

Typing Linear Constraints for Moding $\text{CLP}(\mathcal{R})$ Programs

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- Modes for logic programs assign to every predicate argument an input/output behavior.
 - Input: the predicate argument is ground on calls.
 - Output: the predicate argument is ground on answers.
 - Example: `:- mode append(in, in, out)`.
- Modes can be seen as *lightweight* specifications.
- Groundness is restrictive in the CLP(\mathcal{R}) context. Based on types, we want to extend the notion of moding to upper and/or lower bounds as well.

Definition (types)

A *type* is an element of $\mathcal{BT} = \{\star, \sqcup, \sqcap, \square, !\}$.

- $!$ is intended to type variables that show at most one single value in every solution, a property known as *definiteness*;
- \square is intended to type variables that assume a range of values (hence, lower and upper bounds exist);
- \sqcup (resp., \sqcap) is intended for variables that have a lower bound (resp., an upper bound);
- \star is to be used when no upper or lower bound can be stated.

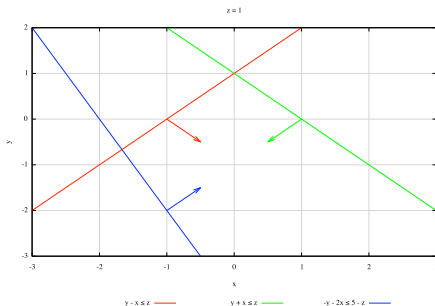
Definition (types assertions)

- An *atomic type declaration* (atd) is an expression $x : \tau$, where x is a variable and $\tau \in \mathcal{BT}$.
- We define $vars(x : \tau) = \{x\}$, and say that x is typed as τ .
- A *type declaration* is a sequence of atd's d_1, \dots, d_n , with $n \geq 0$. We define $vars(d_1, \dots, d_n) = \cup_{i=1..n} vars(d_i)$.
- A *type assertion* is an expression $\mathbf{d}_1 \vdash c \rightarrow \mathbf{d}_2$, where $\mathbf{d}_1, \mathbf{d}_2$ are type declarations and c is a linear constraint.

Example

$z : ! \vdash y - x \leq z, y + x \leq z, -y - 2x \leq 5 - z \rightarrow y : \sqcap, x : \sqcup$
 states that if z has a fixed value then either the set of solutions of the involved constraint is empty or the set of solutions is such that y has an upper bound and x has a lower bound.

The set of solutions for $z = 1$:

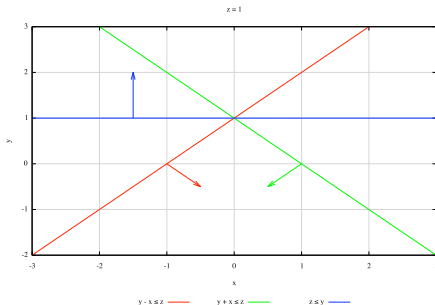


Example

$z \text{ :!} \vdash y - x \leq z, y + x \leq z, z \leq y \rightarrow y \text{ :!}, x \text{ :!}$

states that if z has a fixed value then either the set of solutions of the involved constraint is empty or both x and y assume a unique value in it.

The set of solutions for $z = 1$:



Definition (semantics)

We associate to an atd $d = x : \tau$ a formula $\phi(d)$ over fresh variables $v(d)$, called parameters, as follows:

$$\begin{array}{ll} \phi(x :!) = x = a & v(x :!) = \{a\} \\ \phi(x : \square) = a \leq x \wedge x \leq b & v(x : \square) = \{a, b\} \\ \phi(x : \sqcup) = a \leq x & v(x : \sqcup) = \{a\} \\ \phi(x : \sqcap) = x \leq b & v(x : \sqcap) = \{b\} \\ \phi(x : \star) = \text{true} & v(x : \star) = \emptyset. \end{array}$$

ϕ and v extend to type declarations as follows:

$$\phi(d_1, \dots, d_n) = \bigwedge_{i=1..n} \phi(d_i) \quad v(d_1, \dots, d_n) = \bigcup_{i=1..n} v(d_i).$$

A type assertion $\mathbf{d}_1 \vdash c \rightarrow \mathbf{d}_2$ is valid if for $\mathbf{v} = \text{vars}(c) \cup \text{vars}(\mathbf{d}_1) \cup \text{vars}(\mathbf{d}_2)$, the following formula is true in \mathcal{R} :

$$\forall v(\mathbf{d}_1) \exists v(\mathbf{d}_2) \forall \mathbf{v}. (\phi(\mathbf{d}_1) \wedge c) \rightarrow \phi(\mathbf{d}_2).$$

Example

For the type assertion

$$z :! \vdash y - x \leq z, y + x \leq z, z \leq y \rightarrow y :!, x :!$$

the formula to be proved is:

$$\forall a \exists b, c \forall x, y, z. (z = a \wedge y - x \leq z \wedge y + x \leq z \wedge z \leq y) \\ \rightarrow (y = b \wedge x = c).$$

- Such formulas can be checked by real quantifier elimination methods.
- It allows for generalizing to the *non-linear* case!
For instance: Mathematica, QEPCAD, Redlog.
- But we observe that our formulas represent a quite restricted class.
- Our approach switches from

the *logical* view of constraints-as-formulas
to
a *geometric* view of constraints-as-polyhedra.

Consider a linear constraint c and a type declaration \mathbf{d} .

- c can be equivalently represented as a linear system of inequalities $\mathbf{A}_c \mathbf{v} \leq \mathbf{b}_c$ where $\mathbf{v} = \text{vars}(c) \cup \text{vars}(\mathbf{d})$.
- The linear constraint $\phi(\mathbf{d})$ can be represented as $\mathbf{A}_d \mathbf{v} \leq \mathbf{B}_d \mathbf{a}_d$, where \mathbf{a}_d is the symbolic vector of parameters in $v(\mathbf{d})$.

The resulting system $\phi(\mathbf{d}) \wedge c$ is a parameterized system of linear inequalities \mathcal{P} , where variables in $v(\mathbf{d})$ play the role of parameters:

$$\begin{pmatrix} \mathbf{A}_c \\ \mathbf{A}_d \end{pmatrix} \mathbf{v} \leq \begin{pmatrix} \mathbf{b}_c \\ \mathbf{0} \end{pmatrix} + \begin{pmatrix} \mathbf{0} \\ \mathbf{B}_d \end{pmatrix} \mathbf{a}_d$$

Definition (Parameterized polyhedron)

A *parameterized polyhedron* is a collection of polyhedra defined by fixing the value for parameters in a parameterized system of linear inequalities: $Sol(\mathbf{Ax} \leq \mathbf{b} + \mathbf{Ba}, \mathbf{u}) = \{\mathbf{x} \mid \mathbf{Ax} \leq \mathbf{b} + \mathbf{Bu}\}$.

Example

Let \mathbf{d} be $z \neq 0$ and c be $y - x \leq z, y + x \leq z, -y - 2x \leq 5 - z$. We have that $\phi(\mathbf{d})$ is $z = a$, and $\phi(\mathbf{d}) \wedge c$ is:

$$\begin{pmatrix} -1 & 1 & -1 \\ 1 & 1 & -1 \\ -2 & -1 & 1 \\ 0 & 0 & 1 \\ 0 & 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} \leq \begin{pmatrix} 0 \\ 0 \\ 5 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ -1 \end{pmatrix} a$$

Let $x = \mathbf{v}_i$.

$\mathbf{d} \vdash c \rightarrow x : \tau$ is valid iff

for every \mathbf{u} , $\mathcal{S}_{\mathbf{u}} = \text{Sol}(\mathcal{P}, \mathbf{u})$ either is empty or:

- if $\tau = !$ then $\max\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} = \min\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} \in \mathcal{R}$, namely x assumes a single value;
- if $\tau = \square$ then $\max\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} \in \mathcal{R}$ and $\min\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} \in \mathcal{R}$ namely both an upper and a lower bound exist for x ;
- if $\tau = \sqcup$ then $\min\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} \in \mathcal{R}$, namely a lower bound exists for x ;
- if $\tau = \sqcap$ then $\max\{\mathbf{v}_i \mid \mathbf{v} \in \mathcal{S}_{\mathbf{u}}\} \in \mathcal{R}$, namely an upper bound exists for x ;
- if $\tau = \star$ then we have nothing to show!

From now on, we only consider satisfiable linear constraint (else one can infer assertions of the form $\mathbf{d} \vdash c \rightarrow x \text{ :!}$, for every x).

Lemma

Consider a parameterized polyhedron \mathcal{P} . Let \mathcal{H} be its homogeneous version: $\mathbf{A}_c \mathbf{v} \leq \mathbf{0}$, $\mathbf{A}_d \mathbf{v} \leq \mathbf{0}$.

$\max\{\mathbf{c}^T \mathbf{v} \mid \mathbf{v} \in \text{Sol}(\mathcal{H})\} = 0$ iff for every parameter instance \mathbf{u} , $\text{Sol}(\mathcal{P}, \mathbf{u}) = \emptyset$ or $\max\{\mathbf{c}^T \mathbf{v} \mid \mathbf{v} \in \text{Sol}(\mathcal{P}, \mathbf{u})\} \in \mathcal{R}$.

- When \mathbf{c} is always 0 except for the i^{th} position where it is 1, we have $\mathbf{c}^T \mathbf{v} = v_i$.
- The previous lemma solves the problem of deciding whether $\mathbf{d} \vdash c \rightarrow \mathbf{v}_i : \square$, without having to take into account parameters.
- By reasoning similarly for types \sqcup and \square , we can state an effective procedure, called `LPCHECK`

Input: a type assertion \mathbf{d}_1 , a constraint c and a seq. of vars \mathbf{x} .

Step 0 Define $\mathbf{v} = \text{vars}(c)$, $\mathbf{n} = \text{nf}(\mathbf{d}_1)$, $\mathbf{d} = \mathbf{n}|_{\mathbf{v}}$.

Step 1 Let $\mathbf{A}_c \mathbf{v} \leq \mathbf{b}_c$ be the geometric rep. of c .
Let $\mathbf{A}_d \mathbf{v} \leq \mathbf{B}_d \mathbf{a}_d$ the geometric rep. of $\phi(\mathbf{d})$.

Step 2 If $\text{Sol}(\mathbf{A}_c \mathbf{v} \leq \mathbf{b}_c) = \emptyset$ Then for every x in \mathbf{x} , output “ x :!”

Else

Step 3 for every x in $\mathbf{x} \setminus \mathbf{v}$ and $x : \tau$ in \mathbf{n} , output “ $x : \tau$ ”;

Step 4 for every x in $\mathbf{x} \cap \mathbf{v}$:

- (a) Let $M = \max\{x \mid \mathbf{A}_c \mathbf{v} \leq \mathbf{0}, \mathbf{A}_d \mathbf{v} \leq \mathbf{0}\}$,
 $m = \max\{-x \mid \mathbf{A}_c \mathbf{v} \leq \mathbf{0}, \mathbf{A}_d \mathbf{v} \leq \mathbf{0}\}$.
- (b) Output “ $x : \square$ ” if $M = 0$ and $m = 0$;
- (c) Output “ $x : \sqcup$ ” if $M = \infty$ and $m = 0$;
- (d) Output “ $x : \sqcap$ ” if $M = 0$ and $m = \infty$;
- (e) Else output “ $x : \star$ ”.

Figure: LPCHECK, sound and complete w.r.t. $\{\star, \sqcup, \sqcap, \square\}$.

Minkowski, Motzkin, 1953:

Theorem (Minkowski's decomposition thm)

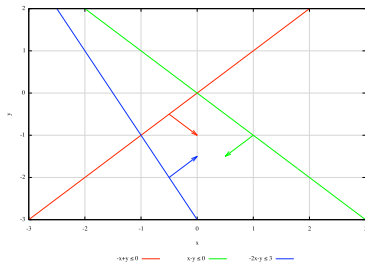
There exists an effective procedure that given $\mathbf{Ax} \leq \mathbf{b}$ decides whether or not the polyhedron $Sol(\mathbf{Ax} \leq \mathbf{b})$ is empty and, if not, it yields a generating matrix \mathbf{R} and a vertex matrix \mathbf{V} such that:

$$Sol(\mathbf{Ax} \leq \mathbf{b}) = \{ \mathbf{x} \mid \mathbf{x} = \mathbf{R}\lambda, \lambda \geq 0 \} + \{ \mathbf{x} \mid \mathbf{x} = \mathbf{V}\gamma, \gamma \geq 0, \Sigma\gamma = 1 \}$$

$$Sol(\mathbf{Ax} \leq \mathbf{0}) = \{ \mathbf{x} \mid \mathbf{x} = \mathbf{R}\lambda, \lambda \geq 0 \}$$

Example

$$\begin{pmatrix} -1 & 1 \\ 1 & 1 \\ -2 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \leq \begin{pmatrix} 0 \\ 0 \\ 3 \end{pmatrix}$$



$$\mathbf{R} = \begin{pmatrix} 1 & 1 \\ -2 & -1 \end{pmatrix} \quad \mathbf{v} = \begin{pmatrix} 0 & -1 \\ 0 & -1 \end{pmatrix}$$

Loechner and Wilde, 1997:

Theorem (Minkowski's thm for parameterized polyhedra)

Every parameterized polyhedron can be expressed by a generating matrix \mathbf{R} and finitely many pairs

$$(\mathbf{v}^a(1), \mathbf{C}_1 \mathbf{a} \leq \mathbf{c}_1), \dots, (\mathbf{v}^a(k), \mathbf{C}_k \mathbf{a} \leq \mathbf{c}_k)$$

where, for $i = 1..k$, $\mathbf{v}^a(i)$ is a vector parametric in \mathbf{a} and $Sol(\mathbf{C}_i \mathbf{a} \leq \mathbf{c}_i) \neq \emptyset$, as follows:

$$Sol(\mathbf{A}\mathbf{x} \leq \mathbf{b} + \mathbf{B}\mathbf{a}, \mathbf{u}) = \{ \mathbf{x} \mid \mathbf{x} = \mathbf{R}\boldsymbol{\lambda}, \boldsymbol{\lambda} \geq 0 \} \\ + ConvexHull(\{ \mathbf{v}^u(i) \mid i = 1..k, \mathbf{C}_i \mathbf{u} \leq \mathbf{c}_i \})$$

and

$$Sol(\mathbf{A}\mathbf{x} \leq \mathbf{0}) = \{ \mathbf{x} \mid \mathbf{x} = \mathbf{R}\boldsymbol{\lambda}, \boldsymbol{\lambda} \geq 0 \}$$

.

Example

$$a + b \geq y, y \geq a, y \geq b, x = a$$

$$\begin{pmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \\ 0 & -1 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \leq \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ -1 & 0 \\ 1 & 1 \\ -1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix}$$

$$\mathbf{R} = \mathbf{0}$$

$$\left(\begin{pmatrix} a \\ b \end{pmatrix}, b \geq a \geq 0 \right) \left(\begin{pmatrix} a \\ a \end{pmatrix}, a \geq b \geq 0 \right) \left(\begin{pmatrix} a \\ a+b \end{pmatrix}, a, b \geq 0 \right)$$

Lemma

Consider the Minkowski's form of a non-empty parameterized polyhedron.

Every $S_{\mathbf{u}} = \{\mathbf{c}^T \mathbf{x} \mid \mathbf{x} \in \text{Sol}(\mathbf{A}\mathbf{x} \leq \mathbf{b} + \mathbf{B}\mathbf{a}, \mathbf{u})\}$ is empty or a singleton iff

- $\mathbf{c}^T \mathbf{R} = \mathbf{0}$
- for $1 \leq m < n \leq k$, $\mathbf{C}_m \mathbf{a} \leq \mathbf{c}_m$, $\mathbf{C}_n \mathbf{a} \leq \mathbf{c}_n \models \mathbf{c}^T \mathbf{v}^{\mathbf{a}}(m) = \mathbf{c}^T \mathbf{v}^{\mathbf{a}}(n)$.

Example

Consider $a + b \geq y, y \geq a, y \geq b, x = a$, and the corresponding parameterized polyhedron defined by $\mathbf{R} = \mathbf{0}$ and

$$\left(\begin{pmatrix} a \\ b \end{pmatrix}, b \geq a \geq 0 \right) \left(\begin{pmatrix} a \\ a \end{pmatrix}, a \geq b \geq 0 \right) \left(\begin{pmatrix} a \\ a+b \end{pmatrix}, a, b \geq 0 \right).$$

Let us reason about definiteness of variables x and y by using the previous lemma

- x is definite, since $a = a$ over any polyhedron.
- y is not definite:
 - fine for the first and second vertex:

$$b \geq a \geq 0, a \geq b \geq 0 \models a = b$$
 - but for the first and third vertex:

$$b \geq a \geq 0, a, b \geq 0 \not\models b = a + b$$

Input: a type assertion \mathbf{d}_1 , a constraint c and a seq. of vars \mathbf{x} .

Step 0 Define $\mathbf{v} = \text{vars}(c)$, $\mathbf{n} = \text{nf}(\mathbf{d}_1)$, $\mathbf{d} = \mathbf{n}|_{\mathbf{v}}$.

Step 1 Let $\mathbf{A}_c \mathbf{v} \leq \mathbf{b}$ be the geometric rep. of c , and $\mathbf{A}_d \mathbf{v} \leq \mathbf{B}_d \mathbf{a}_d$ the geometric rep. of $\phi(\mathbf{d})$.

Step 1 For the parametric polyh. $\mathbf{A}_c \mathbf{v} \leq \mathbf{b}$, $\mathbf{A}_d \mathbf{v} \leq \mathbf{B}_d \mathbf{a}_d$ build the generating matrix \mathbf{R} and the sequence $(\mathbf{v}^a(1), \mathbf{C}_1 \mathbf{a} \leq \mathbf{c}_1), \dots, (\mathbf{v}^a(k), \mathbf{C}_k \mathbf{a} \leq \mathbf{c}_k)$

Step 2 For every $x : \tau$ as output from LPCHECK

Step 3 If $\tau \neq \square$ or $x \notin \mathbf{v}$ then output “ $x : \tau$ ”;

Step 4 Else let i such that $x = \mathbf{v}_i$:

(a) Output “ $x : !$ ” if $\text{row}(\mathbf{R}, i) = \mathbf{0}$ and for $1 \leq m < n \leq k$, $\mathbf{C}_m \mathbf{a} \leq \mathbf{c}_m$,

$\mathbf{C}_n \mathbf{a} \leq \mathbf{c}_n \models \mathbf{v}^a(m)_i = \mathbf{v}^a(n)_i$;

(b) Output “ $x : \square$ ” otherwise.

Figure: POLYCHECK, sound and complete w.r.t. $\{\star, \sqcup, \sqcap, \square, !\}$.

Definition (well-moding)

Let P be a CLP(\mathcal{R}) program. A clause of P :

$p_0(\mathbf{x}_0 : \mu_0 \times \tau_{n+1}) \leftarrow c, p_1(\mathbf{x}_1 : \tau_1 \times \mu_1), \dots, p_n(\mathbf{x}_n : \tau_n \times \mu_n)$
is well-moded if for $i = 1..n + 1$, the type assertion

$$\mathbf{x}_0 : \mu_0, \dots, \mathbf{x}_{i-1} : \mu_{i-1} \vdash c \rightarrow \mathbf{x}_i : \tau_i$$

is valid. P is well-moded if all its clauses are well-moded.

A well-moded program well behaves at run-time.

program	C	A	modes	time
ack	3	6	ack($\star \times \square$, $! \times !$, $\star \times !$)	0.0011
ack	3	6	ack($\star \times \square$, $\square \times \square$, $\star \times \square$)	0.0007
fib	2	4	fib($! \times !$, $\star \times !$)	0.0011
fib	2	4	fib($\star \times \star$, $\star \times !$)	0.0007
mc91	2	4	mc($\square \times \square$, $\star \times \square$)	0.0005
mc91	2	4	mc($! \times !$, $\star \times !$)	0.0006
mortgage	2	3	mortgage($! \times !$, $\square \times !$, $! \times !$, $\star \times !$)	0.0010
schedule	10	21	schedule($! \times !$, $\star \times \square$, $\star \times \square$)	0.0021
tak	3	8	tak($\star \times !$)	0.0015

Table: Time in seconds, Xeon 2.8Ghz, C: # clauses, A: # atoms.

- sources and extended technical report at:
<http://www.di.unipi.it/~ruggieri/software/>
- extensions:
 - $\square_r, r \in \mathcal{R}^+$
 - type inference and terminating modes
 - $CLP(Term + \mathcal{R})$