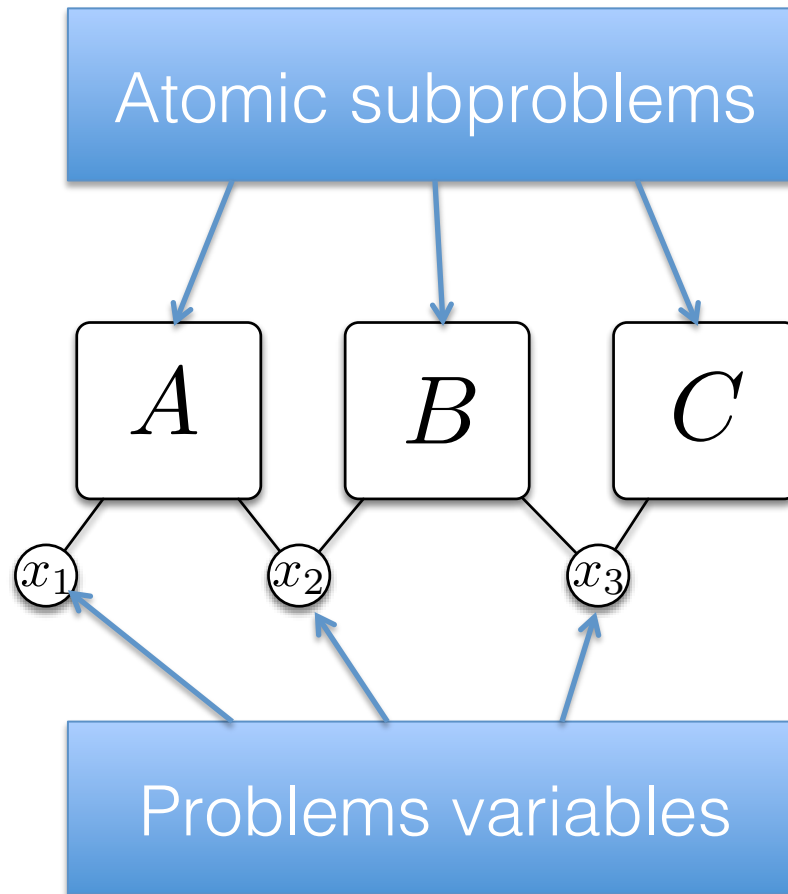


Dynamic Programming on Nominal Graphs

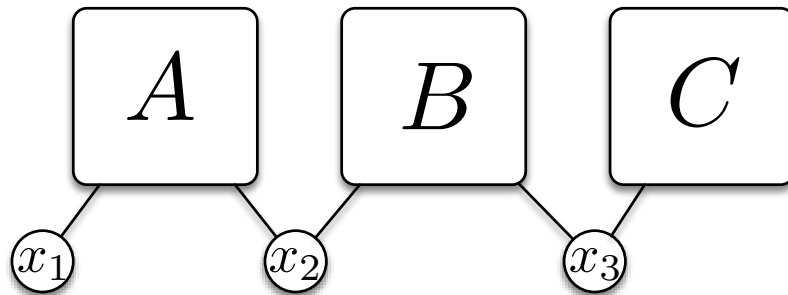
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(joint work with Nicklas Hoch, Ugo Montanari)

Optimization problems as hypergraphs



Optimization problems as hypergraphs



- Lack of algebraic structure for **resolution via structural recursion**
- No indication of variable elimination order (**secondary optimization problem**)

Process calculus-like representation

$$A(x_1, x_2)$$

atomic problem A with
two variables x_1, x_2

$$A(x_1, x_2) \parallel B(x_2, x_3)$$

composite problem

$$(x_1)A(x_1, x_2)$$

- Solved w.r.t. x_1 (**elimination**)
- Solution is parametric in x_2

Permutation algebras

Infinite set of variable names \mathcal{N}

Permutations (bijections) $\pi: \mathcal{N} \rightarrow \mathcal{N}$

Algebras \mathcal{P} including permutations and axioms

$$p \text{ id} \equiv p \quad (p\pi')\pi \equiv p(\pi \circ \pi')$$

Support

$A \subseteq \mathcal{N}$ supports $p \in P$ whenever

$$\forall \pi : \pi|_A = \text{id} \implies p \pi = p$$

Minimal support $\text{supp}(p)$ generalizes free variables

An algebraic specification for optimization problems

$$p, q := p \parallel q \mid (x)p \mid p\pi \mid A(\tilde{x}) \mid nil$$

- Axioms:
- usual process calculi ones
 - permutations distribute over operations

For every term p :

$$supp(p) = fn(p) \text{ (non-restricted variables)}$$

Alfa-conversion

$$(x)p \equiv (y)p[x \mapsto y] \quad y \notin \text{supp}(p)$$

Scope extension

$$(x)(p \parallel q) \equiv (x)p \parallel q \quad x \notin \text{supp}(q)$$

Correct handling of restrictions in all algebras

Hierarchical specification

no scope extension

terms describe solutions of **secondary optimization problem** (variable elimination order)

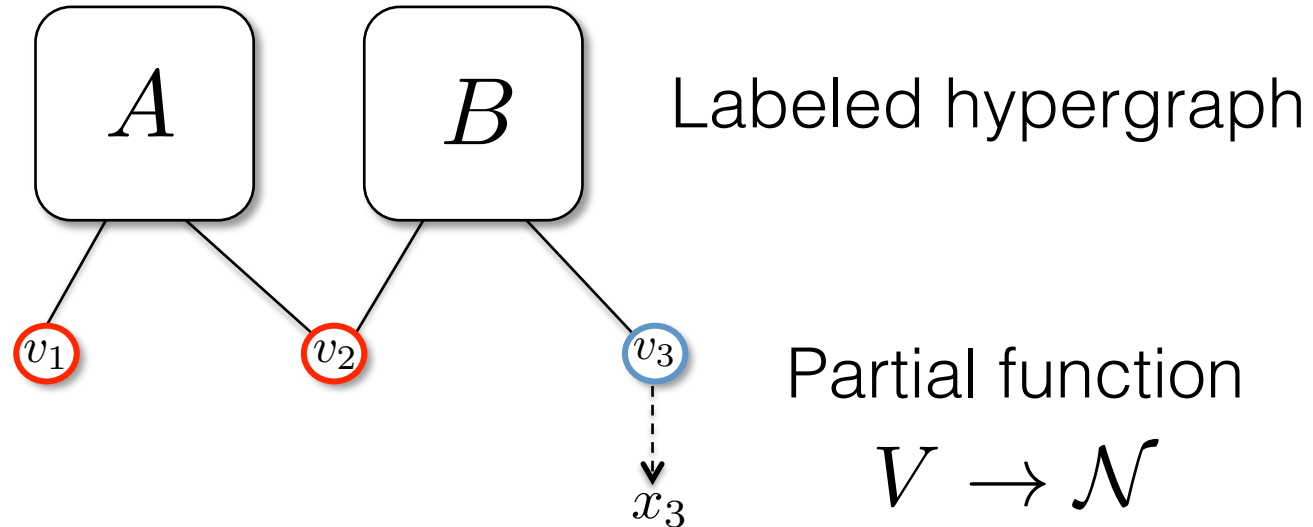
Normal form: elimination **at the end**

$$(x_1)(x_2)(x_3)(A(x_1, x_2) \parallel B(x_2, x_3) \parallel C(x_3))$$

Canonical form: elimination **as soon as possible**

$$(x_2)((x_1)A(x_1, x_2) \parallel (x_3)(B(x_2, x_3) \parallel C(x_3)))$$

Nominal Graphs

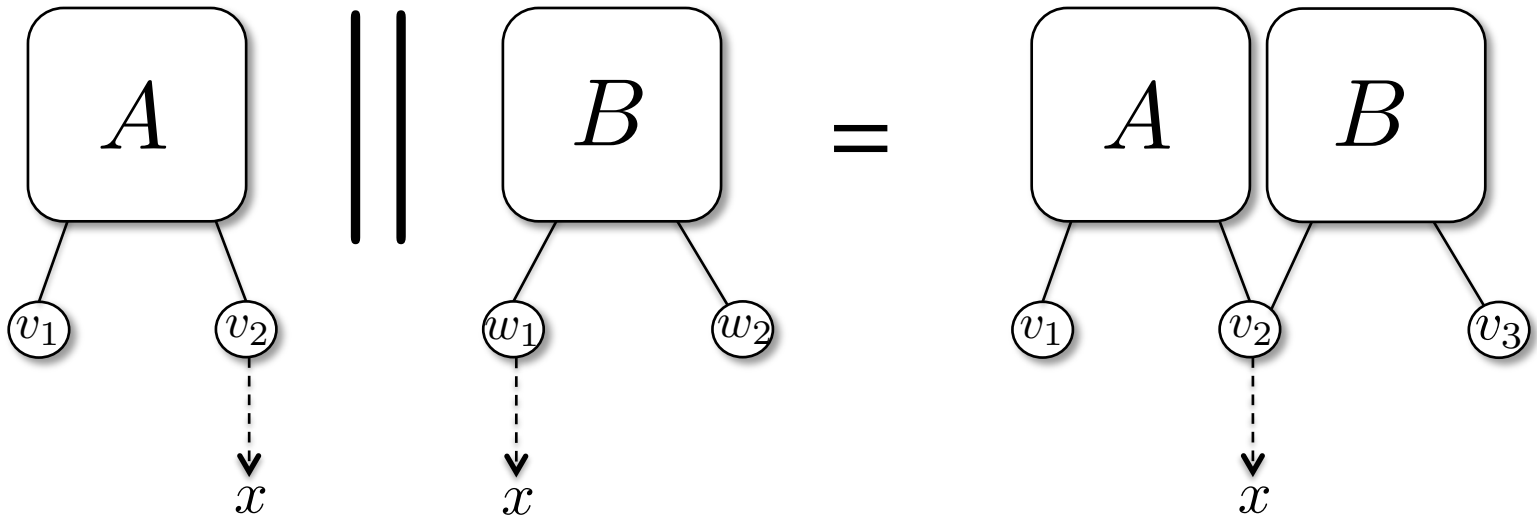


Non-interface vertices =
restricted names
(up to alpha-conversion)

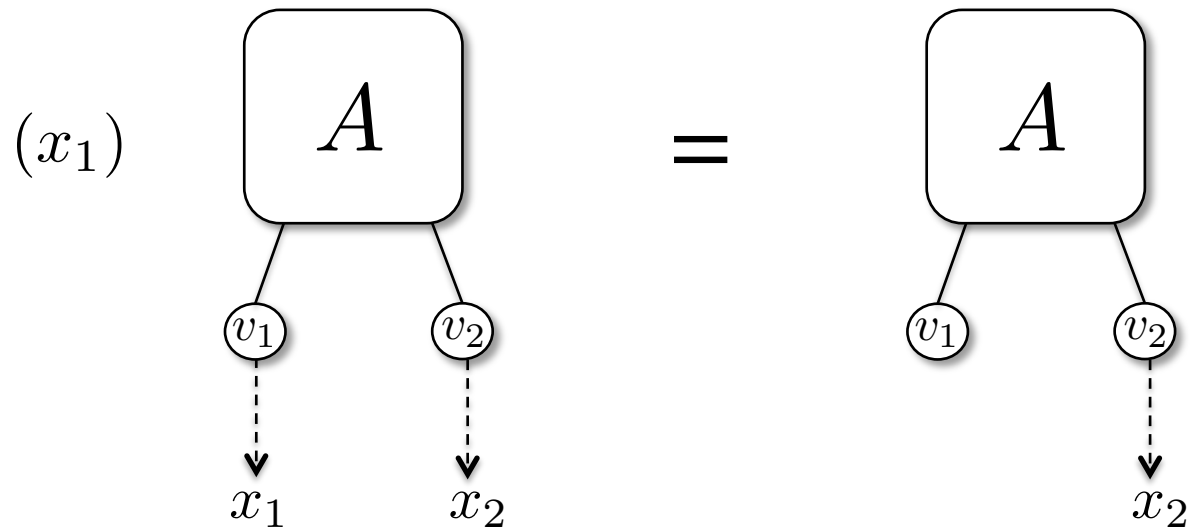
Interface vertices =
free names

$$(\textcolor{red}{x}_1)(\textcolor{red}{x}_2)(A(x_1, x_2) \parallel B(x_2, \textcolor{blue}{x}_3))$$

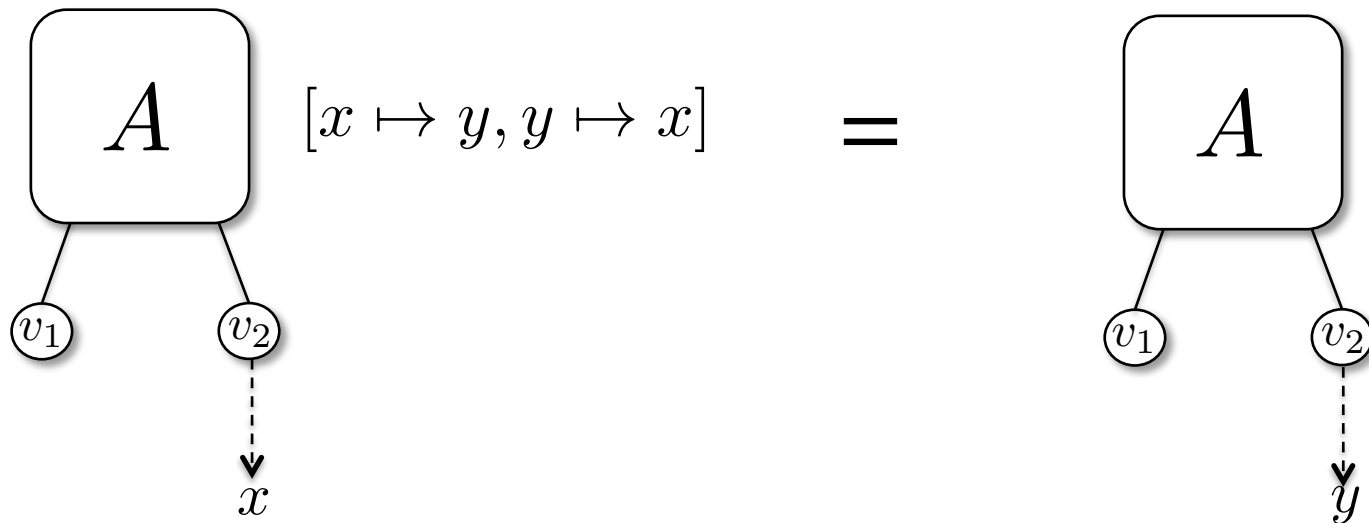
Optimization algebra of nominal graphs



Optimization algebra of nominal graphs



Optimization algebra of nominal graphs



supp = set of interface vertices

Optimization algebra of nominal graphs

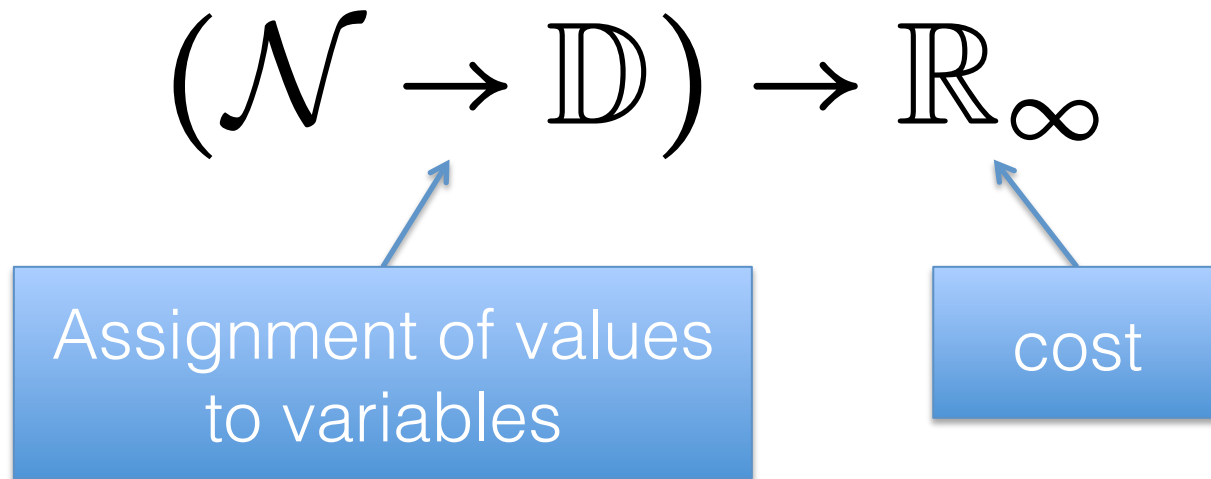
Sound and complete axiomatization

congruent terms

isomorphic nominal graphs

We have recursive evaluations of nominal graphs
in any algebra (initial-algebra semantics)

Optimization algebra of cost functions



$\llbracket p \rrbracket^c$ = evaluation of p as a cost function

By general properties of permutation algebras:

$\llbracket p \rrbracket^c$ determined by assignments to $\text{supp}(p)$

Representable as a finite table

Computed via structural recursion
(“almost” dynamic programming)

Typical optimization problems

One cost function c_A for each atomic problem A

Goal: minimize total cost

$$\llbracket A(x_1, \dots, x_n) \rrbracket^c \rho = c_A(\rho(x_1), \dots, \rho(x_n))$$

$$\llbracket p_1 \parallel p_2 \rrbracket^c \rho = \llbracket p_1 \rrbracket^c \rho + \llbracket p_2 \rrbracket^c \rho$$

$$\llbracket (x)p \rrbracket^c \rho = \min_{v \in \mathbb{D}} \llbracket p \rrbracket^c (\rho[x \mapsto v])$$

Typical optimization problems

One cost function c_A for each atomic problem A

Goal: minimize total cost

$\llbracket A \rrbracket^c$ When all variables are restricted $(\rho_n))$

Cost function = minimal cost

$$\llbracket (x)p \rrbracket^c \rho = \min_{v \in \mathbb{D}} \llbracket p \rrbracket^c (\rho[x \mapsto v])$$

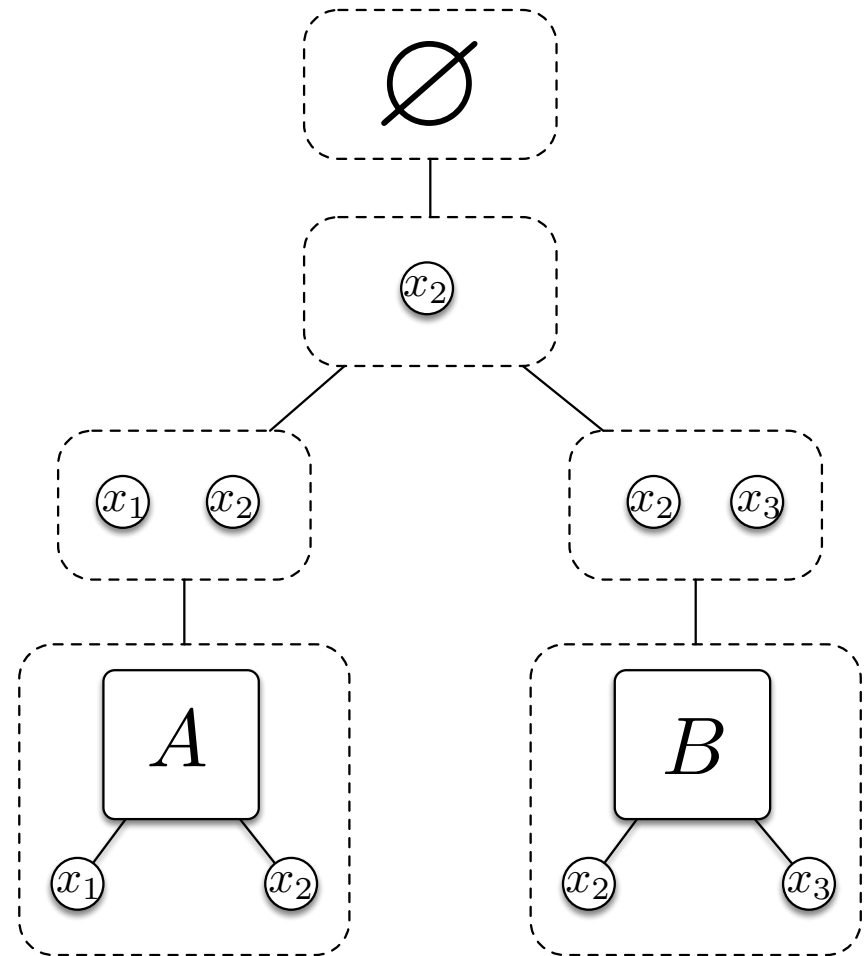
We can compute the cost function for a graph but...

How can we pick a variable elimination strategy?

Hierarchical nominal graphs

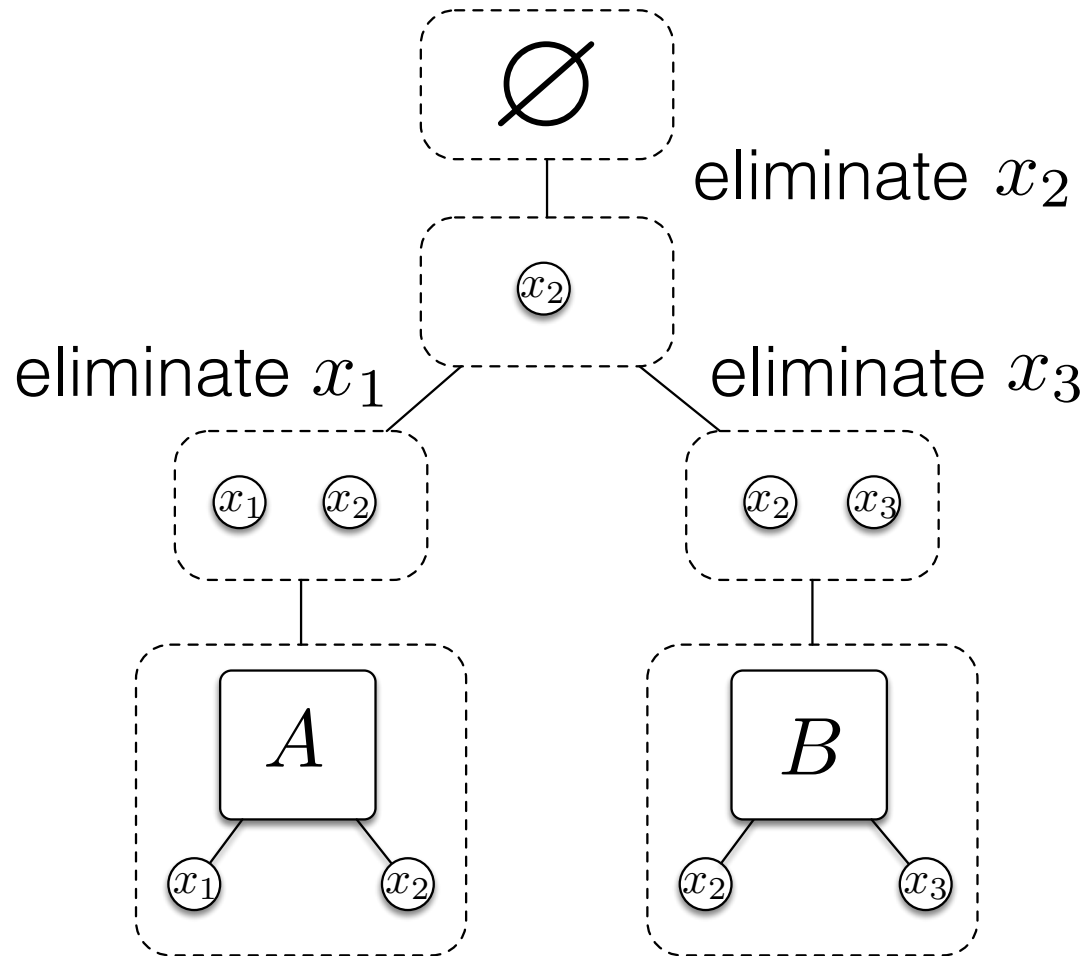
Graphical counterpart of hierarchical terms

$$(x_2)((x_1)A(x_1, x_2) \parallel (x_3)B(x_2, x_3))$$



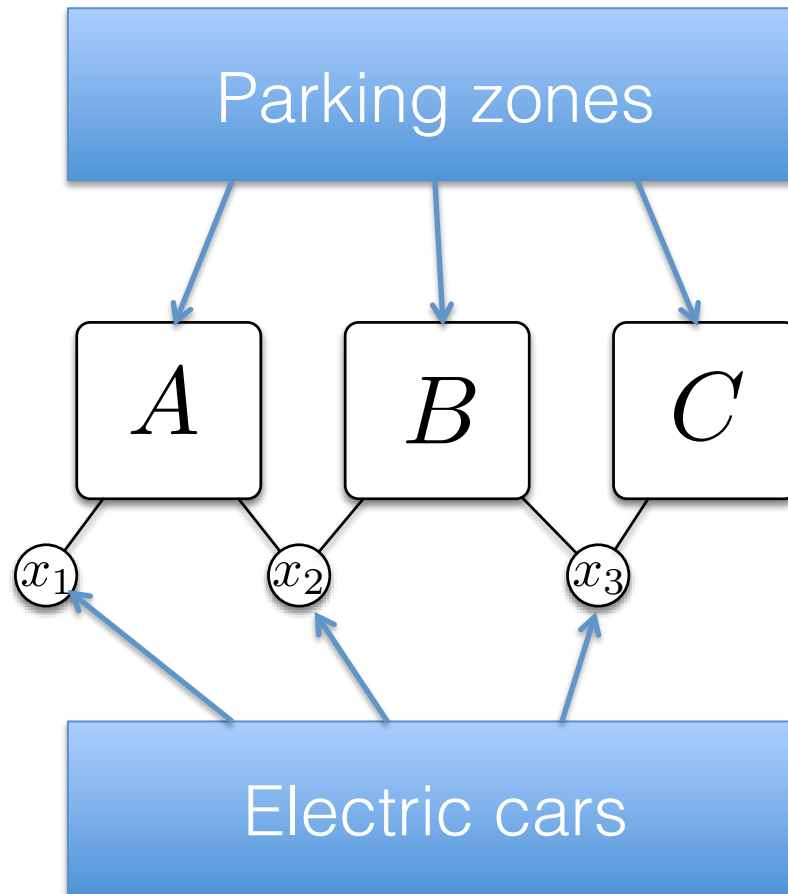
Dynamic programming algorithm

Bottom-up visit of the tree



Compute cost functions (tables) of leaves

ASCENS parking optimization problem



Goal: finding the best allocation of cars to parking zones

Algebraic representation

Boolean variable assignments

– “outside”

✓ “inside”

$A(x_1, \dots, x_n)$

cars x_1, \dots, x_n may
be parked inside A

$(x)p$

x must be parked inside of p

A “more efficient” algebra of cost functions

$$\llbracket p \rrbracket^c : (\mathcal{N} \rightarrow \{-, \checkmark\}) \rightarrow \mathbb{R}_\infty$$

A “more efficient” algebra of cost functions

$$\llbracket p \rrbracket^c : (\mathcal{N} \rightarrow \{-, \checkmark\}) \rightarrow \mathbb{R}_\infty$$



$$X \subseteq fn(p)$$

(other variables do not matter)

A “more efficient” algebra of cost functions

$$\llbracket p \rrbracket^c : (\mathcal{N} \rightarrow \{-, \checkmark\}) \rightarrow \mathbb{R}_\infty$$

$$\llbracket p \parallel q \rrbracket^c X = \min_{\{X_1, X_2\} \in \mathcal{P}_2(X)} \left\{ \llbracket p \rrbracket^c X_1 + \llbracket q \rrbracket^c X_2 \mid \begin{array}{l} X_1 \subseteq fn(p), \\ X_2 \subseteq fn(q) \end{array} \right\}$$

$$\llbracket (x)p \rrbracket^c X = \llbracket p \rrbracket^c (X \cup \{x\})$$

Conclusions

Algebraic specification for optimization problems

- Variable elimination as variable binding
- Correctly handled via permutation algebras

Flat

Nominal terms/graphs

Recursive computation
of cost functions

Hierarchical
(no scope extension)

Hierarchical terms/graphs

Dynamic Programming
algorithm

Future work

- Heuristics
- Dynamic graphs
- Precise correspondence between class of terms and nominal graphs
- Formalizing nominal structure