Constraint Logic Programming

Sylvain Soliman@inria.fr

informatics / mathematics

Project-Team LIFEWARE

MPRI 2.35.1 Course – September–November 2017

Part I: CLP - Introduction and Logical Background



- 2 Examples and Applications
- First Order Logic





Part II: Constraint Logic Programs









Part III: CLP - Operational and Fixpoint Semantics







Part IV: Logical Semantics









Part V: Constraint Solving





Part VI: Practical CLP Programming



20 Optimizing CLP



22 Symmetry Breaking During Search



Part VII: More Constraint Programming





Part VIII: Programming Project









Part IX

Concurrent Constraint Programming

Part IX: Concurrent Constraint Programming







The Paradigm of Constraint Programming

memory of values programming variables memory of constraints mathematical variables



Concurrent Constraint Programs

Class of programming languages $CC(\mathcal{X})$ introduced by Saraswat [Saraswat93mit] as a merge of Constraint and Concurrent Logic Programming.

Processes $P ::= \mathcal{D}.A$ Declarations $\mathcal{D} ::= p(\vec{x}) = A, \mathcal{D} \mid \epsilon$ Agents $A ::= tell(c) \mid \qquad |A \parallel A \mid A + A \mid \exists xA \mid p(\vec{x})$ CC agent CC agent + $A := tell(c) \mid \qquad +$

Constraint Store

Concurrent Constraint Programs

Class of programming languages $CC(\mathcal{X})$ introduced by Saraswat [Saraswat93mit] as a merge of Constraint and Concurrent Logic Programming.



Translating $CLP(\mathcal{X})$ into $CC(\mathcal{X})$ Declarations

 $CLP(\mathcal{X})$ program:

equivalent CC(X) declaration:

 $\begin{aligned} \mathbf{A} &= \textit{tell}(c) \parallel \mathbf{B} \parallel \mathbf{C} + \textit{tell}(d) \parallel \mathbf{D} \parallel \mathbf{E} \\ \mathbf{B} &= \textit{tell}(\mathbf{e}) \end{aligned}$

This is just a process calculus syntax for CLP programs...

```
(CC agent)^{\dagger} = CLP goal
```

 $(tell(c))^{\dagger} =$

 $(CC agent)^{\dagger} = CLP goal$

 $(tell(c))^{\dagger} = c$ $(A \parallel B)^{\dagger} =$

```
(CC agent)^{\dagger} = CLP goal
```

$$\begin{array}{ll} (\textit{tell}(\textit{c}))^{\dagger} & = \textit{c} \\ (\textit{A} \parallel \textit{B})^{\dagger} & = \textit{A}^{\dagger}, \textit{B}^{\dagger} \\ (\textit{A} + \textit{B})^{\dagger} & = \end{array}$$

$$(CC agent)^{\dagger} = CLP goal$$

$$\begin{array}{ll} (tell(c))^{\dagger} &= c \\ (A \parallel B)^{\dagger} &= A^{\dagger}, B^{\dagger} \\ (A+B)^{\dagger} &= p(\vec{x}) \text{ where } \vec{x} = fv(A) \cup fv(B) \text{ and} \\ p(\vec{x}) \leftarrow A^{\dagger} \\ p(\vec{x}) \leftarrow B^{\dagger} \\ (\exists x \ A)^{\dagger} &= \end{array}$$

$$(CC agent)^{\dagger} = CLP goal$$

$$\begin{array}{rcl} (tell(c))^{\dagger} &= c\\ (A \parallel B)^{\dagger} &= A^{\dagger}, B^{\dagger}\\ (A+B)^{\dagger} &= p(\vec{x}) \text{ where } \vec{x} = fv(A) \cup fv(B) \text{ and}\\ p(\vec{x}) \leftarrow A^{\dagger}\\ p(\vec{x}) \leftarrow B^{\dagger}\\ (\exists x \ A)^{\dagger} &= q(\vec{y}) \text{ where } \vec{y} = fv(A) \setminus \{x\} \text{ and}\\ q(\vec{y}) \leftarrow A^{\dagger} \end{array}$$

$$(CC agent)^{\dagger} = CLP goal$$

$$\begin{array}{ll} (tell(c))^{\dagger} &= c \\ (A \parallel B)^{\dagger} &= A^{\dagger}, B^{\dagger} \\ (A+B)^{\dagger} &= p(\vec{x}) \text{ where } \vec{x} = fv(A) \cup fv(B) \text{ and} \\ p(\vec{x}) \leftarrow A^{\dagger} \\ p(\vec{x}) \leftarrow B^{\dagger} \\ (\exists x \ A)^{\dagger} &= q(\vec{y}) \text{ where } \vec{y} = fv(A) \setminus \{x\} \text{ and} \\ q(\vec{y}) \leftarrow A^{\dagger} \\ (p(\vec{x}))^{\dagger} &= p(\vec{x}) \end{array}$$

The ask operation $c \rightarrow A$ has no CLP equivalent.

It is a new synchronization primitive between agents.

CC Computations

Concurrency = communication (shared variables) + synchronization (ask)

Communication channels, i.e., variables, are transmissible by agents (like in π -calculus, unlike CCS, CSP, Occam,...)

Communication is additive (a constraint will never be removed), monotonic accumulation of information in the store (as in CLP, as in Scott's information systems)

Synchronization makes computation both data-driven and goal-directed.

No private communication, all agents sharing a variable will see a constraint posted on that variable.

Not a parallel implementation model.

$CC(\mathcal{X})$ Configurations

Configuration $(\vec{x}; c; \Gamma)$: store *c* of constraints, multiset Γ of agents, modulo \equiv the smallest congruence s.t.:

$$\begin{array}{ll} \mathcal{X}\text{-equivalence} & \frac{c \dashv \mathcal{X} d}{c \equiv d} \\ \\ \alpha\text{-Conversion} & \frac{z \notin fv(A)}{\exists yA \equiv \exists zA[z/y]} \\ \\ \text{Parallel} & (\vec{x}; c; A \parallel B, \Gamma) \equiv (\vec{x}; c; A, B, \Gamma) \\ \\ \text{Hiding} & \frac{y \notin fv(c, \Gamma)}{(\vec{x}; c; \exists yA, \Gamma) \equiv (\vec{x}, y; c; A, \Gamma)} & \frac{y \notin fv(c, \Gamma)}{(\vec{x}, y; c; \Gamma) \equiv (\vec{x}; c; \Gamma)} \end{array}$$

$CC(\mathcal{X})$ Transitions

Interleaving semantics

Procedure call
$$(p(\vec{y}) = A) \in \mathcal{D}$$
 $(\vec{x}; c; p(\vec{y}), \Gamma) \longrightarrow (\vec{x}; c; A, \Gamma)$ Tell $(\vec{x}; c; tell(d), \Gamma) \longrightarrow (\vec{x}; c \land d; \Gamma)$

Ask

 $\begin{array}{ll} \textbf{Blind choice} & (\vec{x}; c; A + B, \Gamma) \longrightarrow (\vec{x}; c; A, \Gamma) \\ \textbf{(local/internal)} & (\vec{x}; c; A + B, \Gamma) \longrightarrow (\vec{x}; c; B, \Gamma) \end{array}$

$CC(\mathcal{X})$ Transitions

Interleaving semantics

Procedure call	$\frac{(\boldsymbol{p}(\vec{\boldsymbol{y}}) = \boldsymbol{A}) \in \mathcal{D}}{(\vec{\boldsymbol{x}}; \boldsymbol{c}; \boldsymbol{p}(\vec{\boldsymbol{y}}), \Gamma) \longrightarrow (\vec{\boldsymbol{x}}; \boldsymbol{c}; \boldsymbol{A}, \Gamma)}$
Tell	$(\vec{x}; c; tell(d), \Gamma) \longrightarrow (\vec{x}; c \wedge d; \Gamma)$
Ask	$\frac{\boldsymbol{c} \vdash_{\mathcal{X}} \boldsymbol{d}[\vec{t}/\vec{y}]}{(\vec{x}; \boldsymbol{c}; \forall \vec{y} (\boldsymbol{d} \rightarrow \boldsymbol{A}), \Gamma) \longrightarrow (\vec{x}; \boldsymbol{c}; \boldsymbol{A}[\vec{t}/\vec{y}], \Gamma)}$
Blind choice (local/internal)	$(\vec{x}; \boldsymbol{c}; \boldsymbol{A} + \boldsymbol{B}, \Gamma) \longrightarrow (\vec{x}; \boldsymbol{c}; \boldsymbol{A}, \Gamma)$ $(\vec{x}; \boldsymbol{c}; \boldsymbol{A} + \boldsymbol{B}, \Gamma) \longrightarrow (\vec{x}; \boldsymbol{c}; \boldsymbol{B}, \Gamma)$

$CC(\mathcal{X})$ extra rules

 $c \vdash_{\mathcal{X}} c_j$ **Guarded choice** $\overline{(\vec{x}; c; \Sigma_i c_i \to A_i, \Gamma)} \longrightarrow (\vec{x}; c; A_i, \Gamma)$ (global/external) $c \vdash_{\mathcal{X}} \neg d$ AskNot $\overline{(\vec{x}; c; \forall \vec{y} (d \to A), \Gamma) \longrightarrow (\vec{x}; c; \Gamma)}$ $(\vec{x}; c; \Gamma) \longrightarrow (\vec{x}; d; \Gamma')$ Sequentiality $\overline{(\vec{\mathbf{x}}:\mathbf{c}:(\Gamma;\Delta),\Phi)\longrightarrow(\vec{\mathbf{x}};\mathbf{d}:(\Gamma';\Delta),\Phi)}$ $(\vec{\mathbf{x}}; \mathbf{C}; (\emptyset; \Gamma), \Delta) \longrightarrow (\vec{\mathbf{x}}; \mathbf{d}; \Gamma, \Delta)$

Properties of CC Transitions (1)

Theorem 1 (Monotonicity)

If $(\vec{x}; c; \Gamma) \longrightarrow (\vec{y}; d; \Delta)$ then $(\vec{x}; c \land e; \Gamma, \Sigma) \longrightarrow (\vec{y}; d \land e; \Delta, \Sigma)$ for every constraint e and agents Σ .

Proof.

Corollary 2

Strong fairness and weak fairness are equivalent.

Properties of CC Transitions (1)

Theorem 1 (Monotonicity)

If $(\vec{x}; c; \Gamma) \longrightarrow (\vec{y}; d; \Delta)$ then $(\vec{x}; c \land e; \Gamma, \Sigma) \longrightarrow (\vec{y}; d \land e; \Delta, \Sigma)$ for every constraint e and agents Σ .

Proof.

tell and ask are monotonic (monotonic conditions in guards).

Corollary 2

Strong fairness and weak fairness are equivalent.

Properties of CC Transitions (2)

A configuration without + is called deterministic.

Theorem 3 (Confluence)

For any deterministic configuration κ with deterministic declarations, if $\kappa \longrightarrow \kappa_1$ and $\kappa \longrightarrow \kappa_2$ then $\kappa_1 \longrightarrow \kappa'$ and $\kappa_2 \longrightarrow \kappa'$ for some κ' .

Corollary 4

Independence of the scheduling of the execution of parallel agents.

Properties of CC Transitions (3)



Proof.

Theorem 6 (Restartability)

If $(\vec{x}; c; \Gamma) \longrightarrow^* (\vec{y}; d; \Delta)$ then $(\vec{x}; \exists \vec{y}d; \Gamma) \longrightarrow^* (\vec{y}; d; \Delta)$.

Proof.

By extensivity and monotonicity.

Properties of CC Transitions (3)

Theorem 5 (Extensivity) If $(\vec{x}; c; \Gamma) \longrightarrow (\vec{y}; d; \Delta)$ then $\exists \vec{y} d \vdash_{\mathcal{X}} \exists \vec{x} c$.

Proof.

For any constraint $e, c \land e \vdash_{\mathcal{X}} c$.

Theorem 6 (Restartability)

If $(\vec{x}; c; \Gamma) \longrightarrow^* (\vec{y}; d; \Delta)$ then $(\vec{x}; \exists \vec{y}d; \Gamma) \longrightarrow^* (\vec{y}; d; \Delta)$.

Proof.

By extensivity and monotonicity.

observing the set of success stores,

observing the set of terminal stores (successes and suspensions),

observing the set of accessible stores,

$$\mathcal{O}_{\infty}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}_{0}) = \{ \sqcup_{?} \{ \exists \vec{\boldsymbol{x}}_{i} \boldsymbol{c}_{i} \}_{i \geq 0} | (\emptyset; \boldsymbol{c}_{0}; \boldsymbol{A}) \longrightarrow (\vec{\boldsymbol{x}_{1}}; \boldsymbol{c}_{1}; \Gamma_{1}) \longrightarrow \dots \}$$

observing the set of success stores,

 $\mathcal{O}_{ss}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{ \exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}}; \boldsymbol{d}; \epsilon) \}$

observing the set of terminal stores (successes and suspensions),

observing the set of accessible stores,

$$\mathcal{O}_{\infty}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}_{0}) = \{ \sqcup_{?} \{ \exists \vec{\boldsymbol{x}}_{i} \boldsymbol{c}_{i} \}_{i \geq 0} | (\emptyset; \boldsymbol{c}_{0}; \boldsymbol{A}) \longrightarrow (\vec{\boldsymbol{x}_{1}}; \boldsymbol{c}_{1}; \Gamma_{1}) \longrightarrow \dots \}$$

observing the set of success stores,

 $\mathcal{O}_{ss}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{ \exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}}; \boldsymbol{d}; \epsilon) \}$

observing the set of terminal stores (successes and suspensions),

$$\mathcal{O}_{ts}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{ \exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}}; \boldsymbol{d}; \Gamma) \not \rightarrow \}$$

observing the set of accessible stores,

$$\mathcal{O}_{\infty}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}_{0}) = \{ \sqcup_{?} \{ \exists \vec{\boldsymbol{x}}_{i} \boldsymbol{c}_{i} \}_{i \geq 0} | (\emptyset; \boldsymbol{c}_{0}; \boldsymbol{A}) \longrightarrow (\vec{\boldsymbol{x}_{1}}; \boldsymbol{c}_{1}; \Gamma_{1}) \longrightarrow \dots \}$$

observing the set of success stores,

 $\mathcal{O}_{ss}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{ \exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}}; \boldsymbol{d}; \epsilon) \}$

observing the set of terminal stores (successes and suspensions),

$$\mathcal{O}_{ts}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{ \exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}}; \boldsymbol{d}; \Gamma) \not \rightarrow \}$$

observing the set of accessible stores,

 $\mathcal{O}_{as}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}) = \{\exists \vec{\boldsymbol{x}} \boldsymbol{d} \in \mathcal{X} \mid (\emptyset;\boldsymbol{c};\boldsymbol{A}) \longrightarrow^{*} (\vec{\boldsymbol{x}};\boldsymbol{d};\Gamma)\}$

$$\mathcal{O}_{\infty}(\mathcal{D}.\boldsymbol{A};\boldsymbol{c}_{0}) = \{ \sqcup_{?} \{ \exists \vec{\boldsymbol{x}}_{i} \boldsymbol{c}_{i} \}_{i \geq 0} | (\emptyset; \boldsymbol{c}_{0}; \boldsymbol{A}) \longrightarrow (\vec{\boldsymbol{x}_{1}}; \boldsymbol{c}_{1}; \Gamma_{1}) \longrightarrow \dots \}$$

CC(*H*) 'append' Program(s)

Undirectional CLP style
Undirectional CLP style $append(A, B, C) = tell(A = []) \parallel tell(C = B)$ $+ tell(A = [X|L]) \parallel tell(C = [X|R]) \parallel append(L, B, R)$

Undirectional CLP style $append(A, B, C) = tell(A = []) \parallel tell(C = B)$ $+ tell(A = [X|L]) \parallel tell(C = [X|R]) \parallel append(L, B, R)$

Directional CC success store style

Undirectional CLP style $append(A, B, C) = tell(A = []) \parallel tell(C = B)$ $+ tell(A = [X|L]) \parallel tell(C = [X|R]) \parallel append(L, B, R)$

Directional CC success store style $append(A, B, C) = (A = [] \rightarrow tell(C = B))$ $+ \forall X, L \ (A = [X|L] \rightarrow tell(C = [X|R]) \parallel append(L, B, R))$

Undirectional CLP style $append(A, B, C) = tell(A = []) \parallel tell(C = B)$ $+ tell(A = [X|L]) \parallel tell(C = [X|R]) \parallel append(L, B, R)$

Directional CC success store style $append(A, B, C) = (A = [] \rightarrow tell(C = B))$ $+ \forall X, L \ (A = [X|L] \rightarrow tell(C = [X|R]) \parallel append(L, B, R))$

Directional CC terminal store style

Undirectional CLP style $append(A, B, C) = tell(A = []) \parallel tell(C = B)$ $+ tell(A = [X|L]) \parallel tell(C = [X|R]) \parallel append(L, B, R)$

Directional CC success store style $append(A, B, C) = (A = [] \rightarrow tell(C = B))$ $+ \forall X, L \ (A = [X|L] \rightarrow tell(C = [X|R]) \parallel append(L, B, R))$

Directional CC terminal store style $append(A, B, C) = A = [] \rightarrow tell(C = B)$ $\parallel \forall X, L \ (A = [X|L] \rightarrow tell(C = [X|R]) \parallel append(L, B, R))$

CC(H) 'merge' Program

Merging streams

$$\begin{split} merge(A, B, C) &= (A = [] \rightarrow tell(C = B)) \\ &+ (B = [] \rightarrow tell(C = A)) \\ &+ \forall X, L(A = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(L, B, R)) \\ &+ \forall X, L(B = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(A, L, R)) \end{split}$$

Good for the observable(s?)

Many-to-one communication: *client*(*C*1,...)

```
 \begin{array}{l} \dots \\ \textit{client}(\textit{Cn}, \dots) \\ \textit{server}([\textit{C1}, \dots, \textit{Cn}], \dots) = \\ \sum_{i=1}^{n} \forall \textit{X}, \textit{L}(\textit{Ci} = [\textit{X}|\textit{L}] \rightarrow \cdots \parallel \textit{server}([\textit{C1}, \dots, \textit{L}, \dots, \textit{Cn}], \dots) \end{array}
```

CC(H) 'merge' Program

Merging streams

$$\begin{split} merge(A, B, C) &= (A = [] \rightarrow tell(C = B)) \\ &+ (B = [] \rightarrow tell(C = A)) \\ &+ \forall X, L(A = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(L, B, R)) \\ &+ \forall X, L(B = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(A, L, R)) \end{split}$$

Good for the \mathcal{O}_{ss} observable

```
Many-to-one communication:

client(C1,...)

...

client(Cn,...)

server([C1,...,Cn],...) =

\sum_{i=1}^{n} \forall X, L(Ci = [X|L] \rightarrow \cdots \parallel server([C1,...,L,...,Cn],...)
```

CC(H) 'merge' Program

Merging streams

$$\begin{split} merge(A, B, C) &= (A = [] \rightarrow tell(C = B)) \\ &+ (B = [] \rightarrow tell(C = A)) \\ &+ \forall X, L(A = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(L, B, R)) \\ &+ \forall X, L(B = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(A, L, R)) \end{split}$$

Good for the \mathcal{O}_{ss} observable

can we get \mathcal{O}_{ts} ?

```
Many-to-one communication: client(C1,...)
```

```
 \begin{array}{l} \dots \\ \textit{client}(\textit{Cn}, \dots) \\ \textit{server}([\textit{C1}, \dots, \textit{Cn}], \dots) = \\ \sum_{i=1}^{n} \forall \textit{X}, \textit{L}(\textit{Ci} = [\textit{X}|\textit{L}] \rightarrow \cdots \parallel \textit{server}([\textit{C1}, \dots, \textit{L}, \dots, \textit{Cn}], \dots) \end{array}
```

Approximating ask condition with the Elimination condition

EL: $\boldsymbol{c} \wedge \Gamma \longrightarrow \Gamma$ if

Approximating ask condition with the Elimination condition

EL: $c \land \Gamma \longrightarrow \Gamma$ if $\mathcal{FD} \models c\sigma$ for every valuation σ of the variables in c by values of their domain.

Suppose access to *min* and *max* indexicals: $ask(X \ge Y + k)$

Approximating ask condition with the Elimination condition

EL: $c \land \Gamma \longrightarrow \Gamma$ if $\mathcal{FD} \models c\sigma$ for every valuation σ of the variables in c by values of their domain.

Suppose access to *min* and *max* indexicals: $ask(X \ge Y + k) \cong min(X) \ge max(Y) + k$

 $asknot(X \ge Y + k)$

Approximating ask condition with the Elimination condition

EL: $c \land \Gamma \longrightarrow \Gamma$ if $\mathcal{FD} \models c\sigma$ for every valuation σ of the variables in c by values of their domain.

Suppose access to *min* and *max* indexicals: $ask(X \ge Y + k) \cong min(X) \ge max(Y) + k$

 $asknot(X \ge Y + k) \cong max(X) < min(Y) + k$

 $ask(X \neq Y)$

Approximating ask condition with the Elimination condition

EL: $c \land \Gamma \longrightarrow \Gamma$ if $\mathcal{FD} \models c\sigma$ for every valuation σ of the variables in c by values of their domain.

Suppose access to *min* and *max* indexicals: $ask(X \ge Y + k) \cong min(X) \ge max(Y) + k$

 $asknot(X \ge Y + k) \cong max(X) < min(Y) + k$

 $ask(X \neq Y)$ $\cong max(X) < min(Y) \lor min(X) > max(Y)$ a better approximation with *dom*:

Approximating ask condition with the Elimination condition

EL: $c \land \Gamma \longrightarrow \Gamma$ if $\mathcal{FD} \models c\sigma$ for every valuation σ of the variables in c by values of their domain.

Suppose access to *min* and *max* indexicals: $ask(X \ge Y + k) \cong min(X) \ge max(Y) + k$

 $asknot(X \ge Y + k) \cong max(X) < min(Y) + k$

 $\begin{array}{ll} ask(X \neq Y) & \cong max(X) < min(Y) \lor min(X) > max(Y) \\ a \ better \ approximation \ with \ dom: \\ \cong (dom(X) \cap dom(Y) = \emptyset) \end{array}$

Basic constraints $(X \ge Y + k) =$

Basic constraints $(X \ge Y + k) = X \text{ in } min(Y) + k \dots \infty \parallel Y \text{ in } 0 \dots max(X) - k$

Reified constraints $(B \Leftrightarrow X = A) =$

Basic constraints $(X \ge Y + k) = X \text{ in } min(Y) + k \dots \infty \parallel Y \text{ in } 0 \dots max(X) - k$

Reified constraints $(B \Leftrightarrow X = A) = B \text{ in } 0..1 \parallel$

Basic constraints $(X \ge Y + k) = X \text{ in } min(Y) + k \dots \infty \parallel Y \text{ in } 0 \dots max(X) - k$

Reified constraints $(B \Leftrightarrow X = A) = B \text{ in } 0..1 \parallel$ $X = A \rightarrow B = 1 \parallel X \neq A \rightarrow B = 0 \parallel$ $B = 1 \rightarrow X = A \parallel B = 0 \rightarrow X \neq A$

Higher-order constraints card(N, L) =

Basic constraints $(X \ge Y + k) = X \text{ in } min(Y) + k \dots \infty \parallel Y \text{ in } 0 \dots max(X) - k$

Reified constraints $(B \Leftrightarrow X = A) = B \text{ in } 0..1 \parallel$ $X = A \rightarrow B = 1 \parallel X \neq A \rightarrow B = 0 \parallel$ $B = 1 \rightarrow X = A \parallel B = 0 \rightarrow X \neq A$

Higher-order constraints $card(N,L) = L = [] \rightarrow N = 0 \parallel$

Basic constraints $(X \ge Y + k) = X \text{ in } min(Y) + k \dots \infty \parallel Y \text{ in } 0 \dots max(X) - k$

Reified constraints $(B \Leftrightarrow X = A) = B \text{ in } 0..1 \parallel$ $X = A \rightarrow B = 1 \parallel X \neq A \rightarrow B = 0 \parallel$ $B = 1 \rightarrow X = A \parallel B = 0 \rightarrow X \neq A$

Higher-order constraints

$$card(N,L) = L = [] \rightarrow N = 0 \parallel L = [C|S] \rightarrow \exists B, M (B \Leftrightarrow C \parallel N = B + M \parallel card(M,S))$$

Andora Principle

"Always execute deterministic computation first".

Disjunctive scheduling:

deterministic propagation of the disjunctive constraints for which one of the alternatives is dis-entailed:

$$card(1, [x \ge y + d_y, y \ge x + d_x])$$

before creating choice points:

$$(x \ge y + d_y) + (y \ge x + d_x)$$

Constructive Disjunction in $CC(\mathcal{FD})$ (1)

$$\forall L \quad \frac{c \vdash_{\mathcal{X}} e \quad d \vdash_{\mathcal{X}} e}{c \lor d \vdash_{\mathcal{X}} e}$$

Intuitionistic logic tells us we can *infer the common information* to both branches of a disjunction without creating choice points!

$$\begin{array}{l} max(X,Y,Z)=(X>Y\parallel Z=X)+(X<=Y\parallel Z=Y)\\ \text{or}\\ max(X,Y,Z)=X>Y\rightarrow Z=X+X<=Y\rightarrow Z=Y.\\ \text{or}\\ max(X,Y,Z)=X>Y\rightarrow Z=X\ \parallel\ X<=Y\rightarrow Z=Y.\\ \text{better? (with indexicals)}\end{array}$$

Constructive Disjunction in $CC(\mathcal{FD})$ (1)

$$\forall L \quad \frac{c \vdash_{\mathcal{X}} e \quad d \vdash_{\mathcal{X}} e}{c \lor d \vdash_{\mathcal{X}} e}$$

Intuitionistic logic tells us we can *infer the common information* to both branches of a disjunction without creating choice points!

$$\begin{array}{l} max(X,Y,Z) = (X > Y \parallel Z = X) + (X <= Y \parallel Z = Y) \\ \text{or} \\ max(X,Y,Z) = X > Y \rightarrow Z = X + X <= Y \rightarrow Z = Y. \\ \text{or} \\ max(X,Y,Z) = X > Y \rightarrow Z = X \parallel X <= Y \rightarrow Z = Y. \\ \text{better? (with indexicals)} \\ max(X,Y,Z) = Z \text{ in } min(X)..\infty \parallel Z \text{ in } min(Y)..\infty \\ \parallel Z \text{ in } dom(X) \cup dom(Y) \parallel \cdots \end{array}$$

Constructive Disjunction in $CC(\mathcal{FD})$ (2)

Disjunctive precedence constraints disjunctive(T1, D1, T2, D2) = $(T1 \ge T2 + D2) + (T2 \ge T1 + D1)$

Using constructive disjunction

Constructive Disjunction in $CC(\mathcal{FD})$ (2)

Disjunctive precedence constraints disjunctive(T1, D1, T2, D2) = $(T1 \ge T2 + D2) + (T2 \ge T1 + D1)$

Using constructive disjunction

 $\begin{array}{l} \textit{disjunctive}(T1, D1, T2, D2) = \\ T1 \textit{ in } (0..max(T2) - D1) \cup (min(T2) + D2..\infty) \parallel \\ T2 \textit{ in } (0..max(T1) - D2) \cup (min(T1) + D1..\infty) \end{array}$

Part X

CC - Denotational Semantics

Part X: CC - Denotational Semantics







35 Non-deterministic Case



Deterministic CC

Agents:

$$A ::= \textit{tell}(c) \mid c \to A \mid A \parallel A \mid \exists xA \mid p(\vec{x})$$

- No choice operator
- Deterministic ask.

Replace non-deterministic pattern matching

 $\forall \vec{x}(c \rightarrow A)$

by deterministic ask and tell:

Deterministic CC

Agents:

$$A ::= \textit{tell}(c) \mid c \to A \mid A \parallel A \mid \exists xA \mid p(\vec{x})$$

- No choice operator
- Deterministic ask.

Replace non-deterministic pattern matching

 $\forall \vec{x}(c \rightarrow A)$

by deterministic ask and tell:

$$(\exists \vec{x}c) \to \exists \vec{x}(\textit{tell}(c) \parallel A)$$

Denotational semantics: input/output function

Input: initial store c_0 Output: terminal store c or *false* for infinite computations

Order the lattice of constraints (\mathcal{C}, \leq) by the information ordering: $\forall c, d \in \mathcal{C} \ c \leq d \text{ iff } d \vdash_{\mathcal{X}} c \text{ iff } \uparrow d \subset \uparrow c \text{ where } \uparrow c = \{d \in \mathcal{C} \mid c \leq d\}.$

$$\llbracket \mathcal{D}.A
rbracket : \mathcal{C}
ightarrow \mathcal{C}$$
 is

- **Q** Extensive: $\forall c \ c \leq [\mathcal{D}.A]c$
- **2** Monotone: $\forall c, d \ c \leq d \Rightarrow \llbracket \mathcal{D}.A \rrbracket c \leq \llbracket \mathcal{D}.A \rrbracket d$
- **3** Idempotent: $\forall c [[\mathcal{D}.A]]c = [[\mathcal{D}.A]]([[\mathcal{D}.A]]c)$
- i.e., $\llbracket \mathcal{D}.A \rrbracket$ is a over (\mathcal{C}, \leq) .

Denotational semantics: input/output function

Input: initial store c_0 Output: terminal store c or *false* for infinite computations

Order the lattice of constraints (\mathcal{C}, \leq) by the information ordering: $\forall c, d \in \mathcal{C} \ c \leq d \text{ iff } d \vdash_{\mathcal{X}} c \text{ iff } \uparrow d \subset \uparrow c \text{ where } \uparrow c = \{d \in \mathcal{C} \mid c \leq d\}.$

 $[\![\mathcal{D}.A]\!]:\mathcal{C}\to\mathcal{C}$ is

- **Q** Extensive: $\forall c \ c \leq [[\mathcal{D}.A]]c$
- **2** Monotone: $\forall c, d \ c \leq d \Rightarrow \llbracket \mathcal{D}.A \rrbracket c \leq \llbracket \mathcal{D}.A \rrbracket d$
- **3** Idempotent: $\forall c [[\mathcal{D}.A]]c = [[\mathcal{D}.A]]([[\mathcal{D}.A]]c)$
- i.e., $[\![\mathcal{D}.A]\!]$ is a closure operator over $(\mathcal{C},\leq).$

Closure Operators

Proposition 7

A closure operator f is characterized by the set of its fixpoints $\mathit{Fix}(f)$

Proof.

Closure Operators

Proposition 7

A closure operator f is characterized by the set of its fixpoints $\mathit{Fix}(f)$

Proof.

We show that $f = \lambda x.min(Fix(f) \cap \uparrow x)$.

Closure Operators

Proposition 7

A closure operator f is characterized by the set of its fixpoints $\mathit{Fix}(f)$

Proof.

We show that $f = \lambda x.min(Fix(f) \cap \uparrow x)$. Let y = f(x). By idempotence and extensivity, $y \in Fix(f) \cap \uparrow x$ By monotonicity $y = f(x) \le f(y')$ for any $y' \in \uparrow x$ Hence, if $y' \in Fix(f) \cap \uparrow x$ then $y \le y'$

Semantic Equations

Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

 $\llbracket \mathcal{D}.tell(c) \rrbracket \\ \llbracket \mathcal{D}.c \to A \rrbracket$

 $\begin{bmatrix} \mathcal{D}.A \parallel B \end{bmatrix} \\ \begin{bmatrix} \mathcal{D}.\exists xA \end{bmatrix} \\ \begin{bmatrix} \mathcal{D}. \exists x \end{bmatrix}$

$$\mathsf{if} \ \boldsymbol{p}(\vec{\boldsymbol{y}}) = \boldsymbol{A} \in \mathcal{D}$$

Theorem 8 ([SRP91popl])

For any deterministic process D.A

$$\mathcal{O}_{ts}(\mathcal{D}.A;c) = \begin{cases} \{min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & if \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & otherwise \end{cases}$$

Semantic Equations

Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

$$\begin{split} \llbracket \mathcal{D}.tell(c) \rrbracket &=\uparrow c & (\simeq \lambda s.s \land c) \\ \llbracket \mathcal{D}.c \to A \rrbracket & \\ \llbracket \mathcal{D}.A \parallel B \rrbracket & \\ \llbracket \mathcal{D}.\exists xA \rrbracket & \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket & \text{if } p(\vec{y}) = A \in \mathcal{D} \end{split}$$

Theorem 8 ([SRP91popl]) For any deterministic process $\mathcal{D}.A$ $(\{ \min([\mathcal{D}.A]] \cap \uparrow c) \}$ if $[\mathcal{D}.A]] \neq 0$

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \begin{cases} \{\min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & \text{if } \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & \text{otherwise} \end{cases}$$
Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

$$\begin{split} \llbracket \mathcal{D}.tell(c) \rrbracket &=\uparrow c & (\simeq \lambda s. s \land c) \\ \llbracket \mathcal{D}.c \to A \rrbracket &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ & (\simeq \lambda s. \text{ if } s \vdash_{\mathcal{C}} c \text{ then } \llbracket \mathcal{D}.A \rrbracket s \text{ else } s) \\ \llbracket \mathcal{D}.A \parallel B \rrbracket \\ \llbracket \mathcal{D}.\exists xA \rrbracket \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket & \text{ if } p(\vec{y}) = A \in \mathcal{D} \end{split}$$

Theorem 8 ([SRP91popl])

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \begin{cases} \{min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & if \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & otherwise \end{cases}$$

Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

$$\begin{split} \llbracket \mathcal{D}.tell(c) \rrbracket &=\uparrow c & (\simeq \lambda s. s \land c) \\ \llbracket \mathcal{D}.c \to A \rrbracket &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ \llbracket \mathcal{D}.A \parallel B \rrbracket &= \llbracket \mathcal{D}.A \rrbracket \cap \llbracket \mathcal{D}.B \rrbracket & (\simeq Y(\lambda s. \llbracket \mathcal{D}.A \rrbracket \llbracket \mathcal{D}.B \rrbracket s)) \\ \llbracket \mathcal{D}.\exists xA \rrbracket & \text{if } p(\vec{y}) = A \in \mathcal{D} \end{split}$$

Theorem 8 ([SRP91popl])

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \begin{cases} \{min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & if \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & otherwise \end{cases}$$

Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

$$\begin{split} \llbracket \mathcal{D}.tell(c) \rrbracket &=\uparrow c & (\simeq \lambda s.s \wedge c) \\ \llbracket \mathcal{D}.c \to A \rrbracket &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ & (\simeq \lambda s. \text{ if } s \vdash_c c \text{ then } \llbracket \mathcal{D}.A \rrbracket s \text{ else } s) \\ \llbracket \mathcal{D}.A \parallel B \rrbracket &= \llbracket \mathcal{D}.A \rrbracket \cap \llbracket \mathcal{D}.B \rrbracket & (\simeq Y(\lambda s. \llbracket \mathcal{D}.A \rrbracket \llbracket \mathcal{D}.B \rrbracket s)) \\ \llbracket \mathcal{D}.\exists xA \rrbracket &= \{d \mid c \in \llbracket \mathcal{D}.A \rrbracket, \exists xc = \exists xd\} (\simeq \lambda s. \exists x \llbracket \mathcal{D}.A \rrbracket \exists xs) \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket & \text{ if } p(\vec{y}) = A \in \mathcal{D} \end{split}$$

Theorem 8 ([SRP91popl])

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \begin{cases} \{min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & if \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & otherwise \end{cases}$$

Let $[]]: \mathcal{D} \times A \to \mathcal{P}(\mathcal{C})$ be a closure operator presented by the set of its fixpoints, and defined as the least fixpoint set of:

$$\begin{split} \llbracket \mathcal{D}.tell(c) \rrbracket &=\uparrow c & (\simeq \lambda 5.5 \land c) \\ \llbracket \mathcal{D}.c \to A \rrbracket &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ &= (\mathcal{C} \setminus \uparrow c) \cup (\uparrow c \cap \llbracket \mathcal{D}.A \rrbracket) \\ \llbracket \mathcal{D}.A \parallel B \rrbracket &= \llbracket \mathcal{D}.A \rrbracket \cap \llbracket \mathcal{D}.B \rrbracket & (\simeq Y(\lambda 5.\llbracket \mathcal{D}.A \rrbracket \mathcal{D}.A \rrbracket 5)) \\ \llbracket \mathcal{D}.\exists xA \rrbracket &= \{d \mid c \in \llbracket \mathcal{D}.A \rrbracket, \exists xc = \exists xd\} (\simeq \lambda 5.\exists x \llbracket \mathcal{D}.A \rrbracket \exists xs) \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket &= \llbracket \mathcal{D}.A[\vec{x}/\vec{y}] \rrbracket & \text{if } p(\vec{y}) = A \in \mathcal{D} \quad (\simeq \lambda 5.\llbracket \mathcal{D}.A \llbracket \vec{x}/\vec{y} \rrbracket 5)$$

Theorem 8 ([SRP91popl])

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \begin{cases} \{min(\llbracket \mathcal{D}.A \rrbracket \cap \uparrow c)\} & if \llbracket \mathcal{D}.A \rrbracket \neq \emptyset \\ \emptyset & otherwise \end{cases}$$

Constraint Propagation and Closure Operators

An environment $E: \mathcal{V} \rightarrow 2^D$ associates a domain of possible values to each variable.

Consider the lattice of environments (\mathcal{E}, \Box) , for the information ordering defined by $E \sqsubset E'$ if and only if $\forall x \in \mathcal{V}, E(x) \supseteq E'(x)$.

The semantics of a constraint propagator c can be defined as a closure operator over \mathcal{E} , noted \overline{c} , i.e., a mapping $\mathcal{E} \to \mathcal{E}$ satisfying

- **(extensivity)** $E \sqsubset \overline{c}(E)$,
- **2** (monotonicity) if $E \sqsubset E'$ then $\overline{c}(E) \sqsubset \overline{c}(E')$
- (idempotence) $\overline{c}(\overline{c}(E)) = \overline{c}(E)$.

Example in $CC(\mathcal{FD})$

Let b = (x > y) and c = (y > x).

Let $E(\mathbf{x}) = [1, 10]$, $E(\mathbf{y}) = [1, 10]$ be the initial environment

we have

$$\overline{b}E(\mathbf{x}) = [2, 10]$$
$$\overline{c}E(\mathbf{x}) = [1, 9]$$
$$(\overline{b} \sqcup \overline{c})E(\mathbf{x}) = [2, 9]$$

The closure operator $\overline{b,c}$ associated to the conjunction of constraints $b \land c$ gives the intended semantics:

$$\overline{b,c}E(x) = Y(\lambda s.\overline{b}(\overline{c}(s)))E(x) = \emptyset$$

Chaotic Iteration of Monotone Operators

Let $L(\Box, \bot, \top, \sqcup, \sqcap)$ be a complete lattice, and $F : L^n \to L^n$ a monotone operator over L^n with n > 0.

The chaotic iteration of *F* from $D \in L^n$ for a fair transfinite choice sequence $\langle J^{\delta} : \delta \in Ord \rangle$ is the sequence $\langle X^{\delta} \rangle$:

$$\begin{split} & X^0 = D, \\ & X_i^{\delta+1} = F_i(X^{\delta}) \text{ if } i \in J^{\delta}, X_i^{\delta+1} = X_i^{\delta} \text{ otherwise,} \\ & X_i^{\delta} = \bigsqcup_{\alpha < \delta} X_i^{\alpha} \text{ for any limit ordinal } \delta. \end{split}$$

Theorem 9 ([CC77popl])

Let $D \in L^n$ be a pre fixpoint of F (i.e., $D \sqsubset F(D)$). Any chaotic iteration of F starting from D is increasing and has for limit the least fixpoint of F above D.

Constraint Propagation as Chaotic Iteration

Corollary 10 (Correctness of constraint propagation)

Let $c = a_1 \land \dots \land a_n$, and E be an environment. Then $\overline{c}(E)$ is the limit of any fair iteration of closure operators $\overline{a}_1, \dots, \overline{a}_n$ from E.

Let $F: L^{n+1} \to L^{n+1}$ be defined by its projections F_i 's:

$$\begin{cases} E_1 = \overline{a}_1(E) = F_1(E_1, \dots, E_n, E) \\ E_2 = \overline{a}_2(E) = F_2(E_1, \dots, E_n, E) \\ \dots \\ E_n = \overline{a}_n(E) = F_n(E_1, \dots, E_n, E) \\ E = E_1 \cap \dots \cap E_n = F_{n+1}(E_1, \dots, E_n, E) \end{cases}$$

The functions F_i 's are obviously monotonic, any fair iteration of $\overline{a}_1, \ldots, \overline{a}_n$ is thus a chaotic iteration of F_1, \ldots, F_{n+1} therefore its limit is equal to the least fixpoint greater than E, i.e., $\overline{c}(E)$.

Denotational Semantics, Non-deterministic CC

Problem: the set of terminal stores of a CC process with one step guarded choice (i.e., *global choice*) is not compositional:

$$A = ask(x = a) \rightarrow tell(y = a)$$

+ ask(true) $\rightarrow tell(false)$
$$B = tell(x = a \land y = a)$$

A and B have the same set of terminal stores

but that is not the case for $\exists xB$ and $\exists xA$

Denotational Semantics, Non-deterministic CC

Problem: the set of terminal stores of a CC process with one step guarded choice (i.e., *global choice*) is not compositional:

$$A = ask(x = a) \rightarrow tell(y = a)$$

+ ask(true) $\rightarrow tell(false)$
$$B = tell(x = a \land y = a)$$

A and B have the same set of terminal stores

$$\uparrow \{ x = a \land y = a \}$$

(with global choice $\mathcal{C} \setminus \uparrow (x = a)$ is not a terminal store for A)

but that is not the case for $\exists xB$ and $\exists xA$

Denotational Semantics, Non-deterministic CC

Problem: the set of terminal stores of a CC process with one step guarded choice (i.e., *global choice*) is not compositional:

$$A = ask(x = a) \rightarrow tell(y = a)$$

+ ask(true) $\rightarrow tell(false)$
$$B = tell(x = a \land y = a)$$

A and B have the same set of terminal stores

$$\uparrow \{ x = a \land y = a \}$$

(with global choice $C \setminus \uparrow (x = a)$ is not a terminal store for A)

but that is not the case for $\exists xB$ and $\exists xA$

y = a is a terminal store for $\exists x B$ and not for $\exists x A$...

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
\begin{split} \llbracket tell(true) \rrbracket &= \\ \llbracket tell(true) + tell(c) \rrbracket &= \\ \mathcal{O}_{ts}(tell(true); true) &= \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \\ \end{split}
```

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
\begin{split} \llbracket tell(true) \rrbracket &= \mathcal{C} \\ \llbracket tell(true) + tell(c) \rrbracket &= \\ \mathcal{O}_{ts}(tell(true); true) &= \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \\ \end{split}
```

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
\begin{split} \llbracket tell(true) \rrbracket &= \mathcal{C} \\ \llbracket tell(true) + tell(c) \rrbracket &= \mathcal{C} \\ \mathcal{O}_{ts}(tell(true); true) &= \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \\ \end{split}
```

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
 \begin{split} \llbracket tell(true) \rrbracket &= \mathcal{C} \\ \llbracket tell(true) + tell(c) \rrbracket &= \mathcal{C} \\ \mathcal{O}_{ts}(tell(true); true) &= \{true\} \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \\ \end{split}
```

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
\begin{split} \llbracket tell(true) \rrbracket &= \mathcal{C} \\ \llbracket tell(true) + tell(c) \rrbracket &= \mathcal{C} \\ \mathcal{O}_{ts}(tell(true); true) &= \{true\} \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \{true, c\} \end{split}
```

The set of terminal stores of a CC process with blind choice can be characterized easily by adding the semantic equation: $[\![\mathcal{D}.A + B]\!] = [\![\mathcal{D}.A]\!] \cup [\![\mathcal{D}.B]\!]$

Theorem 11 ([BGP96sas]) $\llbracket \mathcal{D}.A \rrbracket = \bigcup_{c \in \mathcal{C}} \mathcal{O}_{ts}(\mathcal{D}.A; c)$

but the input-output relation cannot be recovered from $[\![\mathcal{D}.\mathcal{A}]\!]$:

```
\begin{split} \llbracket tell(true) \rrbracket &= \mathcal{C} \\ \llbracket tell(true) + tell(c) \rrbracket &= \mathcal{C} \\ \mathcal{O}_{ts}(tell(true); true) &= \{true\} \\ \mathcal{O}_{ts}(tell(true) + tell(c); true) &= \{true, c\} \end{split}
```

Idea: define $[\![]\!]:\mathcal{D}\times A\to \mathcal{P}(\mathcal{P}(\mathcal{C}))$ to distinguish between branches.

Let $[\![]\!]:\mathcal{D}\times \textbf{\textit{A}}\to \mathcal{P}(\mathcal{P}(\mathcal{C}))$ be the least fixpoint (for \subset) of

 $\llbracket \mathcal{D}.\mathcal{C} \rrbracket =$

Let $[\![]:\mathcal{D}\times \textbf{\textit{A}}\to \mathcal{P}(\mathcal{P}(\mathcal{C}))$ be the least fixpoint (for \subset) of

 $\begin{bmatrix} \mathcal{D}.\mathbf{C} \end{bmatrix} = \{\uparrow \mathbf{C}\}$ $\begin{bmatrix} \mathcal{D}.\mathbf{C} \to \mathbf{A} \end{bmatrix} =$

$$\begin{split} \llbracket \mathcal{D}.c \rrbracket &= \{\uparrow c\} \\ \llbracket \mathcal{D}.c \to A \rrbracket &= \{\mathcal{C} \setminus \uparrow c\} \cup \{\uparrow c \cap X | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \llbracket \mathcal{D}.A \parallel B \rrbracket &= \end{split}$$

$$\begin{bmatrix} \mathcal{D}.\mathbf{C} \end{bmatrix} = \{\uparrow \mathbf{C}\} \\ \begin{bmatrix} \mathcal{D}.\mathbf{C} \to \mathbf{A} \end{bmatrix} = \{\mathcal{C} \setminus \uparrow \mathbf{C}\} \cup \{\uparrow \mathbf{C} \cap \mathbf{X} | \mathbf{X} \in \llbracket \mathcal{D}.\mathbf{A} \rrbracket\} \\ \begin{bmatrix} \mathcal{D}.\mathbf{A} \parallel \mathbf{B} \end{bmatrix} = \{\mathbf{X} \cap \mathbf{Y} \mid \mathbf{X} \in \llbracket \mathcal{D}.\mathbf{A} \rrbracket, \ \mathbf{Y} \in \llbracket \mathcal{D}.\mathbf{B} \rrbracket\} \\ \llbracket \mathcal{D}.\mathbf{A} + \mathbf{B} \rrbracket = \begin{bmatrix} \mathbf{D}.\mathbf{A} \parallel \mathbf{A} \rrbracket \end{bmatrix} = \begin{bmatrix} \mathcal{D}.\mathbf{A} \parallel \mathbf{A} \rrbracket \end{bmatrix}$$

$$\begin{bmatrix} \mathcal{D}.c \end{bmatrix} = \{\uparrow c\} \\ \begin{bmatrix} \mathcal{D}.c \to A \end{bmatrix} = \{\mathcal{C} \setminus \uparrow c\} \cup \{\uparrow c \cap X | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \begin{bmatrix} \mathcal{D}.A \parallel B \end{bmatrix} = \{X \cap Y \mid X \in \llbracket \mathcal{D}.A \rrbracket, Y \in \llbracket \mathcal{D}.B \rrbracket\} \\ \begin{bmatrix} \mathcal{D}.A + B \end{bmatrix} = \llbracket \mathcal{D}.A \rrbracket \cup \llbracket \mathcal{D}.B \rrbracket \\ \end{bmatrix}$$

$$\begin{bmatrix} \mathcal{D}.c \end{bmatrix} = \{\uparrow c\} \\ \begin{bmatrix} \mathcal{D}.c \to A \end{bmatrix} = \{\mathcal{C} \setminus \uparrow c\} \cup \{\uparrow c \cap X | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \llbracket \mathcal{D}.A \parallel B \rrbracket = \{X \cap Y \mid X \in \llbracket \mathcal{D}.A \rrbracket, Y \in \llbracket \mathcal{D}.B \rrbracket\} \\ \llbracket \mathcal{D}.A + B \rrbracket = \llbracket \mathcal{D}.A \rrbracket \cup \llbracket \mathcal{D}.B \rrbracket \\ \llbracket \mathcal{D}.\exists xA \rrbracket = \{\{d \mid \exists xc = \exists xd, c \in X\} | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket =$$

Let $[\![]:\mathcal{D}\times \textbf{\textit{A}}\to \mathcal{P}(\mathcal{P}(\mathcal{C}))$ be the least fixpoint (for \subset) of

$$\begin{bmatrix} \mathcal{D}.c \end{bmatrix} = \{\uparrow c\} \\ \begin{bmatrix} \mathcal{D}.c \to A \end{bmatrix} = \{\mathcal{C} \setminus \uparrow c\} \cup \{\uparrow c \cap X | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \begin{bmatrix} \mathcal{D}.A \parallel B \end{bmatrix} = \{X \cap Y \mid X \in \llbracket \mathcal{D}.A \rrbracket, Y \in \llbracket \mathcal{D}.B \rrbracket\} \\ \begin{bmatrix} \mathcal{D}.A + B \end{bmatrix} = \llbracket \mathcal{D}.A \rrbracket \cup \llbracket \mathcal{D}.B \rrbracket \\ \begin{bmatrix} \mathcal{D}.\exists xA \rrbracket = \{\{d \mid \exists xc = \exists xd, c \in X\} | X \in \llbracket \mathcal{D}.A \rrbracket\} \\ \llbracket \mathcal{D}.p(\vec{x}) \rrbracket = \llbracket \mathcal{D}.A [\vec{x}/\vec{y}] \rrbracket$$

Theorem 12 ([FGMP97tcs])

For any process $\mathcal{D}.A$, $\mathcal{O}_{ts}(\mathcal{D}.A; c) = \{d | \text{ there exists } X \in \llbracket \mathcal{D}.A \rrbracket \text{ s.t. } d = min(\uparrow c \cap X)\}.$

'merge' Example Revisited

$$\begin{array}{l} \text{Merging streams} \\ \textit{merge}(A,B,C) = \\ & (A = [] \rightarrow \textit{tell}(C = B)) \parallel \\ & (B = [] \rightarrow \textit{tell}(C = A)) \parallel \\ & (\forall X, L(A = [X|L] \rightarrow \textit{tell}(C = [X|R]) \parallel \textit{merge}(L,B,R)) + \\ & \forall X, L(B = [X|L] \rightarrow \textit{tell}(C = [X|R]) \parallel \textit{merge}(A,L,R))) \end{array}$$

Do we have the expected terminal stores?

'merge' Example Revisited

Merging streams $merge(A, B, C) = (A = [] \rightarrow tell(C = B)) \parallel (B = [] \rightarrow tell(C = A)) \parallel (\forall X, L(A = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(L, B, R)) + \forall X, L(B = [X|L] \rightarrow tell(C = [X|R]) \parallel merge(A, L, R)))$

Do we have the expected terminal stores? No!

for merge(X, [1|Y], Z) we don't necessarily get 1 in Z, the merging is not *greedy*...

Sequentiality

Let us define a new operator, •, as follows:

$$\frac{(X; \boldsymbol{c}; \boldsymbol{A}) \longrightarrow (Y; \boldsymbol{d}; \boldsymbol{B})}{(X; \boldsymbol{c}; \boldsymbol{A} \bullet \boldsymbol{C}, \Gamma) \longrightarrow (Y; \boldsymbol{d}; \boldsymbol{B} \bullet \boldsymbol{C}, \Gamma)} \qquad (X; \boldsymbol{c}; \emptyset \bullet \boldsymbol{A}) \longrightarrow (X; \boldsymbol{c}; \boldsymbol{A})$$

We can characterize completely the observables of any CC_{seq} program, $\mathcal{D}.A$, by those of a new CC (without •) program, $\mathcal{D}^{\bullet}.A^{\bullet}$, in a new constraint system, \mathcal{C}^{\bullet} .

Idea

Let ok be a new relation symbol of arity one. C^{\bullet} is the constraint system C to which ok is added, without any non-logical axiom. The program $\mathcal{D}^{\bullet}.A^{\bullet}$ is defined inductively as follows:

$$(p(\vec{y}) = A)^{\bullet} = p^{\bullet}(x, \vec{y}) = A_x^{\bullet}$$

$$A^{\bullet} = \exists x A_x^{\bullet}$$

$$tell(c)_x^{\bullet} = tell(c \land ok(x))$$

$$p(\vec{y})_x^{\bullet} = p^{\bullet}(x, \vec{y})$$

$$(A \parallel B)_x^{\bullet} = \exists y, z(A_y^{\bullet} \parallel B_z^{\bullet} \parallel (ok(y) \land ok(z)) \rightarrow ok(x))$$

$$(A + B)_x^{\bullet} = A_x^{\bullet} + B_x^{\bullet}$$

$$\forall \vec{y}(c \rightarrow A))_x^{\bullet} = \forall \vec{z}(c[\vec{z}/\vec{y}] \rightarrow A[\vec{z}/\vec{y}]_x^{\bullet}) \text{ with } x \notin \vec{z}$$

$$(\exists y A)_x^{\bullet} = \exists z A[z/y]_x^{\bullet} \text{ with } z \neq x$$

$$(A \bullet B)_x^{\bullet} =$$

Idea

Let ok be a new relation symbol of arity one. C^{\bullet} is the constraint system C to which ok is added, without any non-logical axiom. The program $\mathcal{D}^{\bullet}.A^{\bullet}$ is defined inductively as follows:

$$(p(\vec{y}) = A)^{\bullet} = p^{\bullet}(x, \vec{y}) = A_{x}^{\bullet}$$

$$A^{\bullet} = \exists x A_{x}^{\bullet}$$

$$tell(c)_{x}^{\bullet} = tell(c \land ok(x))$$

$$p(\vec{y})_{x}^{\bullet} = p^{\bullet}(x, \vec{y})$$

$$(A \parallel B)_{x}^{\bullet} = \exists y, z(A_{y}^{\bullet} \parallel B_{z}^{\bullet} \parallel (ok(y) \land ok(z)) \rightarrow ok(x))$$

$$(A + B)_{x}^{\bullet} = A_{x}^{\bullet} + B_{x}^{\bullet}$$

$$\forall \vec{y}(c \rightarrow A))_{x}^{\bullet} = \forall \vec{z}(c[\vec{z}/\vec{y}] \rightarrow A[\vec{z}/\vec{y}]_{x}^{\bullet}) \text{ with } x \notin \vec{z}$$

$$(\exists y A)_{x}^{\bullet} = \exists y(A_{y}^{\bullet} \parallel ok(y) \rightarrow B_{x}^{\bullet})$$