# An Implementation of the Barvinok-Woods Integer Projection Algorithm 

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## Outline

(1) Integer Projection: Motivation and Introduction

- Optimization and Game Theory
- Application in Program Transformations
(2) Operations on Generating Functions
- Projection/Summation
- Integer Projection: The Case of Several Variables
(3) Lattice Widths
- Definition
- Lattice Width Computation: The Eisenbrand-Shmonin Method


## Multicriterion integer linear programming problems

De Loera, Hemmecke, K., 2007
Let $A=\left(a_{i j}\right)$ be an integral $m \times n$-matrix and $\mathbf{b} \in \mathbf{Z}^{m}$ such that the convex polyhedron $P=\left\{\mathbf{u} \in \mathbf{R}^{n}: A \mathbf{u} \leq \mathbf{b}\right\}$ is bounded. Given $k$ linear functionals $f_{1}, f_{2}, \ldots, f_{k} \in \mathbf{Z}^{n}$, we consider the multicriterion integer linear programming problem

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\begin{aligned}
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& \text { vmin }\left(f_{1}(\mathbf{u}), \ldots, f_{k}(\mathbf{u})\right) \\
& \text { subject to } \mathbf{A u} \leq \mathbf{b} \\
& \mathbf{u} \in \mathbf{Z}^{n}
\end{aligned} \\
& \text { An outcome vector } \\
& \mathrm{f}(\mathrm{u})=\left(f_{1}(\mathbf{u}) \ldots, f_{k}(\mathrm{u})\right) \\
& \text { is a Pareto optimum if and } \\
& \text { only if there is no other } \\
& \text { feasible point } \tilde{u} \text { such that } \\
& f_{i}(\tilde{u}) \leq f_{i}(\mathbf{u}) \text { for all } i \text { and } \\
& f_{j}(\tilde{u})<f_{j}(\mathbf{u}) \text { for at least } \\
& \text { one index } j \text {. }
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$D$

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## Bilevel integer linear optimization <br> K., Queyranne, Ryan, 2008

## Bilevel integer linear programming problem

Consider the bilevel integer linear programming problem

$$
\begin{array}{ll}
\min & f(\mathbf{x})+h(\mathbf{z}) \\
\text { s.t. } & \mathbf{z} \in Z \subseteq \mathbf{R}^{d} \\
& \mathbf{x} \in \operatorname{Argmin}\left\{\boldsymbol{\psi}^{\top} \mathbf{x}: A \mathbf{x} \leq \pi(\mathbf{z}), \mathbf{x} \in \mathbf{Z}^{n}\right\}
\end{array}
$$

where $Z$ is a polytope, $\psi \in \mathbf{Q}^{n}$, and $\pi: \mathbf{R}^{d} \rightarrow \mathbf{R}^{m}, f: \mathbf{R}^{n} \rightarrow \mathbf{R}$, and $h: \mathbf{R}^{d} \rightarrow \mathbf{R}$ are affine-linear functions. We assume that all sets $\left\{\mathbf{x} \in \mathbf{Z}^{n}: A \mathbf{x} \leq \boldsymbol{\pi}(\mathbf{z})\right\}$ are contained in a polytope.

## Hardness

Not even in NP.

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## Pure Nash equilibria

Consider the set of pure $k$-player Nash equilibria, for instance in an integer-splittable weighted network congestion game.
Let the strategy of each player $i \in\{1, \ldots, k\}$ be described by a vector $\mathbf{u}_{i} \in U_{i} \cap \mathbf{Z}^{d_{i}}$, let $d=d_{1}+\cdots+d_{k}$, where $U_{i} \subseteq \mathbf{R}^{d_{i}}$ is a polytope.
Let $F=U_{1} \times \cdots \times U_{k}$.
Let the payoff function of each player $i$ be a piecewise linear concave function $\psi_{i}\left(\mathbf{u}_{1}, \ldots, \mathbf{u}_{k}\right)$, which the player seeks to maximize.
A pure Nash equilibrium $\overline{\mathbf{u}}=\left(\overline{\mathbf{u}}_{1}, \ldots, \overline{\mathbf{u}}_{k}\right)$ is characterized by the following conditions.
(i) $\overline{\mathbf{u}}$ is a feasible strategy combination, i.e.,

$$
\begin{equation*}
\overline{\mathbf{u}}=\left(\overline{\mathbf{u}}_{1}, \ldots, \overline{\mathbf{u}}_{k}\right) \in F \tag{1}
\end{equation*}
$$

(ii) For all $i=1, \ldots, k$,

$$
\begin{align*}
& \nexists \mathbf{u}_{i} \in \mathbf{Z}^{d_{i}} \text { with } \tilde{\mathbf{u}}^{(i)}:=\left(\overline{\mathbf{u}}_{1}, \ldots, \overline{\mathbf{u}}_{i-1}, \mathbf{u}_{i}, \overline{\mathbf{u}}_{i+1}, \ldots, \overline{\mathbf{u}}_{k}\right) \in F  \tag{2}\\
& \quad \text { and } \psi_{i}\left(\tilde{\mathbf{u}}^{(i)}\right)>\psi_{i}(\overline{\mathbf{u}})
\end{align*}
$$

## Program Transformations in Optimizing Compilers

Verdoolaege et al., 2004-2007
Perform program transformations to

- reduce total memory requirements or working set
- parallelize

Need to count

- number of memory elements, communication volume, ...


## Example: Count array elements accessed in a nested loop

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

Equal to the number of elements in


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S_{p}=\{\ell \in \mathbf{Z} \mid \exists i, j \in \mathbf{Z}: \ell=6 i+9 j-7,1 \leq j \leq p, 1 \leq i \leq 8\}
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\# S_{p}= \begin{cases}8 & \text { if } p=1 \\
3 p+10 & \text { if } p \geq 2\end{cases}
\end{gathered}
$$

## Counting Example

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## Barvinok's short rational generating functions

## Generating functions

$$
g_{P}(z)=z^{0}+z^{1}+z^{2}+z^{3}+z^{4}
$$

## Theorem (Alexander Barvinok, 1994)

Let the dimension $d$ be fixed. There is a
for computing a representation of the generating-

## function



## Corollary

In particular,
$N=\left|P \cap \mathbf{Z}^{d}\right|=g_{P}(1)$
can be computed in
(in
fixed dimension)
of the integer points $P \cap \mathbf{Z}^{d}$ of a polyhedron $P \subset \mathbf{R}^{d}$ (given
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## Barvinok's short rational generating functions

## Generating functions

| $\boldsymbol{\phi}$ | $\bullet$ | $\bullet$ | $\bullet$ | $\boldsymbol{\phi}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| 0 | 1 | 2 | 3 | 4 |$\quad$| $g_{P}(z)$ | $=z^{0}+z^{1}+z^{2}+z^{3}+z^{4}$ |
| :---: | :---: |
|  | $=\frac{1-z^{5}}{1-z}$ |$\quad$ for $z \neq 1$

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## Corolary

 algorithm for computing a representation of the generating function$$
g_{P}\left(z_{1}, \ldots, z_{d}\right)=\sum_{\left(\alpha_{1}, \ldots, \alpha_{d}\right) \in P \cap z^{d}} z_{1}^{\alpha_{1}} \cdots z_{d}^{\alpha_{d}}=\sum_{\alpha \in P \cap z^{d}} z^{\alpha}
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## First Practical Implementation: LattE

## From the website http: //www.math.ucdavis.edu/~latte/

IattF is a computer software dedicated to the problems of counting and detecting lattice points inside convex polytopes, and the solution of integer programs. LattE contains the first ever implementation of Barvinok's algorithm. LattE stands for Lattice point Enumeration.


## Developed 2002-3 by:

Jesís De Inera
David Haws
Raymond Hemmecke
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## Capabilities

- Count lattice points
- Compute multivariate generating functions
- Compute Ehrhart series
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## S. Verdoolaege's barvinok library (2005-)

## Parametric counting problems

## LattE macchiato - an improved version of LattE (K., 2006-)

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## Boolean Operations: Products of Generating Functions

## Set Intersection

$$
f(\mathbf{x})=\sum_{\mathbf{s} \in \mathbf{Z}^{d}} a(\mathbf{s}) \mathbf{x}^{\mathbf{s}} \quad g(\mathbf{x})=\sum_{\mathbf{s} \in \mathbf{Z}^{d}} b(\mathbf{s}) \mathbf{x}^{\mathbf{s}}
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Hadamard product:

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f(\mathbf{x}) \star g(\mathbf{x})=\sum_{\mathbf{s} \in \mathbf{Z}^{d}} a(\mathbf{s}) b(\mathbf{s}) \mathbf{x}^{\mathbf{s}}
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Multi-sets: counts pairs of occurrences

## Union and set difference (not for multisets)



## Theorem (Barvinok and Woods (2003))

The rational generating function of arbitrary fixed-length Boolean combinations of sets
given by Barvinok-style rational generating functions can be computed in
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$$
\begin{aligned}
& {[\mathbf{s} \in S \backslash T]=[\mathbf{s} \in S]-[\mathbf{s} \in S \cap T]} \\
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## Integer Projection: The Codimension-1 Case



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$$
\begin{gathered}
P \\
P+\mathbf{e}_{n+d+1} \\
S^{\prime}=P \backslash\left(P+\mathbf{e}_{n+d+1}\right) \\
S=\pi_{n+d} S^{\prime} \\
\text { (one-to-one projection) }
\end{gathered}
$$

## Integer Projection: The Case of Several Variables

## Shifting once in second direction is not enough



We need to subtract as many shifted copies as the size of the biggest gap

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## Lattice Widths



# width $_{(1,0)} P=9$ <br> width(0,1) $P=6$ <br> width $_{(1,1)} P=15$ <br> width $_{(-1,2)} P=3$ <br> width $_{(-2, \text { 3) }} P=3$ <br> width $_{(1,-1)} P=3$ 

Width in direction $\mathbf{c}$ :

$$
\operatorname{width}_{\mathbf{c}} P(\mathbf{p})=\max \{\langle\mathbf{c}, \mathbf{x}\rangle \mid \mathbf{x} \in P(\mathbf{p})\}-\min \{\langle\mathbf{c}, \mathbf{x}\rangle \mid \mathbf{x} \in P(\mathbf{p})\}
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## Lattice width of $P$

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\text { width } P(\mathbf{p})=\min _{\mathbf{c} \in \mathbf{Z}^{d} \backslash\{0\}} \text { width }_{c} P(\mathbf{p})
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## Lattice Widths



$$
\operatorname{width}_{(1,0)} P=9
$$

$$
\begin{aligned}
& \text { width }_{(0,1)} P=6 \\
& \text { width }_{(1,1)} P=15
\end{aligned}
$$

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\text { width }_{(-1,2)} P=3
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## Flatness Theory and Small Gaps

As a consequence of flatness theory for lattice-point-free convex bodies, it is possible to construct directions of small gaps for the parametric polytopes.

## Theorem (Small-gaps theorem; cf. Barvinok and Woods (2003))

Let $\kappa \geq 1$ and let $\mathbf{c} \in \mathbf{Z}^{m}$ be a $\kappa$-approximative lattice width direction, i.e.,

$$
\text { width }_{\mathrm{c}} P_{\mathrm{s}, \mathrm{t}} \leq \kappa \cdot \text { width } P_{\mathrm{s}, \mathrm{t}}
$$

Then the image
$Y=\left\{\langle\mathbf{c}, \mathbf{x}\rangle \mid \mathbf{x} \in P_{\mathrm{s}, \mathbf{t}} \cap \mathbf{Z}^{d}\right\}$ does not have gaps larger than $\kappa \cdot \omega(m)$.

Here $\omega(m)$ is the flatness constant; it only depends on the dimension $m$ (Lagarias et al., 1990; Barvinok, 2002; Banaszczyk et al., 1999).

## Theorem (Kannan (1992))

Let the total dimension $n+d+m$ be fixed. Then there exists a polynomial-time algorithm for the following problem.
Given as input, in binary encoding,
$\left(\mathrm{I}_{1}\right)$ inequalities describing a rational polytope $P \subset \mathbf{R}^{n+d+m}$,
output, in binary encoding,
$\left(\mathrm{O}_{1}\right)$ inequality systems describing partially open polyhedra $\tilde{Q}_{1}, \ldots, \tilde{Q}_{M} \subset \mathbf{R}^{n+d}$ that form a partition of the projection, $Q$, of $P$ onto the first $n+d$ coordinates,
$\left(\mathrm{O}_{2}\right)$ integer vectors $\mathbf{c}_{1}, \ldots, \mathbf{c}_{M} \in \mathbf{R}^{m}$, such that $\mathbf{c}_{i}$ is a 2-approximative lattice width direction for every polytope $P_{\mathrm{s}, \mathrm{t}}$ when $(\mathbf{s}, \mathbf{t}) \in \tilde{Q}_{i}$.

## Flatness Theory and Small Gaps: Strengthening

As a consequence of flatness theory for lattice-point-free convex bodies, it is possible to construct directions of small gaps for the parametric polytopes.

## Theorem (Small-gaps theorem; cf. Barvinok and Woods (2003))

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\text { width }_{\mathrm{c}} P_{\mathrm{s}, \mathrm{t}} \leq \kappa \cdot \text { width } P_{\mathrm{s}, \mathrm{t}}
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Here $\omega(m)$ is the flatness constant; it only depends on the dimension $m$ (Lagarias et al., 1990; Barvinok, 2002; Banaszczyk et al., 1999).

## Theorem (Eisenbrand-Shmonin, 2008)

Let the dimension $n+d$ be fixed; $m$ is allowed to vary. Then there exists a polynomial-time algorithm for the following problem. Given as input, in binary encoding,
( $\mathrm{I}_{1}$ ) inequalities describing a rational polytope $P \subset \mathbf{R}^{n+d+m}$,
output, in binary encoding,
$\left(\mathrm{O}_{1}\right)$ inequality systems describing partially open polyhedra $\tilde{Q}_{1}, \ldots, \tilde{Q}_{M} \subset \mathbf{R}^{n+d}$ that form a partition of the projection, $Q$, of $P$ onto the first $n+d$ coordinates,
$\left(\mathrm{O}_{2}\right)$ integer vectors $\mathbf{c}_{1}, \ldots, \mathbf{c}_{M} \in \mathbf{R}^{m}$, such that $\mathbf{c}_{i}$ is an exact lattice width direction for every polytope $P_{\mathrm{s}, \mathrm{t}}$ when $(\mathbf{s}, \mathbf{t}) \in \tilde{Q}_{i}$.

## Small Gaps in Lattice Width Direction



Apply unimodular transformation

$$
T=\left[\begin{array}{cc}
-1 & 2 \\
0 & 1
\end{array}\right]
$$

$\Rightarrow$ small gaps $(=1$ if $m=2)$

## Small Gaps in Lattice Width Direction



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## Lattice Widths Computation: The Eisenbrand-Shmonin Method

Width in direction $\mathbf{c} \in \mathbf{Z}^{d}$ :

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\text { width }_{\mathbf{c}} P(\mathbf{p})=\max \{\langle\mathbf{c}, \mathbf{x}\rangle \mid \mathbf{x} \in P(\mathbf{p})\}-\min \{\langle\mathbf{c}, \mathbf{x}\rangle \mid \mathbf{x} \in P(\mathbf{p})\}
$$

Lattice width of $P$

$$
\text { width } P(\mathbf{p})=\min _{\mathbf{c} \in \mathbf{Z}^{d} \backslash\{0\}} \text { width }_{\mathbf{c}} P(\mathbf{p})
$$

min and max occur at vertices (extremal points of polytope)
Consider all pairs of (parametric) vertices
Compute candidate width directions for each pair
Compute smallest overall width

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$\Rightarrow$ Consider all pairs of (parametric) vertices
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## Lattice Widths Computation: The Eisenbrand-Shmonin Method II



Candidates for vertex pair ( $\left.\mathbf{v}_{1}(\mathbf{p}), \mathbf{v}_{2}(\mathbf{p})\right)$

$$
\begin{aligned}
\text { width }_{\mathbf{c}} P(\mathbf{p}) & =\left\langle\mathbf{c}, \mathbf{v}_{2}(\mathbf{p})\right\rangle-\left\langle\mathbf{c}, \mathbf{v}_{1}(\mathbf{p})\right\rangle \\
& =\left\langle\mathbf{c}, \mathbf{v}_{2}(\mathbf{p})-\mathbf{v}_{1}(\mathbf{p})\right\rangle
\end{aligned}
$$

width $P(\mathbf{p})=\min _{\mathbf{c} \in \mathrm{Z}^{d} \backslash\{\mathbf{0}\}}$ width $_{\mathbf{c}} P(\mathbf{p})$
$\Rightarrow \mathbf{c} \in\left(C_{1}^{*} \cap-C_{2}^{*} \cap \mathbf{Z}^{d}\right) \backslash\{\mathbf{0}\}$
$\Rightarrow$ vertices of integer hull of $\left(C_{1}^{*} \cap-C_{2}^{*}\right) \backslash\{\mathbf{0}\}$

Polynomial-time due to Cook et al. (1992)

## Practical implementation

- Compute the convex hull based on a linear optimization oracle
- implemented using binary search and Generalized Basis Reduction
- implement using CPLEX?
- Use Hilbert basis computation?


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## Computational Results

|  |  | Problem |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | ex1 | woods | pugh | p-pugh | scarf1 |
| Parameter variables s | $n$ | 0 | 1 | 0 | 1 | 2 |
| Variables t | d | 0 | 0 | 0 | 0 | 0 |
| Existentially quant. variables u | $m$ | 2 | 2 | 2 | 2 | 2 |
| Inequalities |  | 4 | 4 | 4 | 4 | 5 |
| Parametric Vertices |  | 4 | 4 | 4 | 4 | 6 |
| Width directions |  | 7 | 3 | 8 | 8 | 6 |
| Distinct width directions |  | 4 | 2 | 7 | 7 | 4 |
| Chambers |  | 1 | 2 | 1 | 6 | 2 |
| LPs solved in gen. basis red. |  | 8 |  | 49 | 43 | 4 |
| Without exploiting small gaps in dimension 2 |  |  |  |  |  |  |
| Time (CPU seconds) |  | 0.11 | 29.2 | 0.09 | 797 | 126 |
| Exploiting small gaps in dimension 2 |  |  |  |  |  |  |
| Time (CPU seconds) |  | 0.08 | 2.7 |  | 18.0 | 1.1 |

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