

Dynamic Resource Exchange with CoinOR-CBC in Cyclus, a Nuclear Fuel Cycle Simulator

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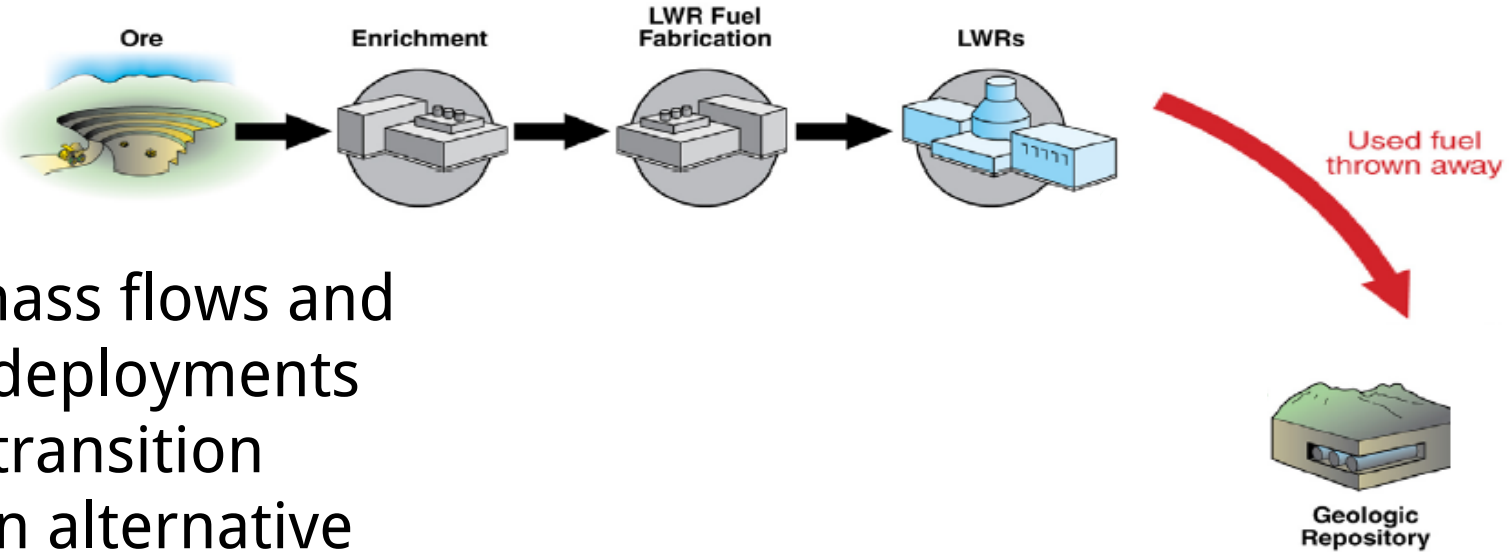
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Overview



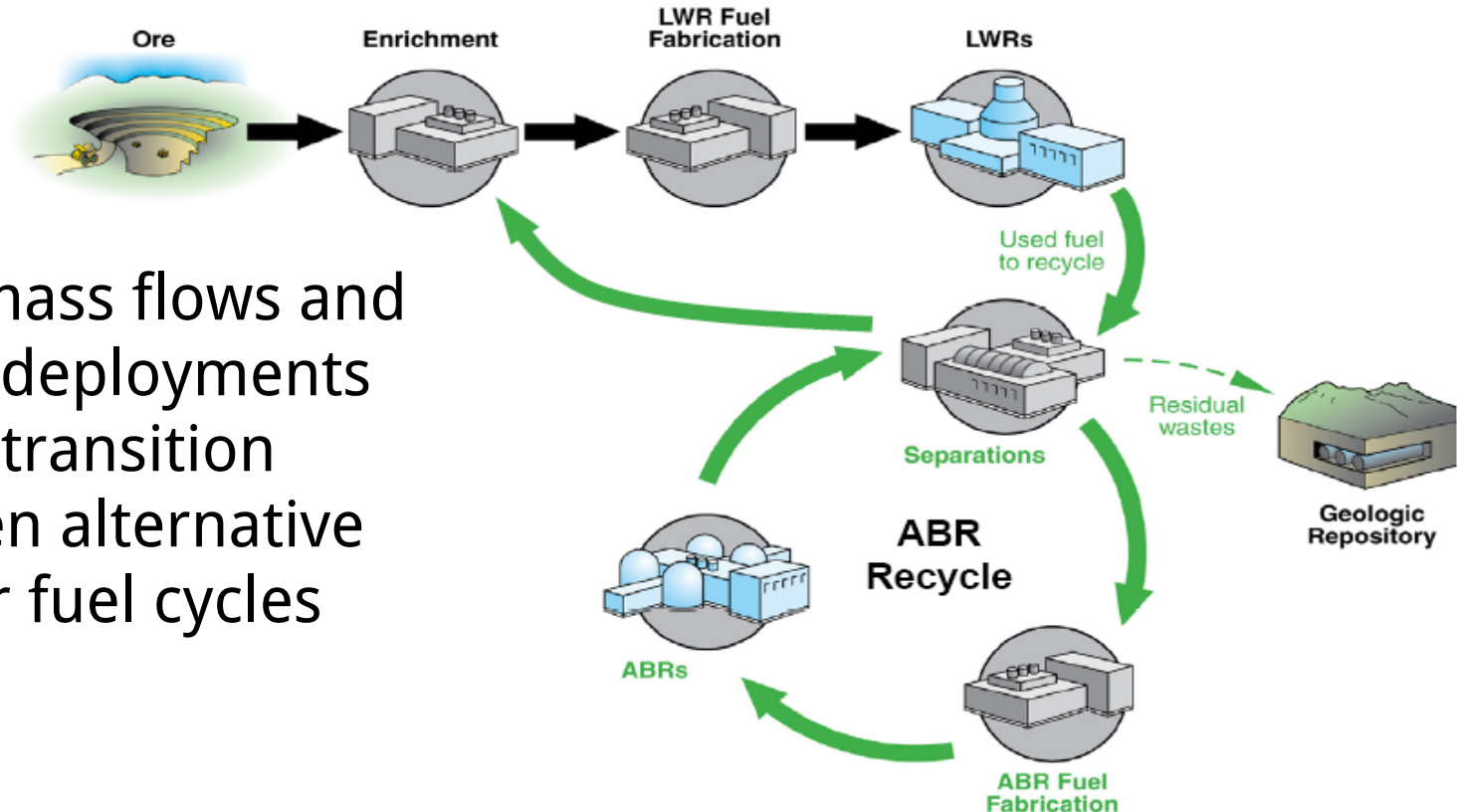
- What is a nuclear fuel cycle?
- How do we simulate it?
- What is a Dynamic Resource Exchange (DRE) and how does it work?
- How does the DRE perform in practice?

Fuel Cycle Simulator - Purpose



- Track mass flows and facility deployments during transition between alternative nuclear fuel cycles

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Difficulties



- Reactor performance (e.g., cycle length, safety margins) depend on fuel isotopics
- Commodities (i.e., elements, isotopes) are fungible
- Supply chain with recycling
- Reactors use fuel assemblies

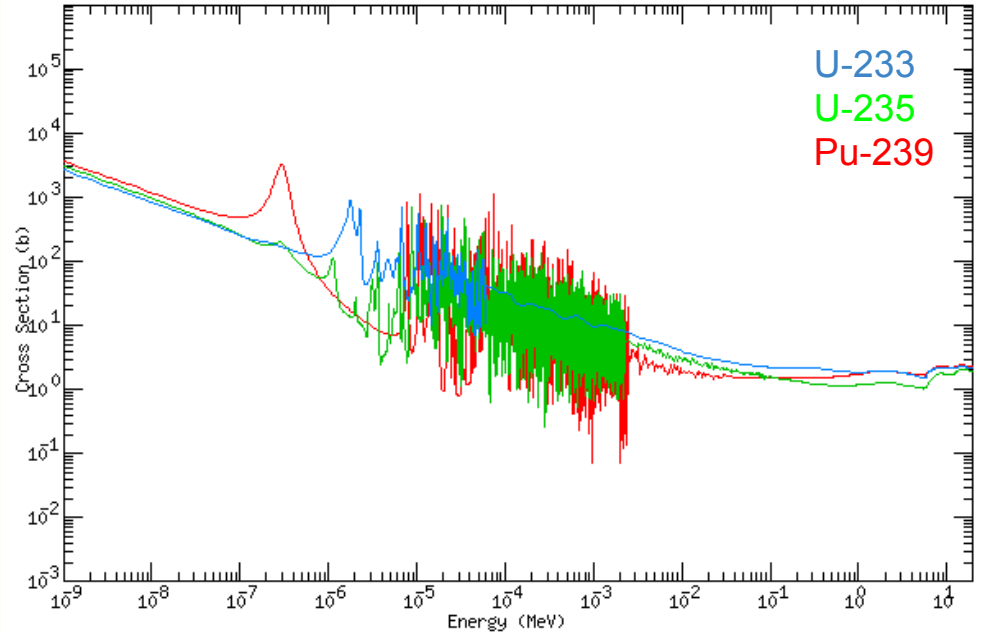


Fig: Fission Cross Section [1]

Difficulties



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Fig: Used Fuel Assemblies [2]

Dynamic Analysis

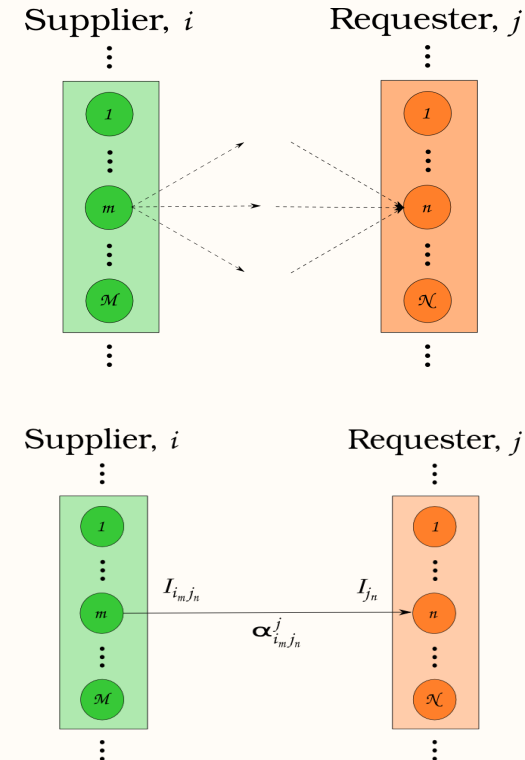


- Properties
 - Number, type of reactors is function of time
 - Fuel inventories is a function of simulation history
 - “endless number” of possible fuel cycles
 - Regional/Institutional effects
- Cyclus Approach
 - Discrete time
 - Agent-based modeling (individual facilities)
 - Resources have a quality (isotopics)
 - Discrete resource flow
 - Resource flow determined dynamically
 - <http://fuelcycle.org/>

Dynamic Resource Exchange



- DRE: Core algorithm for fuel cycle simulation
- Recomputed at each time step
- Solves economic problem dynamically; no hard-coded supply-demand behavior
- Treats arbitrarily complicated fuel cycles



DRE Phases



Request for Bids

Queries each requesting Agent in the simulation that ***demands*** a resource

Response to Request for Bids

Queries each responding Agent in the simulation that ***supplies*** a resource

Preference Adjustment

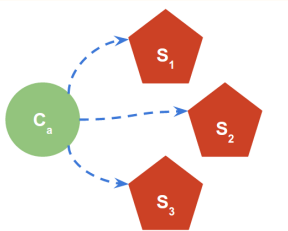
Agent ***reviews*** all matches; opportunity for preference adjustment

Solution

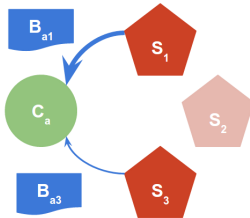
Matches ***selected*** for satisfiable requests

Trade Execution

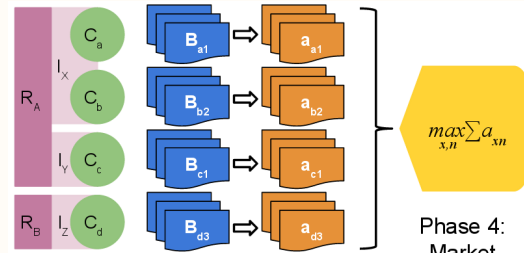
Material ***transaction*** takes place



Phase 1: Request for bids



Phase 2: Response to request for bids



Phase 3: Preference Adjustment

Phase 4:
Market
Resolution

DRE Phases



Request for Bids

Queries each requesting Agent in the simulation that ***demands*** a resource

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Preference Adjustment

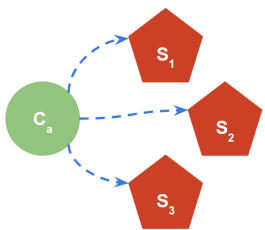
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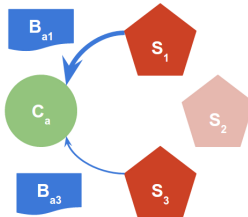
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Trade Execution

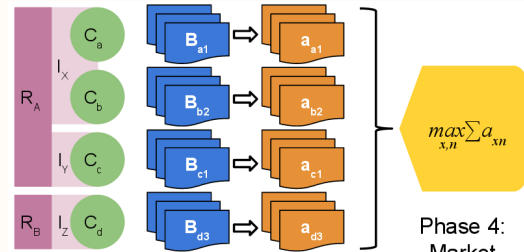
Material ***transaction*** takes place



Phase 1: Request for bids



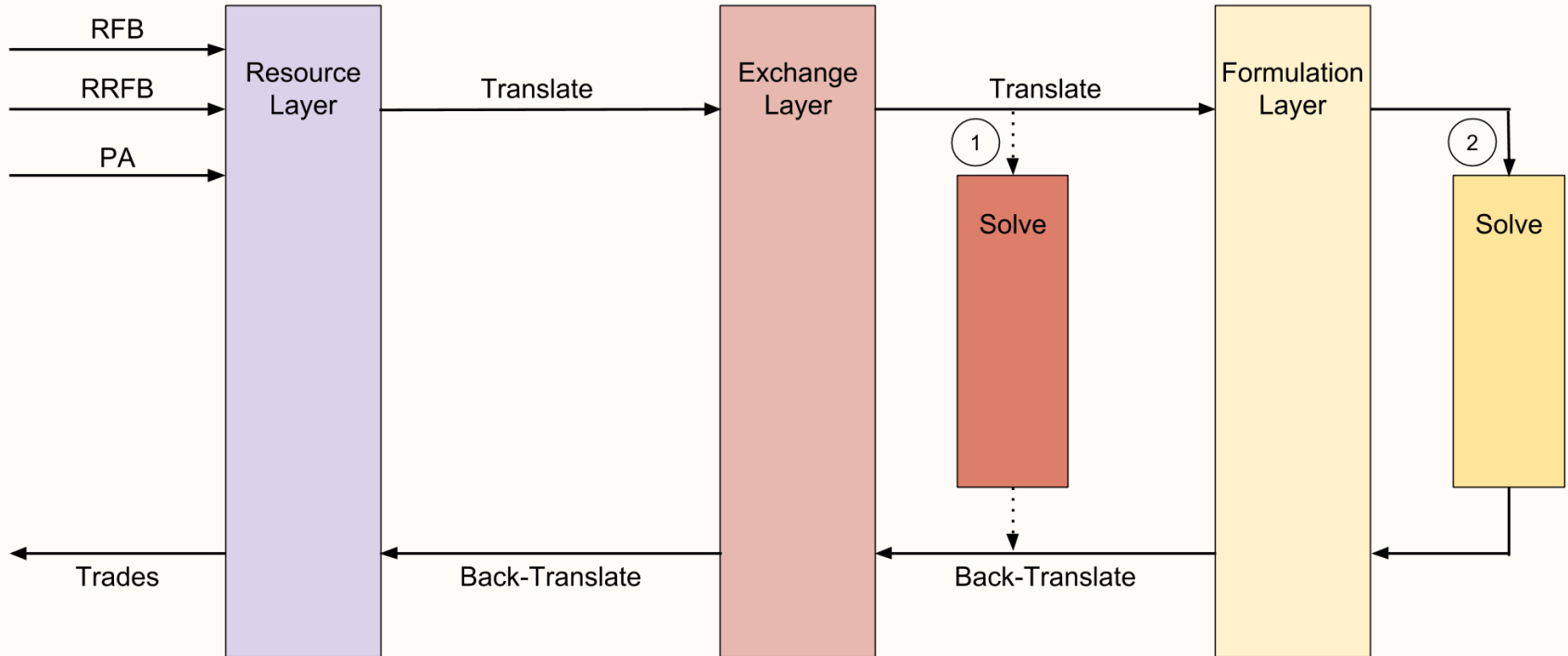
Phase 2: Response to request for bids



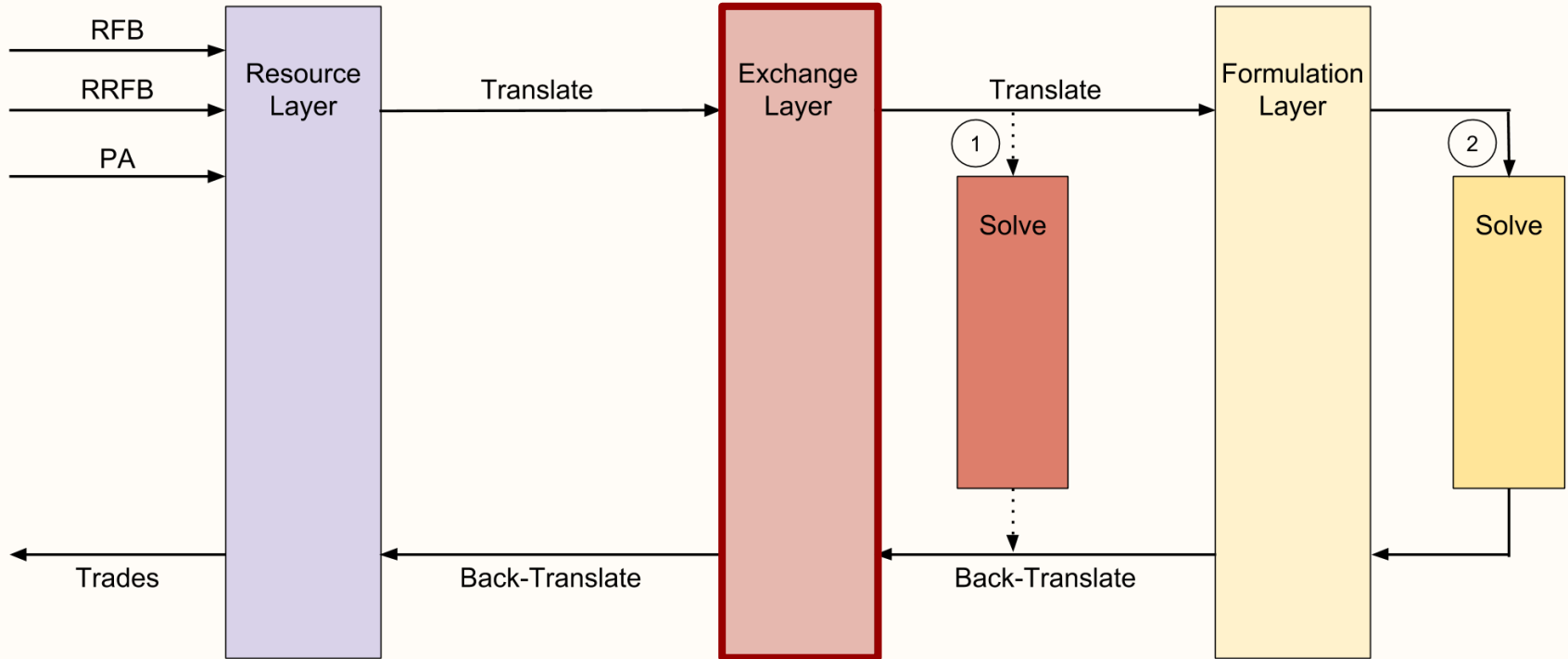
Phase 3: Preference Adjustment

Phase 4:
Market
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DRE Framework



DRE Framework

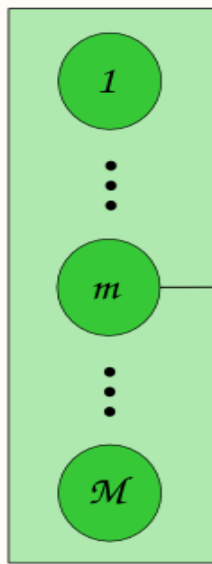


DRE Bipartite Graph



Supplier, i

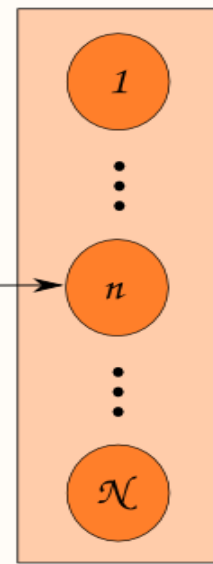
\vdots



\vdots

Requester, j

\vdots



\vdots

$I_{i_m j_n}$

I_{j_n}

$\alpha_{i_m j_n}^j$

DRE Constraints



- Agents define conversion functions for constraint coefficients
 - input: proposed resource quality
 - output: unit capacity coefficient
- Allows arbitrary physics/chemistry fidelity

$$\sum_{j \in J} f_{SWU}(\varepsilon_j) x_{i,j}^{EU} \leq s_{i,SWU}$$



$$\sum_{j \in J} \beta_{i,k}(q_j^h) x_{i,j}^h \leq s_{i,k}$$



$$\sum_{j \in J} f_{NU}(\varepsilon_j) x_{i,j}^{EU} \leq s_{i,NU}$$

$$\sum_{j \in J} a_{i,j}^k x_{i,j} \leq b_i^k$$

DRE Constraints



- Requests and bids can be *mutually exclusive*
- Given set of possible connections, only one allowed

$$\sum_{(i,j) \in M_s} y_{i,j} \leq 1$$

$$\sum_{(i,j) \in M_r} y_{i,j} \leq 1$$

DRE Solution



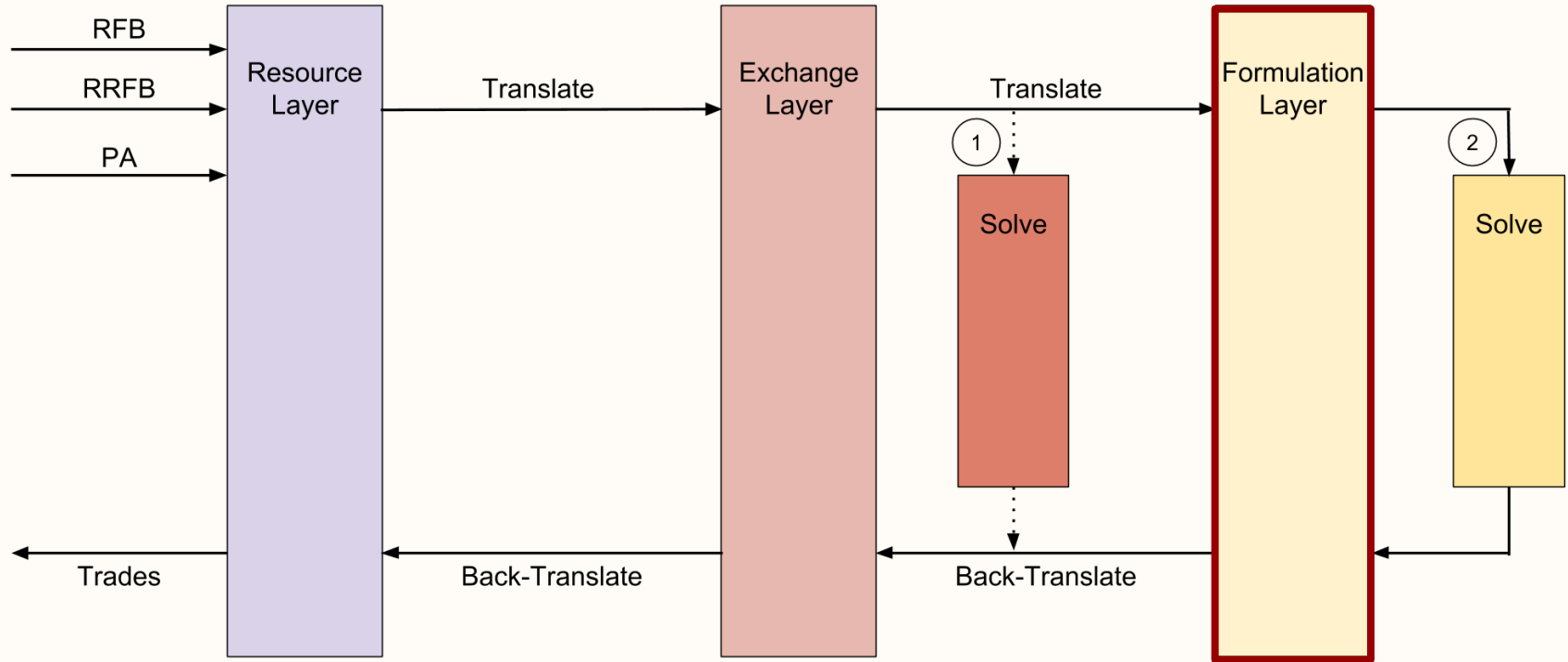
- Using Heuristics
 - Match arcs subject to supply constraints
 - Allows for exclusive trades
- Formulated as Multicommodity Transportation Problem (MTP) Variant
 - Use false arcs to guarantee feasible solution
 - Translate preferences to costs
 - LP or MILP
 - MILP allows for exclusive trades

Greedy Solver Algorithm



```
order request portfolios by average preference;
forall the request portfolios do
    order requests by average preference;
    matched  $\leftarrow 0$ ;
    while  $matched \leq q_J$  and  $\exists$  a request do
        get next request;
        order arcs by preference;
        while  $matched \leq q_J$  and  $\exists$  an arc do
            get next arc;
            remaining  $\leftarrow q_J - matched$ ;
            to_match  $\leftarrow \min\{remaining, Capacity(arc)\}$ ;
            matched  $\leftarrow matched + to\_match$ ;
        end
    end
end
```

DRE Framework



LP Formulation



Variable	Description
S, s	Suppliers
I, i	Bids
R, r	Requesters
J, j	Requests
K, k	Capacities
c	Cost of commodity
x	Decision variable
a	Capacity coeff.
b_s	Supply capacity
b_r	Demand capacity

$$\min_x z = \sum_{i \in I} \sum_{j \in J} c_{i,j} x_{i,j}$$

$$\text{s.t.} \quad \sum_{i \in I_s} \sum_{j \in J} a_{i,j}^k x_{i,j} \leq b_s^k$$

$$\forall k \in K_s, \forall s \in S$$

$$\sum_{j \in J_r} \sum_{i \in I} a_{i,j}^k x_{i,j} \geq b_r^k$$

$$\forall k \in K_r, \forall r \in R$$

$$x_{i,j} \in [0, \tilde{x}_j]$$

$$\forall i \in I, \forall j \in J$$

MILP vs. LP



Strict subsets of suppliers and consumers:

1. those that require *exclusive* orders
2. those that allow *partial* orders

$$A = \bigcup_{r \in R} A_{p_r} \cup A_{e_r}$$

Introduce a **binary variable**, $y_{i,j}$:

- 1 if resource is sent by supplier i to consumer j

MILP Formulation



Variable	Description
A_p	Partial-order arcs
A_e	Exclusive-order arcs
M	Mutually-exclusive arcs
$\sim x$	Exclusive-arc quantity
x, y	Decision variable

$$\min_{x,y} z = \sum_{(i,j) \in A_p} c_{i,j} x_{i,j} + \sum_{(i,j) \in A_e} c_{i,j} \tilde{x}_j y_{i,j}$$

$$\text{s.t.} \quad \sum_{(i,j) \in A_{ps}} a_{i,j}^k x_{i,j} + \sum_{(i,j) \in A_{es}} a_{i,j}^k \tilde{x}_j y_{i,j} \leq b_s^k$$

$$\forall k \in K_s, \forall s \in S$$

$$\sum_{(i,j) \in M_s} y_{i,j} \leq 1$$

$$\forall s \in S$$

$$\sum_{(i,j) \in A_{pr}} a_{i,j}^k x_{i,j} + \sum_{(i,j) \in A_{er}} a_{i,j}^k \tilde{x}_j y_{i,j} \geq b_r^k$$

$$\forall k \in K_r, \forall r \in R$$

$$\sum_{(i,j) \in M_r} y_{i,j} \leq 1$$

$$\forall r \in R$$

$$x_{i,j} \in [0, \tilde{x}_j]$$

$$\forall (i,j) \in A_p$$

$$y_{i,j} \in \{0,1\}$$

$$\forall (i,j) \in A_e$$

Experimental Framework



- Two exchange types: Front-End, Back-End
- Three fuel cycles ($f_{fc}: 0, 1, 2$)
 - Once through
 - MOX recycle with thermal and fast reactors
 - MOX/THOX recycle with thermal and fast reactors
- Reactors modeled using assemblies or batches ($f_{rx}: 0, 1$)
- Commodity and location-based ($f_{loc}: 0, 1, 2$)
- Solve with Greedy Heuristic, Clp, Cbc

Experimental Framework



- Stochasticity
 - Location assignment
 - Reactor fuel enrichment
- Constraint coefficients function of commodity and enrichment
- Objective coefficients function of commodity and location

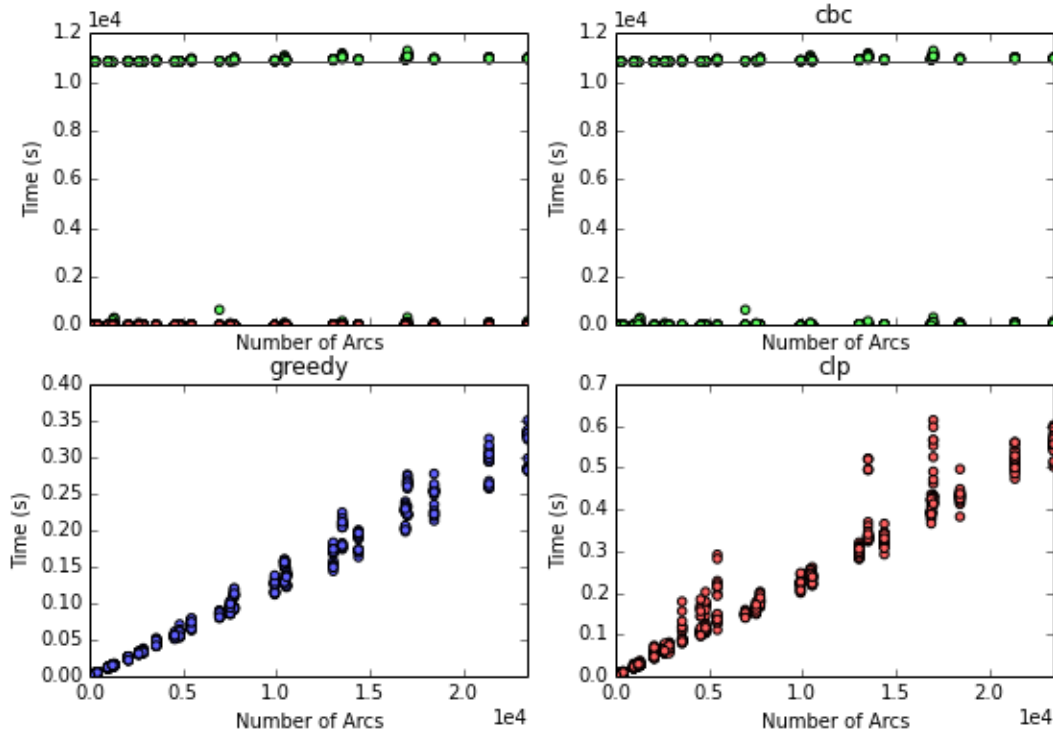
$$p(i, j) = p_c(i, j) + r_{l,c} p_l(i, j)$$

$$p_l(i, j) = \delta_{\text{reg}} \frac{\exp(-|\text{reg}_i - \text{reg}_j|) + \delta_{\text{loc}} \exp(-|\text{loc}_i - \text{loc}_j|)}{1 + \delta_{\text{loc}}}$$

Scoping Study, $f_{\text{rxtr}} = 0$



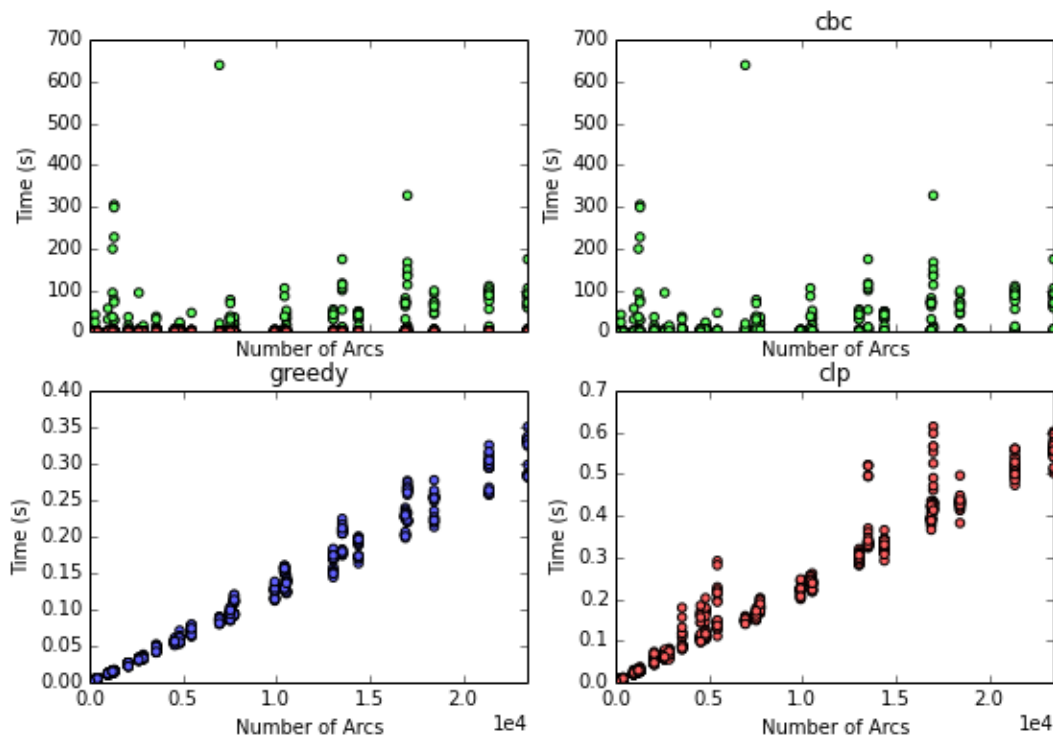
Solution Times for Front-End Exchanges



Scoping Study, $f_{rxtr} = 0$



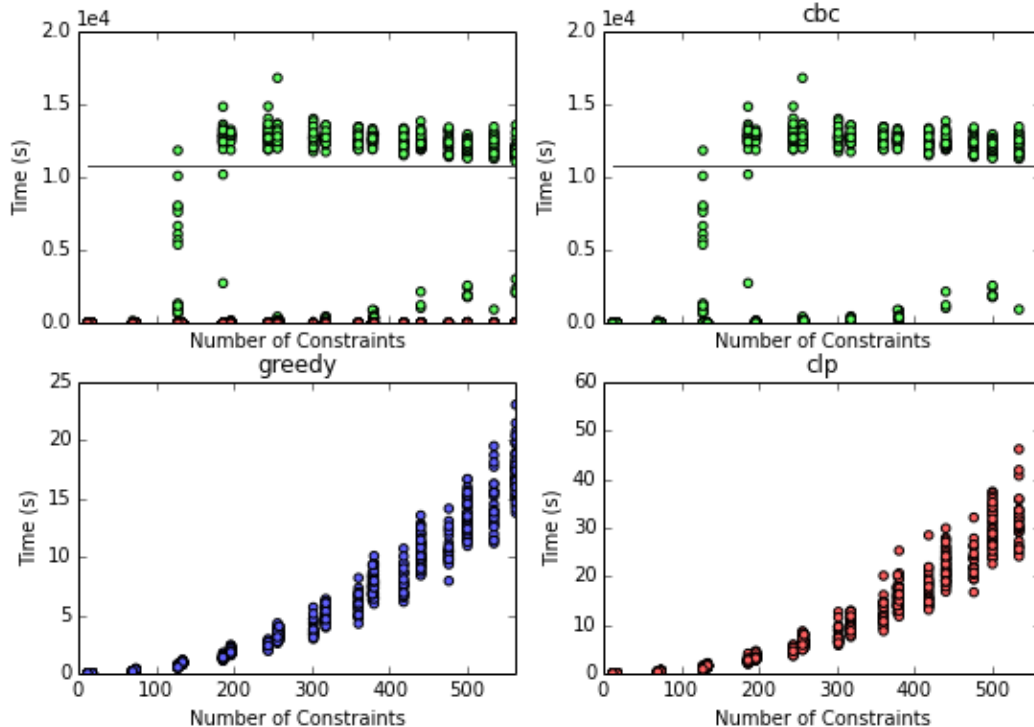
Solution Times for Front-End Exchanges



Scoping Study, $f_{rxtr} = 1$



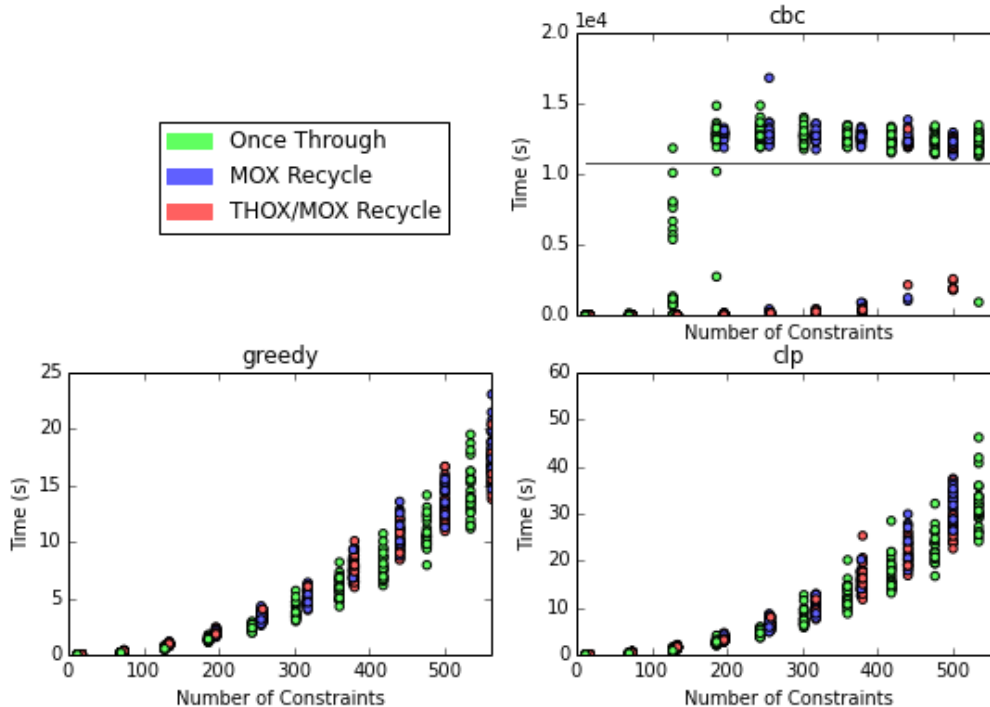
Solution Times for Front-End Exchanges



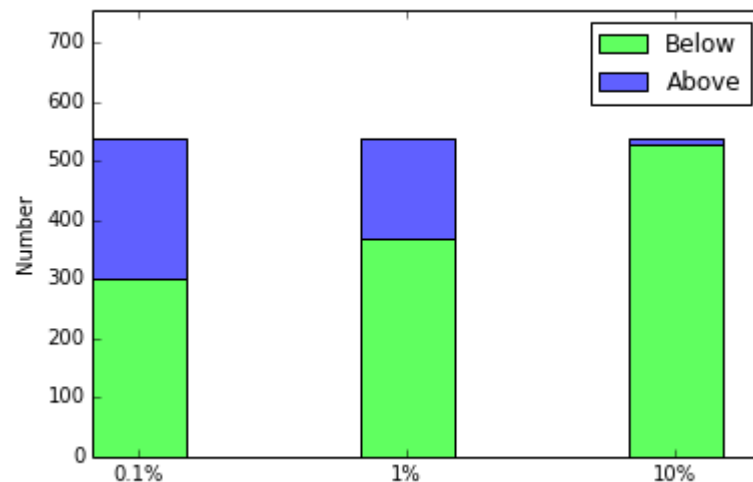
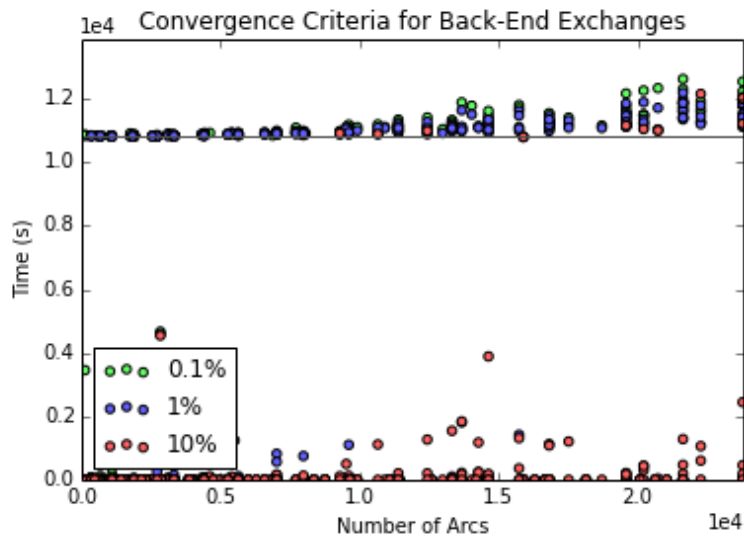
Scoping Study, $f_{rxtr} = 1$



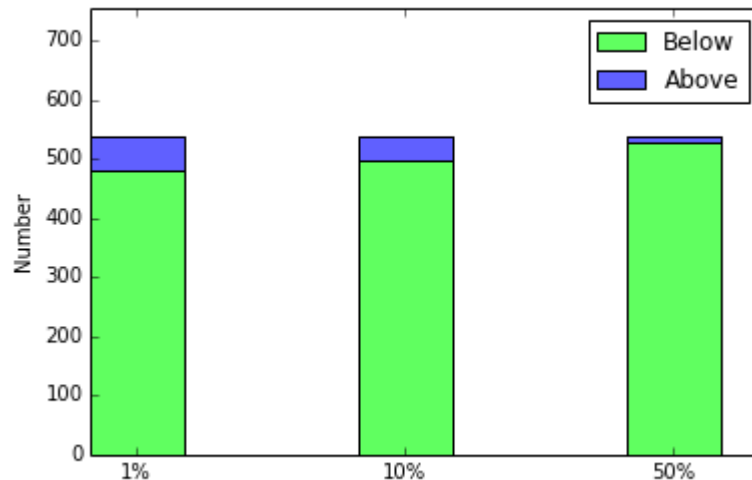
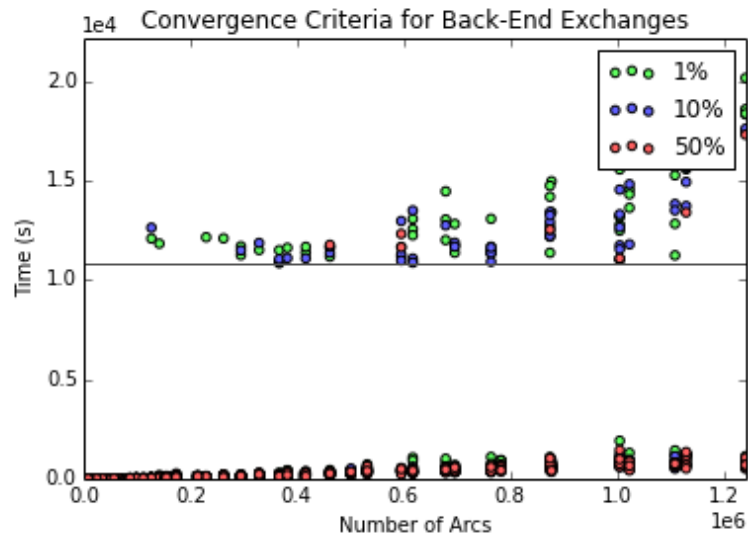
Fuel Cycle Effects on Solution Times for Front-End Exchanges



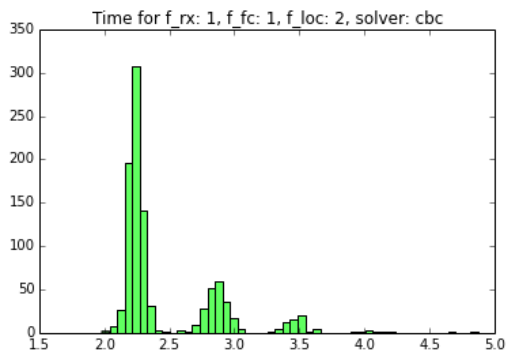
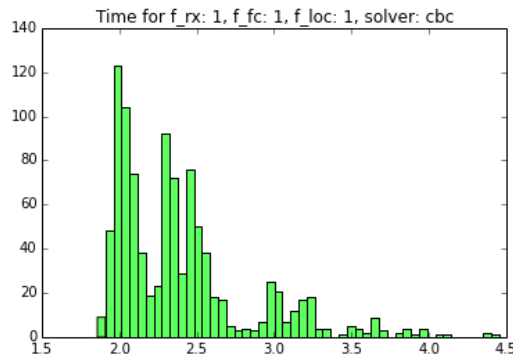
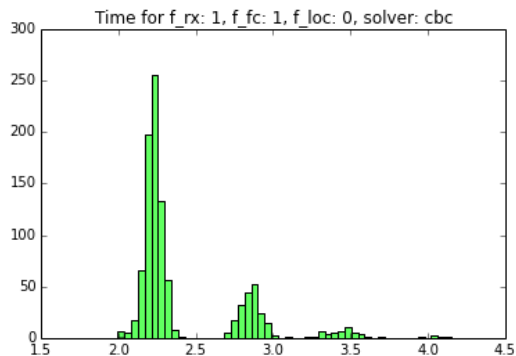
Convergence Study, $f_{\text{rxtr}} = 0$



Convergence Study, $f_{rxtr} = 1$

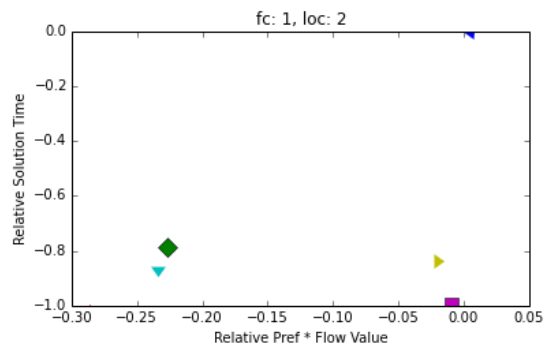
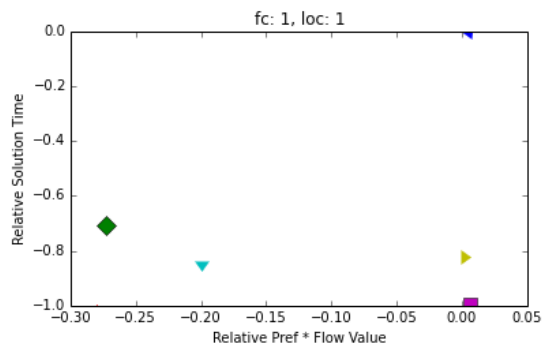
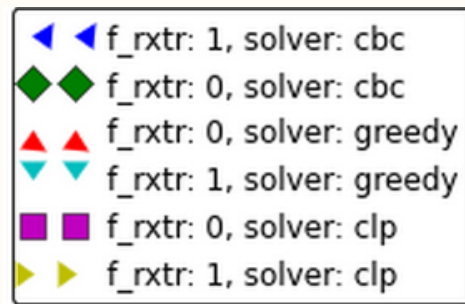
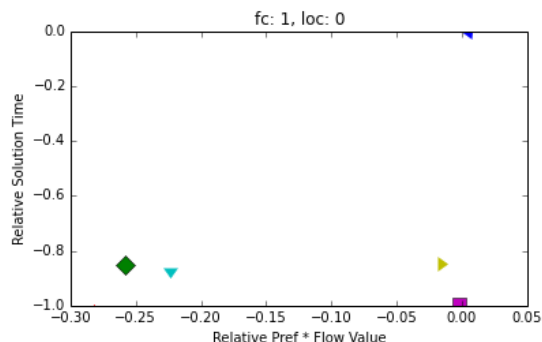


Stochastic Study, Front-End



- ~17,500 variables
- ~100 constraints

Stochastic Study, Front-End



Conclusions



- Dynamic, physics-informed supply and demand framework designed and implemented
- Large-scale, high-throughput experimental apparatus designed and implemented
- Characteristics of solutions can vary greatly based on fuel cycle, reactor fidelity, and how preferences are modeled
- Time for some simulation!

Acknowledgements



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Citations



[1] Korea Atomic Energy Institute Table of Nuclides, <http://atom.kaeri.re.kr/>, accessed 29-10-2014

[2] <http://www.energy-net.org/>

[3] Essential Physics for Fuel Cycle Modeling & Analysis. Scopatz, A. Dissertation. Dec., 2011.