

An Integer Set Library for Program Analysis

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Program Analysis and Transformation

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manipulation of *arrays* inside *loops*
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⇒ need for *compact representation* of
iterations of a loop/elements of an array

⇒ + *efficient to manipulate*

⇒ Integer points in polyhedra (“polyhedral model”)

⇒ More generally: sets of integers bounded by affine inequalities

Representation Example: Iteration Domain

```
#define N 5
for (i = 1; i <= N; ++i)
    for (j = 1; j <= i; ++j)
        a[i][j]=
```

Assumptions on sequential code:

- ▶ iterators are integers
- ▶ loops with affine bounds
- ▶ affine conditions
- ▶ affine index expressions

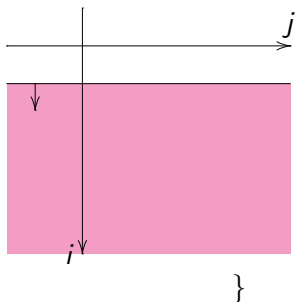
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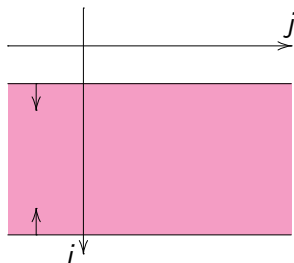
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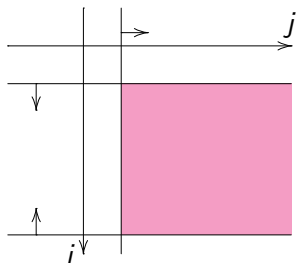
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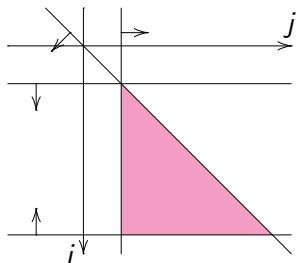
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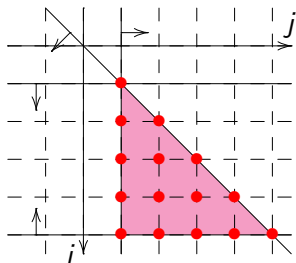
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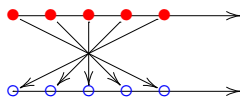
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for (i = 0; i <= N; ++i)
    a[i] = ...
for (i = 0; i <= N; ++i)
    b[i] = f(a[N-i])
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Execution order: top-down, left-right

- , ○ Statement iteration
- Data flow dependence
- Executed statement
- Data in memory

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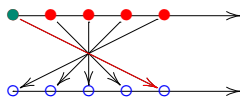


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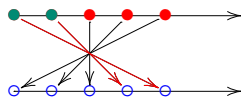


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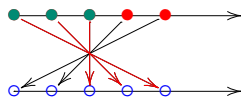


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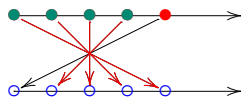


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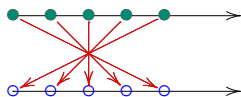


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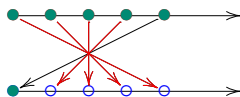


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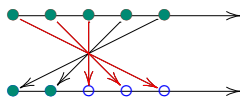


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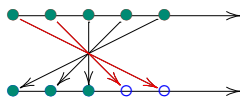


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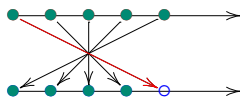


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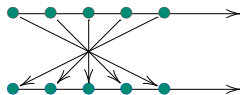


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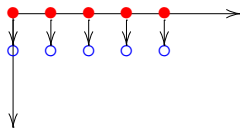
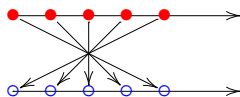
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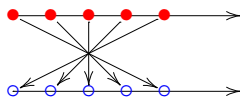
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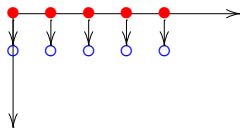


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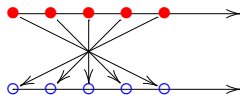


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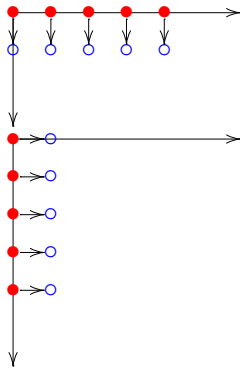


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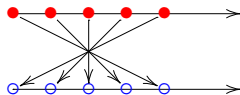


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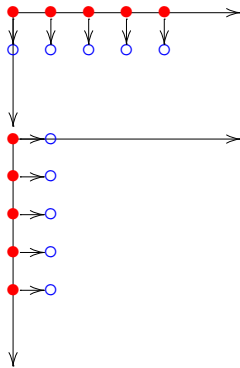


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Two Program Analysis and Transformation Tools

Why do we need an integer set library?

- ▶ Equivalence checker
 - ▶ Checks the equivalence of two programs represented in the polyhedral model
 - ▶ Proves output is the same given that input is the same
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Requirements

- ▶ manipulations on *integer* sets/maps
- ▶ explicit support for existentially quantified variables

Equivalence Checking Example

Given $in1 == in2$, can we prove $out1 == out2$?

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a1[0] = in1;
```

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for (i = 1; i <= N; ++i)
```

```
    a1[i] = f(a1[i - 1]);
```

```
out1 = a1[N];
```

```
a2[0] = in2;
```

```
for (i = 1; i <= N; ++i)
```

```
    a2[i] = f(a2[i - 1]);
```

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out2 = a2[N];
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\Rightarrow requires $a1[i] == a2[i]$ for $1 \leq i \leq N$

\Rightarrow induction for $2 \leq i \leq N$ + requires $a1[0] = a2[0]$

\Rightarrow **Integer affine hull** of $\{(N, N), (N-1, N-1)\}$: $\{(i, i)\}$

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Requirements

- ▶ manipulations on *integer* sets/maps
- ▶ explicit support for existentially quantified variables
- ▶ CLooG
 - ▶ Generates code for scanning integer points in polyhedra (iteration domains)

Requirements

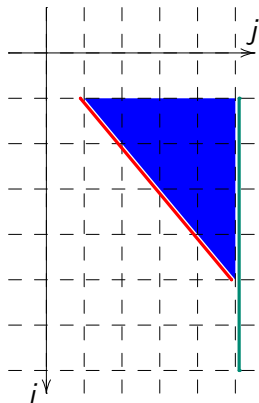
- ▶ manipulations on *integer* sets
 - ⇒ remove redundant constraints/code
- ▶ explicit support for existentially quantified variables
 - ⇒ replace some loops by guards

CLooG Example

$$S1: \{(i, j) \mid 1 \leq i \leq n \leq m \wedge j = i\}$$

$$S2: \{(i, j) \mid 1 \leq i \leq n \leq m \wedge i \leq j \leq n\}$$

$$S3: \{(i, j) \mid 1 \leq i \leq m \wedge j = n \leq m\}$$

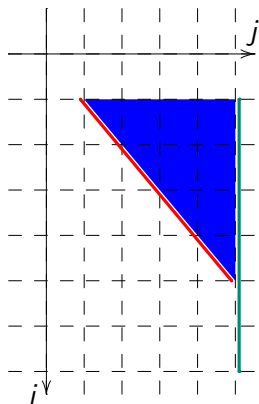


CLooG Example

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S3: $\{(i, j) \mid 1 \leq i \leq m \wedge j = n \leq m\}$



```
for (i=1; i<=m; i++) {  
    if (i <= n) {  
        S1(i, i);  
    }  
    for (j=i; j<=n; j++) {  
        S2(i);  
    }  
    S3(i, n);  
}
```

Required Operations

- ▶ Basic operations
 - ▶ Union
 - ▶ Intersection
 - ▶ Set difference
 - ▶ ...
- ▶ Operations required by equivalence checking
 - ▶ Integer affine hull
 - ▶ ...
- ▶ Operations required by CLoG
 - ▶ Projection (rational)
 - ▶ Ordering
 - ▶ Convex hull (rational)
 - ▶ Simplification
 - ▶ ...

Why not use a double description based library?

E.g., PolyLib, PPL, (polymake)

- ▶ Who needs vertices anyway?
 - ▶ Very useful for LattE macchiato/barvinok style counting (but neither equivalence checking or CLoog needs any counting)
 - ▶ Some operations can be performed more efficiently on explicit representation

But:

- ▶ Computing the dual can be costly
- ▶ Double description requires more space

⇒ trade-off

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- ⇒ trade-off
- (sets used in equivalence checking and CLoog usually have few constraints)
- ▶ Usually focus on *rational* values
 - ▶ Little/no support for existentially quantified variables

Why do we need existentially quantified variables?

- ▶ Modeling some problems

Which array elements are accessed in this loop?

```
for (j = 1; j <= p; ++j)
  for (i = 1; i <= 8; ++i)
    a[6i+9j-7] = a[6i+9j-7] + 5;
```

$$S(s) = \{l \in \mathbb{Z} \mid \exists i, j \in \mathbb{Z} : l = 6i + 9j - 7 \wedge 1 \leq j \leq s \wedge 1 \leq i \leq 8\}$$

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- ▶ Especially integer divisions/reminders

E.g., $i \% 10 \leq 6$

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Why do we need existentially quantified variables?

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Which array elements are accessed in this loop?

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for (j = 1; j <= p; ++j)
  for (i = 1; i <= 8; ++i)
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- ▶ May appear in original code
- ▶ May be introduced by (PIP-based) dependence analysis

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```
# Omega Calculator v2.1 (based on Omega Library 2.1, Ju
```

```
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```

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```

⇒ completely unacceptable for equivalence checking

Internal Representation

$$S(\mathbf{s}) = \{ \mathbf{x} \in \mathbb{Z}^d \mid \exists \mathbf{z} \in \mathbb{Z}^e : \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{s} + \mathbf{D}\mathbf{z} \geq \mathbf{c} \}$$

$$R(\mathbf{s}) = \{ (\mathbf{x}_1, \mathbf{x}_2) \in \mathbb{Z}^{d_1} \times \mathbb{Z}^{d_2} \mid \exists \mathbf{z} \in \mathbb{Z}^e : \mathbf{A}_1\mathbf{x}_1 + \mathbf{A}_2\mathbf{x}_2 + \mathbf{B}\mathbf{s} + \mathbf{D}\mathbf{z} \geq \mathbf{c} \}$$

- ▶ “basic” types: “convex” sets and maps (relations)
 - ▶ equality + inequality constraints
 - ▶ parameters \mathbf{s}
 - ▶ (optional) explicit representation of existentially quantified variables as integer divisions
 - ⇒ useful for aligning dimensions when performing set operations (e.g., set difference)
 - ⇒ can be computed using PIP
 - ⇒ already available if obtained from PIP-based dependence analysis
- ▶ union types: sets and maps
 - ⇒ (disjoint) unions of basic sets/maps

Parametric Integer Programming

$$R(\mathbf{s}) = \{ (\mathbf{x}_1, \mathbf{x}_2) \in \mathbb{Z}^{d_1} \times \mathbb{Z}^{d_2} \mid \exists \mathbf{z} \in \mathbb{Z}^e : A_1 \mathbf{x}_1 + A_2 \mathbf{x}_2 + B\mathbf{s} + D\mathbf{z} \geq \mathbf{c} \}$$

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$$\text{lexmin } R = \{ (\mathbf{x}_1, \mathbf{x}_2) \in R \mid \forall \mathbf{x}'_2 \in R(\mathbf{s}, \mathbf{x}_1) : \mathbf{x}_2 \preceq \mathbf{x}'_2 \}$$

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- ▶ explicit representation of range variables

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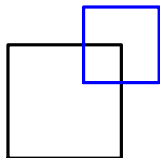
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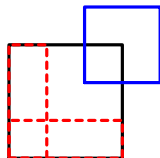
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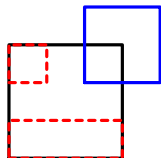
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- ▶ with existentially quantified variables
⇒ compute explicit representation

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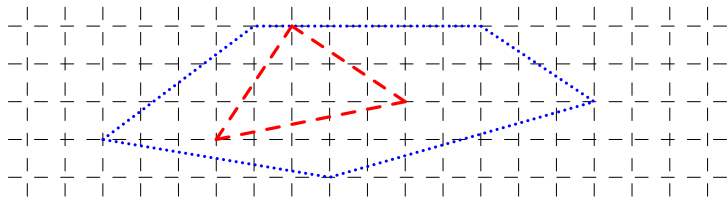
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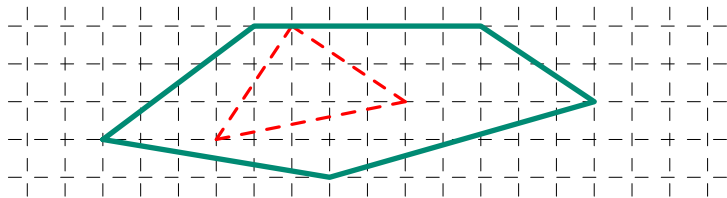
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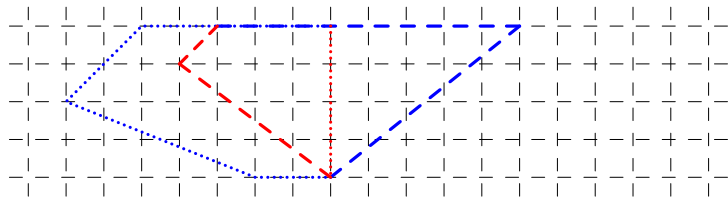
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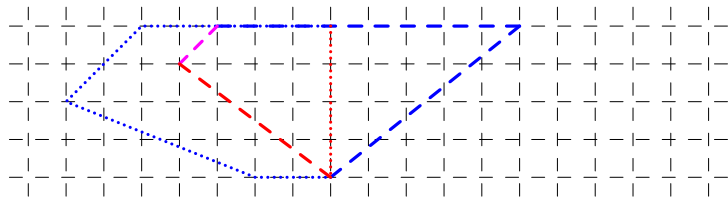
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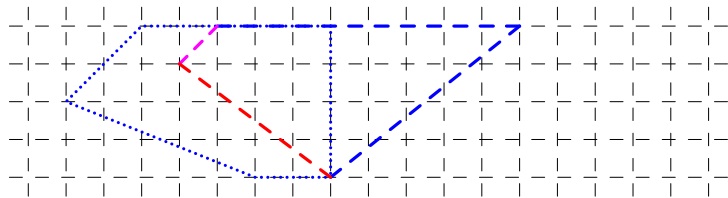
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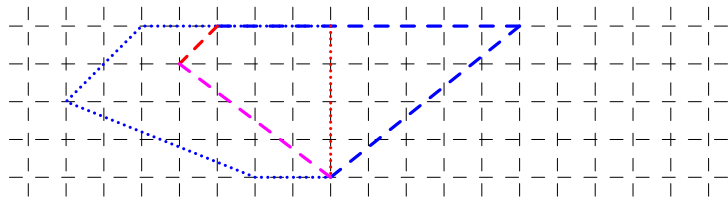
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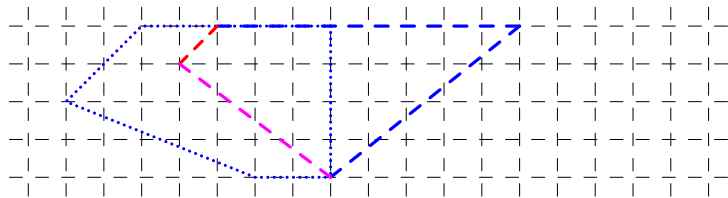
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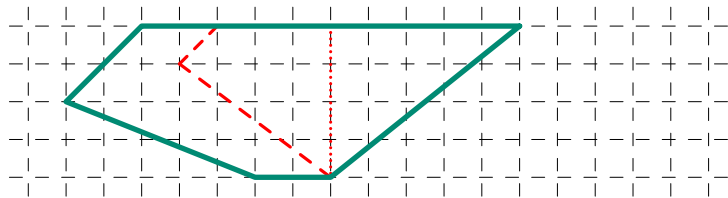
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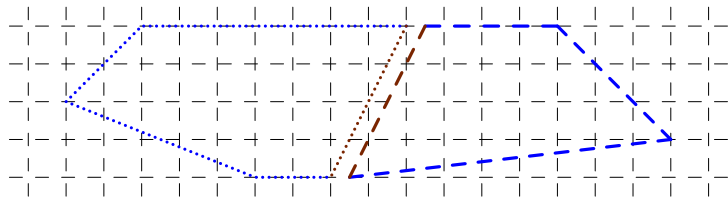
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- ▶ no separating constraints and cut constraints of S_2 are valid for cut facets of S_1 (similar to BFT2001)
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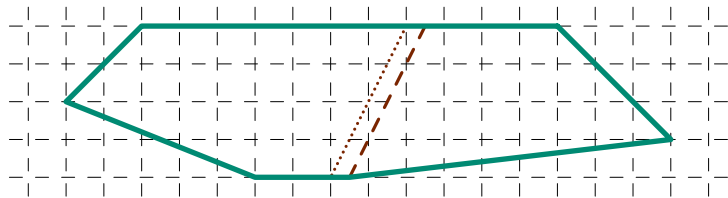
isl Operation: Set Coalescing



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is1 Operation: Set Coalescing

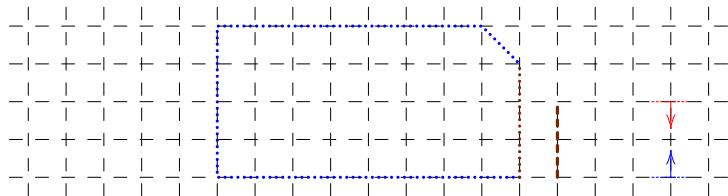
1. Classify constraints

- ▶ redundant: $\min \langle \mathbf{a}_i, \mathbf{x} \rangle > c_i - 1$ over remaining constraints
- ▶ valid: $\min \langle \mathbf{a}_i, \mathbf{x} \rangle > c_i - 1$ over S_2
- ▶ separating: $\max \langle \mathbf{a}_i, \mathbf{x} \rangle < c_i$ over S_2
special cases:
 - ▶ adjacent to equality: $\langle \mathbf{a}_i, \mathbf{x} \rangle = c_i - 1$ over S_2
 - ▶ adjacent to inequality: $\langle (\mathbf{a}_i + \mathbf{b}_j), \mathbf{x} \rangle = (c_i + d_j) - 1$ over S_2
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- ▶ single adjacent pair of an inequality (S_1) and an equality (S_2)
+ other constraints of S_1 are valid
+ constraints of S_2 valid for facet of relaxed inequality
 \Rightarrow drop S_2 and relax adjacent inequality of S_1

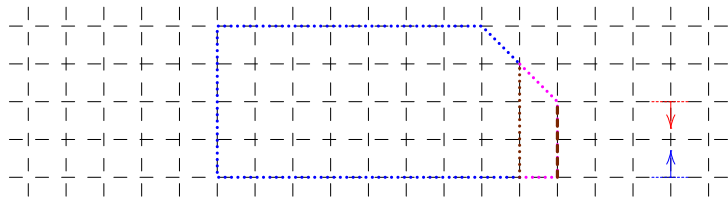
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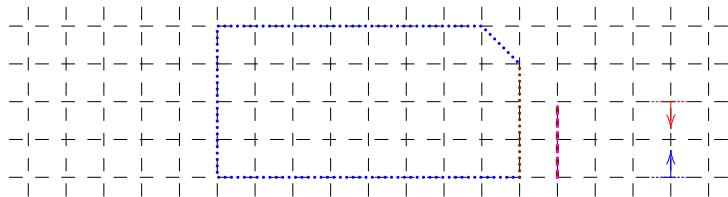
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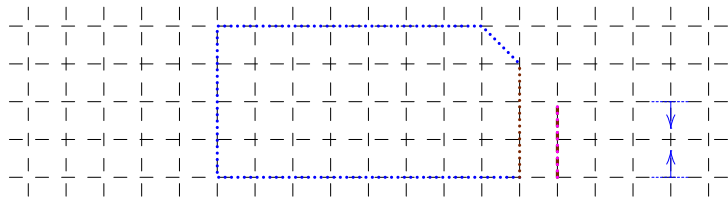
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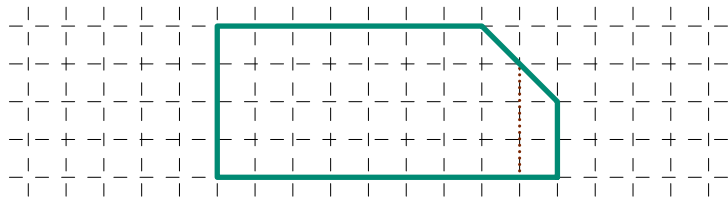
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is1 Operation: Closed Convex Hull

$$H = \text{conv.hull}(S_1 \cup S_2) \quad S_1 = \{ \mathbf{x} \mid A\mathbf{x} \geq \mathbf{c} \} \quad S_2 = \{ \mathbf{x} \mid B\mathbf{x} \geq \mathbf{d} \}$$

1. using elimination

- ▶ convex hull of polyhedra
⇒ sum of cones in homogeneous space

$$H = \{ \mathbf{x} \mid \exists \mathbf{x}_1, \mathbf{x}_2, z_1, z_2 : \mathbf{x} = \mathbf{x}_1 + \mathbf{x}_2 \wedge \mathbf{1} = z_1 + z_2 \wedge \\ A\mathbf{x}_1 \geq \mathbf{c}z_1 \wedge z_1 \geq 0 \wedge B\mathbf{x}_2 \geq \mathbf{d}z_2 \wedge z_2 \geq 0 \}$$

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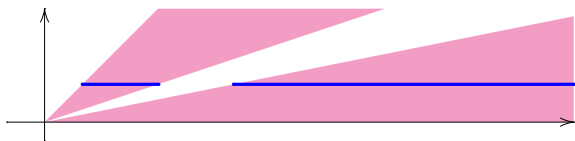
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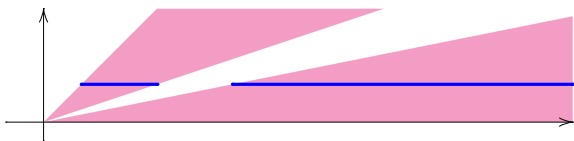
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⇒ very inefficient!

2. using “wrapping”

- ▶ S_1 and S_2 are polytopes
⇒ wrap facets around ridges until all facets found (FLL2000)
- ▶ H is pointed
⇒ change perspective
- ▶ S_1 and S_2 are pointed (R_i recession cone of S_i)
⇒ project out lineality $H = \text{lin.hull}(R_1 \cap -R_2)$
- ▶ S_1 or S_2 has non-trivial lineality space
⇒ project out lineality S_1 and lineality S_2

is1 Operation: Closed Convex Hull—Wrapping

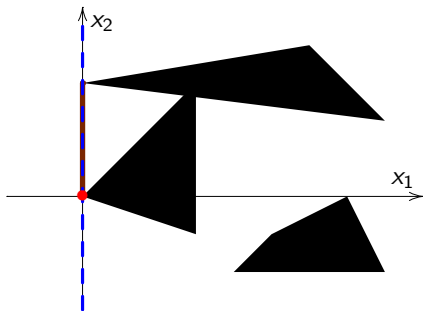
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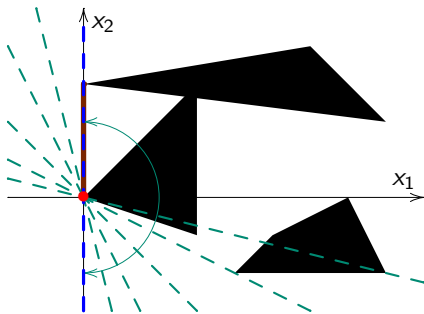
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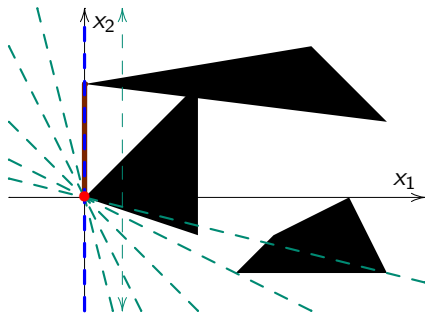
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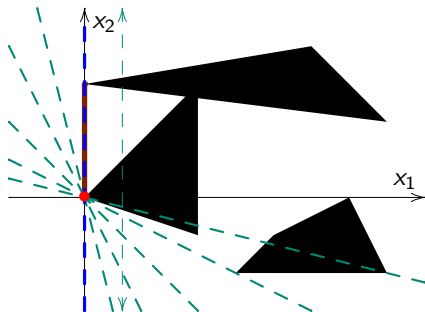
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Compute $a = \min x_2 + y_2$ s.t.

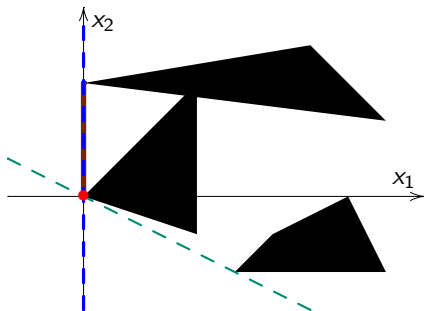
$$x_1 + y_1 = 1 \wedge A\mathbf{x} \geq \mathbf{c}x_0 \wedge x_0 \geq 0 \wedge B\mathbf{y} \geq \mathbf{d}y_0 \wedge y_0 \geq 0$$

(Cone of hull is sum of cones in homogeneous space)

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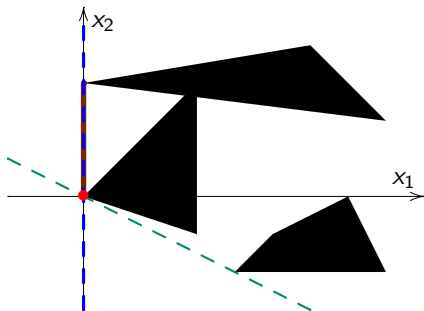
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- ▶ Repeat for all ridges
- ▶ Ridges found through recursive application
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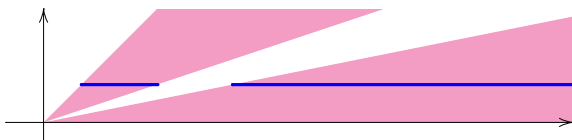
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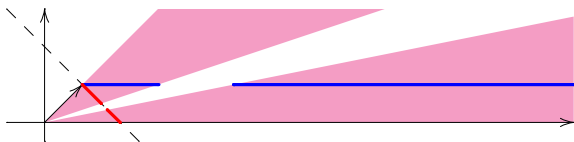


- ▶ Consider cones in homogeneous space

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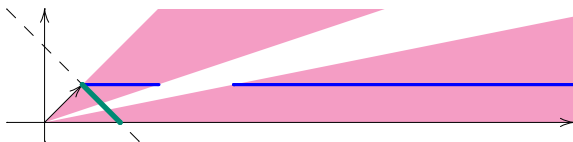


- ▶ Consider cones in homogeneous space
- ▶ Take other homogeneous direction \Rightarrow union of polytopes

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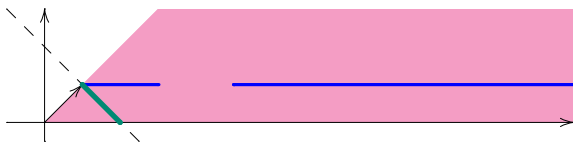


- ▶ Consider cones in homogeneous space
- ▶ Take other homogeneous direction \Rightarrow union of polytopes
- ▶ Compute convex hull

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- ▶ Consider cones in homogeneous space
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- ▶ Convert back

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is1 Operation: Closed Convex Hull—Wrapping

$$H = \text{conv.hull}(S_1 \cup S_2) \quad S_1 = \{ \mathbf{x} \mid A\mathbf{x} \geq \mathbf{c} \} \quad S_2 = \{ \mathbf{x} \mid B\mathbf{x} \geq \mathbf{d} \}$$

- ▶ S_1 and S_2 are polytopes (FLL2000)
 - ▶ Assume $x_1 \geq 0$ defines a facet and $x_2 \geq 0$ a ridge on the facet
 - ▶ Wrap facet around ridge \Rightarrow new facet constraint $x_2 \geq ax_1$
 - ▶ Repeat for all ridges
 - ▶ Ridges found through recursive application
 - ▶ Repeat for new facets until all facets found
- ▶ H is pointed \Rightarrow change perspective



- ▶ Consider cones in homogeneous space
- ▶ Take other homogeneous direction \Rightarrow union of polytopes
- ▶ Compute convex hull
- ▶ Convert back

is1 Operation: Closed Convex Hull—Wrapping

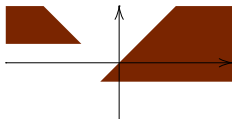
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 - ▶ Convert back
- ▶ S_1 and S_2 are pointed (R_i recession cone of S_i)
 \Rightarrow project out lineality $H = \text{lin.hull}(R_1 \cap -R_2)$

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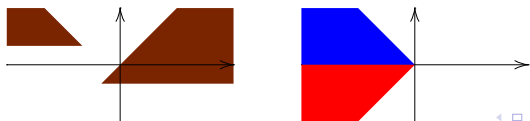
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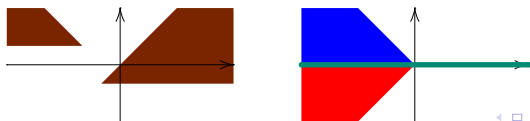
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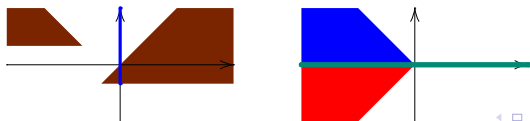
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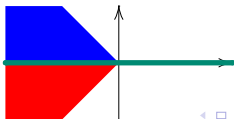
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- ▶ S_1 and S_2 are pointed (R_i recession cone of S_i)
 \Rightarrow project out lineality $H = \text{lin.hull}(R_1 \cap -R_2)$
- ▶ S_1 or S_2 has non-trivial lineality space
 \Rightarrow project out lineality S_1 and lineality S_2

Improved Code Generation using CLooG

Using PolyLib as a backend:

```
for (p1=0;p1<=floord(8*N+63,32);p1++) {
  for (p3=max(max(max(max(0,ceild(-32*p1-27,4)),
    ceild(512*p1-128*N-975,16)),ceild(28*p1-7*N-20,36)),
    ceild(60*p1-15*N-44,68));
    p3<=min(min(floord(4*M+47,16),floord(24*p1+5*M+36,20)),
    floord(136*p1+31*M+224,124));p3++) {
    if ((p1 >= 0) && (p1 <= floord(N-1,4))) {
      for (p5=max(0,4*p3);p5<=min(M-1,4*p3+3);p5++) {
        for (p7=max(0,4*p1);p7<=min(N-1,4*p1+3);p7++) {
          S9(p3,p5,p1,p7); /* ... */
        }
      }
    }
  }
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```

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          S9(p3,p5,p1,p7); /* ... */
        }
      }
    }
  }
}
```

Using isl as a backend:

```
for (p1=0;p1<=floord(N+7,4);p1++) {
  for (p3=max(0,ceild(4*p1-N+1,4));
    p3<=min(floord(M+11,4),floord(4*p1+M+3,4));p3++) {
    if (p1 <= floord(N-1,4)) {
      for (p5=4*p3;p5<=min(M-1,4*p3+3);p5++) {
        for (p7=4*p1;p7<=min(N-1,4*p1+3);p7++) {
          S9(p3,p5,p1,p7); /* ... */
        }
      }
    }
  }
}
```

CLooG Speed Comparison

	PolyLib-64	PolyLib-gmp	isl-gmp
Example from previous slide (from Harald Devos)	0.15s	0.31s	0.18s
CLooG test suite	5.1s	11.4s	7.5s
Simple tiling example	1.11s	2.63s	1.11s
Extreme tiling example	14.6s	28.5s	5.15s
LU example	0.86s	1.88s	0.35s
Sobel example (from Harald Devos)	0.62s	1.64s	0.15s

(Tiling examples from Uday K Bondhugula)

Conclusion

- ▶ isl: a new integer set library
- ▶ currently used in
 - ▶ equivalence checking tool
 - ▶ CLooG
 - ▶ Produces better code than PolyLib backend
 - ▶ Comparable in speed or faster than PolyLib backend
- ▶ explicit support for existentially quantified variables
- ▶ uses PIP for solving (P)ILP problems
- ▶ all computations in exact integer arithmetic using GMP
- ▶ built-in incremental LP solver
- ▶ released under LGPL license