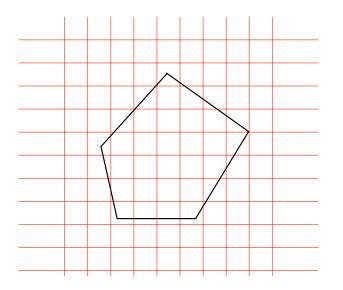
### Generating Functions Algorithms in Integer Optimization

### **LECTURE I** Generating functions and Lattice Points

### Jesús Antonio De Loera University of California, Davis



### Lattice Point Problems

Given a subset X of  $\mathbb{R}^d$ , there are a number of basic problems about lattice points:

- Decide whether  $X \cap \mathbb{Z}^d$  is non empty.
- If X is bounded, count how many lattice points are in X.
- Given a norm, such as the  $l_\infty$  or  $l_p$  norms, find the shortest lattice vector of X.
- Given a linear functional  $c \cdot x$  we wish to optimize it over the lattice points of X, i.e. find the lattice point in X that maximizes (minimizes) cx.

- Given a polynomial  $f(x) \in \mathbb{Z}[x_1, \ldots, x_d]$ , find  $y \in X \cap \mathbb{Z}^d$  which maximizes the value f(y).
- How to generate a lattice point in X uniformly at random?
- Find a Hilbert bases for a polyhedral cone X.

We present a non-traditional algebraic-analytic point of view: GENERATING FUNCTIONS!!

# BARVINOK's ENCODING

### The Generating Function Encoding

Given  $K \subset \mathbb{R}^d$  we **WANT** to compute the generating function

$$f(K) = \sum_{\alpha \in K \cap \mathbb{Z}^d} z_1^{\alpha_1} z_2^{\alpha_2} \dots z_n^{\alpha_n}.$$

Think of the lattice points as monomials!!! EXAMPLE: (7, 4, -3) is  $z_1^7 z_2^4 z_3^{-3}$ .

f(K) has inside **all lattice points** of K. But it is too long! In fact, this is an infinite formal power series if K is not bounded, but if K is a polytope it is a (Laurent) polynomial.

### We need a SHORT REPRESENTATION !!!

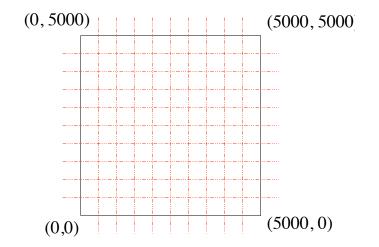
### BARVINOK's ANSWER:

When K is a rational convex polyhedron, i.e.  $K = \{x \in \mathbb{R}^n | Ax = b, Bx \leq b'\}$ , where A, B are integral matrices and b, b' are integral vectors, The generating function f(K), and thus ALL the lattice points of the polyhedron K, can be encoded in a "short" sum of rational functions!!!

**EXAMPLE 1:** Suppose my polyhedron is the infinite half-line  $P = \{x | x \ge 0\}$ 

### Example 2

Let P be the square with vertices  $V_1 = (0,0)$ ,  $V_2 = (5000,0)$ ,  $V_3 = (5000, 5000)$ , and  $V_4 = (0, 5000)$ .



The generating function f(P) has over 25,000,000 monomials,  $f(P) = 1 + z_1 + z_2 + z_1^1 z_2^2 + z_1^2 z_2 + \cdots + z_1^{5000} z_2^{5000}$ ,

### But it has only four rational functions in its Barvinok's encoding.

$$\frac{1}{(1-z_1)(1-z_2)} + \frac{z_1^{5000}}{(1-z_1^{-1})(1-z_2)} + \frac{z_2^{5000}}{(1-z_2^{-1})(1-z_1)} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1)} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})(1-z_1^{-1})} + \frac{z_1^{5000}z_2^{5000}}{(1-z_1^{-1})(1-z_1^{$$

### Barvinok's Original Algorithm (1993 Barvinok)

Assume the dimension d is fixed. Let P be a rational convex d-dimensional polytope. Then, in polynomial time on the size of the input, we can write the generating function  $f(P) = \sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$ . as a polynomial-size sum of rational functions of the form:

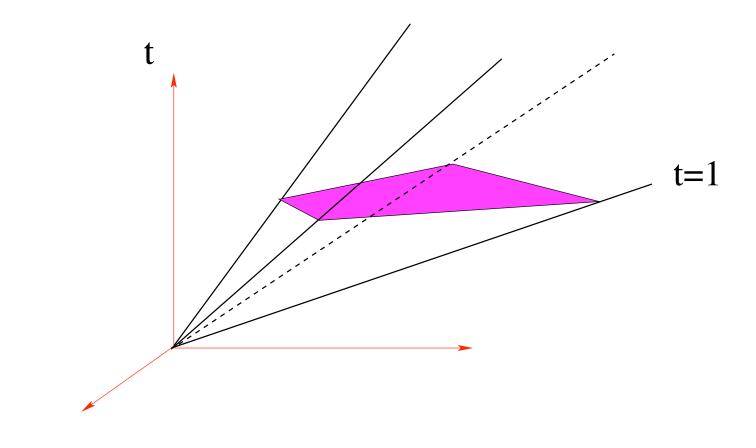
$$\sum_{i \in I} E_i \frac{z^{u_i}}{\prod_{j=1}^d (1 - z^{v_{ij}})},$$
(1)

where I is a polynomial-size indexing set, and where  $E_i \in \{1, -1\}$  and  $u_i, v_{ij} \in \mathbb{Z}^d$  for all i and j.

We present an improved algorithm (2002 De Loera et al.)

### **Enough to do it for CONES**

Set your polytope P inside the hyperplane t = 1. What we want is the generating function of the lattice points in the cone.



### **Enough to do it for SIMPLE CONES**

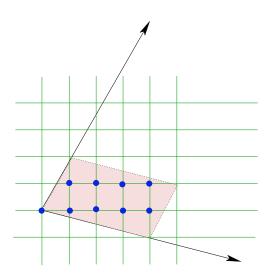
By the INCLUSION-EXCLUSION principle, we can just add the generating functions of the simplicial pieces!

### Simple Cones are Easy

For a simple cone  $K \subset \mathbb{R}^d$ ,

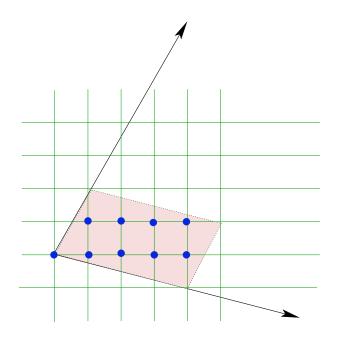
$$f(K) = \frac{\sum_{u \in \Pi \cap \mathbb{Z}^d} z^u}{(1 - z^{c_1})(1 - z^{c_2}) \dots (1 - z^{c_d})}$$

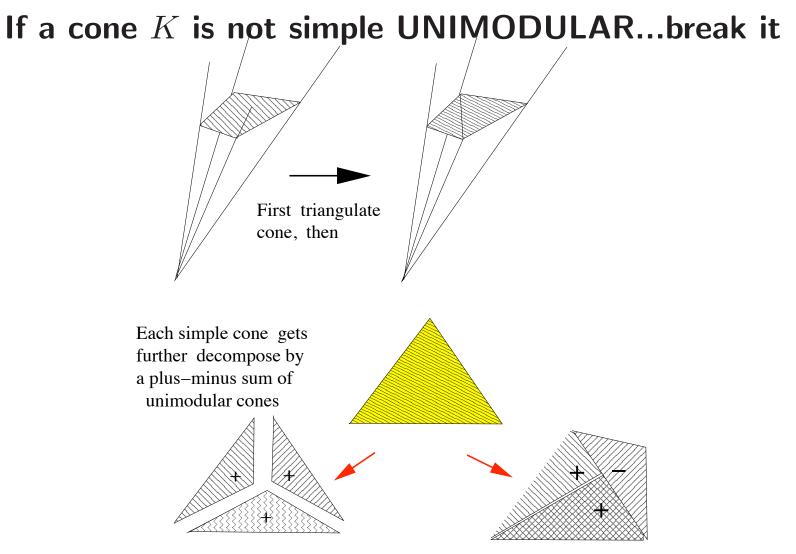
 $\Pi$  is the half open parallelepiped  $\{x | x = \alpha_1 c_1 + \cdots + \alpha_d c_d, 0 \le \alpha_i < 1\}.$ 



**Example** In this case, we have d = 2 and  $c_1 = (1, 2)$ ,  $c_2 = (4, -1)$ . We have:

$$f(K) = \frac{z_1^4 z_2 + z_1^3 z_2 + z_1^2 z_2 + z_1 z_2 + z_1^4 + z_1^3 + z_1^2 + z_1 + 1}{(1 - z_1 z_2^2)(1 - z_1^4 z_2^{-1})}.$$



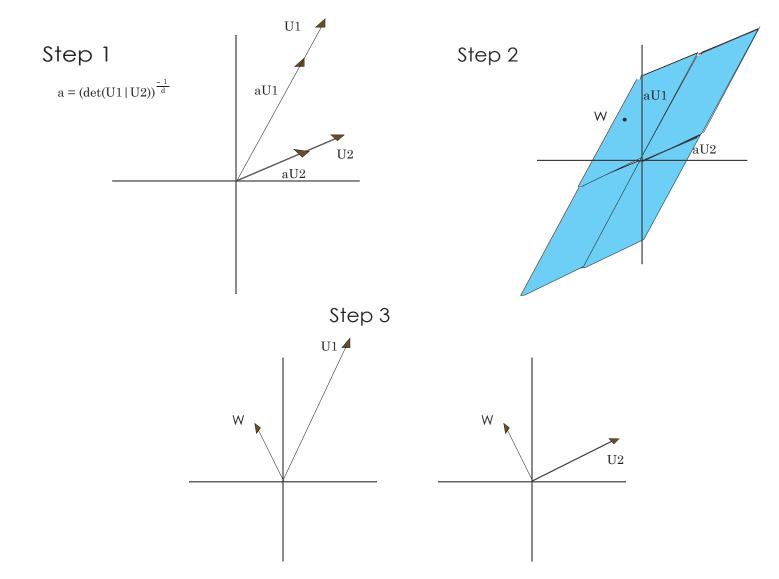


### Barvinok's cone decomposition lemma

**Theorem** [Barvinok] Fix the dimension d. Then there exists a polynomial time algorithm which decomposes a rational polyhedral cone  $K \subset \Re^d$  into unimodular cones  $K_i$  with numbers  $\epsilon_i \in \{-1, 1\}$  such that

$$f(K) = \sum_{i \in I} \epsilon_i f(K_i), \ |I| < \infty.$$

**Main idea** Triangulation is TOO expensive, allow simplicial cones's rays to be outside the original cone. Rays are short integer vectors inside a convex body, apply Minkowski's theorem!



### **SUMMARY** of Homogenized Barvinok Algorithm.

**Input** is a full-dimensional convex rational convex polytope P in  $\mathbb{R}^d$  specified by linear inequalities and linear equations.

- 1. Place the polytope P into the hyperplane defined by  $x_{d+1} = 1$  in  $\mathbb{R}^{d+1}$ . Let K be the d+1-dimensional cone over P, that is,  $K = cone(\{(p, 1) : p \in P\})$ .
- 2. We can triangulate K and reduce everything to simple cones  $\sigma_1, \sigma_2, \ldots, \sigma_r$ . Apply Barvinok's decomposition of  $\sigma_i$  into unimodular cones. We get a **signed** unimodular cone decomposition of K.
- 3. Retrieve a signed sum of multivariate rational functions, one per cone, which represents the series  $\sum_{a \in K \cap \mathbb{Z}^n} x^a$ .

4. If we call the variable  $x_{d+1} = t$  then we obtain the expression of the generating function of  $\sum_{n=0}^{\infty} \left( \sum_{\alpha \in nP \cap \mathbb{Z}^d} z^{\alpha} \right) t^n$ ,

### **EXAMPLE**

For the triangle  $\sigma$  with vertices  $V_0=(-1,-1),\ V_1=(2,-1),$  and  $V_2=(-1,2)$  we have

$$(1-x)^{-1}(1-y)^{-1}\left(1-\frac{t}{xy}\right)^{-1} + (1-x^{-1})^{-1}\left(1-\frac{y}{x}\right)^{-1}\left(1-\frac{x^{2}t}{y}\right)^{-1} + (1-y^{-1})^{-1}\left(1-\frac{x}{y}\right)^{-1}\left(1-\frac{y^{2}t}{x}\right)^{-1}$$

### **Counting Lattice Points FAST!**

**LEMMA:** The number of lattice points in P is the limit when the vector  $(x_1, \ldots, x_n)$  goes to  $(1, 1, \ldots, 1)$ .

**TROUBLE**: The vector (1, 1, ..., 1) is a pole in all the rational functions, a singularity, because the Barvinok rational functions are

$$\frac{z^a}{\prod_{i=1}^k (1-z_i^v)}$$

HOW TO COMPUTE THIS LIMIT???? Shall I expand into monomials??? The singularity gets resolved that way...right?

# **NO WAY!** Never fully expand the rational functions into ALL monomials!

## USE NUMERICAL COMPLEX ANALYSIS 101 TO EVALUATE THE RATIONAL FUNCTIONS!!

### **Computation of Residues for rational functions**

This reduces to computing a residue at a pole  $z_0$ .

If  $f(z) = \sum_{k=-m}^{\infty} a_n (z-z_0)^k$ , the residue is defined as

 $Res(f(z_0)) = a_{-1}.$ 

Given a rational function  $f(z) = \frac{p(z)}{q(z)}$ , and a pole  $z_0$  we use

**THEOREM** Henrici's Algorithm for the residue: If p(z), q(z) have degree no more than d, then residue at  $z_0$  can be computed in no more than  $0(d^2)$ arithmetic operations.

### Algorithm

(CASE 1) If  $z_0$  is a simple pole is TRIVIAL, then  $Resf(z_0) = \frac{p(z_0)}{q'(z_0)}$ . (CASE 2) Else  $z_0$  is a pole of order m > 1,

(A) Write  $f(z) = \frac{p(z)}{(z-z_0)^m q_1(z)}$ .

(B) Expand  $p, q_1$  in powers of  $(z - z_0)$ 

 $p(z) = a_0 + a_1(z - z_0) + a_2(z - z_0)^2 + \dots \quad q_1(z) = b_0 + b_1(z - z_0) + b_2(z - z_0)^2 + \dots$ 

(C) The Taylor expansion of  $p(z)/q_1(z)$  at  $z_0$  is  $c_0 + c_1(z - z_0) + c_2(z - z_0)^2 + c_3(z - z_0)^3 + \dots$  where

$$c_0 = \frac{a_0}{b_0}$$
, and  $c_k = \frac{1}{b_0}(a_k - b_1c_{k-1} - b_2c_{k-2} - \dots - b_kc_0)$   
(D) OUTPUT  $Res(f(z_0)) = c_{m-1}$ .

### **Monomial Substitution**

**Lemma:** Let us fix k, the number of binomials in the denominator of a rational function. Given a rational function sum g of the form

$$g(x) = \sum_{i \in I} \alpha_i \frac{x^{u_i}}{\prod_{j=1}^k (1 - x^{v_{ij}})},$$

where  $u_i, v_{ij}$  are integral *d*-dimensional vectors, and a monomial map  $\psi : \mathbb{C}^n \longrightarrow \mathbb{C}^d$  given by the variable change  $x_i \rightarrow z_1^{l_{i1}} z_2^{l_{i2}} \dots z_n^{l_{in}}$  whose image does not lie entirely in the set of poles of g(x), then there exists a polynomial time algorithm which, computes the function  $g(\psi(z))$  as a sum of rational functions of the same shape as g(z).

### **Corollary: Random Generation of Lattice Points**

How to pick a random lattice point? Markov chain methods have been around for some time, but they work on some "roundness" assumptions!! Not working well for all polytopes! (work by Dyer, Frieze, Kannan, Lovasz, Simonovits and others)

**THEOREM** (Barvinok-Pak) Let P be a convex rational polytope in  $\mathbb{R}^d$ . Then using  $O(d^2 \log(size(P)))$  calls to Barvinok's counting algorithm, one can in polynomial time can sample uniformly from set  $P \cap \mathbb{Z}^d$ .

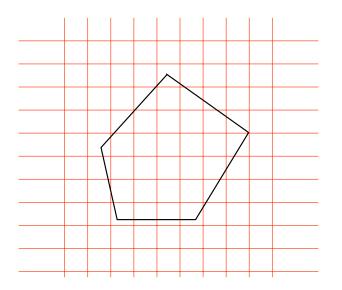
# LattE

- Our goal was to implement and develop algebraic-analytic algorithms. Members: J. De Loera, R. Hemmecke, R. Yoshida, D. Haws, P. Huggins, J. Tauzer.
- First implementation of Barvinok's encoding algorithm. Software implemented in C++.
- We used also libraries from CDD, NTL.
- We use BOTH geometric computing AND symbolic-algebraic manipulations!!

### Generating Functions Algorithms in Integer Optimization

### **LECTURE II** Contributions to Integer Linear Programming

Jesús Antonio De Loera University of California, Davis



### **Integer Linear Programming**

Given the a convex polyhedron  $X = P \cap \mathbb{Z}^d$ , and a linear functional  $c \cdot x$  we wish to optimize it over the lattice points of X, i.e. find the lattice point in X that maximizes (minimizes) cx.

We take the point of view: GENERATING FUNCTIONS.

### **Recall: Barvinok's Theorem**

Assume the dimension d is fixed. Let P be a rational convex d-dimensional polytope. Then, in polynomial time on the size of the input, we can write the generating function  $f(P) = \sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$ . as a polynomial-size sum of rational functions of the form:

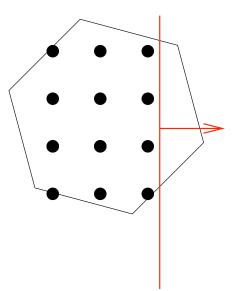
$$\sum_{i \in I} E_i \frac{z^{u_i}}{\prod_{j=1}^d (1 - z^{v_{ij}})},$$
(2)

where I is a polynomial-size indexing set, and where  $E_i \in \{1, -1\}$  and  $u_i, v_{ij} \in \mathbb{Z}^d$  for all i and j.

# INTEGER LINEAR PROGRAMS

### **ALGORITHM: Barvinok + Binary Search** Input: $A \in \mathbb{Z}^{m \times d}, b \in \mathbb{Z}^m, c \in \mathbb{Z}^d$ .

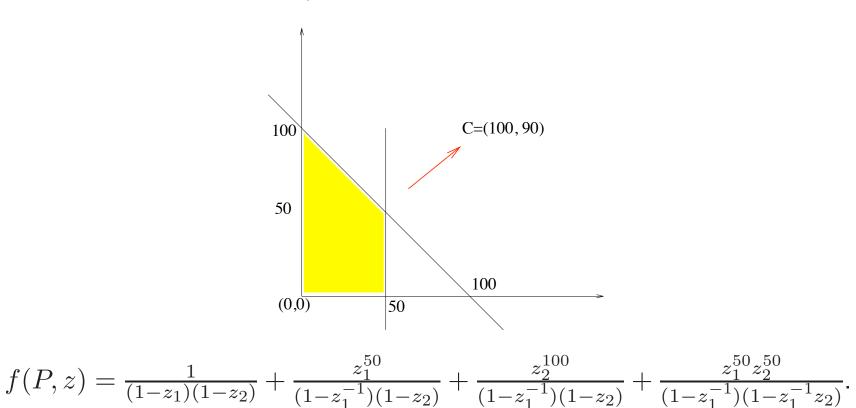
**Output:** The optimal value of maximize  $\{c \cdot x : Ax \leq b, x \geq 0, x \in \mathbb{Z}^d\}$ .



For fixed d, this algorithm runs in polynomial time (on the input size) by using the polynomiality of Barvinok's counting algorithm.

### **Toward More Direct Algorithms:**

Barvinok's algorithm computes the function  $f(P, z) = \sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$ , in the form:  $f(P, z) = \sum_{i \in I} \epsilon_i \frac{z^{u_i}}{\prod_{j=1}^n (1-z^{v_{ij}})}$ .



- Integer Optimization -

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### Changing Variables is IMPORTANT!!

 $f(P,z) = \sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$ , in the form:

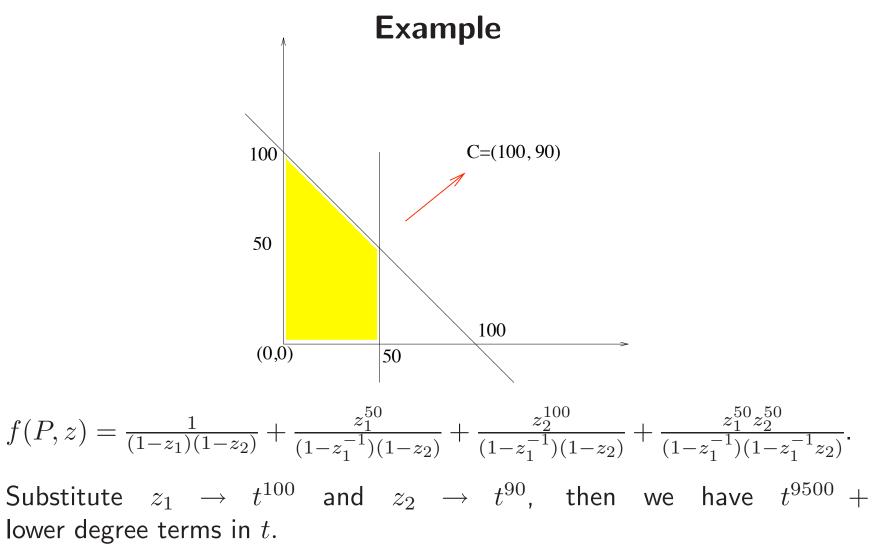
$$f(P, z) = \sum_{i \in I} \epsilon_i \frac{z^{u_i}}{\prod_{j=1}^n (1 - z^{v_{ij}})}.$$

If we make the substitutions  $z_i 
ightarrow t^{c_i}$ , then we have  $z^lpha 
ightarrow t^{c\cdot lpha}$ ,

$$f(P,z) \to \sum_{\alpha \in P \cap \mathbb{Z}^d} t^{c \cdot \alpha}$$

 $= t^M + ($ lower degree terms in t)

M is the optimal value of the integer linear programming problem!



- Integer Optimization -

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### **Monomial Substitution**

**Lemma:** (Barvinok-Woods) Let us fix k, the number of binomials in the denominator of a rational function. Given a rational function sum g of the form

$$g(x) = \sum_{i \in I} \alpha_i \frac{x^{u_i}}{\prod_{j=1}^k (1 - x^{v_{ij}})},$$

where  $u_i, v_{ij}$  are integral *d*-dimensional vectors, and a monomial map  $\psi : \mathbb{C}^n \longrightarrow \mathbb{C}^d$  given by the variable change  $x_i \rightarrow z_1^{l_{i1}} z_2^{l_{i2}} \dots z_n^{l_{in}}$  whose image does not lie entirely in the set of poles of g(x), then there exists a polynomial time algorithm which, computes the function  $g(\psi(z))$  as a sum of rational functions of the same shape as g(z).

### **A** Reformulation of Integer Linear Programming:

**GOAL:** Given  $A \in \mathbb{Z}^{m \times d}$ ,  $b \in \mathbb{Z}^m$ ,  $c \in \mathbb{Z}^d$ , and assume that number of variables d is fixed. Wish to solve the integer programming problem

maximize  $(c \cdot x)$  subject to  $x \in \{x | Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \},\$ 

In our setting this is

DETECTING THE HIGHEST DEGREE COEFFICIENT OF A POLYNOMIAL!

THE POLYNOMIAL IS GIVEN AS A SUM OF RATIONAL FUNCTIONS.

### Several different ways to do this!

### **Digging Algorithm: Laurent Series Expansion**

Input:  $A \in \mathbb{Z}^{m \times d}, b \in \mathbb{Z}^m, c \in \mathbb{Z}^d$ .

**Output:** The optimal value of maximize  $\{c \cdot x : Ax \leq b, x \geq 0, x \in \mathbb{Z}^d\}$ .

(A) Using Barvinok's algorithm and monomial substitution compute the rational function expression

$$\sum_{i \in I} E_i \frac{t^{c \cdot u_i}}{\prod_{j=1}^d (1 - t^{c \cdot v_{ij}})}.$$
 (3)

(B) Use the identity

$$\frac{1}{1 - t^{c \cdot v_{ij}}} = \frac{-t^{-c \cdot v_{ij}}}{1 - t^{-c \cdot v_{ij}}}$$

as necessary to enforce that all  $v_{ij}$  in (3) satisfy  $c \cdot v_{ij} < 0$ . So now the terms of the series are given in decreasing order with respect to the degree of t.

(3) For each of the rational functions in the sum compute a Laurent series expansion of the form

$$E_i t^{c \cdot u_i} \prod_{j=1}^d (1 + t^{c \cdot v_{ij}} + (t^{c \cdot v_{ij}})^2 + \ldots).$$

multiply out the factors and add the terms, group together those of the same degree in t. Thus we find a term expansion. Proceed in decreasing order with respect to the degree of t.

(4) Continue until a degree n of t is found such that for some the coefficient is non-zero in the expansion. Return n as the optimal value.

<sup>–</sup> Integer Optimization –

# Number of roots of a polynomial

We have a black box polynomial p(z). Assume coefficients are all 1 or 0. This happens for generic cost vectors!!

**LEMMA: (Argument principle)** Let C be a simple closed curve in the complex plane that contains no root of p(z) itself. Then

number of roots of p(z) (with multiplicity) inside  $C = \frac{1}{2\pi i} \left( \int_C \frac{p'(z)}{p(z)} dz \right).$ 

(A) Find an upper bound M on the absolute value of the roots. M=2 sufficient by a result of Cauchy.

(B) Let C be the square centered at the origin of size 2M

(C) Perform the integration numerically. High accuracy not necessary because the answer is an integer.

<sup>-</sup> Integer Optimization -

## Example

$$\begin{split} f(z) &= z^{25} + z^{24} + z^{23} + z^{22} + z^{21} + z^{20} + z^{19} + z^{18} + z^{17} + z^{16} + z^{15} + z^{14} + z^{13} + z^{12} + z^{11} + z^{10} + z^9 + z^8 + z^7 + z^6 + z^5 + z^4 + z^3 + z^2 + z \end{split}$$

p:=unapply(D(f)(z)/f(z),z);

Digits:=14:

evalf(Int(p(t-2\*I), t=-2..2,method=\_CCquad)+ Int(p(2+I\*t)\*I, t -Int(p(t+2\*I), t=-2..2, method=\_CCquad)-Int(p(-2+t\*I)\*I,

-13 INTEGRALS EQUAL 0.6 10 + 157.07963267950 I -1 > evalf(%/(2\*Pi\*I)) -> 25.0000000002 - 0.95492965855137 10

# **Boolean operations on rational functions**

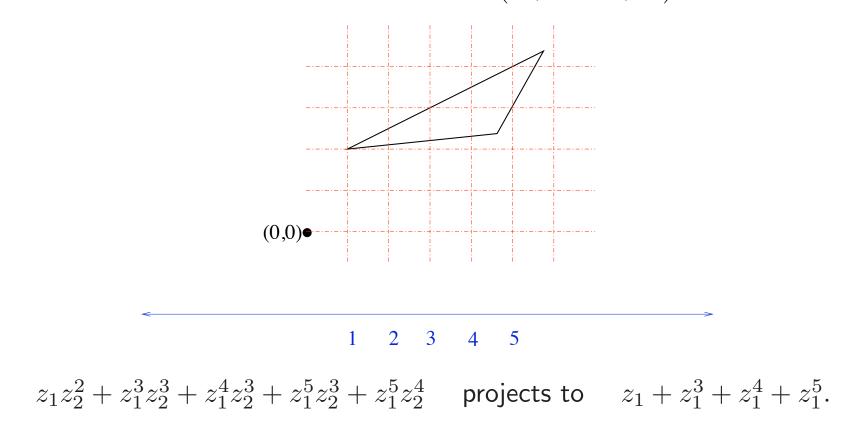
**Lemma:** Let  $S_1, S_2$  be finite subsets of  $\mathbb{Z}^n$  and let  $f(S_1, x)$  and  $f(S_2, x)$  be the corresponding generating functions, represented as short rational functions with at most k binomials in each denominator. Then there exist a polynomial time algorithm, which, given  $f(S_i, x)$ , computes

$$f(S_1 \cap S_2, x) = \sum_{i \in I} \gamma_i \frac{x^{u_i}}{(1 - x^{v_{i1}}) \dots (1 - x^{v_{is}})}$$

with  $s \leq 2k$  and  $\gamma_i$  rational numbers,  $u_i, v_{ij}$  nonzero integers.

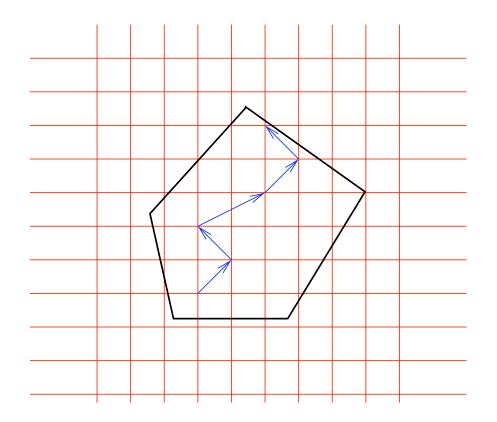
Same with finite unions or complements!

**The Projection Lemma** Lemma Consider a rational polytope  $P \subset R^n$  and a linear map T:  $\mathbb{Z}^n \to \mathbb{Z}^k$ . There is a polynomial time algorithm which computes a short representation of the generating function  $f(T(P \cap \mathbb{Z}^n), x)$ .



# **TEST SETS**

A TEST SET is a finite collection of integral vectors with the property that every feasible non-optimal solution of an integer program in the can be improved by adding a vector in the test set.



**Examples of test sets and augmentation methods:** Graver and Gröbner bases, Hilbert bases, integral basis method. Work by Hemmecke, Graver, Scarf, Thomas, Sturmfels, Weismantel et al. and others.

**TROUBLE** Test sets can be exponentially large even in fixed dimension!

**THEOREM:** (Barvinok-Woods 2003) When the dimension is fixed, Barvinok's rational functions can compute Hilbert bases or Graver bases in polynomial time (on the size of the input), as rational functions.

**THEOREM:** (LattE team 2004 + Sturmfels) When the dimension is fixed, Barvinok's rational functions can compute reduced Gröbner bases in polynomial time (on the size of the input), as rational functions.

## **Experimental Results**

The cost vector c, we choose the first d components of the vector

(213, -1928, -11111, -2345, 9123, -12834, -123, 122331, 0, 0).

	-										
Problem					a						
prob1	25067	49300	4 9717	62124	87608	88025	11 3673	119169			3
prob2	11948	23330	30635	44197	92754	123389	136951	140745			1
prob3	39559	61679	79625	99658	133404	137071	159757	173977			5
prob4	48709	55893	62177	65919	86271	87692	102881	109765			6
prob5	28637	48198	80330	91980	102221	135518	165564	176049			6
prob6	20601	40429	40429	45415	53725	61919	64470	69340	78539	95043	2
prob7	18902	26720	34538	34868	49201	49531	65167	66800	84069	137179	2
prob8	17035	45529	48317	48506	86120	100178	112464	115819	125128	129688	2
prob9	3719	20289	29067	60517	64354	65633	76969	102024	106036	119930	1
prob10	45276	70778	86911	92634	97839	125941	134269	141033	147279	153525	10

Table 1: We implemented the *BBS algorithm* and the *digging algorithm* in LattE. We solved several challenging knapsack problems by Aardal, Lenstra, and Lenstra

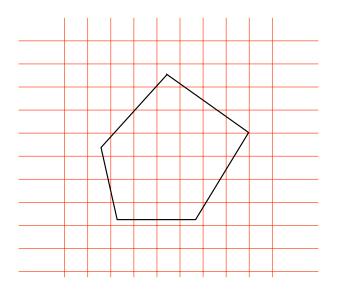
Problem	Value	Solution	Digging	BBS	CPLEX 6.6
prob1	9257735	[966 5 0 0 1 0 0 74]	51.4 sec.	> 3h	> 1h
prob2	3471390	[853 2 0 4 0 0 0 27]	24.8 sec.	> 10h	> 0.75h
prob3	21291722	[708 0 2 0 0 0 1 173]	48.2 sec.	> 12h	> 1.5h
prob4	6765166	[1113 0 7 0 0 0 0 54]	34.2 sec.	> 5h	> 1.5h
prob5	12903963	[1540 1 2 0 0 0 0 103]	34.5 sec.	> 5h	> 1.5h
prob6	2645069	[1012 1 0 1 0 1 0 20 0 0]	143.2 sec.	> 4h	> 2h
prob7	22915859	[782 1 0 1 0 0 0 186 0 0]	142.3 sec.	> 4h	> 1h
prob8	3546296	[1 385 0 1 1 0 0 35 0 0]	469.9 sec.	> 3.5h	> 2.5h
prob9	15507976	[31 11 1 1 0 0 0 127 0 0]	1,408.2 sec.	> 11h	4.7 sec.
prob10	47946931	[0 705 0 1 1 0 0 403 0 0]	250.6 sec.	> 11h	> 1h

Table 2: Optimal values, solutions, and running times for each problem.

# Generating Functions Algorithms in Integer Optimization

## **LECTURE III** Contributions to Integer Non-Linear Programming

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# Integer Semi-algebraic Optimization

**Problem:** Let  $f, g_i$  are d-variate polynomials with integral coefficients.

maximize  $f(x_1, \ldots, x_d)$  subject to  $g_i(x_1, \ldots, x_d) \ge 0, x \in \mathbb{Z}$ 

Also called Integer Semialgebraic Optimization.

# Question: What happens if we assume the number of variables is fixed?

Positive Notes: problem contains *Integer Linear Programming*, Lenstra's Algorithm guarantees is solvable in polynomial time for fixed dimension. Also, *Integer Semidefinite Programming* runs in polynomial time in fixed dimension by Khachiyan and Porkolab's work.

Negative Notes: continuous polynomial optimization over polytopes, without fixed dimension, is NP-hard and no FPTAS is possible! the max-cut problem can be modeled as minimizing a quadratic form over the cube  $[-1,1]^d$ .

## The whole picture

Table 3: Computational complexity of polynomial integer problems in fixed dimension.

	Ту	pe of objective functi	on
Type of constraints	linear	convex polynomial	arbitrary polynomial
Linear constraints,	polytime (*) . 介	$\Leftarrow  polytime \ (**) \\ \uparrow$	NP-hard (a) ↓
Convex semialgebraic constraints,	polytime (**)	$\Leftarrow$ polytime (**)	NP-hard (c)
Arbitrary polynomial constraints,	undecidable (b) =	$\Rightarrow$ undecidable (d) $\Rightarrow$	> undecidable (e

## Integer Polynomial Optimization over a Polytope

**Problem:** Let f be a d-variate polynomial with integral coefficients. Now the  $g_i(x)$  are *linear inequalities*.

maximize  $f(x_1, \ldots, x_d)$  subject to  $g_i(x_1, \ldots, x_d) \ge 0, x \in \mathbb{Z}$ 

**Example:** Consider this problem from *MINLPLIB* library

$$\max \quad 100 \left( \frac{1}{2} + i_2 - \left( \frac{3}{5} + i_1 \right)^2 \right)^2 + \left( \frac{2}{5} - i_1 \right)^2$$
(4)  
s.t.  $i_1, i_2 \in [0, 200] \cap \mathbb{Z}.$ 

Its optimal solution is  $i_1 = 1$ ,  $i_2 = 2$  with an objective value of 0.72.

# Integer Polynomial Optimization over a Polytope

**Theorem** (D,Hemmecke,Koeppe,Weismantel) Let the number of variables d be fixed. Let  $f(x_1, \ldots, x_d)$  be a polynomial of maximum total degree D with integer coefficients, and let P be a convex rational polytope defined by linear inequalities in d variables.

(1) We can construct an increasing sequence of lower bounds  $\{L_k\}$  and a decreasing sequence of upper bounds  $\{U_k\}$  to the optimal value

$$f^* = \text{maximize } f(x_1, x_2, \dots, x_d) \text{ subject to } x \in P \cap \mathbb{Z}^d.$$
 (5)

The bounds  $L_k$ ,  $U_k$  can be computed in time polynomial in k, the input size of P and f, and the maximum total degree D and they satisfy the inequality  $U_k - L_k \leq f^* \cdot (\sqrt[k]{|P \cap \mathbb{Z}^d|} - 1)$ .

(2) Moreover, if f is positive semidefinite over the polytope (i.e.  $f(x) \ge 0$  for all  $x \in P$ ), there exists a fully polynomial-time approximation scheme (FPTAS) for the optimization problem (5).

The construction of the bounds and algorithm uses **Barvinok's rational functions.** 

## **Polynomial Evaluation Lemma**

**Lemma:** Given a Barvinok rational function f(S), representing a finite set of lattice points S, and a polynomial g with integer coefficients we can compute, in time polynomial on the input size a Barvinok rational function for the generating function

$$f(S, g, z) = \sum_{a \in S} g(a) z^a.$$

NOTE: This is *independent* of the degree of g.

## **Differential Operators give the coefficients:**

We can define the basic differential operator associated to  $f(x) = x_r$ 

$$z_r \frac{\partial}{\partial z_r} \cdot \sum_{\alpha \in P \cap \mathbb{Z}^d} z^\alpha = \sum_{\alpha \in P \cap \mathbb{Z}^d} z_r \frac{\partial}{\partial z_r} z^\alpha = \sum_{\alpha \in P \cap \mathbb{Z}^d} \alpha_r z^\alpha$$

Next if  $f(z) = c \cdot z_1^{\beta_1} \cdot \ldots \cdot z_d^{\beta_d}$ , then we can compute again a rational function representation of  $g_{P,f}(z)$  by repeated application of basic differential operators:

$$c\left(z_1\frac{\partial}{\partial z_1}\right)^{\beta_1}\cdot\ldots\cdot\left(z_d\frac{\partial}{\partial z_d}\right)^{\beta_d}\cdot g_P(z)=\sum_{\alpha\in P\cap\mathbb{Z}^d}c\cdot\alpha^{\beta}z^{\alpha}$$

# Sketch of proof of theorem/algorithm

*Input:* A rational convex polytope  $P \subset \mathbb{R}^d$ , a polynomial objective  $f \in \mathbb{Z}[x_1, \ldots, x_d]$  of maximum total degree D.

*Output:* An increasing sequence of lower bounds  $L_k$ , and a decreasing sequence of upper bounds  $U_k$  reaching the maximal function value  $f^*$  of f over all lattice points of P.

**W.I.o.g:** We can assume f is positive semidefinite. Else translate it!

Via Barvinok's algorithm compute a short rational function expression for the generating function  $g_P(z) = \sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$ . From  $g_P(z)$  compute the number  $|P \cap \mathbb{Z}^d| = g_P(1)$  of lattice points in P. Can be done in polynomial time.

### How to define such sequences to approximate the maximum?

**Lemma:** For a collection  $S = \{s_1, \ldots, s_r\}$  of non-negative real numbers,  $\max\{s_i | s_i \in S\}$  equals  $\lim_{k \to \infty} \sqrt[k]{\sum_{j=1}^r s_j^k}$ .

From the rational function representation  $g_P(z)$  of the generating function  $\sum_{\alpha \in P \cap \mathbb{Z}^d} z^{\alpha}$  we can compute the rational function representation of  $g_{P,f^k}(z)$ of  $\sum_{\alpha \in P \cap \mathbb{Z}^d} f^k(\alpha) z^{\alpha}$  in polynomial time by application of the Polynomial Evaluation Lemma.

define

$$L_k := \sqrt[k]{g_{P,f^k}(1)/g_{P,f^0}(1)}$$
 and  $U_k := \sqrt[k]{g_{P,f^k}(1)}.$ 

If you want the optimal value compute the sequences  $L_k, U_k$  until  $\lfloor U_k \rfloor - \lfloor L_k \rceil < 1$  stop and return  $\lfloor L_k \rceil = \lfloor U_k \rfloor$  as the optimal value.

## **Example**

maximize  $x^3y$  subject to

$$\{(x,y)|3991 \le 3996 x - 4y \le 3993, \ 1/2 \le x \le 5/2, \ \text{integer}\}.$$

Region contains only 2 lattice points. The sum of rational functions encoding the lattice points is

$$x^{2}y^{1000}\left(1-\frac{1}{xy^{999}}\right)^{-1}\left(1-y^{-1}\right)^{-1} + xy\left(1-xy^{999}\right)^{-1}\left(1-y^{-1}\right)^{-1} + \frac{xy}{\left(1-xy^{999}\right)\left(1-y\right)} + x^{2}y^{1000}\left(1-\frac{1}{xy^{999}}\right)^{-1}\left(1-y\right)^{-1}.$$

The true optimal value is 8000. Here are a few iterations:

Iteration	Lower bound	Upper bound
1	4000.500000	8001.
2	5656.854295	8000.000063
3	6349.604210	8000.00000
4	6727.171325	8000.00000
5	6964.404510	8000.00000
6	7127.189745	8000.00000
7	7245.789315	8000.00000
8	7336.032345	8000.00000
9	7406.997700	8000.00000
10	7464.263930	8000.00000
11	7511.447285	8000.00000
12	7550.994500	8000.00000
13	7584.620115	8000.00000
14	7613.561225	8000.00000
15	7638.732830	8000.00000

## Example

### Recall

$$\max \quad 100 \left( \frac{1}{2} + i_2 - \left( \frac{3}{5} + i_1 \right)^2 \right)^2 + \left( \frac{2}{5} - i_1 \right)^2$$
(6)  
s.t.  $i_1, i_2 \in [0, 200] \cap \mathbb{Z}$ 

Using the bounds on  $i_1$  and  $i_2$  we obtain an upper bound of  $165 \cdot 10^9$  for the objective function. Use it to convert the problem into one where all feasible points have a non-negative objective value.

Expanding the new objective function and translating it into a differential

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### operator yields

$$\frac{4124999999947}{25} \text{Id} - 28z_2 \frac{\partial}{\partial z_2} + \frac{172}{5} z_1 \frac{\partial}{\partial z_1} - 117 \left( z_1 \frac{\partial}{\partial z_1} \right)^{(2)} - 100 \left( z_2 \frac{\partial}{\partial z_2} \right) \\ + 240 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right) + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(2)} - 240 \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} - 100 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} - 100 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(2)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(2)} - 240 \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} - 100 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_2} \right)^{(2)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(2)} - 240 \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} - 100 \left( z_2 \frac{\partial}{\partial z_2} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right) \left( z_1 \frac{\partial}{\partial z_1} \right)^{(3)} + 200 \left( z_1 \frac{\partial}{\partial z_2} \right)^{(3)} + 200 \left( z_2 \frac{\partial}{\partial z_2} \right)^{(3)} + 200$$

The short generating function can be written as  $g(z_1, z_2) = \left(\frac{1}{1-z_1} - \frac{z_1^{201}}{1-z_1}\right) \left(\frac{1}{1-z_2} - \frac{z_2^{201}}{1-z_2}\right).$ 

In this example, the number of lattice points is  $|P \cap \mathbb{Z}^2| = 40401$ . The first bounds are  $L_1 = 139463892042.292155534$ ,  $U_1 = 28032242300500.723262442$ . After 30 iterations the bounds become  $L_{30} = 164999998845.993553019$  and  $U_{30} = 16500000475.892451381$ . The new optimal objective value is 16499999999.28.

## **Mixed Integer Case**

What is the computational complexity, of the *non-linear* mixed integer problem?

$$\max f(x_1, \dots, x_{d_1}, z_1, \dots, z_{d_2}):$$
(7a)

$$Ax + Bz \le b \tag{7b}$$

$$x_i \in \mathbb{R}$$
 for  $i = 1, \dots, d_1$ , (7c)

$$z_i \in \mathbb{Z} \qquad \qquad \text{for } i = 1, \dots, d_2, \qquad (7d)$$

where f is a polynomial function of maximum total degree D with rational coefficients, and  $A \in \mathbb{Z}^{m \times d_1}$ ,  $B \in \mathbb{Z}^{m \times d_2}$ ,  $b \in \mathbb{Z}^m$  (here we assume that  $Ax + Bz \leq b$  describes a convex polytope, which we denote by P).

**Theorem** Let the dimension  $d = d_1 + d_2$  be fixed.

There exists a fully polynomial time approximation scheme (FPTAS) for the mixed integer polynomial optimization problem for all polynomial functions  $f \in \mathbb{Q}[x_1, \ldots, x_{d_1}, z_1, \ldots, z_{d_2}]$  that are non-negative on the feasible region.

Moreover, the restriction to non-negative polynomials is necessary, as there does not even exist a polynomial time approximation scheme (PTAS) for the maximization of *arbitrary* polynomials over mixed-integer sets in polytopes, even for fixed dimension  $d \ge 2$ .

# OTHER RATIONAL FUNCTION TECHNIQUES

# THE PROBLEMS

Given a  $d \times n$  integral matrix A and integral d-vectors c, b. Solve:

• General

maximize cx subject to  $x \in P = \{x | Ax = b, x \ge 0\} \cap \mathbb{Z}^d$ ,

## • Binary

maximize cx subject to  $x \in P = \{x | Ax = b, x_i \in \{0, 1\}\},\$ 

**Two Non-linear Models LEMMA** Let  $A_i$  denote the columns of the matrix A.

$$\prod_{j=1}^{n} (1 + z^{A_j} t^{c_j}) = \sum_{\text{over feasible } b} \left( \sum_{\alpha \in P} t^{c\alpha} \right) z^b.$$

There is a monomial  $t^{\beta}x^{b}$  in expansion if and only if there is a 0/1 vertex of  $P = \{x | Ax = b, 0 \le x \le 1, \}$  of cost value  $\beta$ .

**LEMMA** Let  $A_i$  denote the columns of the matrix A.

$$\frac{1}{\prod_{j=1}^{n} (1 - z^{A_j} t^{c_j})} = \sum t^{\beta} z^b.$$

There is a monomial  $t^{\beta}x^{b}$  if and only if  $P = \{x | Ax = b, x \ge 0, integer\}$  has a lattice point of cost value  $\beta$ .

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## Example

maximize x + 2y + z

subject to  $\{(x, y, z) \in \mathbb{Z}^3 \mid 3x + 5y + 17z = b, x \ge 0, y \ge 0, z \ge 0\}$ 

It is encoded, for any right-hand-side into

$$(1+x^{3}t)(1+x^{5}t^{2})(1+x^{17}t)$$

This is a COMPACT representation of any optimal value. In expanded form

$$1 + x^{17}t + x^5t^2 + x^{22}t^3 + x^3t + x^{20}t^2 + x^8t^3 + x^{25}t^4.$$

Suppose now it is not bounded,

$$\frac{1}{(1-x^3t)\left(1-x^5t^2\right)\left(1-x^{17}t\right)}.$$

Its Multivariate Taylor Series expansion is

 $1 + x^{3}t + x^{5}t^{2} + x^{6}t^{2} + x^{8}t^{3} + x^{9}t^{3} + x^{10}t^{4} + x^{11}t^{4} + x^{12}t^{4} + x^{13}t^{5} + x^{14}t^{5} + (t^{6} + t^{5})x^{15} + x^{16}t^{6} + (t^{6} + t)x^{17} + (t^{7} + t^{6})x^{18} + x^{19}t^{7} + (t^{8} + t^{7} + t^{2})x^{20} + (t^{8} + t^{7})x^{21} + (t^{8} + t^{3})x^{22} + (t^{9} + t^{8} + t^{3})x^{23} + (t^{9} + t^{8})x^{24} + (t^{10} + t^{9} + t^{4})x^{25} + (t^{10} + t^{9} + t^{4})x^{26} + (t^{10} + t^{9} + t^{5})x^{27} + (t^{11} + t^{10} + t^{5})x^{28} + \dots$ 

Note that we have  $t^{10}x^{28}$  because we have ONE Knapsack solution x = 6, y = 2, z = 0

**IMPORTANT** Note that if t = 1, we COUNT lattice points.

# SELECTING A COEFFICIENT!!

Let  $\phi_A(b)$  be the coefficient of  $z^b := z_1^{b_1} \cdots z_m^{b_m}$  of the function

$$f(z) = \frac{1}{(1 - z^{A_1}) \cdots (1 - z^{A_d})}$$

expanded as a power series centered at z = 0.

# Fast Fourier Transforms

**THEOREM** The coefficients of the product of n polynomials of degree  $d_i$  can be computed by FFT in O(dln(d)ln(n)) arithmetic operations where  $d = \sum d_i$ .

IDEA:

- Polynomials represented as monomials are wasteful!
- Represent polynomials  $p_1(x), p_2(x)$  of degree d sets of 2d + 1 points  $(y_j, p_i(y_j))$ .
- Pointwise multiply these values to get  $p_1(x)p_2(x)$ . We get the point representation of  $p_1(x)p_2(x)$ . Takes O(d)
- Choose the values of evaluation cleverly  $y_i$  comes from roots of unity! This can be done in  $O(d \log(d))$  using the **Fast Fourier Transform**.

- Interpolate to create the coefficient representation of the polynomial  $p_1(x)p_2(x)$  through an application of FFT transform, again takes  $O(d\log(d))$  operations.
- VERY fast code available now, parallelizable.
- Idea goes back at least to the 1970's when Statisticians used it for enumerating contingency tables with given margins. See Diaconis-Gangolli 1995 survey.

# A Case Study: 0/1-Knapsack Problems

**THEOREM** Using dynamic programming one can solve the knapsack problem

maximize cx subject to  $a_1x_1 + a_2x_2 + \cdots + a_nx_n = b \ x_i \in \{0, 1\},\$ 

in O(nb) steps.

**THEOREM** (Nesterov 2004) The same knapsack problem, using Fast Fourier Transforms, can be solved in  $O(b \log^2(n))$  steps.

QUESTION: How does this idea behave in practice? Can one improve the complexity further?

## Multivariate Complex Analysis view

Work by Beck et al., Lasserre et al., Pemantle et al.

$$\phi_A(b) = \frac{1}{(2\pi i)^m} \int_{|z_1|=\epsilon_1} \cdots \int_{|z_m|=\epsilon_m} \frac{z_1^{-b_1-1} \cdots z_m^{-b_m-1}}{(1-z^{A_1}) \cdots (1-z^{A_d})} dz$$

Here  $0 < \epsilon_1, \ldots, \epsilon_m < 1$  are different numbers such that we can expand all the  $\frac{1}{1-z^{M_k}}$  into the power series about 0.

# VISIT: www.math.ucdavis.edu/~latte

# www.math.ucdavis.edu/~totalresidue

## with lots of nice stuff about lattice points on polytopes...

# THANK YOU!