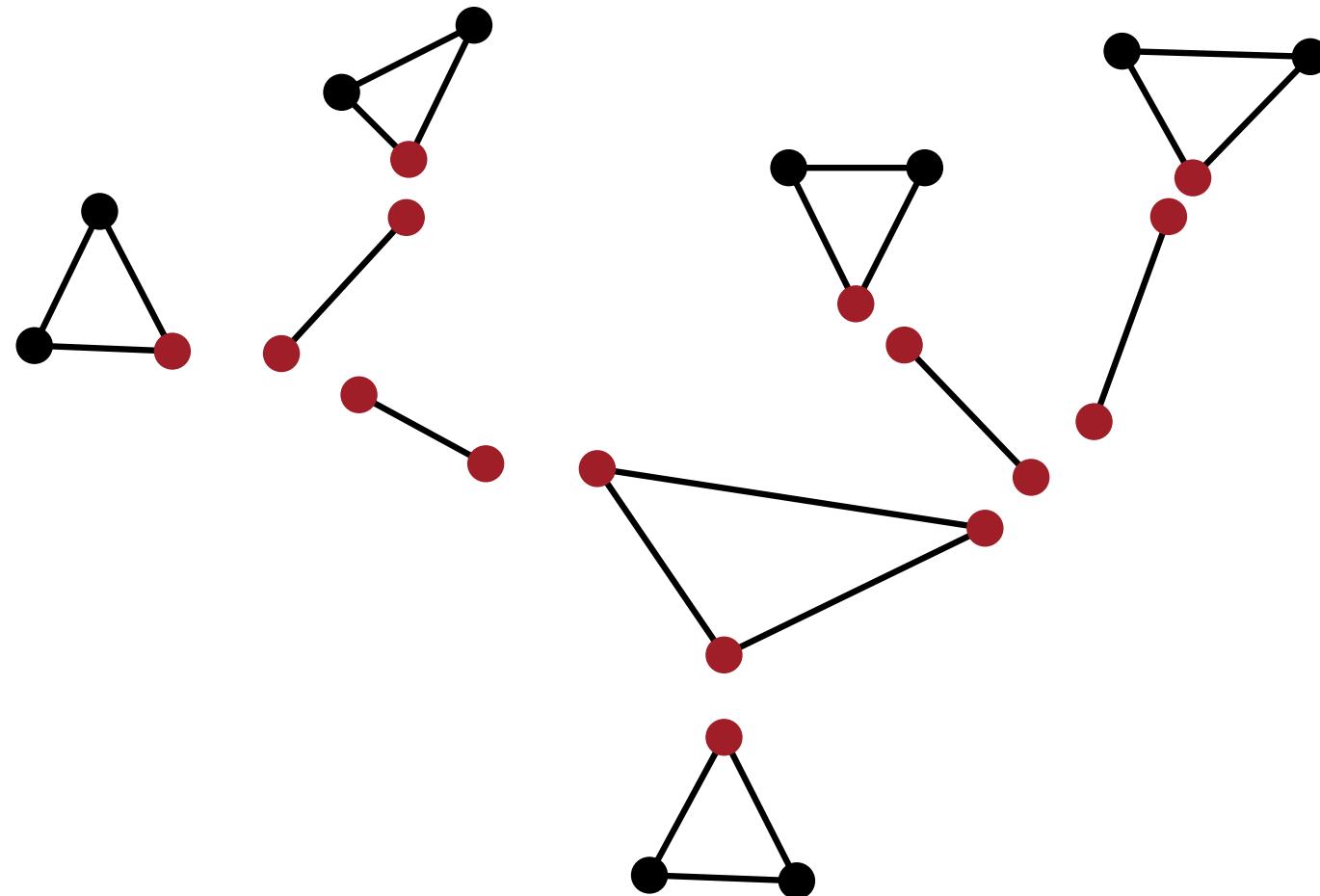


# Idea of Proof for the Feedback Cactus Set



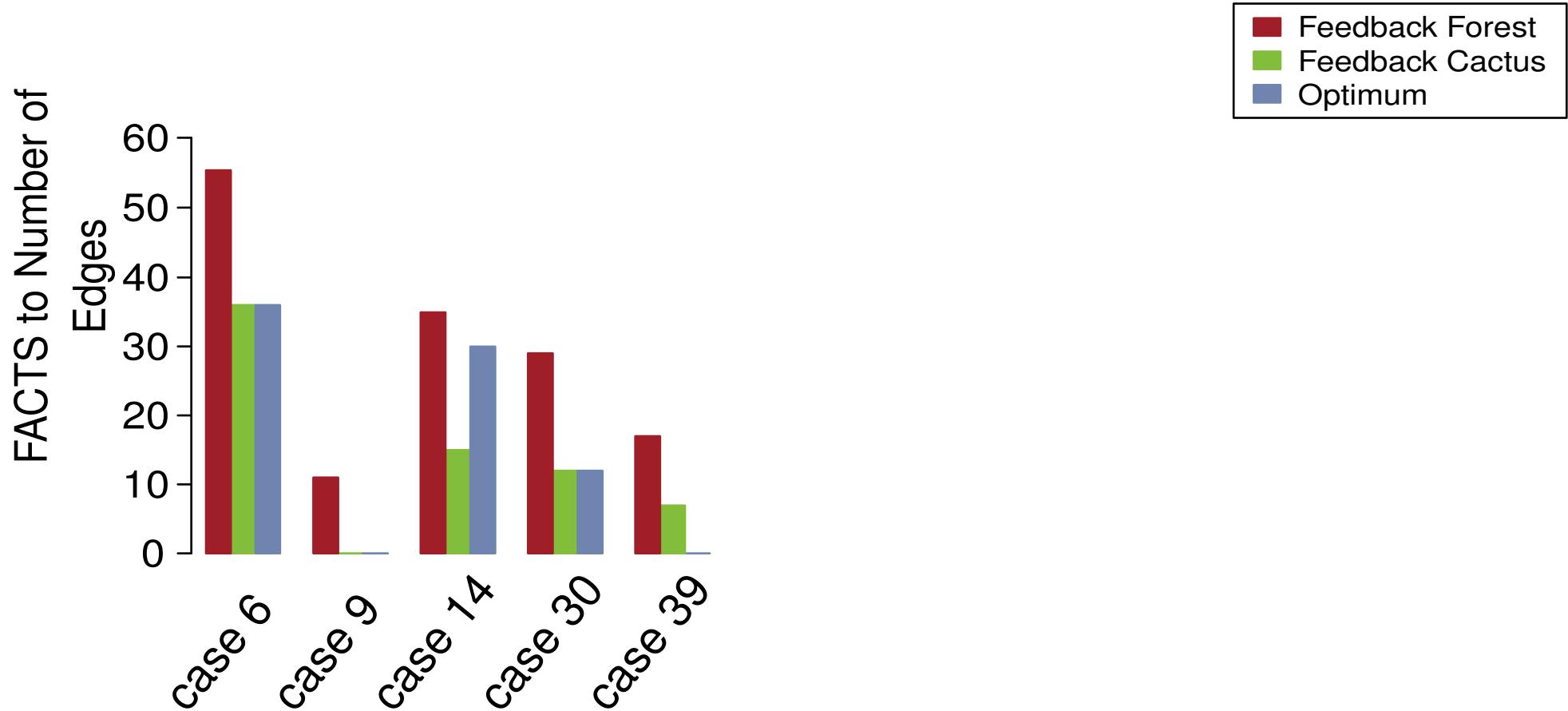
⇒ Cacti can be cut into single *edges* and single *cycles*

# Idea of Proof for the Feedback Cactus Set

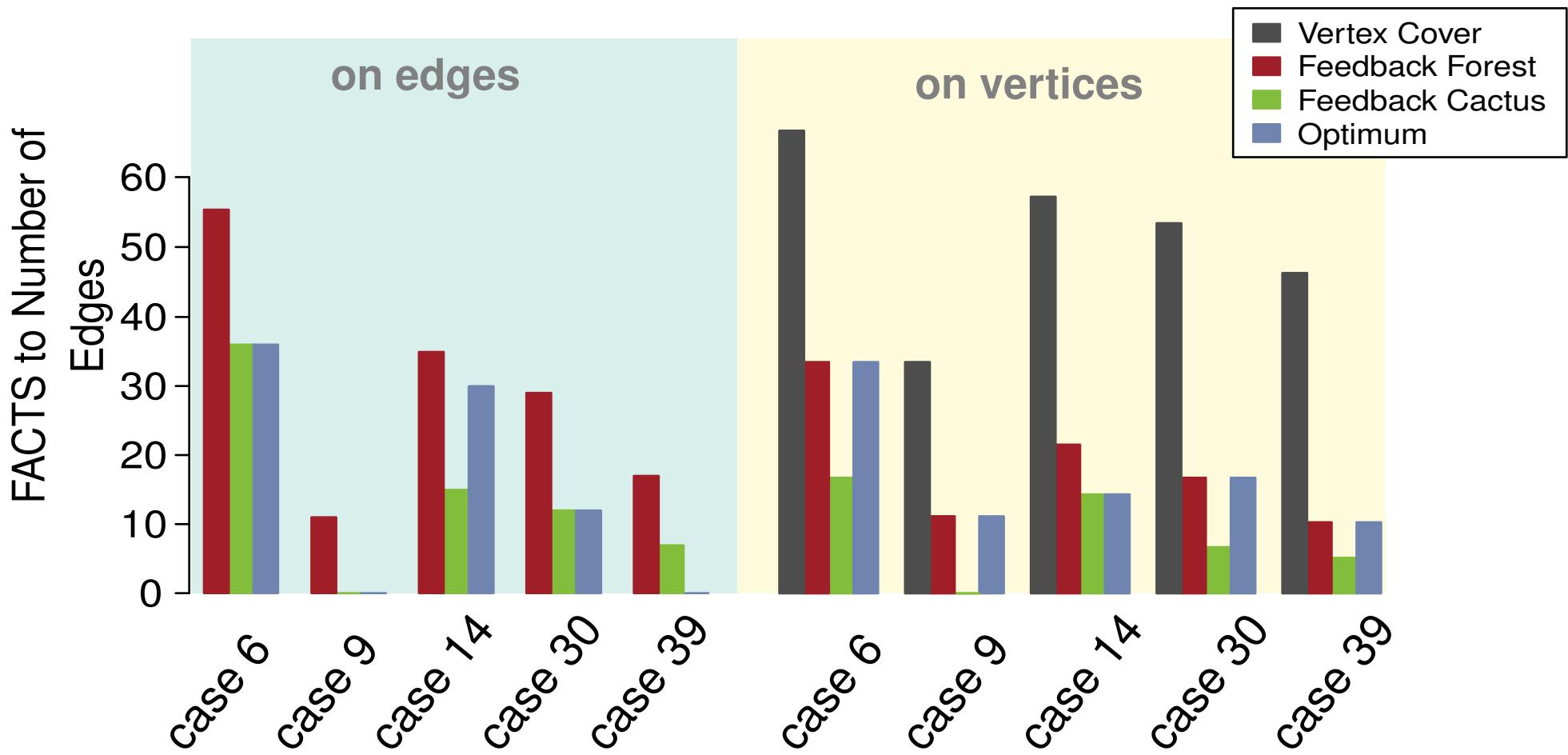


⇒ For every *edge* and every *cycle* always exists a solution!

# Number FACTS $\leftrightarrow$ structural Results

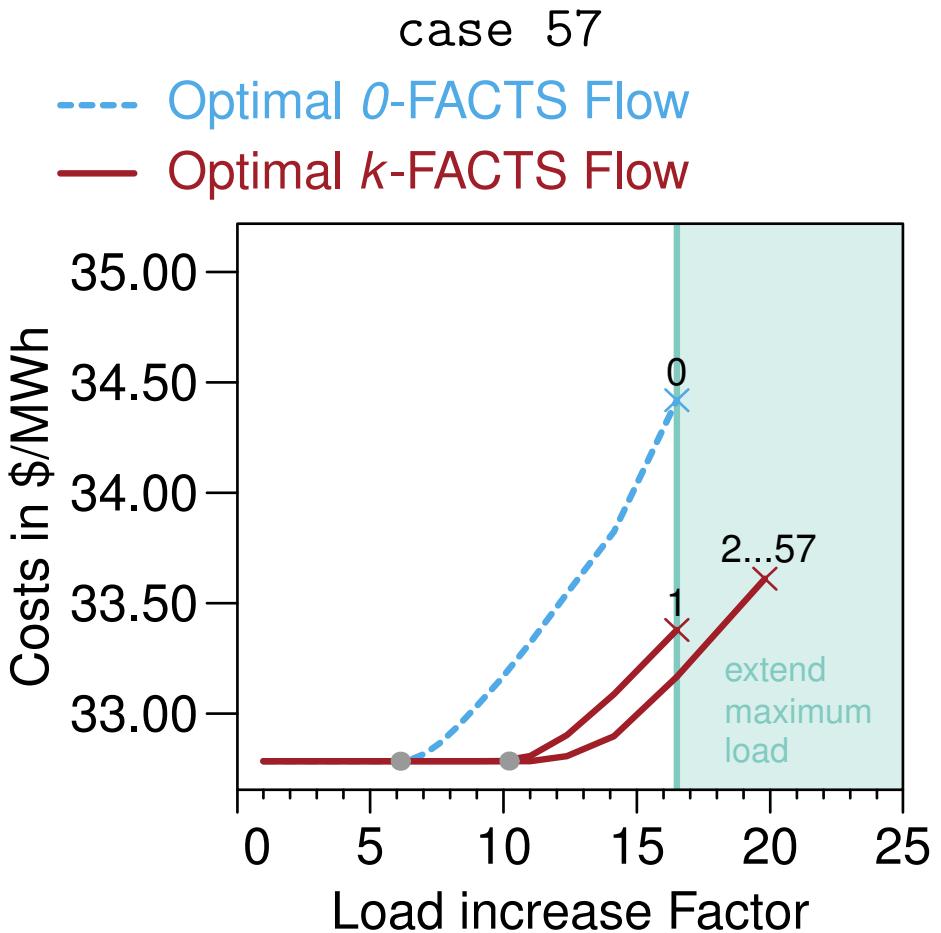
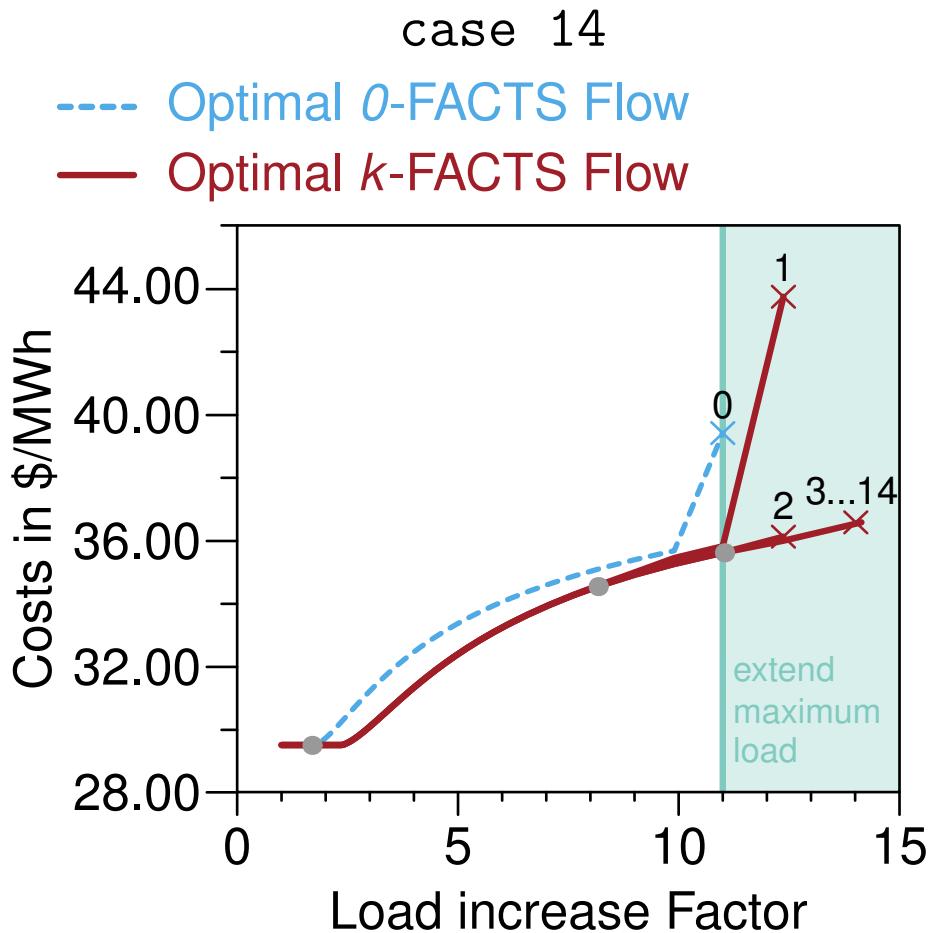


# Number FACTS $\leftrightarrow$ structural Results

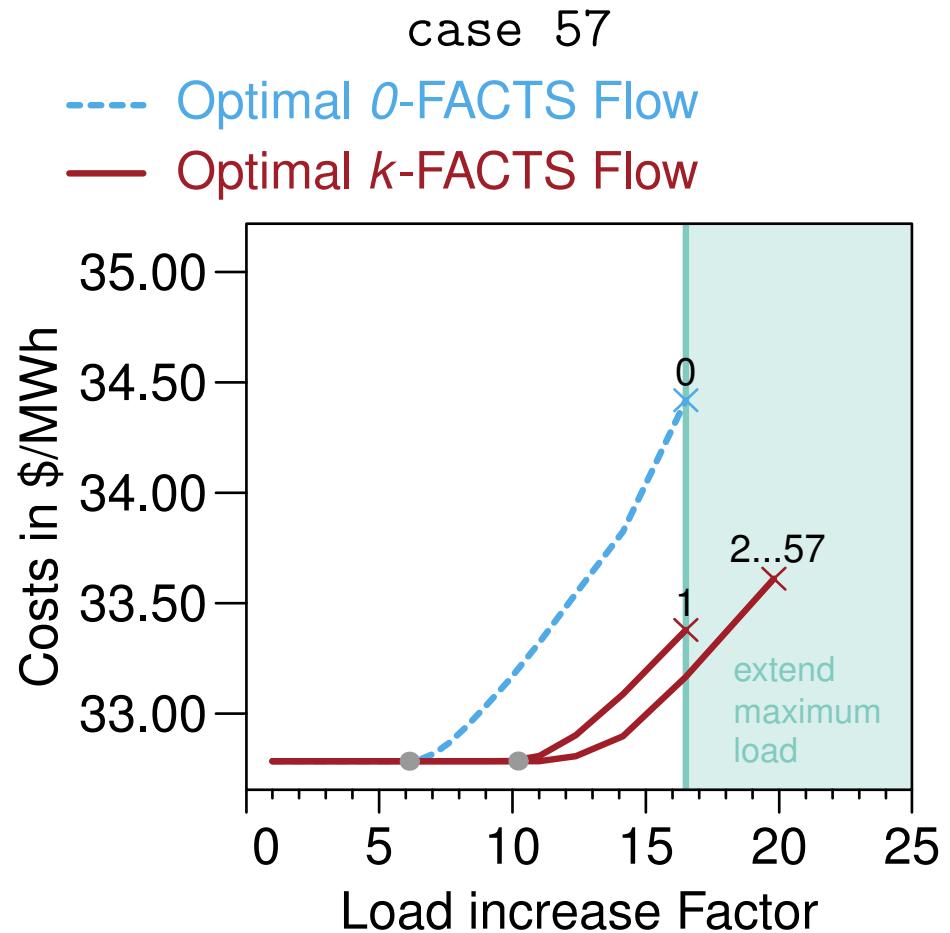
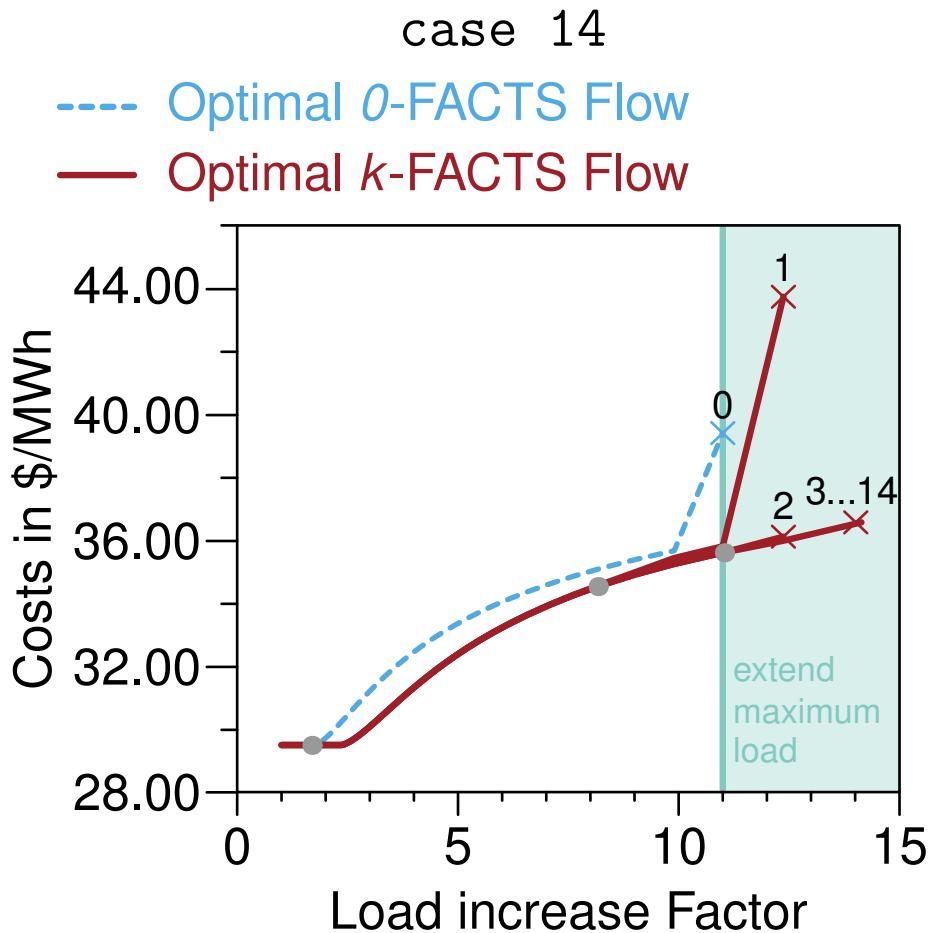


Often a **small number** of ideal FACTS suffices for **matching cost** of the flow model.

# Operation under Increasing Loads

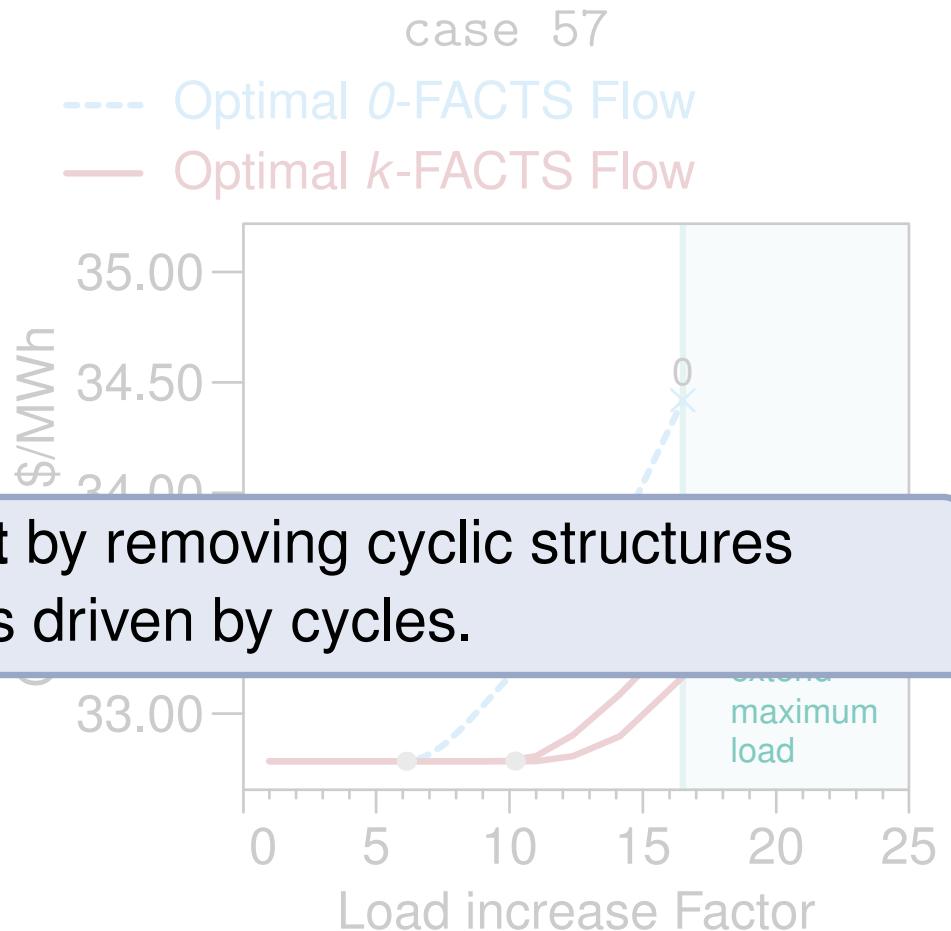
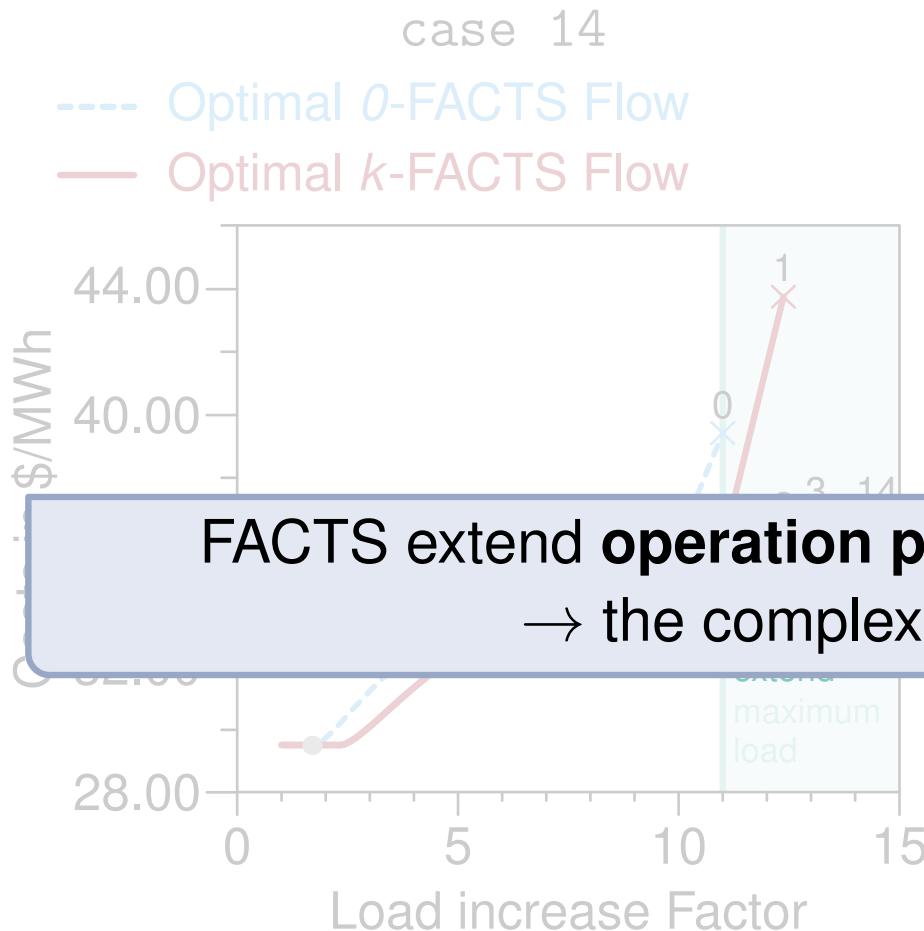


# Operation under Increasing Loads



Very few flow control branches extend **maximum load** at **lower cost**.

# Operation under Increasing Loads



**FACTS extend operation point by removing cyclic structures**  
 → the complexity is driven by cycles.

Very few flow control branches extend **maximum load** at lower cost.

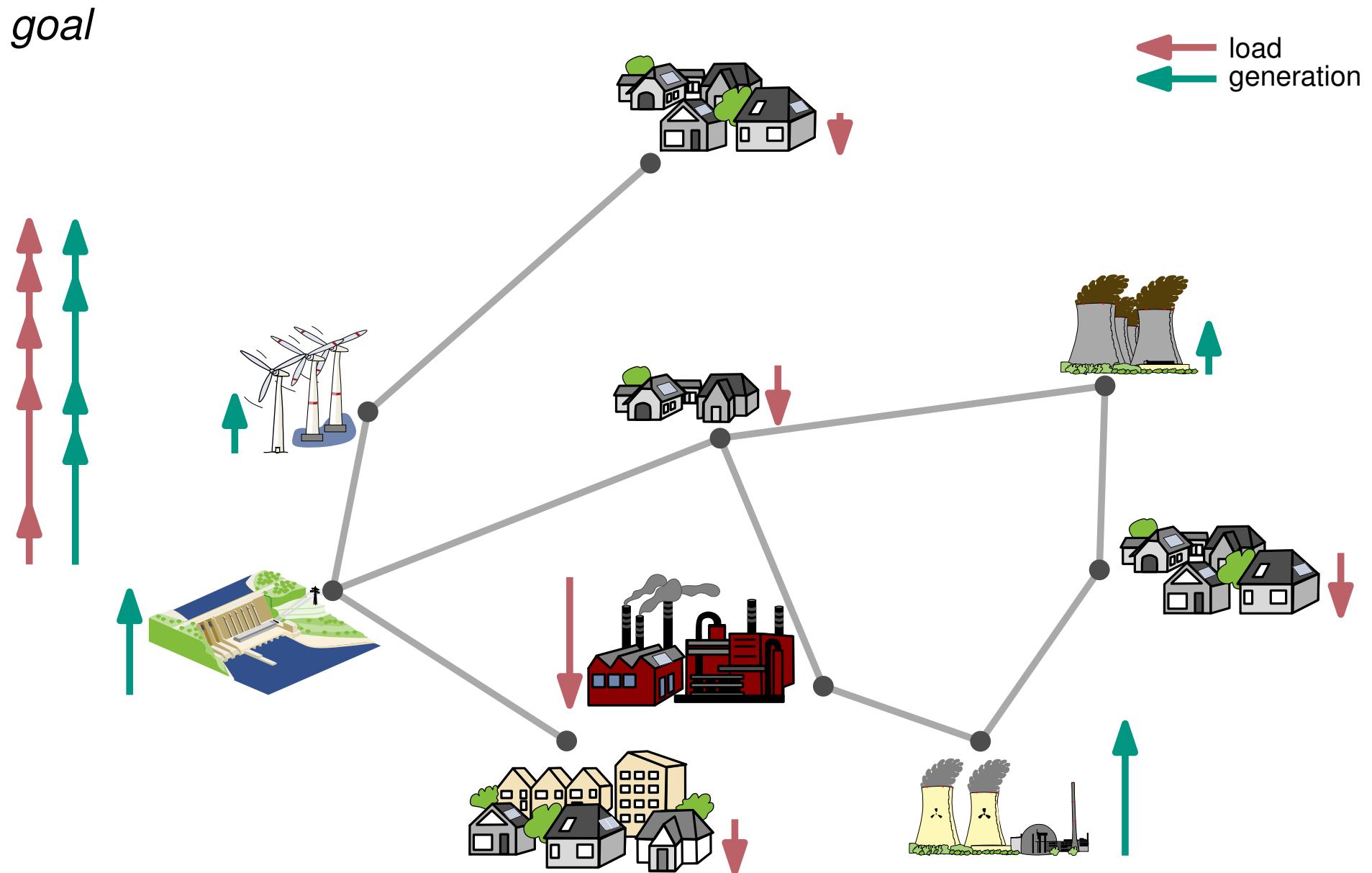


## From ideal FACTS to real FACTS

Figures FLTR:

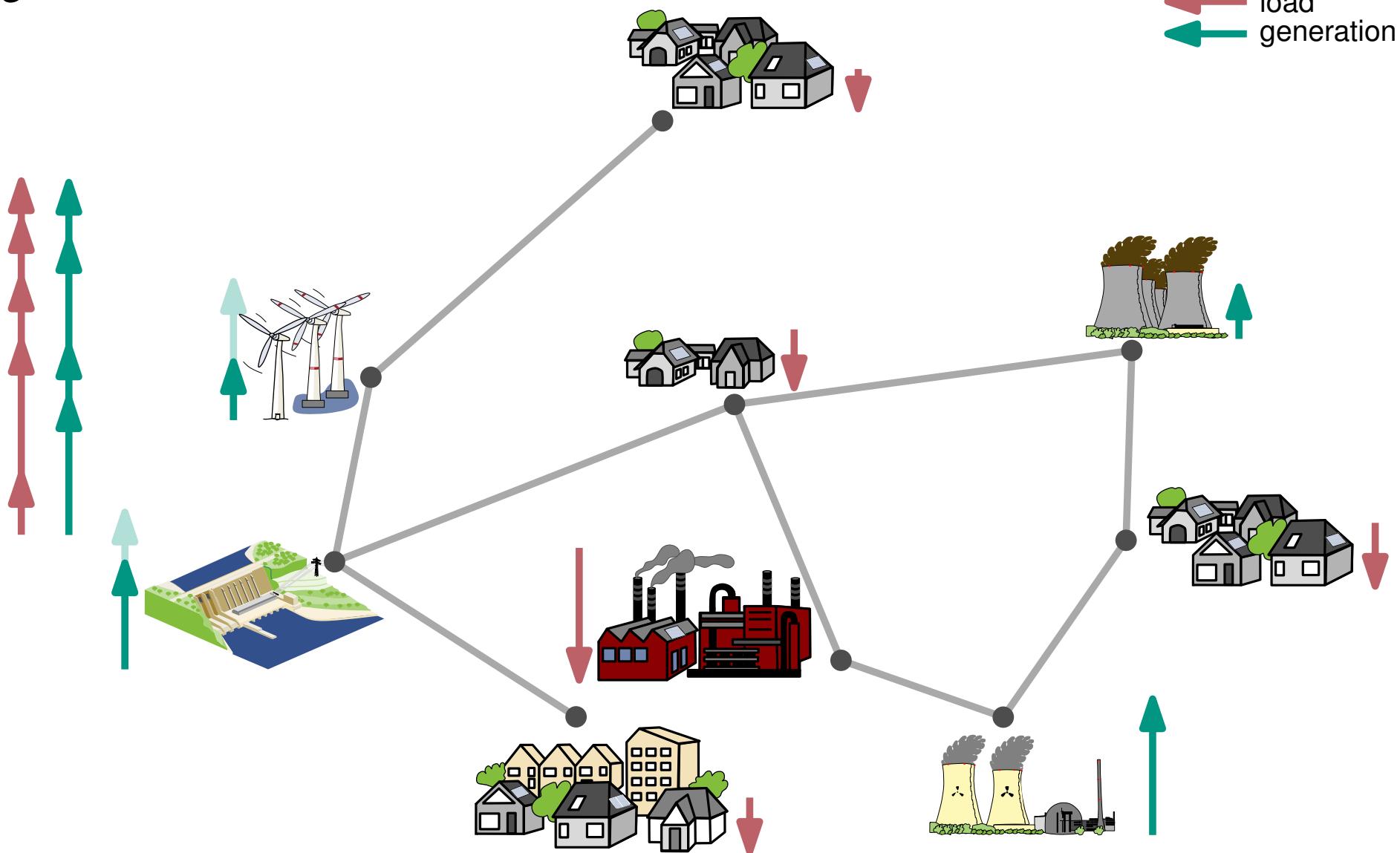
- <sup>3</sup> <http://www.abb.com/cawp/seitp202/c36f4e62da52ab46c1257670003690d3.aspx>
- <sup>4</sup> <http://electrical-engineering-portal.com/facts-flexible-ac-transmission-systems>

# AC Conservation of Flow in Power Grids

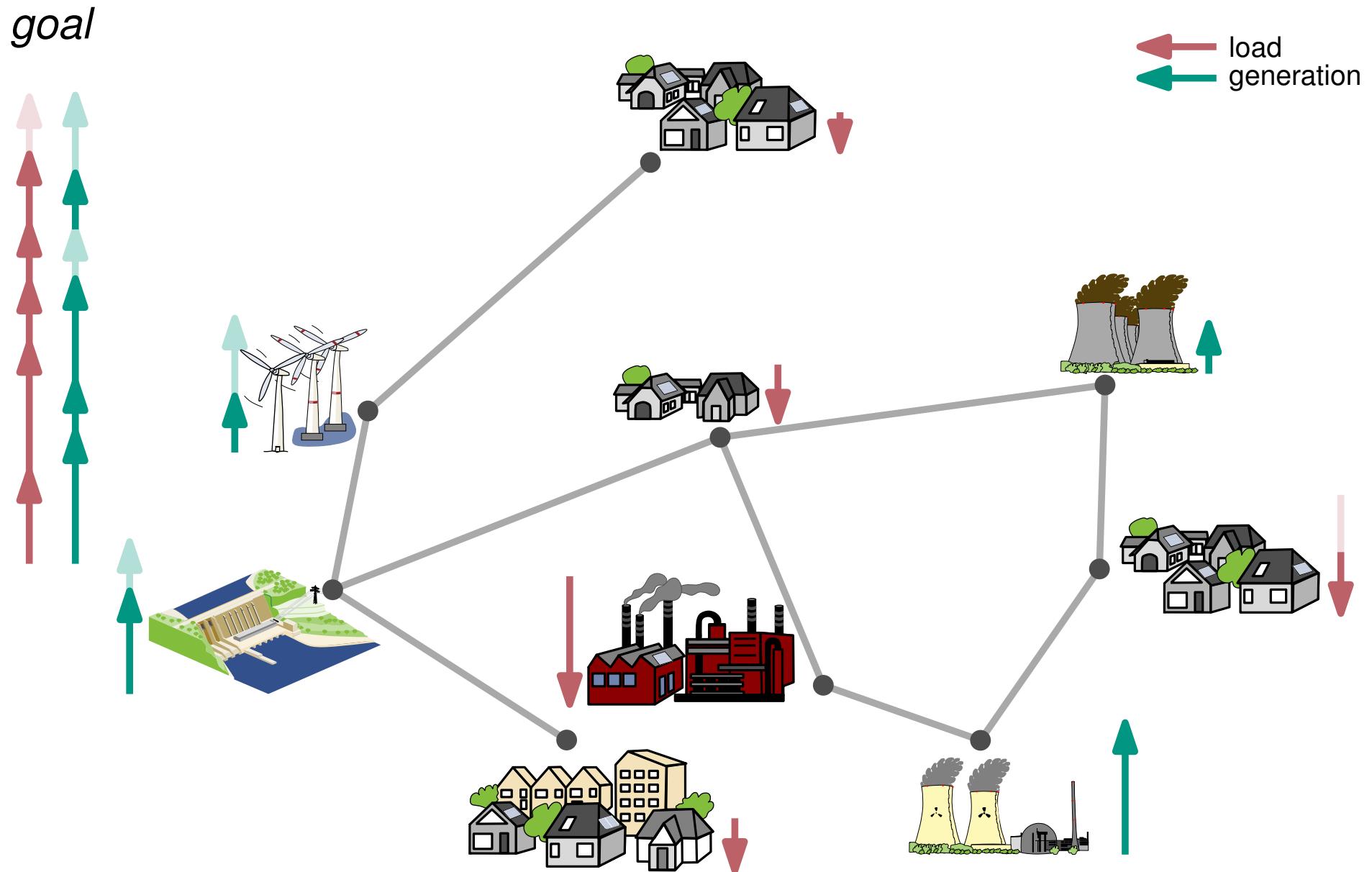


# AC Conservation of Flow in Power Grids

## *goal*



# AC Conservation of Flow in Power Grids

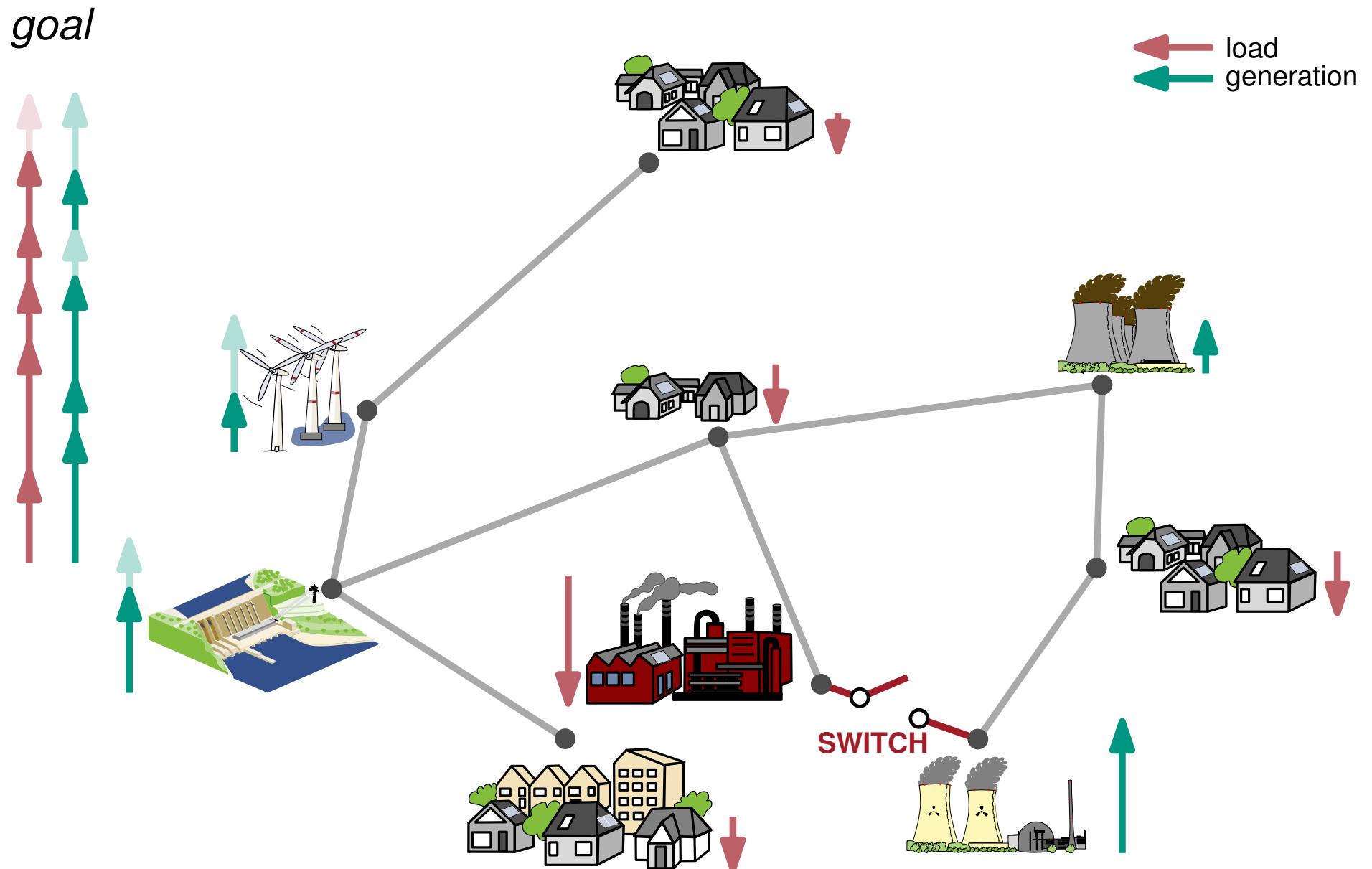


# AC Conservation of Flow in Power Grids

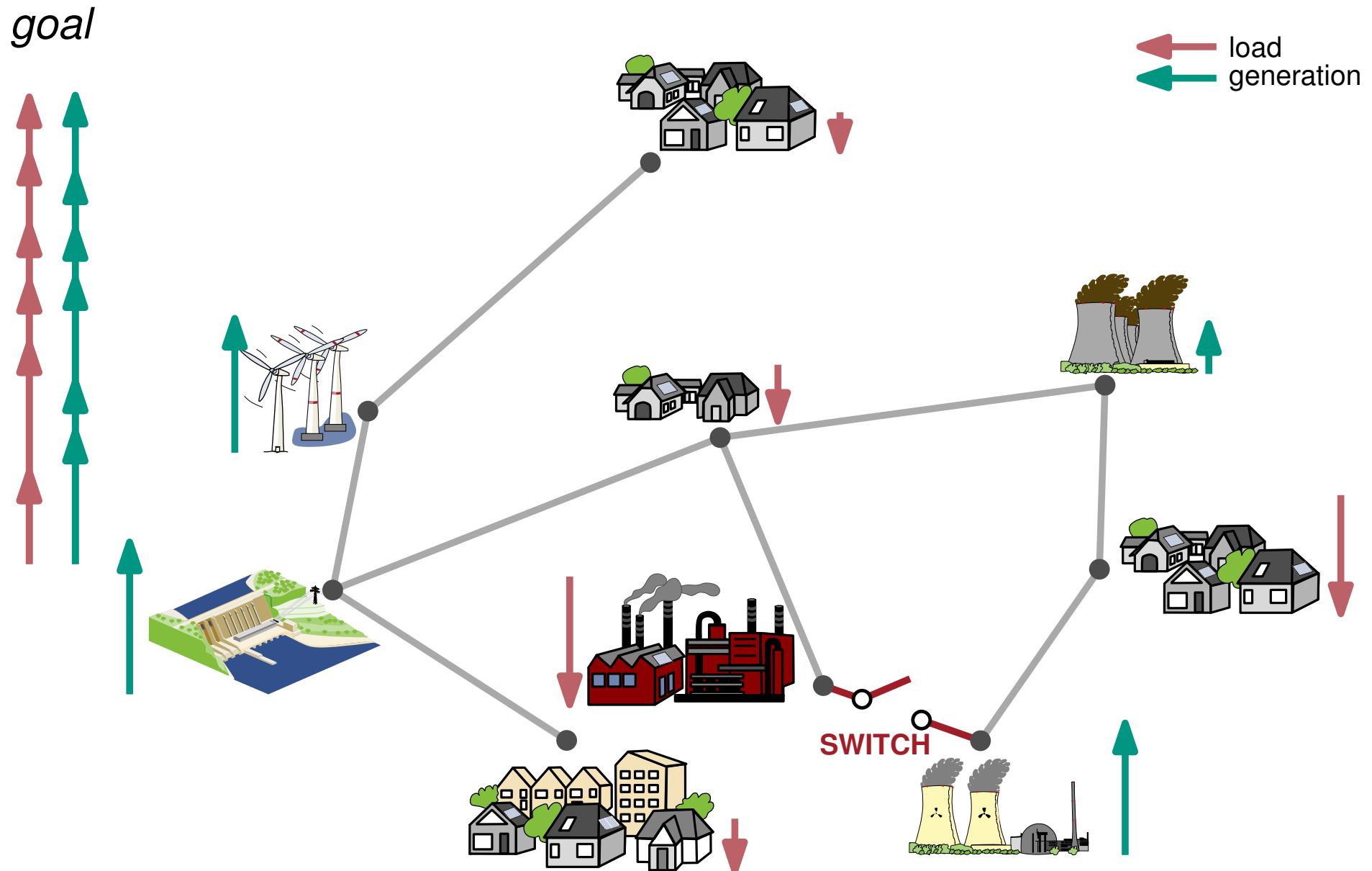
*goal*



# Switches



# Switches



# Switches

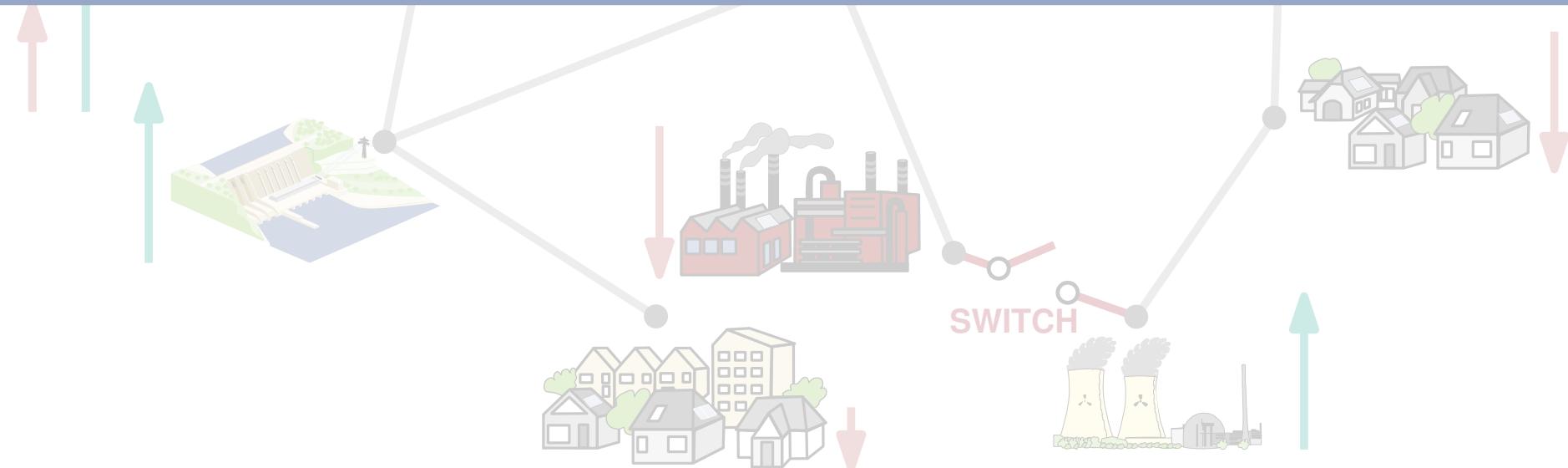
*goal*



load  
generation

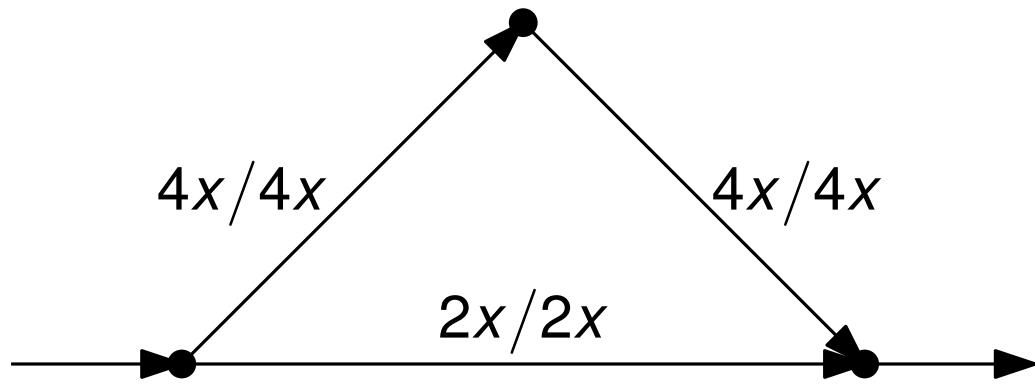


AC Flow Conservation  $\subseteq$  Maximum Switching Flow  $\subseteq$  Optimal Switching Flow  
(FEAS)  $\subseteq$  (MSF)  $\subseteq$  (OSF)

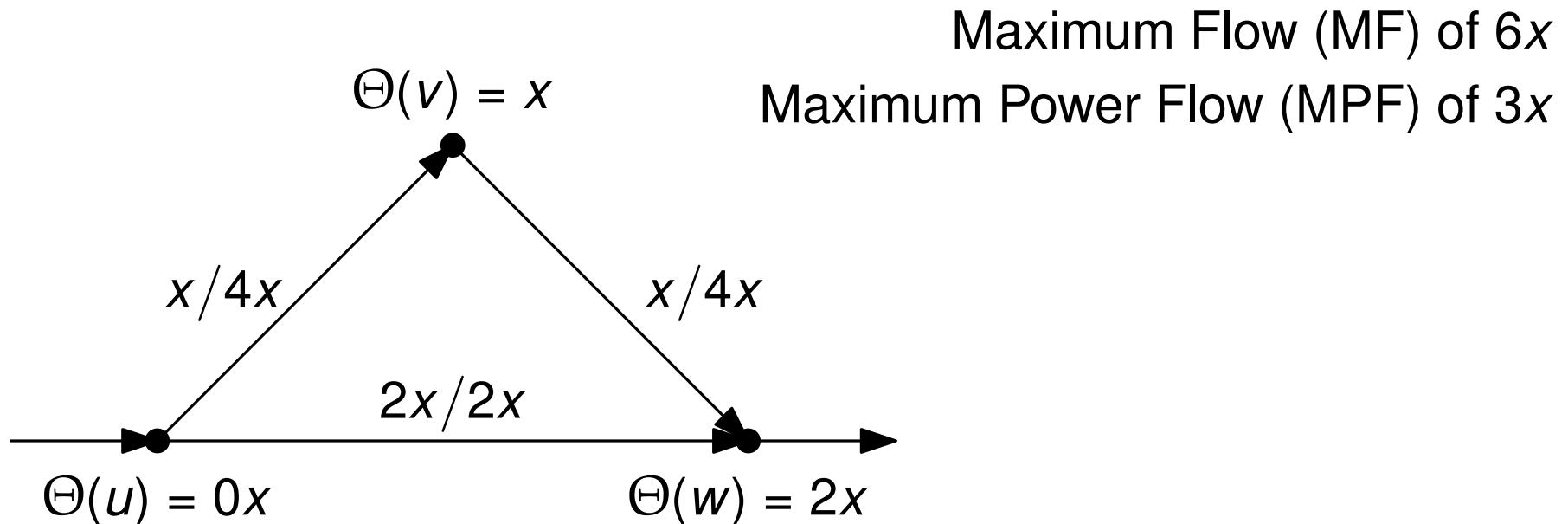


# Maximum Switching Flow (MSF)

Maximum Flow (MF) of  $6x$

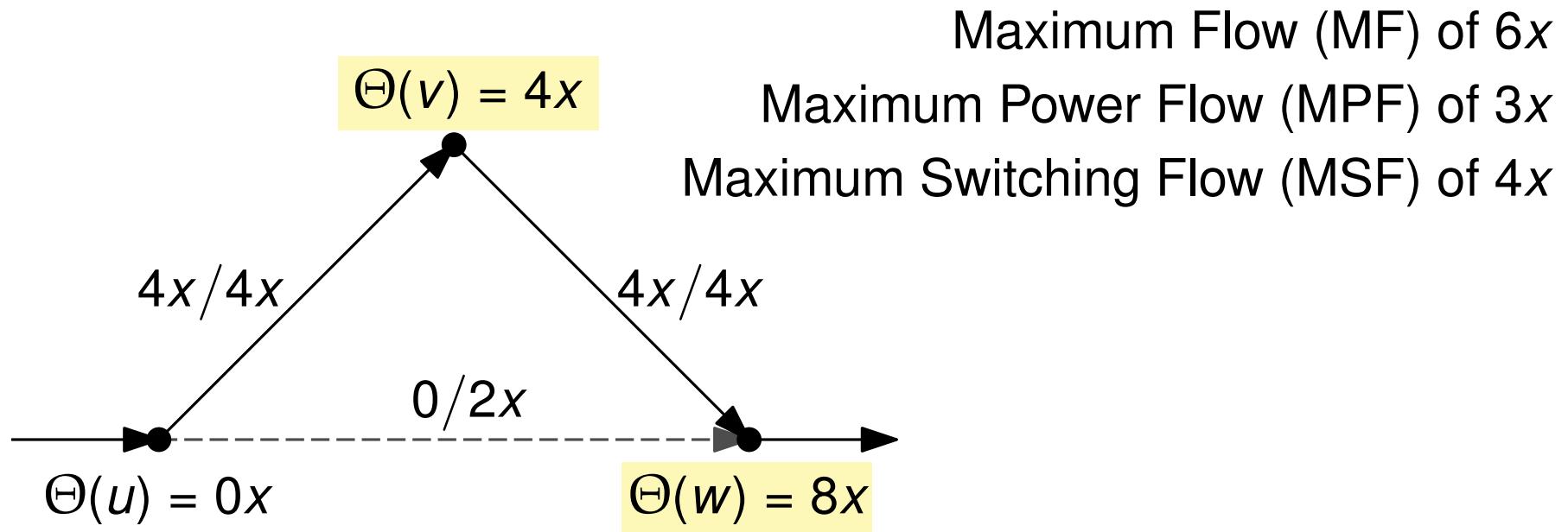


# Maximum Switching Flow (MSF)



$$P(u, v) = B(u, v)(\Theta(u) - \Theta(w))$$

# Maximum Switching Flow (MSF)



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# Complexity of the Maximum Switching Flow (MSF)

Graph Structure	Complexity
cacti with max degree 3	NP-hard
2-level trees	NP-hard
planar graphs with max degree of 3	strong NP-hard
arbitrary graphs	Non-APX

# Complexity of the Maximum Switching Flow (MSF)

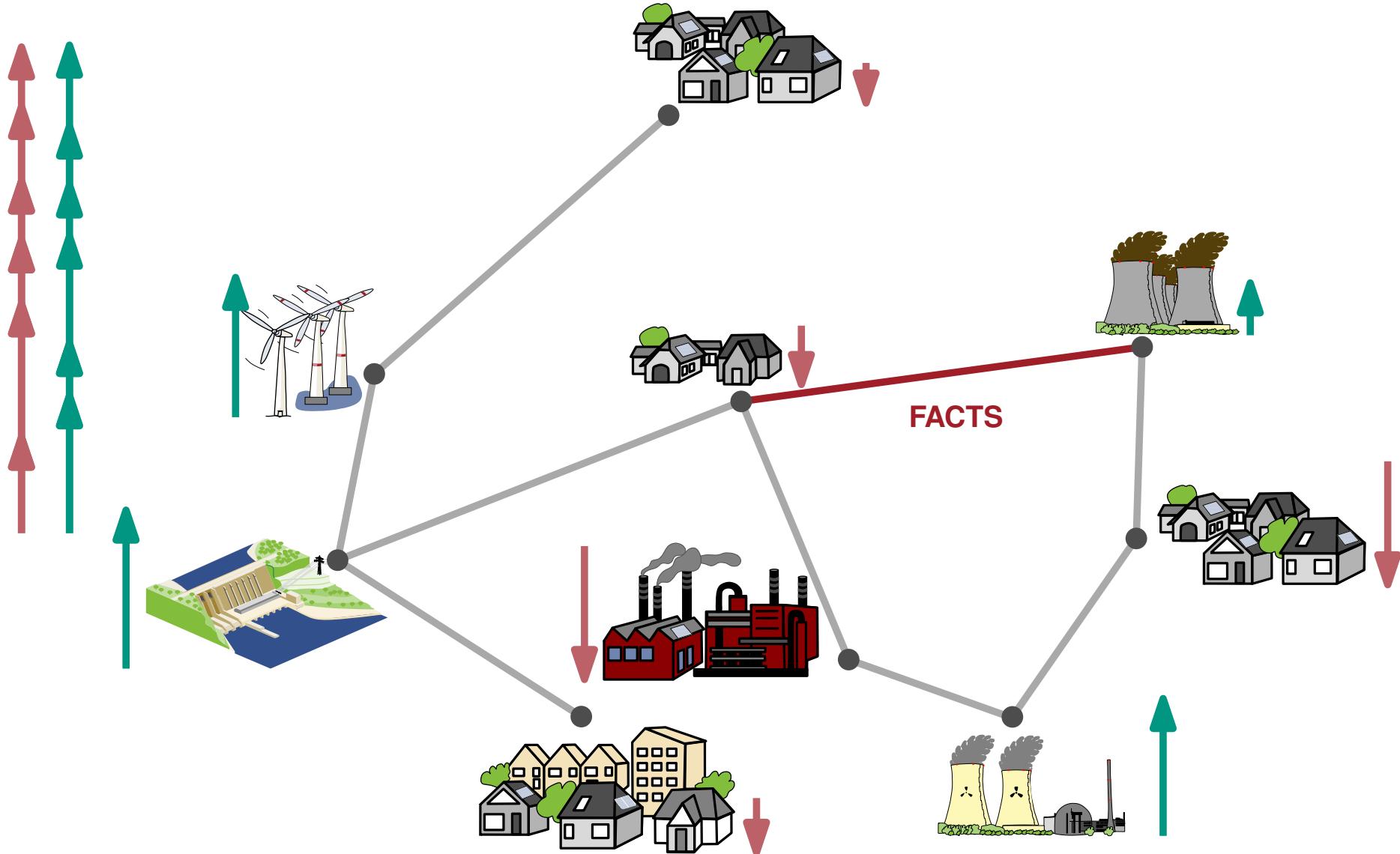
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## Theorem 3

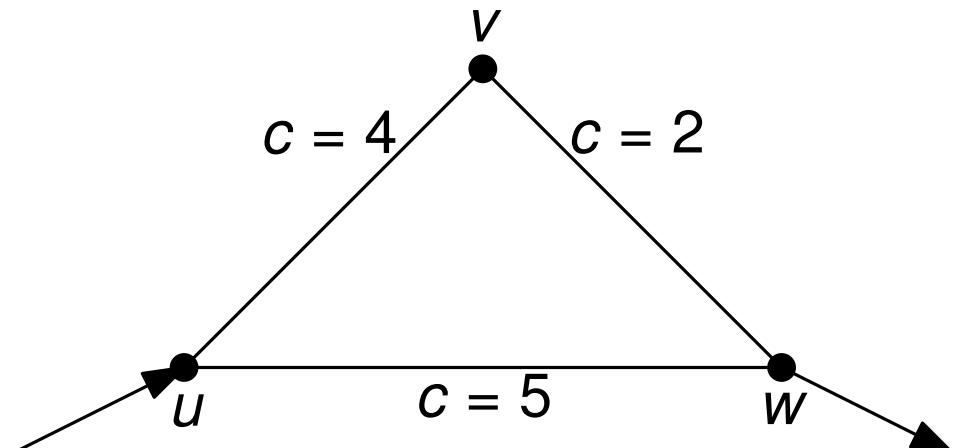
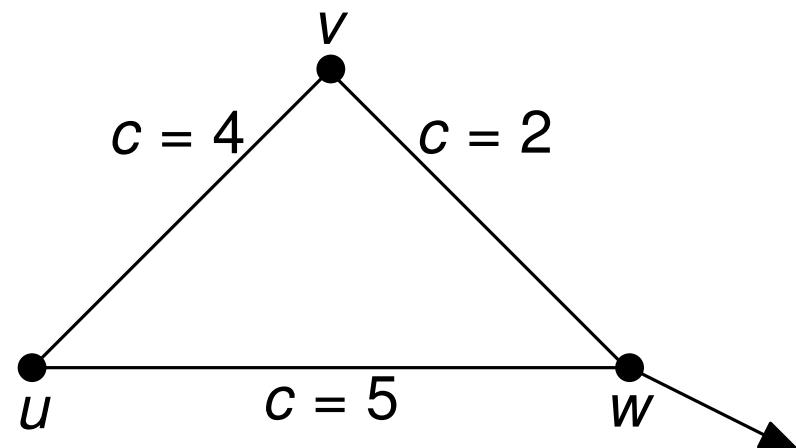
The maximum switching flow (MSF) is 2-approximatable on cacti.

# FACTS with Susceptance Interval

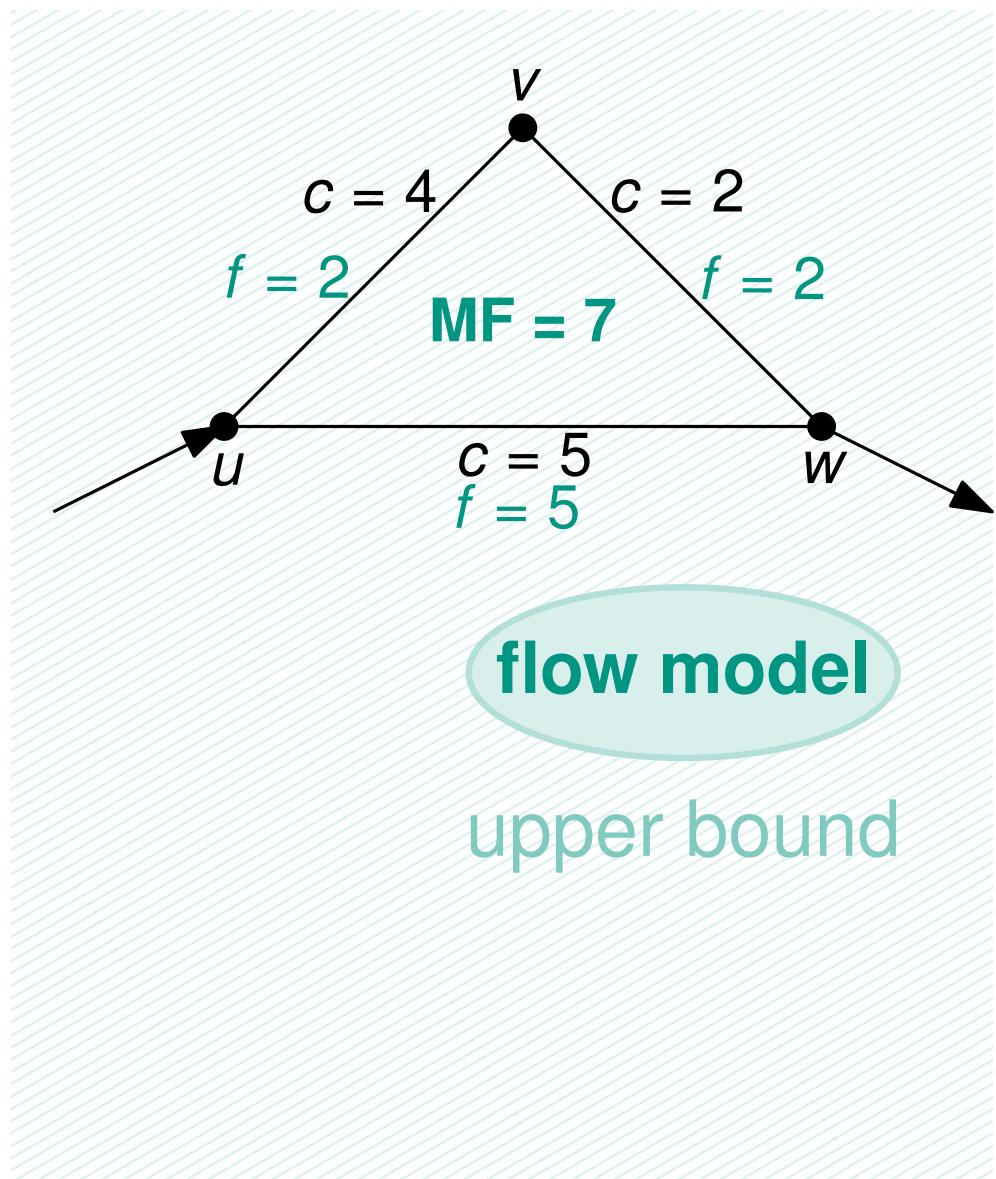
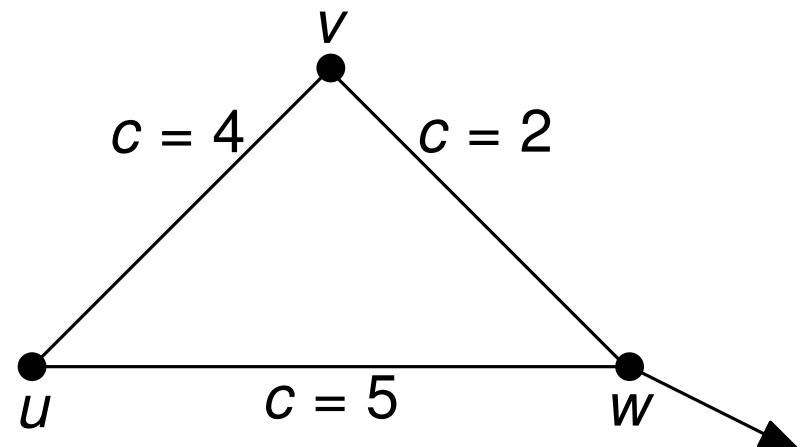
*goal*



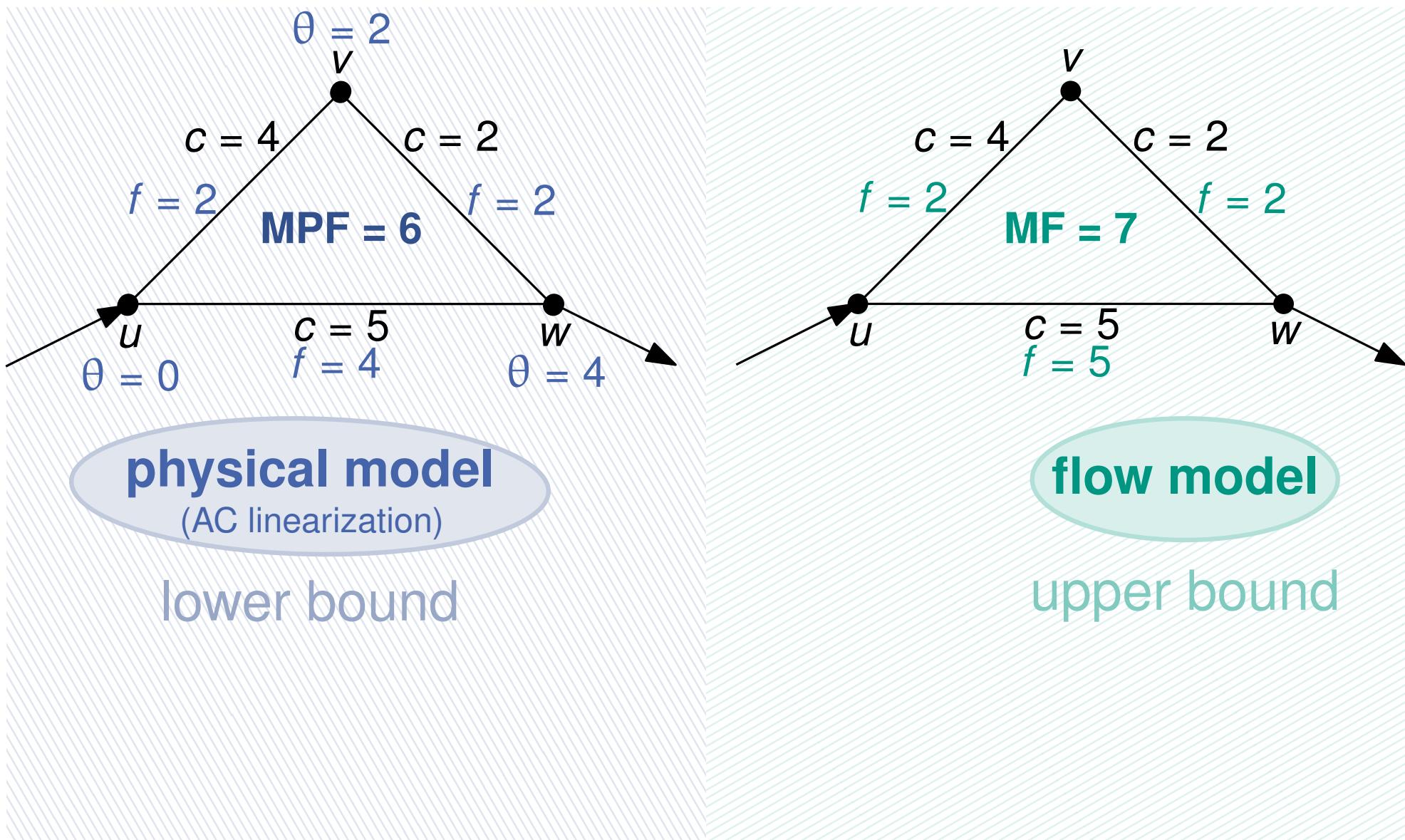
# Maximum FACTS Flow



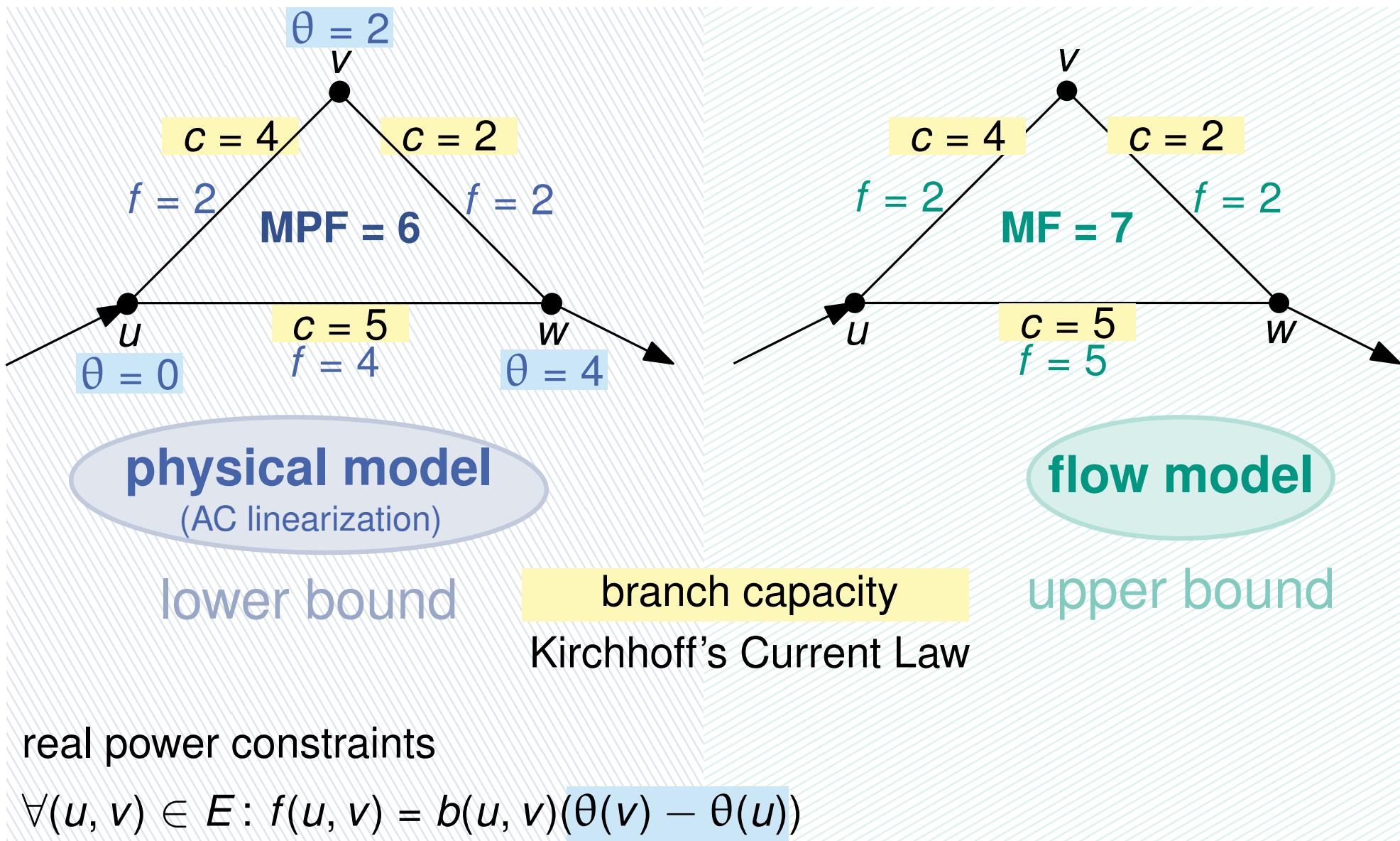
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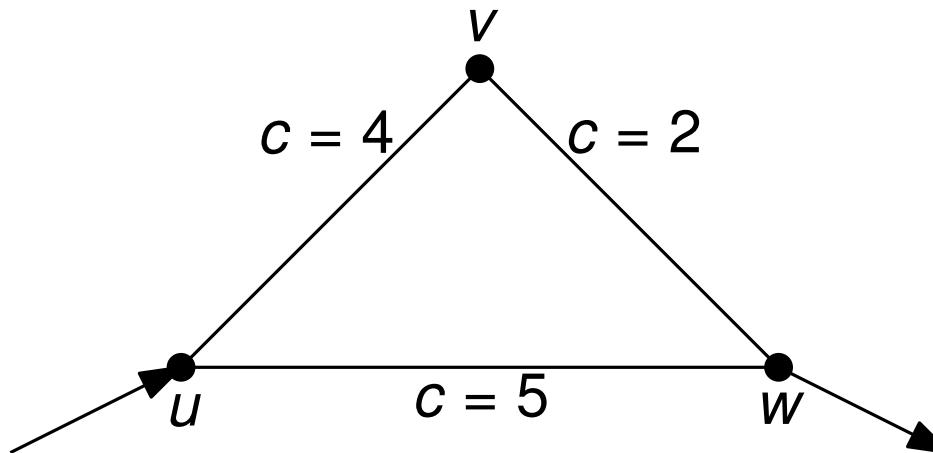
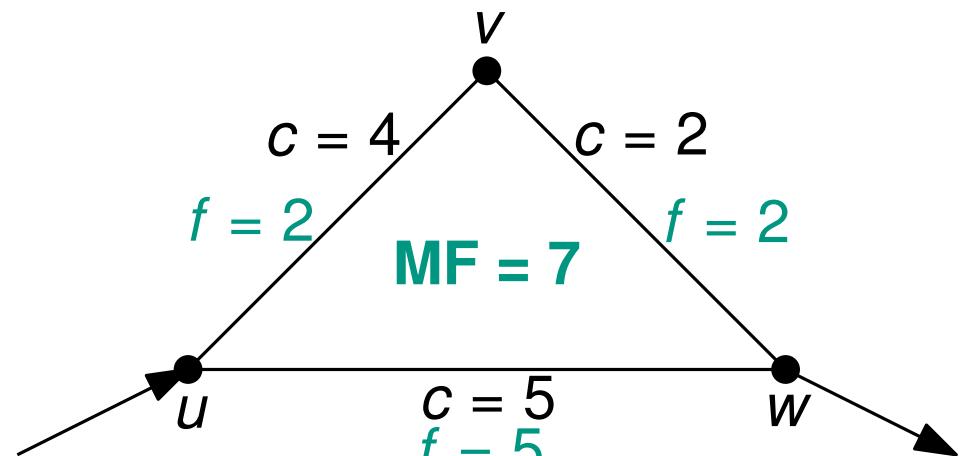
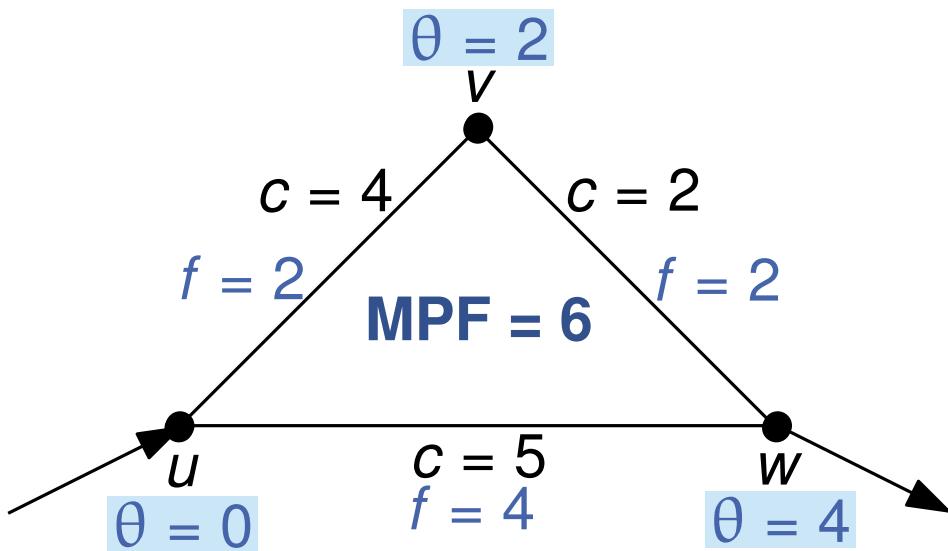
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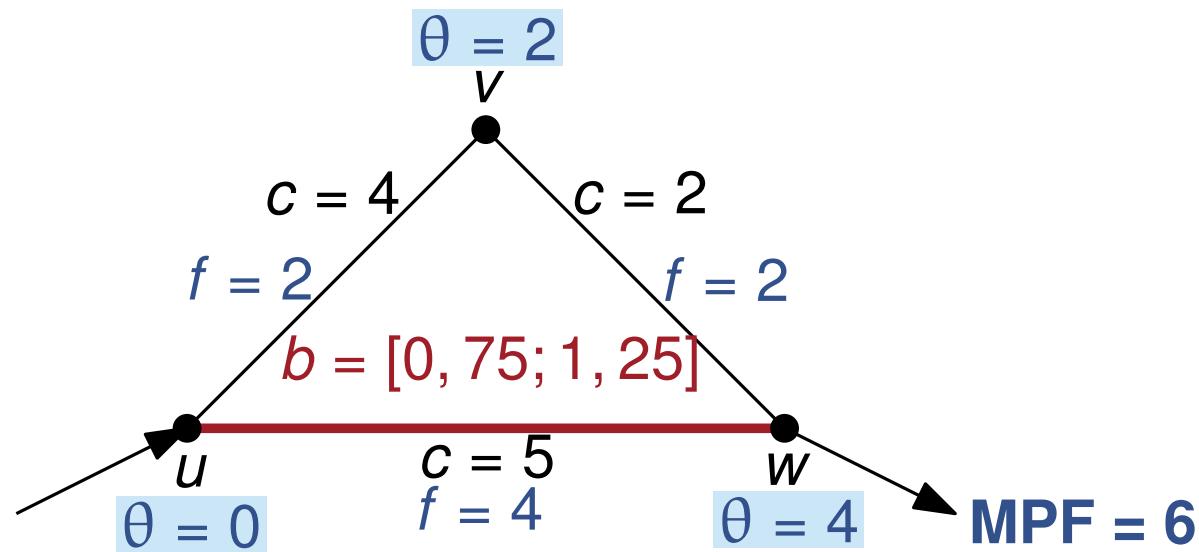
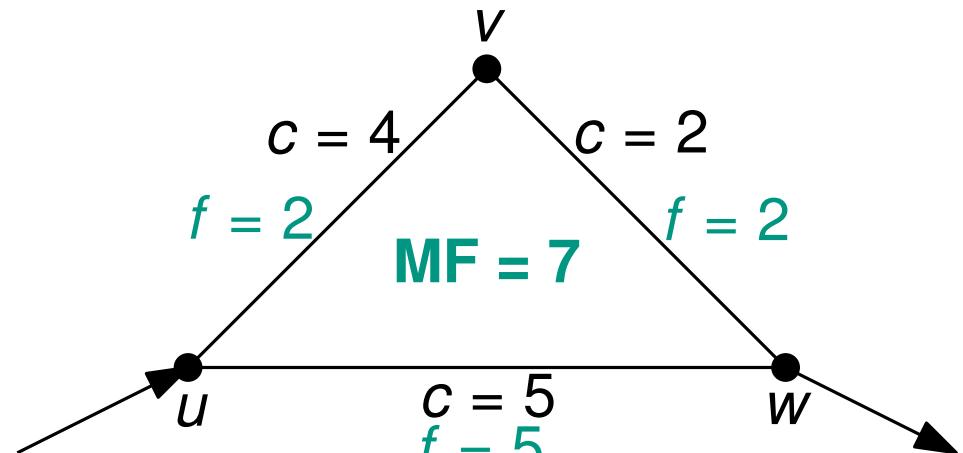
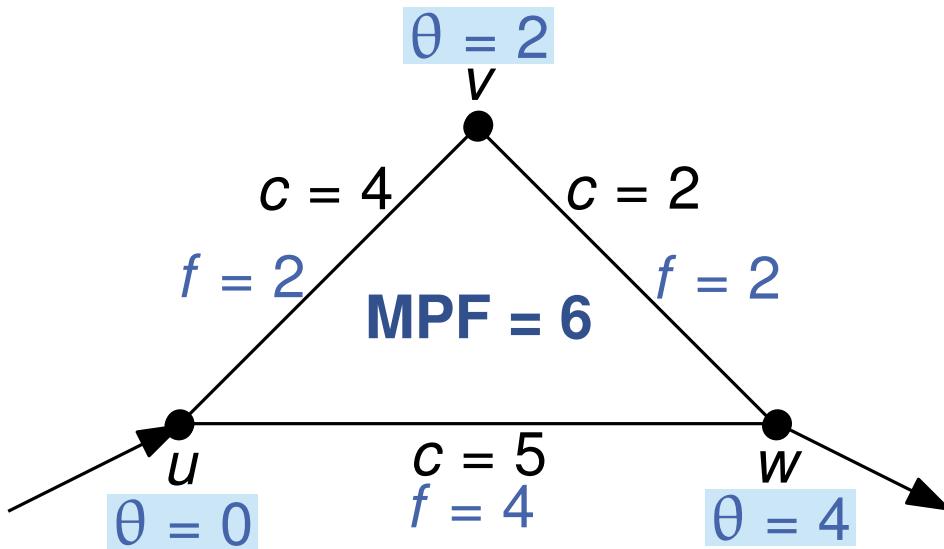
# Maximum FACTS Flow



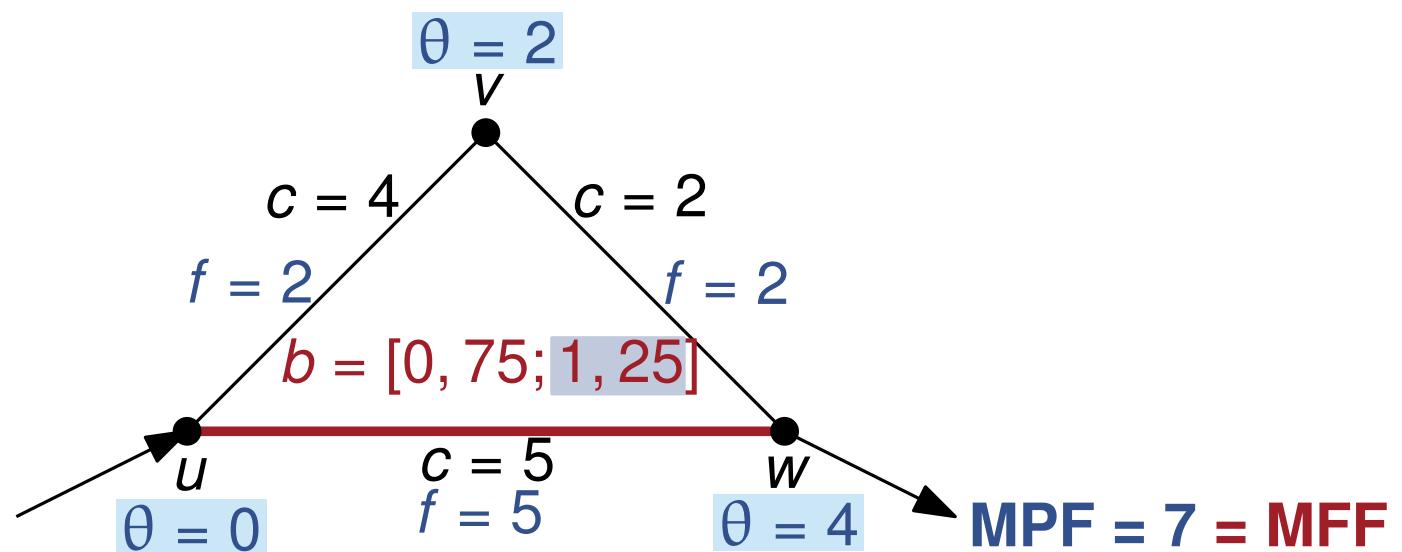
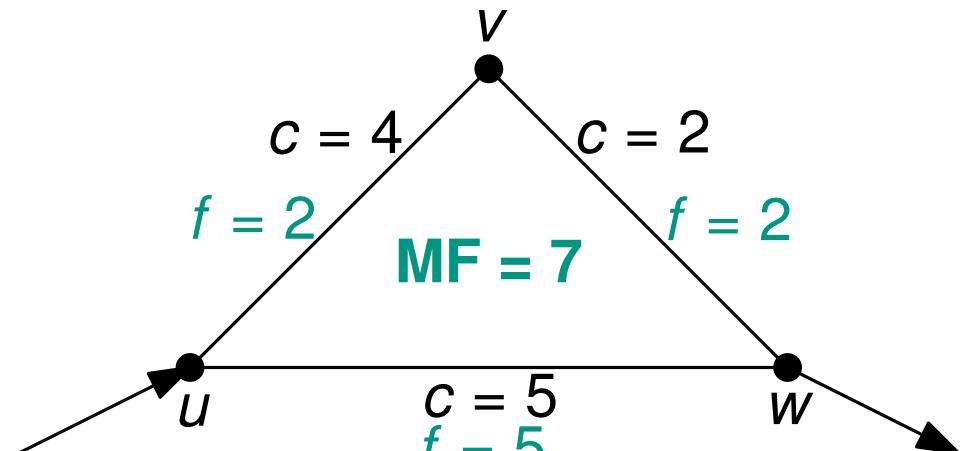
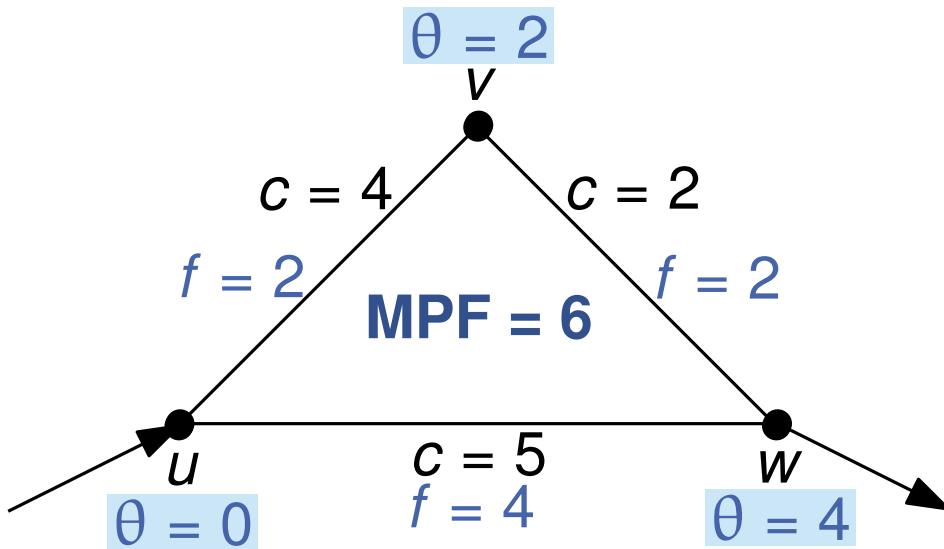
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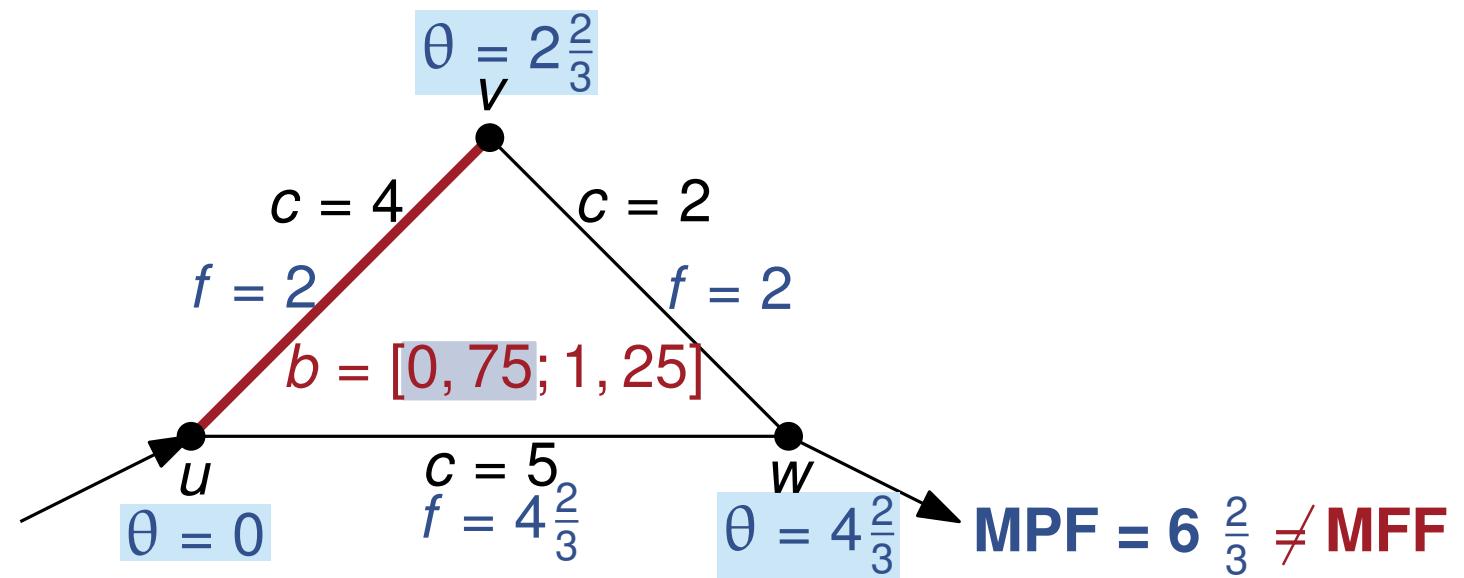
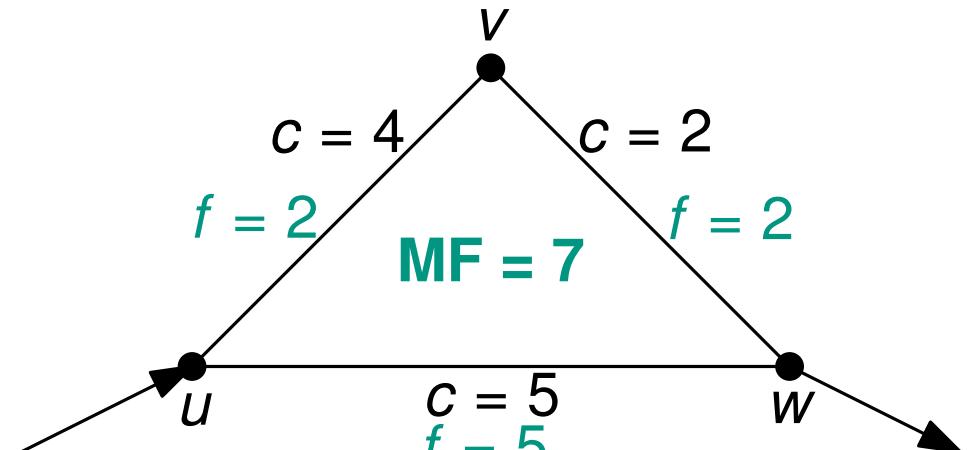
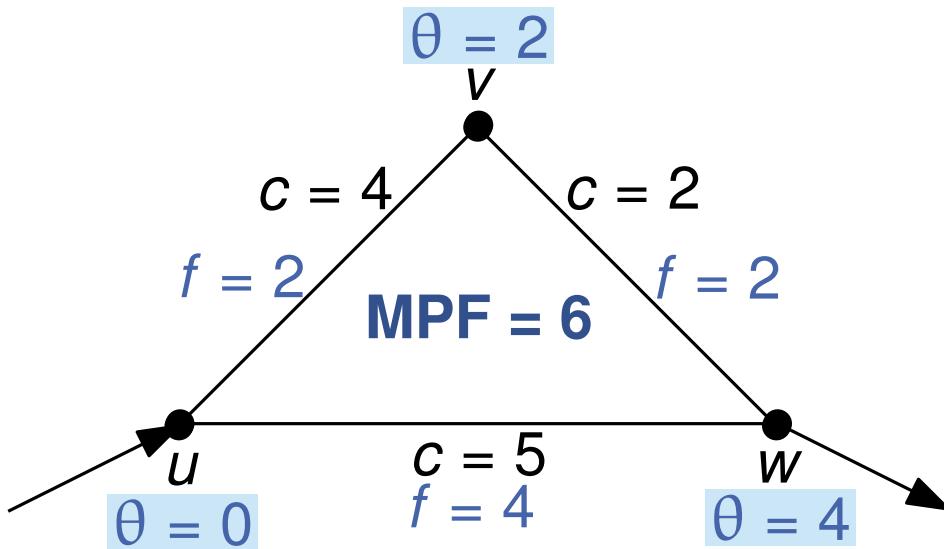
# Maximum FACTS Flow



# Maximum FACTS Flow



# Maximum FACTS Flow



# Problem Statement: Maximum FACTS Flow

- maximum FACTS flow with  $k$  FACTS

$$\mathbf{MFF}(k) = \max_{E' \subseteq E, b} \mathbf{MPF}(G) \quad |E'| = k$$

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$$\mathbf{MFF}(G) = \max_k \mathbf{MFF}(k)$$

$$\forall (u, v) \in E: f(u, v) = b(u, v)(\theta(v) - \theta(u))$$

variables

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variables

$$b(u, v)\theta(v) - b(u, v)\theta(u)$$

$$B(b(u, v), \theta(v)) - B(b(u, v), \theta(u))$$

||

$$B(b(u, v), \theta(v) - \theta(u))$$

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$$\mathbf{MFF}(G) = \max_k \mathbf{MFF}(k)$$

- minimum number of FACTS to reach the **MFF**

$$\mathbf{MNF}_{\mathbf{MFF}}(G) = \min_{\mathbf{MFF}(k)=\mathbf{MFF}(G)} k$$

# Relationship Optimal iFACTS Flow (OiFF)

**physical model**  
(AC linearization)

**flow model**

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## How to approach the graph theoretical flow?

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How to approach the graph theoretical flow?

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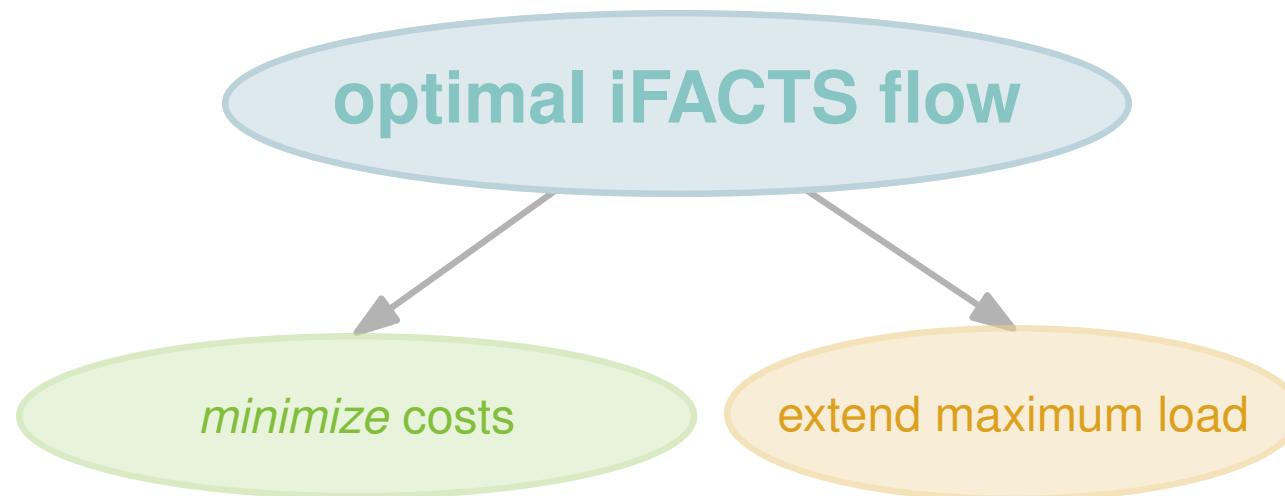
*minimize costs*

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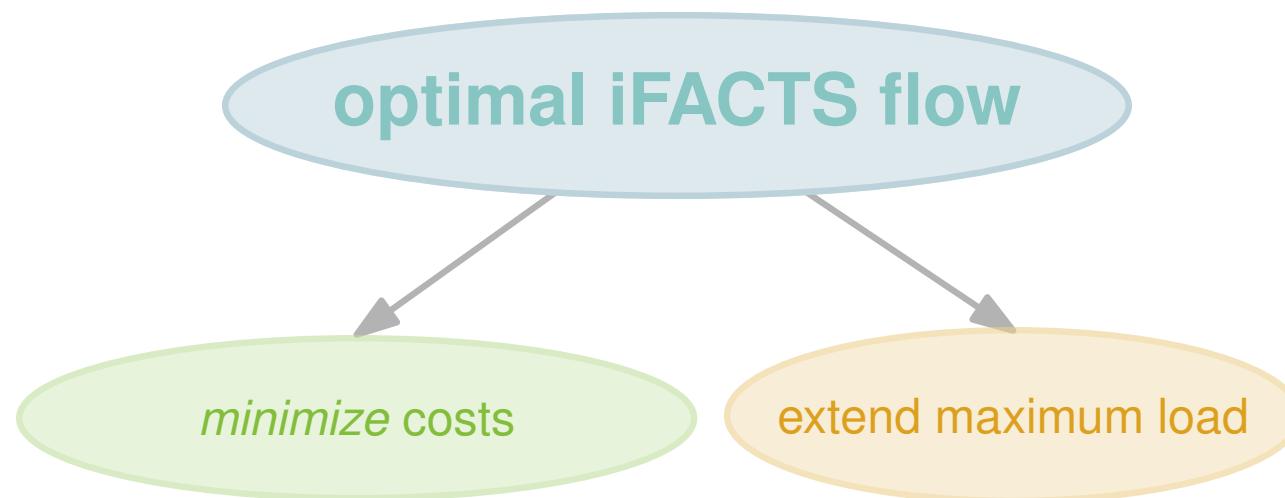


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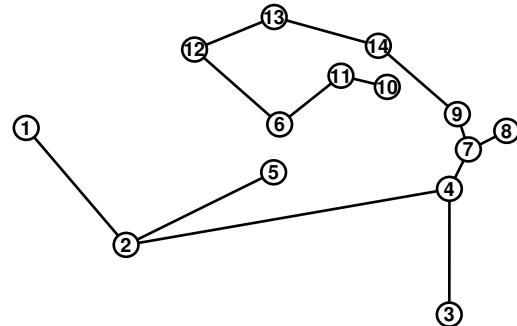
number of FACTS between  
**FFS** and **FCS**

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**optimal iFACTS flow**

*minimize costs*

*extend maximum load*

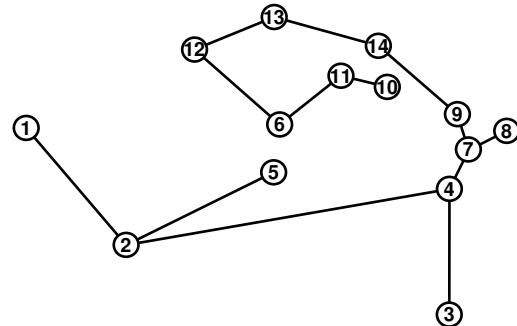
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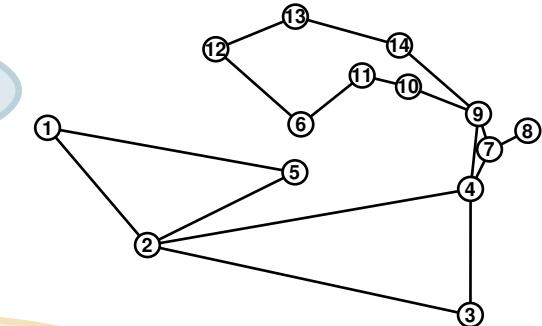
**flow model**

## How to approach the graph theoretical flow?



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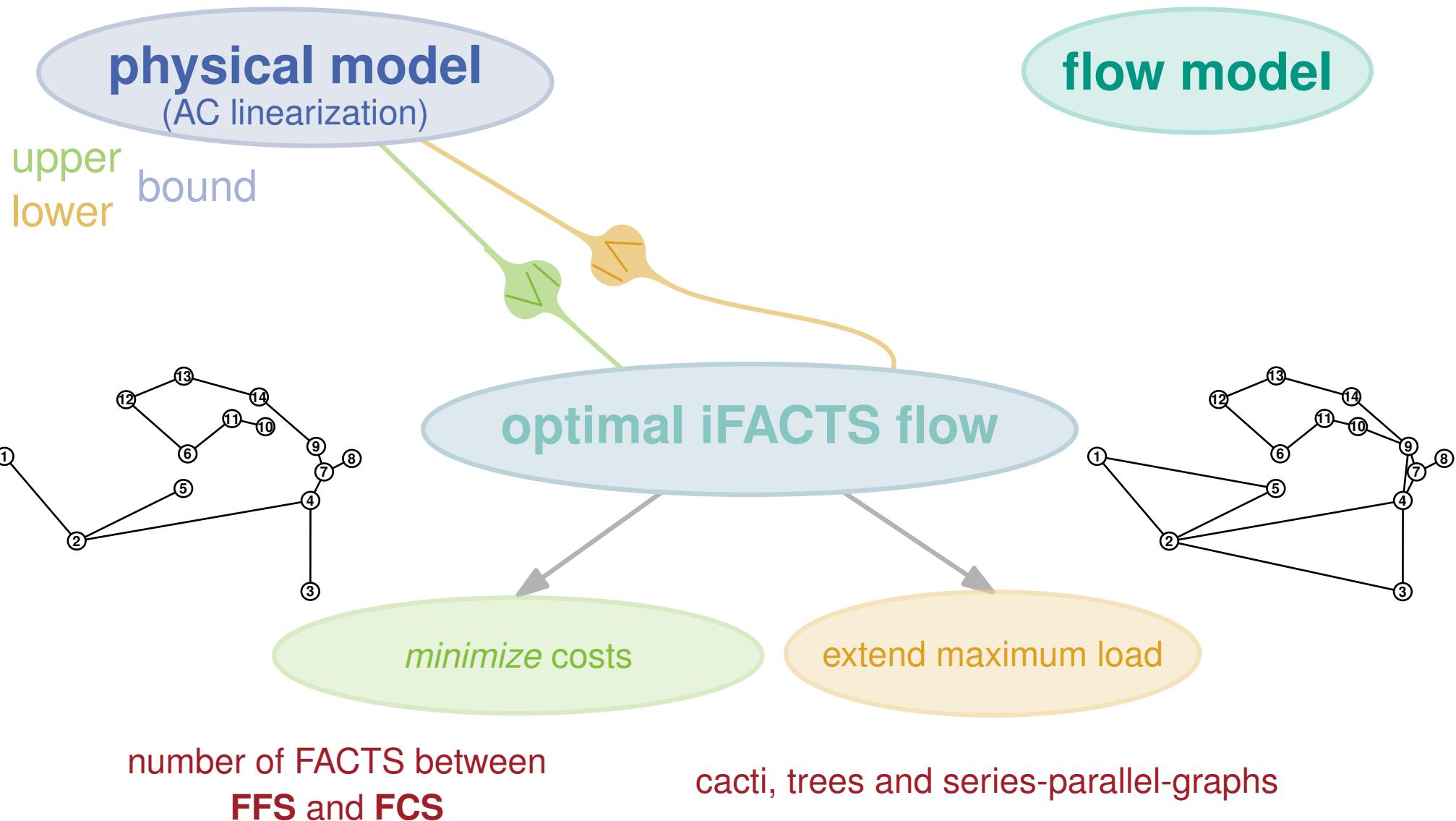


*extend maximum load*

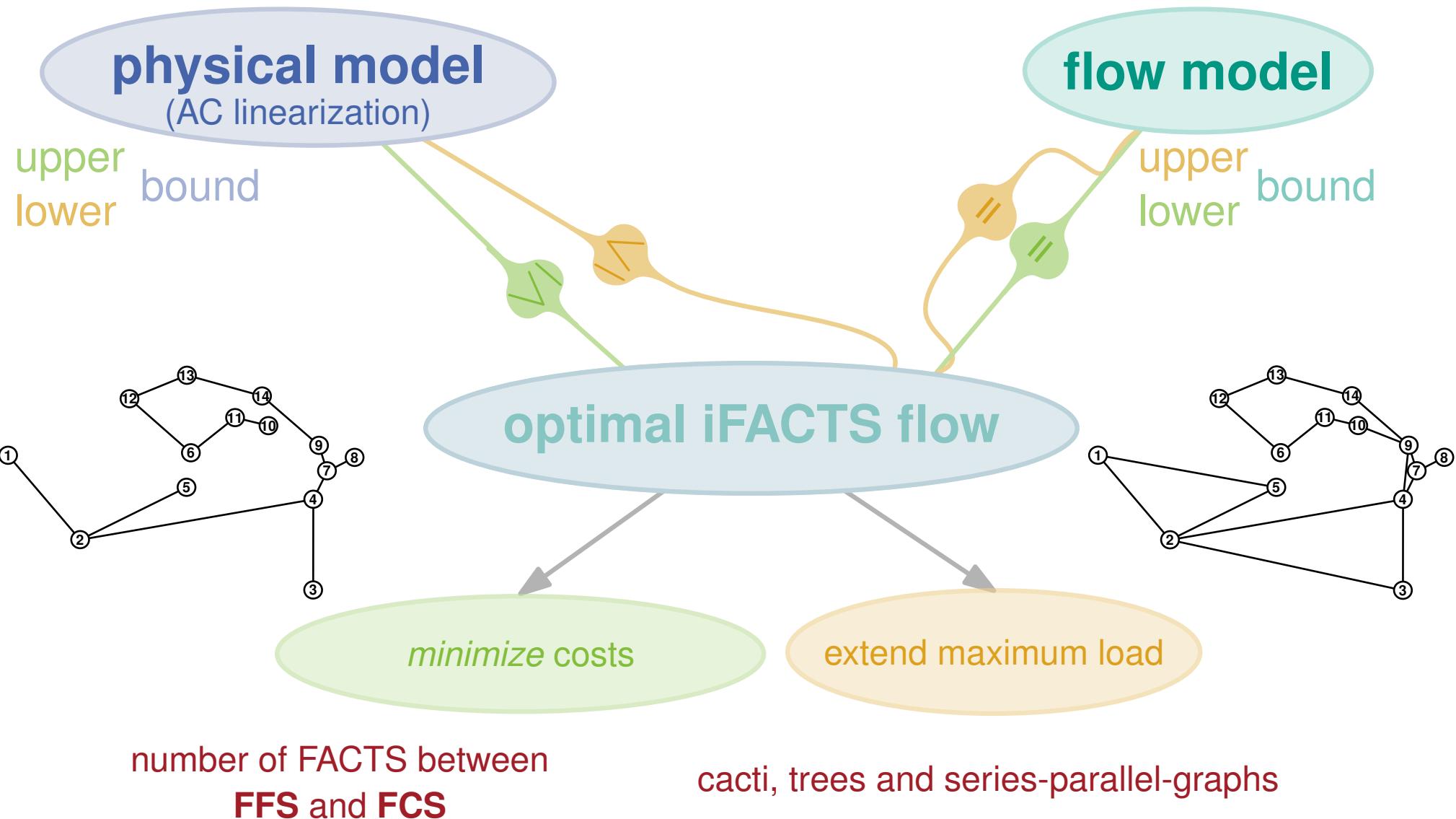
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cacti, trees and series-parallel-graphs

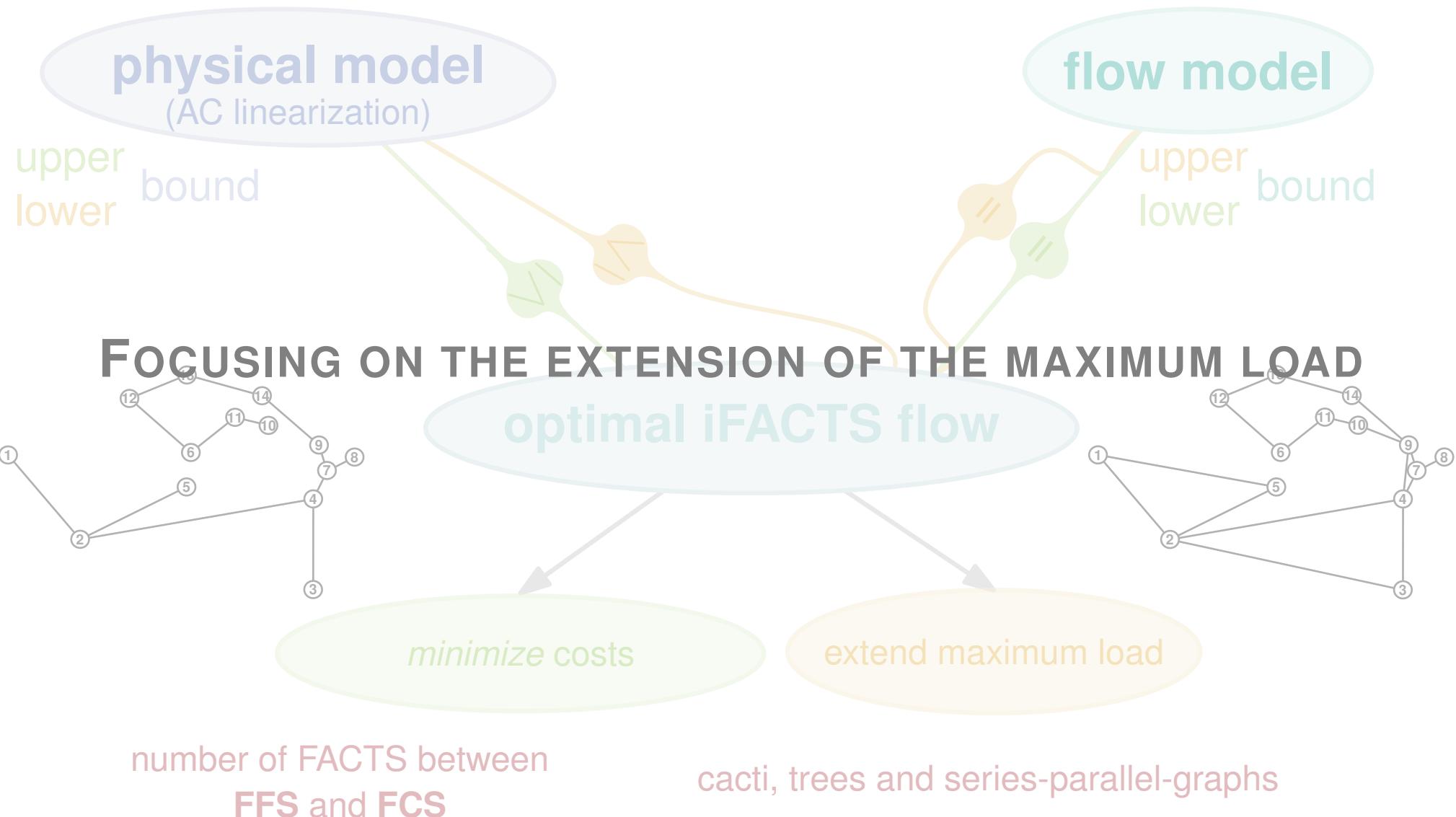
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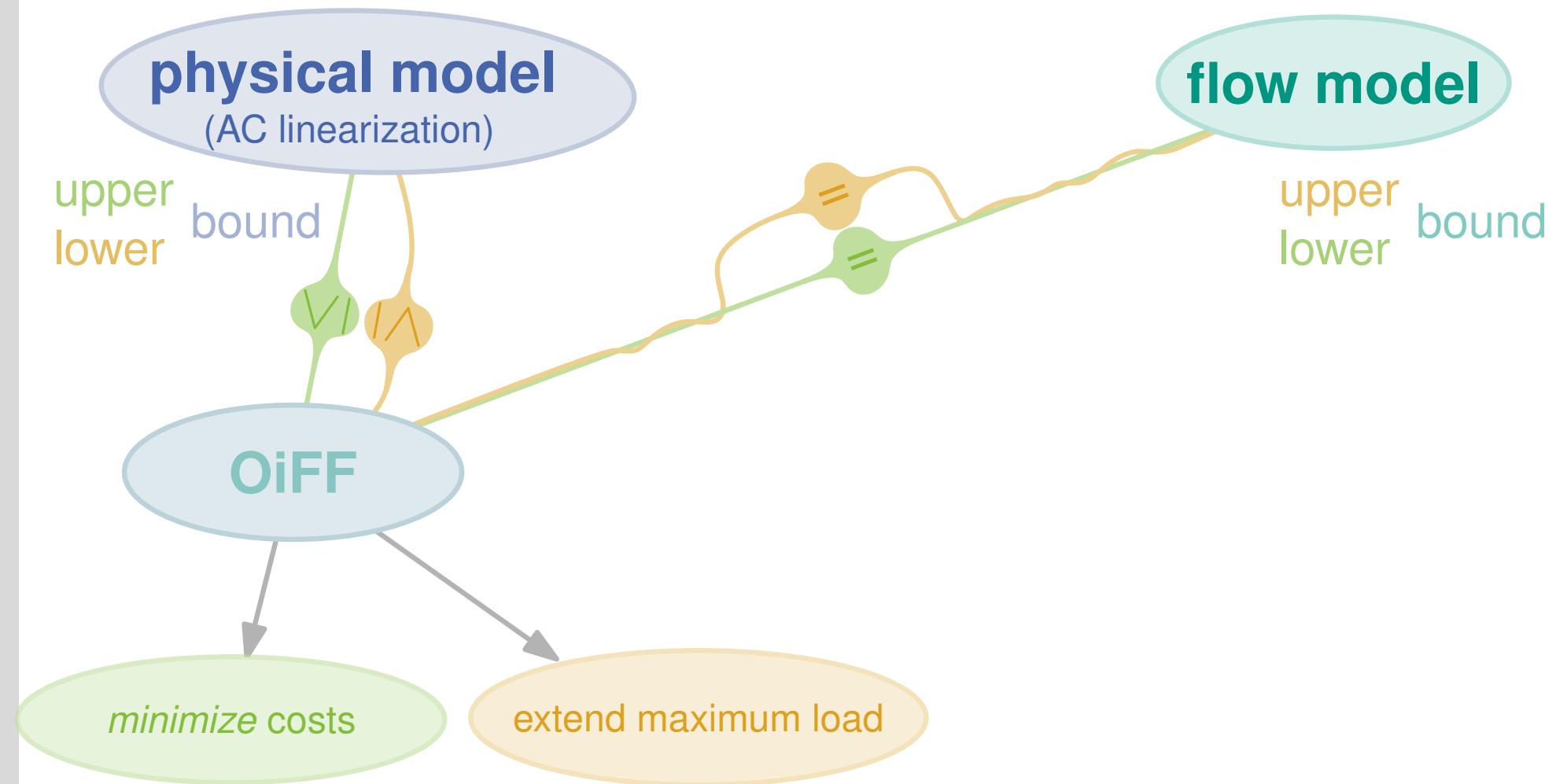
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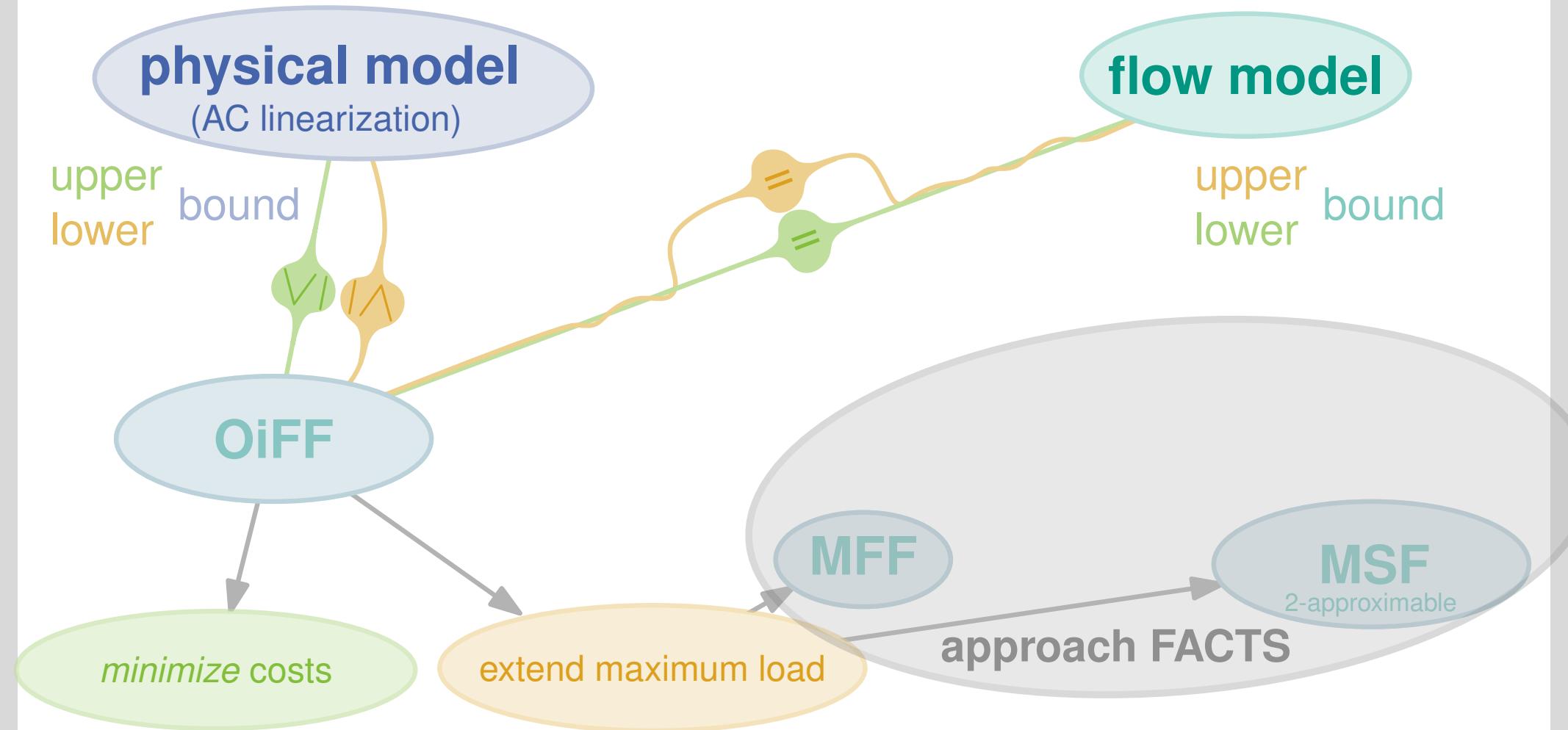


# Conclusion



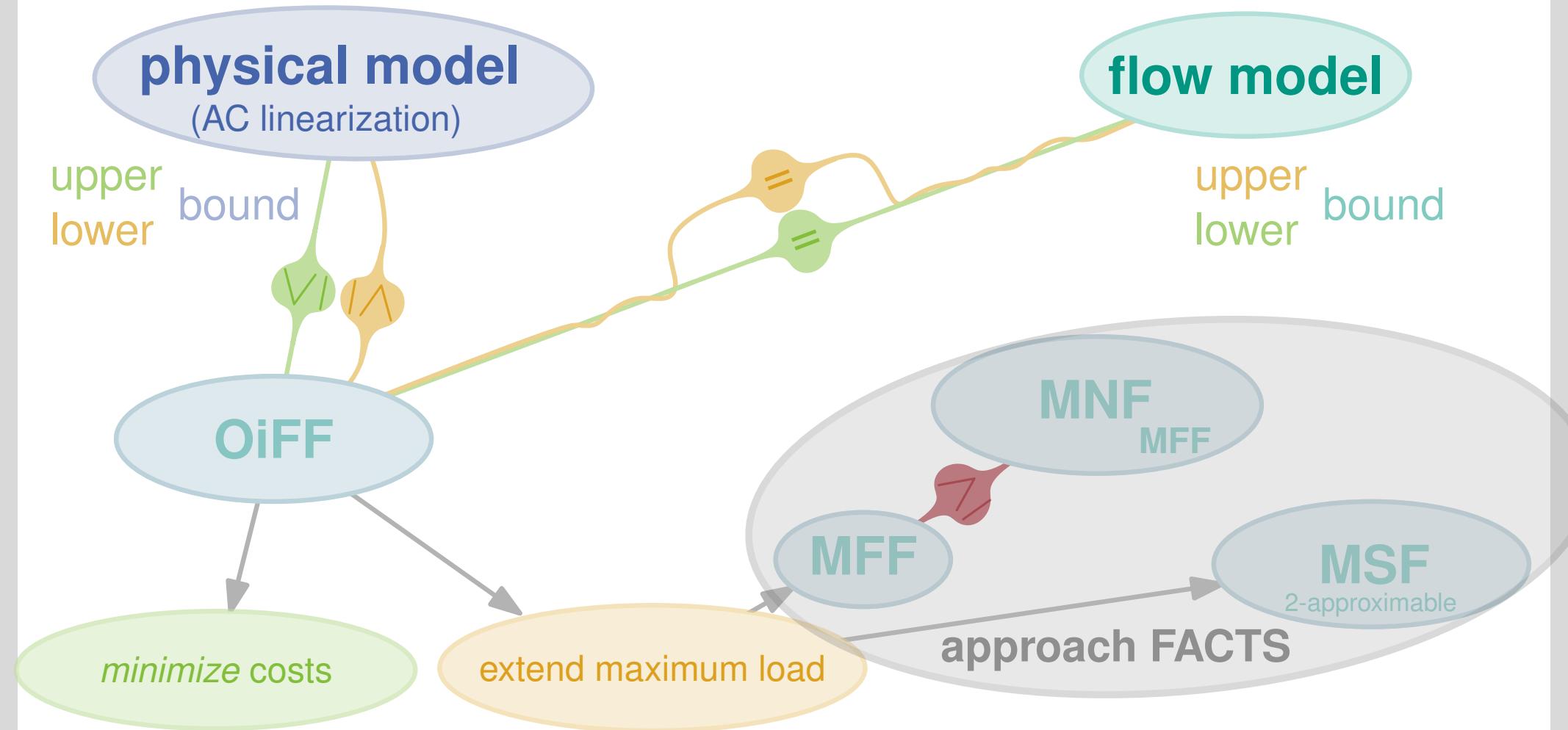
- number of FACTS between **FFS** and **FCS**
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# Conclusion



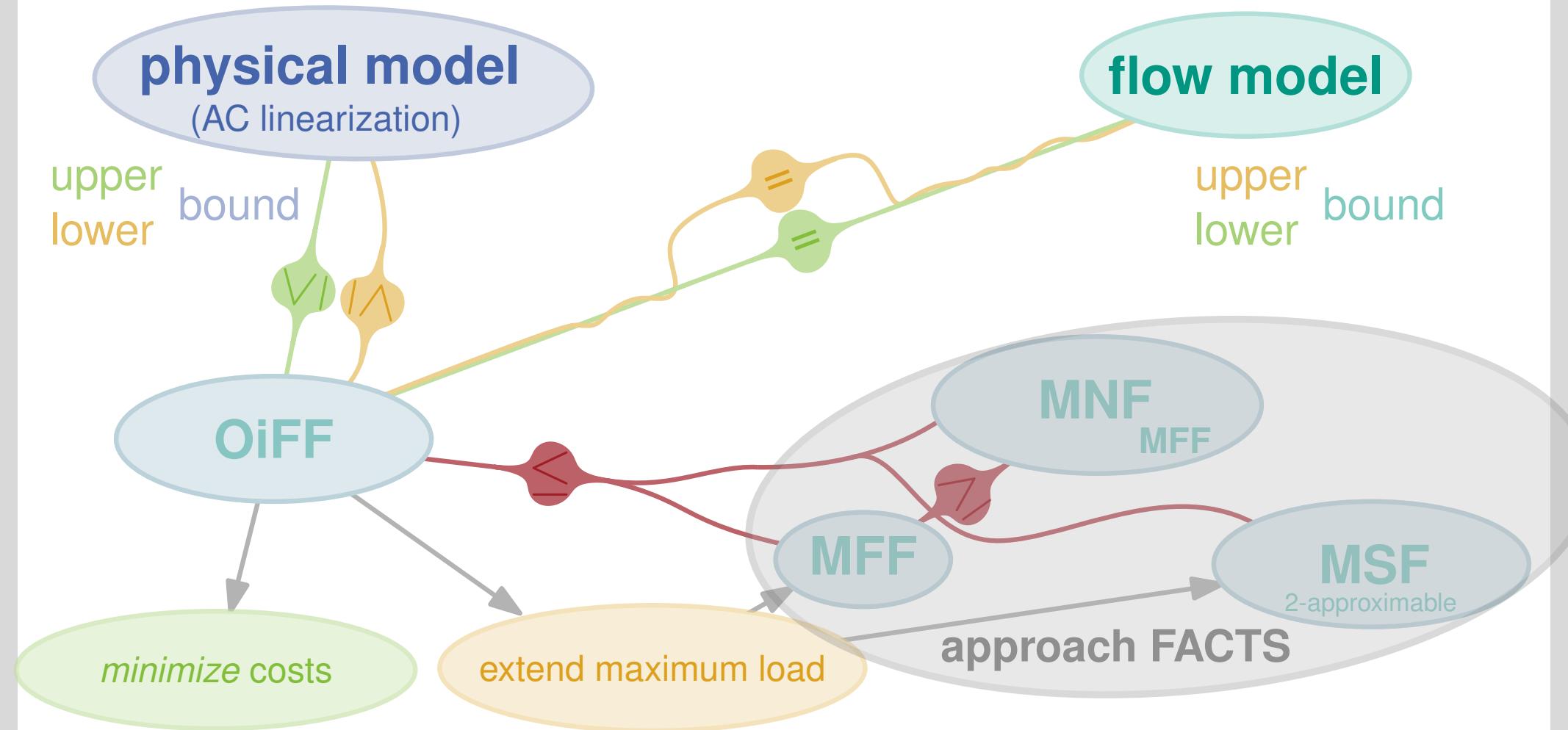
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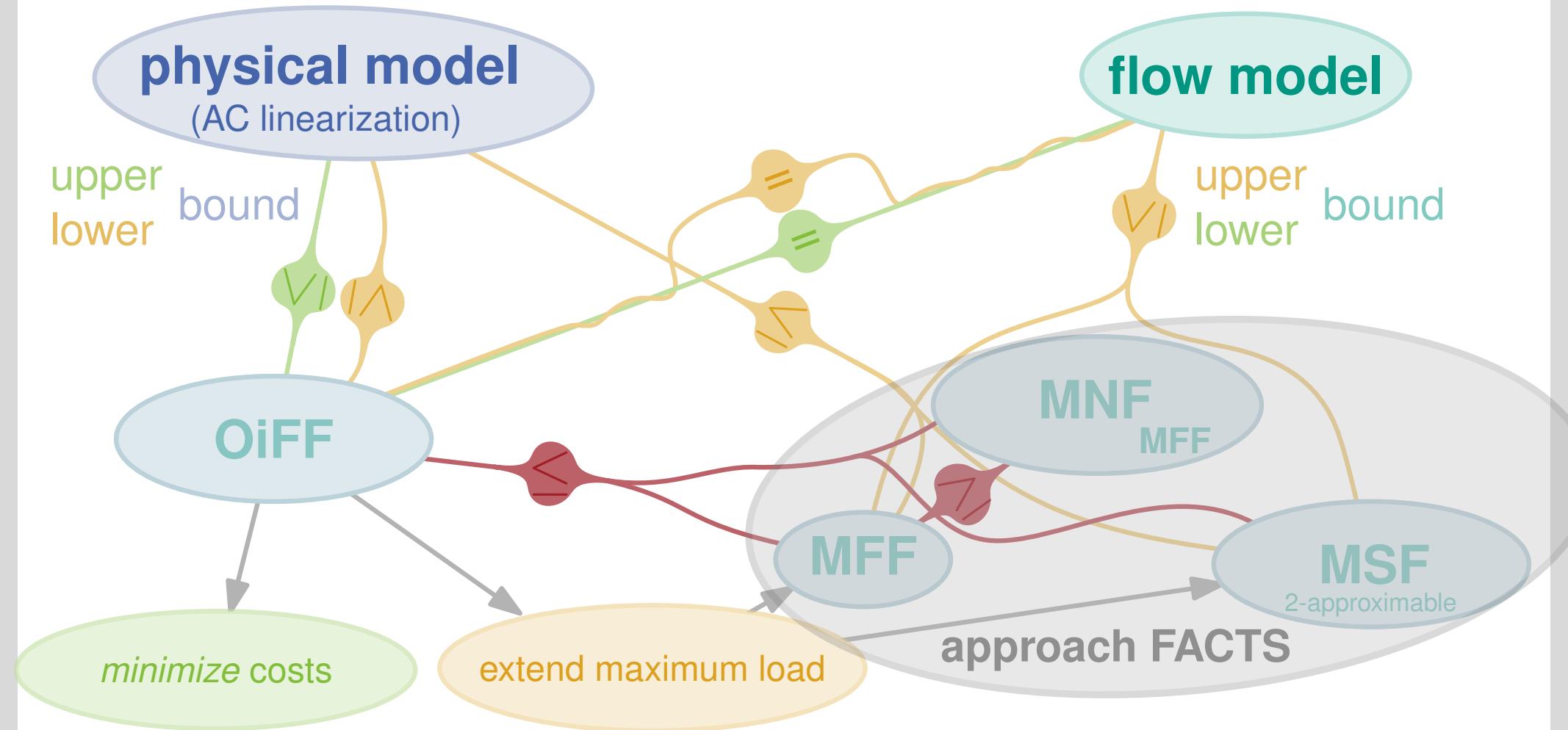
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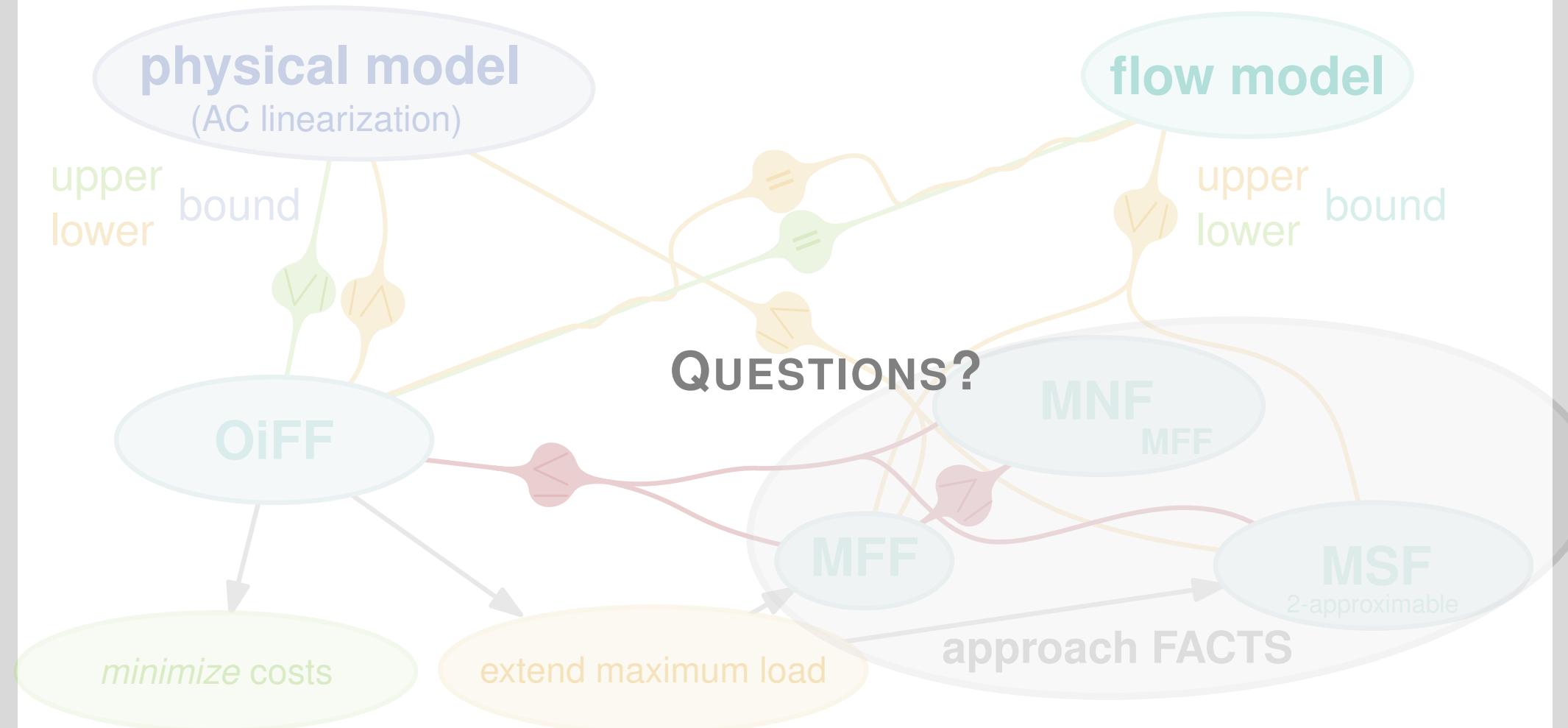
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