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Introducing Plasmo.jl

A Package for Graph-Based Modeling using JuMP

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Plasmo.jl - What is it?

Platform for **S**calable **M**odeling and **O**ptimization

A Graph-based modeling and optimization framework

Key Features:

- Component models associated with nodes **and** edges
- Facilitates construction of hierarchical graphs (uses subgraphs)
- Modularization of component models
- Manipulate graph structure for solver interface
- Ease of modeling complex systems

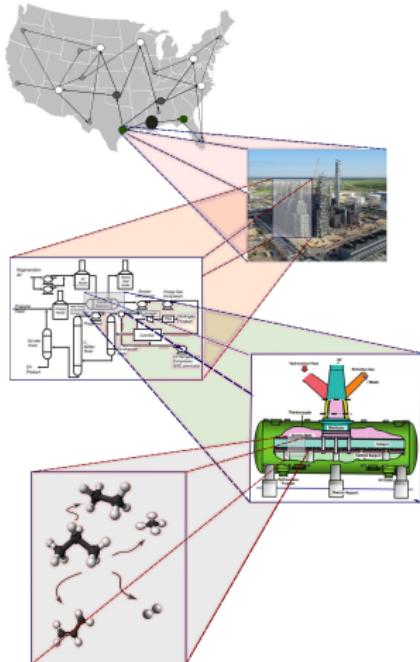


Overview

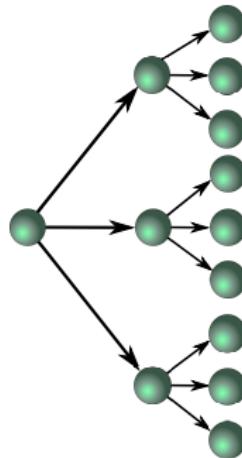
- Motivation - Complex systems
- Modeling Systems with Components (Graphs)
- Applications
- Design considerations
- Goals right now



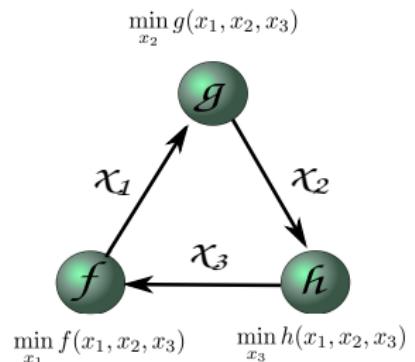
Group Research Theme:Complex Systems



Multi-scale systems



Multi-stage stochastic programs

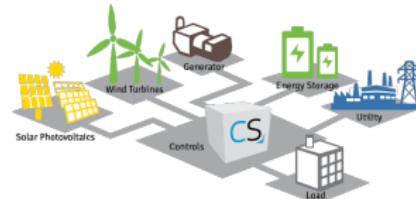


Asynchronous systems

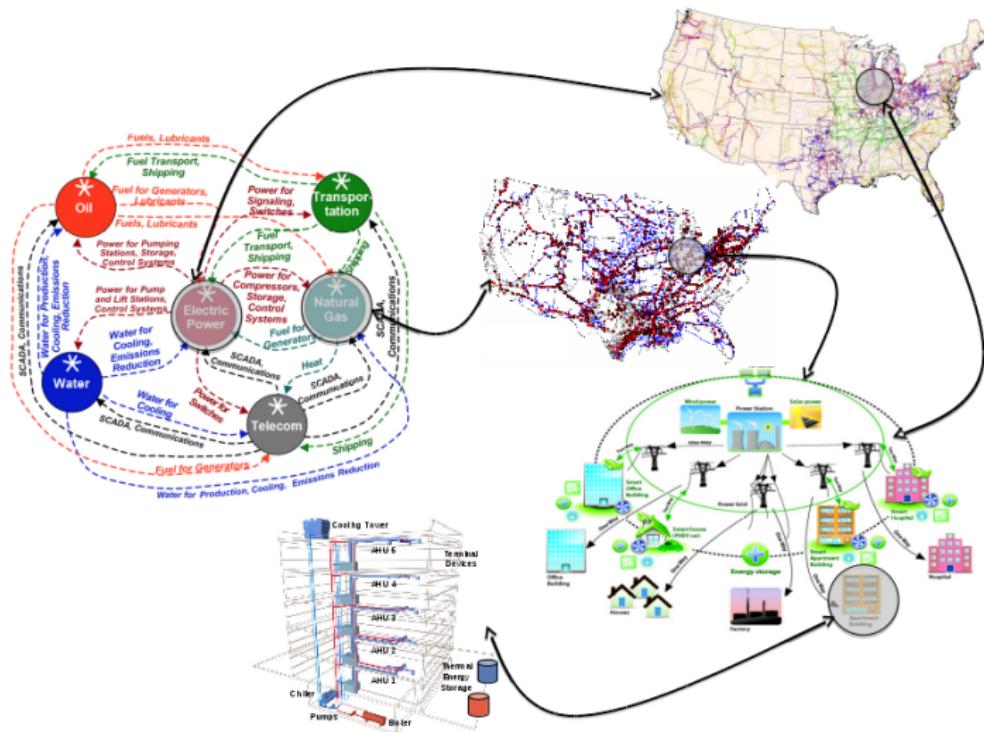


Types of Problems

- Nonlinear (nonconvex) optimization
- Stochastic programming
- Model predictive control
- Some Applications
 - ▶ Energy storage systems
 - ▶ Connected infrastructure
 - ▶ Microgrids

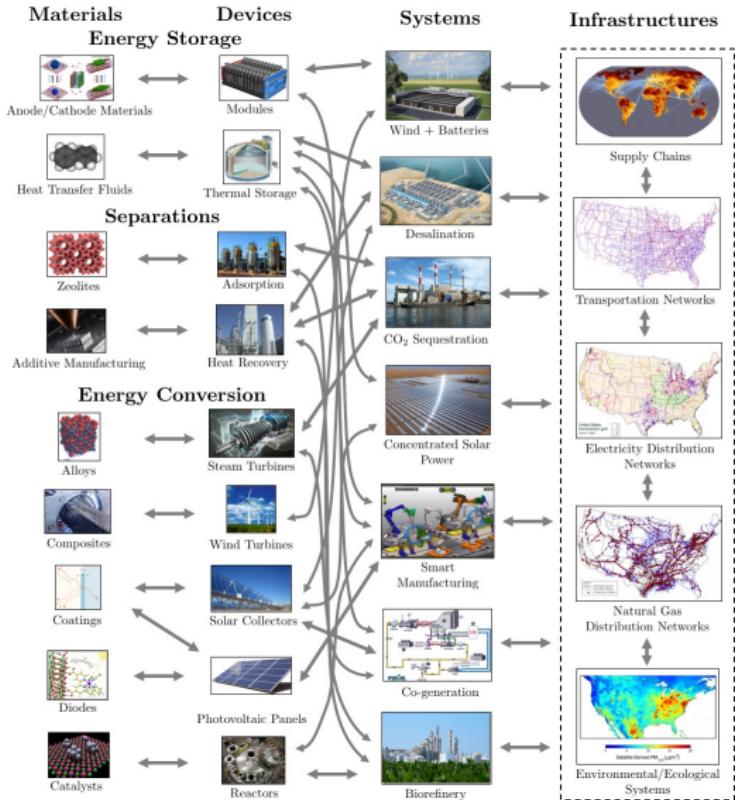


Hierarchical Networks



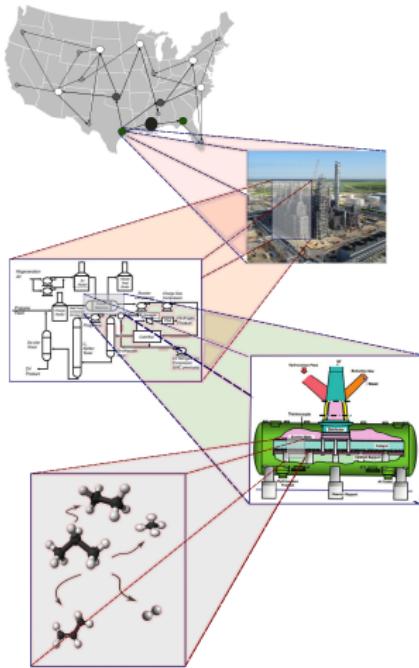


Technology Landscapes





Challenges with complex physical systems



- Millions of constraints and variables can make the computation intractable
 - ▶ Generally apply ad-hoc methods to perform some model reduction
- Millions of system connections makes model instantiation non-trivial
 - ▶ Multiple scenarios
 - ▶ Solution inspection
- Modeling asynchronicity in large communicating systems is non trivial
 - ▶ Decentralized control



Some existing modeling frameworks

- Modelica
 - ▶ Components, hierarchies, architectures (highly abstracted)
 - ▶ Designed for simulation (Optimica extension does some optimization)
 - ▶ Write connectors for coupling (I always found this difficult)



- gProms

- ▶ Equation oriented chemical flowsheeting software
- ▶ Custom modeling language
- ▶ Commercial



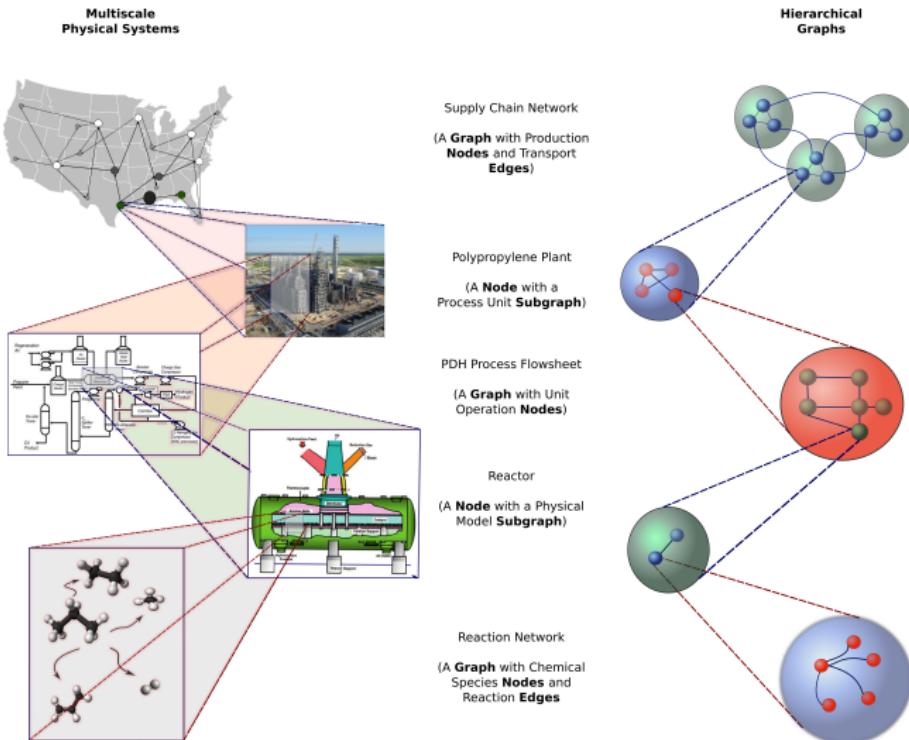


Revisiting our Goals

- Model encapsulation
- Modularity and reuse
- Navigate solutions to complex optimization problems
- Facilitate modeling of communicating systems



The Power of Abstraction - Graphs

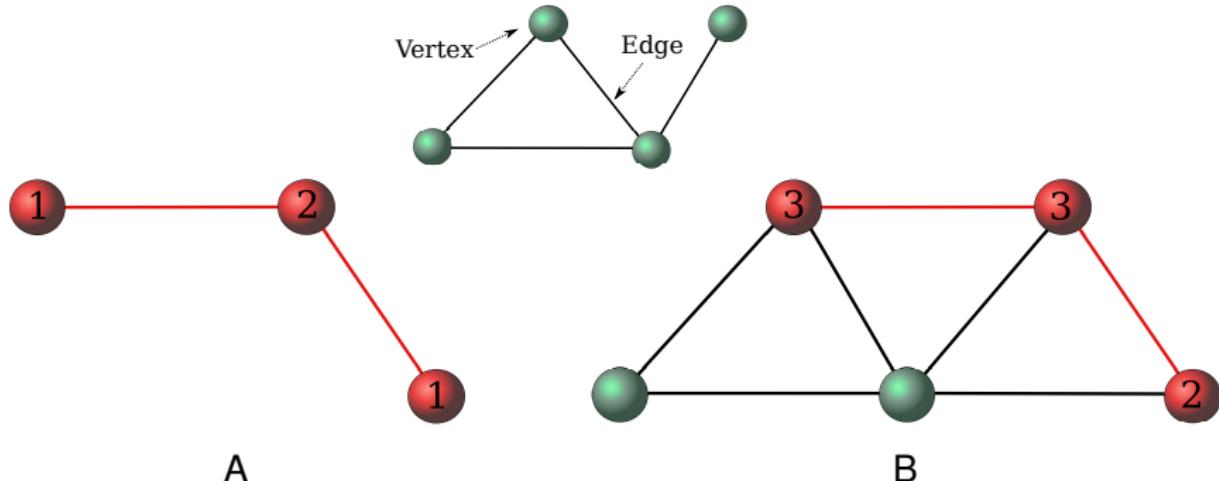




Relevant Graph Concepts

Graph Definition

A graph (G) is a finite set $V(G)$ of vertices (nodes) and a finite family $E(G)$ of pairs of elements of $V(G)$ called edges



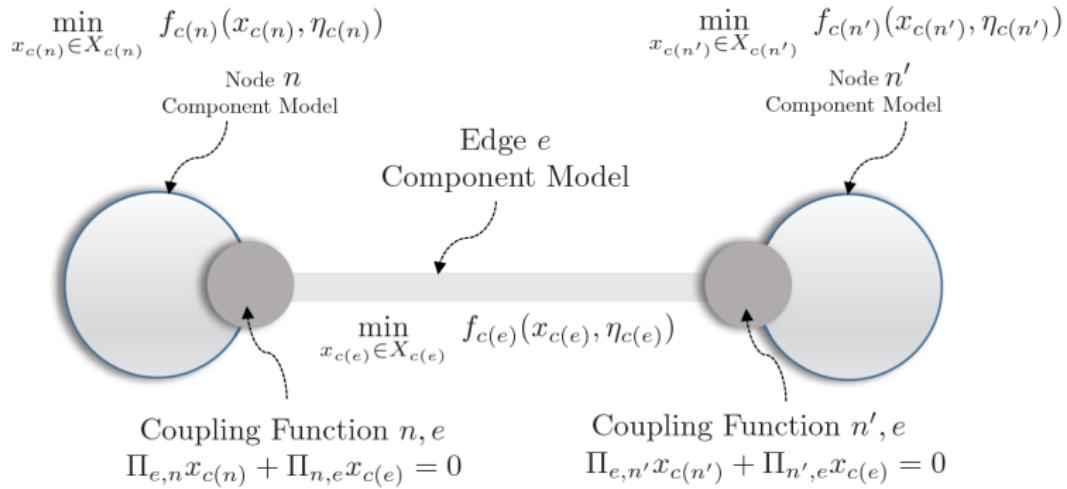
- A is a subgraph of B

- The degree of a node is specific to its graph



Graph Based Modeling

Plasmo associates model components with nodes and edges





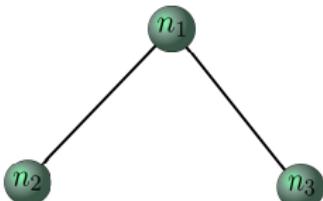
Plasmo.jl - Key Features

Key Features (some in progress):

- Associates models (JuMP Models) and linking information (constraints) with nodes and edges within a graph
- Exploits a subgraph abstraction to enable hierarchies of models (multiple graphs defined on a set of nodes)
- Uses [LightGraphs.jl](#) as the graph backend
- Accesses model information on nodes and edges
- Provides interfaces with structured solvers (PIPS, etc...)



Plasmo - Old syntax (still supported)



```
using Plasmo
graph = PlasmaGraph()      #Create a graph
n1 = add_node!(graph)
n2 = add_node!(graph)
n3 = add_node!(graph)
edge1 = add_edge!(graph, n1, n2)
edge2 = add_edge!(graph, n1, n3)
#Set component models
setmodel!(n1, simple_model())
setmodel!(n2, simple_model())
setmodel!(n3, simple_model())
#provide linking information
setcouplingfunction!(graph, edge1, couplenodes)
setcouplingfunction!(graph, edge2, couplenodes)
model = generate_model(graph)
setsolver(model, IpoptSolver())
solve(model)

function couplenodes(m::Model, graph, edge)
    @constraint(m, getconnectedfrom(graph, edge)[:, x]
    == getconnectedto(graph, edge[:, x]))
end
```

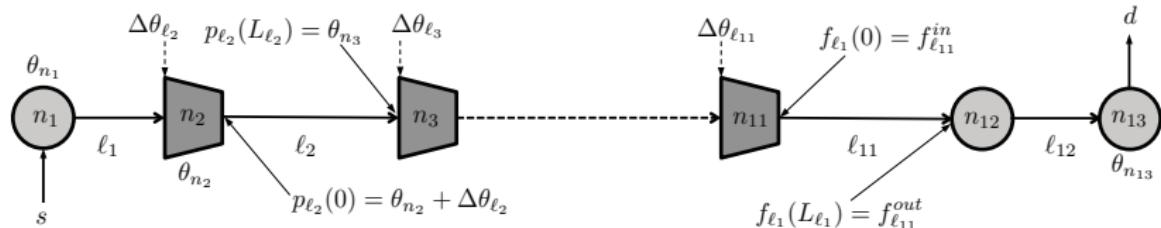


New Syntax - In Progress

```
using Plasmo
#Create a graph
m = GraphModel(solver = IpoptSolver())
n1 = add_node!(m)
n2 = add_node!(m)
n3 = add_node!(m)
edge1 = add_edge!(m, n1, n2)
edge2 = add_edge!(m, n1, n3)
#Set component models
setmodel!(n1, simple_model())
setmodel!(n2, simple_model())
setmodel!(n3, simple_model())
#link the two models
@linkconstraint(edge1, getconnectedfrom(m, edge1)[:, x]
== getconnectedto(edge1)[:, x])
@linkconstraint(edge2, getconnectedfrom(m, edge2)[:, x]
== getconnectedto(edge2)[:, x])
solve(m) #solve with Ipopt
#solve_pips(m, n1, [n2, n3]) #solve with PIPS NLP
```



Gas Networks



- \mathcal{N} : Set of nodes in the gas network (junctions)
- \mathcal{L} : Set of links (pipelines)
- $\mathcal{S} \subseteq \mathcal{N}$: Set of gas supplies
- $\mathcal{D} \subseteq \mathcal{N}$: Set of gas demands
- $\mathcal{L}_a \subseteq \mathcal{L}$: Set of active links (pipelines with compressors)
- $\mathcal{L}_p \subseteq \mathcal{L}$: Set of passive links (pipelines without compressors)



Gas Networks

Mass and Momentum Balances on a Network

$$\frac{\partial p_\ell(t, x)}{\partial t} + \frac{c^2}{A_\ell} \frac{\partial f_\ell(t, x)}{\partial x} = 0, \quad \ell \in \mathcal{L}$$

$$\frac{\partial f_\ell(t, x)}{\partial t} + \frac{2c^2 f_\ell(t, x)}{A_\ell p_\ell(t, x)} \frac{\partial f_\ell(t, x)}{\partial x} - \frac{c^2 f_\ell(t, x)^2}{A_\ell p_\ell(t, x)^2} \frac{\partial p_\ell(t, x)}{\partial x} + A_\ell \frac{\partial p_\ell(t, x)}{\partial x} = -\frac{8c^2 \lambda A_\ell}{\pi^2 D_\ell^5} \frac{f_\ell(t, x)}{p_\ell(t, x)} |f_\ell(t, x)|, \quad \ell \in \mathcal{L}$$

Compressor Power

$$P_\ell(t) = c_p \cdot T \cdot f_{in,\ell} \left(\left(\frac{p_{in,\ell}(t) + \Delta p_\ell(t)}{p_{in,\ell}(t)} \right)^{\frac{\gamma-1}{\gamma}} - 1 \right), \quad \ell \in \mathcal{L}_a$$

Node Conservation

$$\sum_{\ell \in \mathcal{L}_n^{rec}} f_{out,\ell}(t) - \sum_{\ell \in \mathcal{L}_n^{snd}} f_{in,\ell}(t) + \sum_{i \in \mathcal{S}_n} g_i(t) - \sum_{j \in \mathcal{D}_n} d_j^{gas}(t) = 0, \quad n \in \mathcal{N}$$

Supply and Demand

$$f_{deliver,n}(t) \leq f_{demand,n}(t), \quad n \in \mathcal{D}$$

Boundary Conditions

$$p_\ell(0, t) = p_{in,\ell}(t) + \Delta p_\ell(t), \quad \ell \in \mathcal{L}_a$$

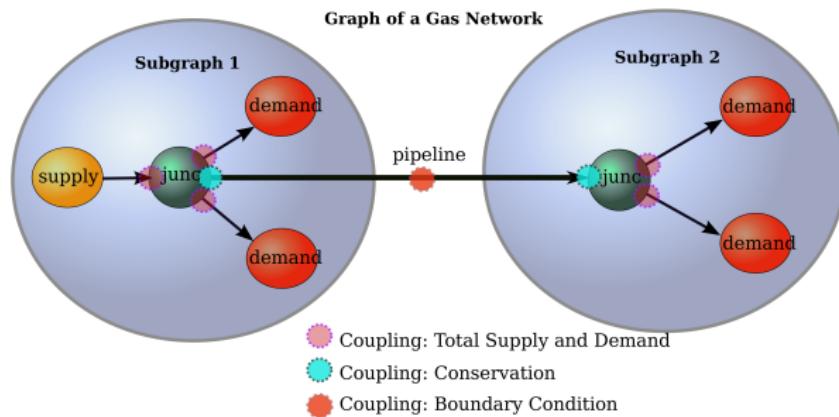
$$p_\ell(0, t) = p_{in,\ell}(t), \quad \ell \in \mathcal{L}_p$$

$$p_\ell(L_\ell, t) = p_{out,\ell}(t), \quad \ell \in \mathcal{L}$$



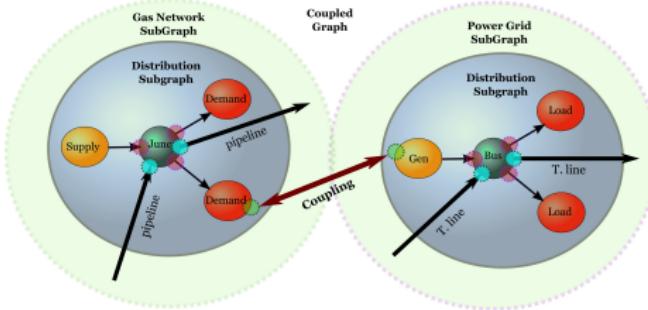
Graph Based Modeling - Gas Networks

- The subgraph abstraction allows multiple couplings on the same node
- This can be used to build modular systems and couple them at higher levels





Graph Based Modeling - Coupled Networks



```
m = GraphModel()
graph = getgraph(m)
add_subgraph!(graph, power_network)
add_subgraph!(graph, gas_network)
generator = getnode(power_network, :gen)
demand = getnode(gas_network, :demand)
link = add_edge!(graph, generator, demand)
@linkconstraint(link, getconnectedfrom(graph, link)[:Pgnd] <=
getconnectedto(graph, link)[:fdeliver])
```



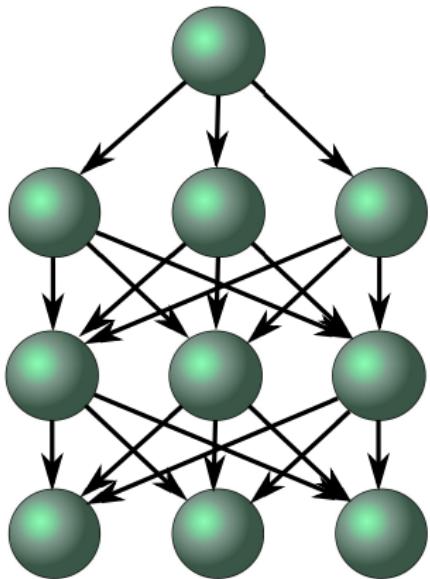
Graph Based Modeling - Coupled Networks

Key Findings:

- Infrastructure models (graphs) can be developed independently and coupled within larger systems (graphs)
- Illinois Case Study: 7% more gas delivered to generators; 27% revenue increase versus uncoordinated case
- Uncoordinated case simulated by solving successive optimization problems



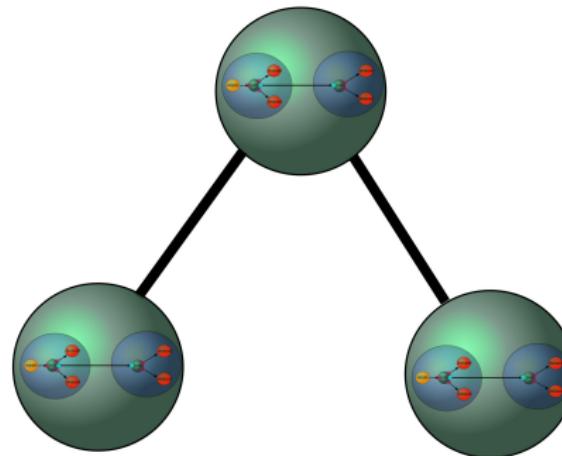
Multistage Stochastic Programming



- Our graph abstraction corresponds to the node-based abstraction in multistage stochastic programming
- Component models associated within nodes (scenarios)
- Link constraints propagate transition from stage to stage



Embedding Graph Models

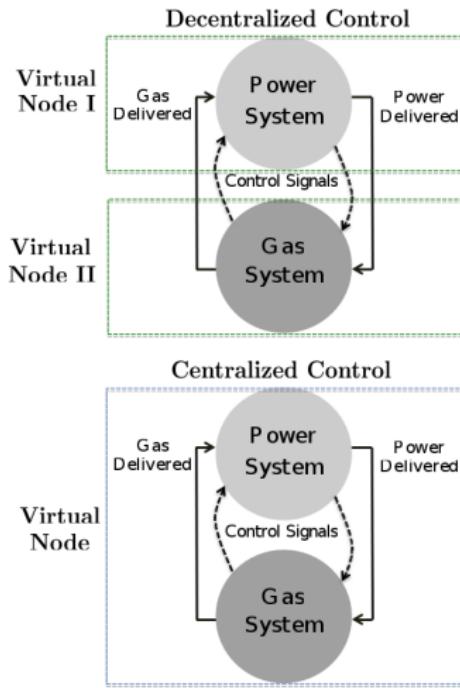


- Graph models can themselves be embedded as models in nodes or edges
- Simplifies construction of multiple layers in systems



Future Direction

- Generalize the model interface if possible (strictly uses JuMP)
- Find suitable abstraction for computational workflows
 - ▶ Decentralized control
 - ▶ Algorithmic strategies (e.g. scheduling and operations)
 - ▶ Graph partitioning and model reduction
 - ▶ Initialization strategies
- Simulation interfaces





Goals right now

- Finalize physical model abstraction
- Push first version to github
- Figure out a suitable graph communication abstraction