

LABORATOIRE d'ANALYSE et d'ARCHITECTURE des SYSTEMES

DUALITY AND INTEGER PROGRAMMING

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Current solvers (CPLEX, XPRESS-MP) are rather efficient and can solve many large size problems with thousands of variables. However, the very small **5-variables knapsack** problem

$$\text{max} \quad 213x_1 - 1928x_2 - 11111x_3 - 2345x_4 + 9123x_5$$

$$12223x_1 + 12224x_2 + 36674x_3 + 61119x_4 + 85569x_5 = 89643482$$

x_1, x_2, x_3, x_4, x_5 nonnegative integers

still resists all efficient solvers (takes **HOURS** on CPLEX 8.1 and XPRESS-MP !) Optimal solution $x^* = (7334, 0, 0, 0, 0)$...!

Any insight on integer problems is welcome

The integer program

$$\max\{c'x \mid Ax = b; \quad x \in \mathbf{N}^n\}$$

with $A = [A_1, \dots, A_n] \in \mathbf{Z}^{m \times n}$, $b \in \mathbf{Z}^m$ has a dual problem

$$\min_{f \in \Gamma} \{f(b) \mid f(A_j) \geq c_j, \quad j = 1, \dots, n\}$$

where Γ is the space of functions $f : \mathbf{R}^m \rightarrow \mathbf{R}$ that are **super-additive** ($f(a + b) \geq f(a) + f(b)$), and with $f(0) = 0$. (Jeroslow, Wolsey, etc ...)

Elegant but rather abstract and not very practical; however, used to derive valid inequalities. Moreover, the celebrated **Gomory cuts** used in Solvers (CPLEX, XPRESS-MP, ..) can be interpreted as such superadditive functions.

CONTINUOUS OPTIM.

$$f(b, c) := \max c'x$$

$$\text{s.t.} \begin{cases} Ax \leq b \\ x \in \mathbf{R}_+^n \end{cases}$$

INTEGRATION

$$\hat{f}(b, c) := \int_{\Omega} e^{c'x} dx$$

$$\Omega := \begin{cases} Ax \leq b \\ x \in \mathbf{R}_+^n \end{cases}$$

DISCRETE OPTIM.

$$f_d(b, c) := \max c'x$$

$$\text{s.t.} \begin{cases} Ax \leq b \\ x \in \mathbf{N}^n \end{cases}$$

SUMMATION

$$\hat{f}_d(b, c) := \sum_{\Omega} e^{c'x}$$

$$\Omega := \begin{cases} Ax \leq b \\ x \in \mathbf{N}^n \end{cases}$$

$$e^{f(b,c)} = \lim_{r \rightarrow \infty} \widehat{f}(b, rc)^{1/r}; \quad e^{f_d(b,c)} = \lim_{r \rightarrow \infty} \widehat{f}_d(b, rc)^{1/r}.$$

or, equivalently

$$f(b, c) = \lim_{r \rightarrow \infty} \frac{1}{r} \ln \widehat{f}(b, rc); \quad f_d(b, c) = \lim_{r \rightarrow \infty} \frac{1}{r} \ln \widehat{f}_d(b, rc).$$

Note in passing that we have the **mean-value** theorem

$$\widehat{f}(b, c) = e^{c'x^*} \times \lambda(\Omega) = e^{c'x^*} \times \text{vol}(\Omega).$$

and a **Maslov “max” mean-value theorem**

$$f(b, c) = c'x^* = c'x^* \times \mu_{maslov}(\Omega) = c'x^*.$$

Maslov duality for “linear” functionals in the Max-Plus algebra

$f \mapsto L(f)$ a “linear” functional in the “max +” algebra, i.e.,

$$L(f \vee h) = L(f) \vee L(h) \quad \forall f, h \Rightarrow L(f) = \int f d\mu_{maslov}.$$

with

$$\mu_{maslov}(A \cup B) = \mu_{maslov}(A) \vee \mu_{maslov}(B) \quad \forall A, B \in \mathcal{B}.$$

CONTINUOUS OPTIM.

—
**Legendre-Fenchel
Duality**



INTEGRATION
—
**Laplace-Transform
Duality**

DISCRETE OPTIM.

—
??



SUMMATION
—
**Z-Transform
Duality**



Legendre-Fenchel duality : $f : \mathbf{R}^n \rightarrow \mathbf{R}$ convex; $f^* : \mathbf{R}^n \rightarrow \mathbf{R}$.

$$\lambda \mapsto f^*(\lambda) = \mathcal{F}(f)(\lambda) := \sup_y \{\lambda'y - f(y)\}.$$

(One-sided) Laplace-Transform: $f : \mathbf{R}_+^n \rightarrow \mathbf{R}$; $F : \mathbf{C}^n \rightarrow \mathbf{C}$.

$$\lambda \mapsto F(\lambda) = \mathcal{L}(f)(\lambda) := \int_{\mathbf{R}_+^n} e^{-\lambda'y} f(y) dy.$$

(One-sided) Z-Transform: $f : \mathbf{Z}_+^n \rightarrow \mathbf{R}$; $F : \mathbf{C}^n \rightarrow \mathbf{C}$.

$$\lambda \mapsto F(z) = \mathcal{Z}(f)(z) := \sum_{m \in \mathbf{Z}_+^n} z^{-m} f(m).$$

Observe :

$$\mathcal{L}(e^f)(\lambda) = \int_{\mathbf{R}_+^n} e^{-(\lambda'y - f(y))} dy$$

and

$$e^{\mathcal{F}(f)(\lambda)} = \begin{cases} e^{\sup_y (\lambda'y - f(y))} = \sup_y \{e^{(\lambda'y - f(y))}\} \\ \oint_{\text{Maslov}} e^{(\lambda'y - f(y))} dy \end{cases}$$

$$\text{exponential (Fenchel (f))} = \text{Laplace (exponential (f))}$$

in “max , × ” algebra

with $\Omega := \{x \in \mathbf{R}^n \mid Ax \leq b; x \geq 0\}$

Fenchel-duality:

$$f(b, c) := \max_{x \in \Omega} c'x$$

$$f^*(\lambda, c) := \min_y \{\lambda'y - f(y, c)\}$$

with : $A'\lambda - c \geq 0; \lambda \geq 0$

Laplace-duality

$$\hat{f}(b, c) := \int_{\Omega} e^{c'x} dx$$

$$\hat{F}(\lambda, c) := \int e^{-\lambda y} \hat{f}(y, c) dy$$

$$= \int_{x \geq 0} e^{c'x} \left[\int_{Ax \leq y} e^{-\lambda y} dy \right] dx$$

$$= \frac{1}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A'\lambda - c)_k}$$

with : $\Re(A'\lambda - c) > 0; \Re(\lambda) > 0$

Fenchel-duality:

$$f(b, c) = \min_{\lambda} \{ \lambda' b - f^*(\lambda, c) \}$$
$$= \min_{\lambda} \{ b' \lambda \mid A' \lambda \geq c; \lambda \geq 0 \}$$

Laplace-duality

$$\hat{f}(b, c) = \frac{1}{(2i\pi)^m} \int_{\Gamma} e^{\lambda' b} \hat{F}(\lambda, c) d\lambda$$
$$= \int_{\Gamma} \frac{(2i\pi)^{-m} e^{\lambda' b}}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A' \lambda - c)_k} d\lambda$$

where $\Gamma \subset \mathbf{C}^m$

and $\int_{\Gamma} = \int_{u_1 - i\infty}^{u_1 + i\infty} \cdots \int_{u_m - i\infty}^{u_m + i\infty}$
with $u \in \mathbf{R}^m$; $A' u > c$; $u > 0$.

Polytope volume: – > Brion, Brion and Vergnes, Barvinok and Pommersheim, Lasserre and Zeron, ...

Multidimensional integration over a simplex, ellipsoid – > Lasserre and Zeron.

Remark:

Logarithmic Barrier function of $\min\{\lambda'b \mid A'\lambda - c \geq 0; \lambda \geq 0\}$

$$\lambda \mapsto \varphi(r, \lambda) := r\lambda'b - \sum_{k=1}^n \ln(A'\lambda - c)_k - \sum_{j=1}^m \ln(\lambda_j).$$

$$\begin{aligned} \hat{f}(b, rc) &= \frac{1}{r^n} \hat{f}(rb, c) = \frac{1}{r^n (2i\pi)^m} \int_{\Gamma} \frac{e^{r\lambda'b}}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A'\lambda - c)_k} d\lambda \\ &= \frac{1}{r^n (2i\pi)^m} \int_{\Gamma} e^{\varphi(r, \lambda)} d\lambda \end{aligned}$$

Hence,

$$\lim_{r \rightarrow \infty} e^{\min_{\lambda} \varphi(r, \lambda)} = \lim_{r \rightarrow \infty} \min_{\lambda} e^{\varphi(r, \lambda)} = \lim_{r \rightarrow \infty} \frac{1}{r^n (2i\pi)^m} \int_{\Gamma} e^{\varphi(r, \lambda)} d\lambda$$

Fenchel-duality:

$$f(b, c) = \min_{\lambda} \{ \lambda' b - f^*(\lambda, c) \}$$
$$= \min_{\lambda} \{ b' \lambda \mid A' \lambda \geq c; \lambda \geq 0 \}$$

Simplex Algorithm

Vertices of the polyhedron

$$\{ Ax \leq b; x \geq 0 \}$$

Laplace-duality

$$\hat{f}(b, c) = \frac{1}{(2i\pi)^m} \int_{\Gamma} e^{\lambda' b} \hat{F}(\lambda, c) d\lambda$$
$$= \int_{\Gamma} \frac{(2i\pi)^{-m} e^{\lambda' b}}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A' \lambda - c)_k} d\lambda$$

where $\int_{\Gamma} = \int_{u_1 - i\infty}^{u_1 + i\infty} \cdots \int_{u_m - i\infty}^{u_m + i\infty}$
with $u \in \mathbf{R}^m$; $A' u > c$; $u > 0$.

Cauchy 's Residue Technique

Poles of the function

$$\frac{1}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A' \lambda - c)_k}$$

THE SAME!

Laplace-Duality algorithm: Cauchy 's Residue Technique

$$\hat{f}(b, c) = \frac{1}{(2i\pi)^m} \int_{\Gamma} \frac{e^{\lambda'b}}{\prod_{j=1}^m \lambda_j \prod_{k=1}^n (A'\lambda - c)_k} d\lambda.$$

Integration w.r.t. $\lambda_1, \lambda_2, \dots, \lambda_m$ (one variable at a time) by repeated application of Cauchy 's residue theorem.

Yields a tree of depth m whose level k is integration w.r.t. λ_k on the integration path $\{u_k - i\infty, u_k + i\infty\}$ while the other variables $\lambda_{k+1}, \lambda_{k+2}, \dots, \lambda_m$ are fixed, on their respective integration path $\{u_j - i\infty, u_j + i\infty\}$. Application of Cauchy 's Residue Theorem yields a rational fraction in the variables $\lambda_{k+1}, \dots, \lambda_m$, and so on, until level $m - 1$ where one obtains a rational fraction in the single variable λ_m .

Brion and Vergne 's continuous formula

Terminology of LP in standard form:

Let $A_\sigma := [A_{\sigma_1} | \dots | A_{\sigma_m}]$ be a **basis** of $\max\{c'x \mid Ax = b; x \geq 0\}$, with $x(\sigma)$ the corresponding **vertex**, $\pi^\sigma := c'_\sigma A_\sigma^{-1}$ the associated **dual variable** and the **reduced cost** vector $c_k - \pi^\sigma A_k$, $k \notin \sigma$. Then :

$$\hat{f}(b, c) = \sum_{x(\sigma): \text{vertex of } \Omega(b)} \frac{e^{c'x(\sigma)}}{\det(A_\sigma) \prod_{k \notin \sigma} (-c_k + \pi^\sigma A_k)}$$

from which it easily follows that

$$\log \left[\lim_{r \rightarrow \infty} \hat{f}(b, rc)^r \right]^{1/r} = \max_{\text{vertex of } \Omega(b)} c'x(\sigma).$$

Let $\Omega(y) := \{x \in \mathbf{R}^n \mid Ax = y; \quad x \geq 0\}$.

Continuous Laplace-duality Discrete Z-duality

$$\hat{f}(y, c) := \int_{\Omega(y)} e^{c'x} d\mu$$

$$\hat{f}_d(y, c) := \sum_{x \in \Omega(y) \cap \mathbf{Z}^n} e^{c'x}$$

$$\hat{F}(\lambda, c) := \int_{\mathbf{R}_+^m} e^{-\lambda y} \hat{f}(y, c) dy$$

$$\hat{F}_d(z, c) := \sum_{y \in \mathbf{Z}^m} z^{-y} \hat{f}_d(y, c)$$

$$= \frac{1}{\prod_{k=1}^n (A'\lambda - c)_k}$$

$$= \prod_{k=1}^n \frac{1}{1 - e^{c_k} z_1^{-A_{1k}} \dots z_m^{-A_{mk}}}$$

with $\Re((A'\lambda)_k) > c_k \quad \forall k$

with $|z_1^{A_{1k}} \dots z_m^{A_{mk}}| > e^{c_k} \quad \forall k$

Continuous Laplace-duality

I. $\hat{F}(\lambda, c) \longrightarrow f(b, c)$

by Inverse Laplace Transform
and Cauchy 's Residue Th.

II. Data appear as

COEFFICIENTS

of dual variables λ in $\hat{F}(\lambda, c)$

III. The poles of $\hat{F}(\lambda, c)$ are

REAL

Discrete Z-duality

I. $\hat{F}_d(z, c) \longrightarrow f_d(b, c)$

by Inverse Z-Transform
and Cauchy 's Residue Th.

II. Data appear as

EXPONENTS !

of dual variables z in $\hat{F}_d(z, c)$.

III. The poles of $\hat{F}_d(z, c)$ are

COMPLEX

and much more numerous !

Let $\sigma := [A_{\sigma_1} | \cdots | A_{\sigma_m}]$ be a feasible **basis** of $\{Ax = b, x \geq 0\}$, with $\mu(\sigma) := \det(A_\sigma)$, and “dual” variable $\pi^\sigma A_\sigma = c_\sigma$.

Continuous Laplace-duality Discrete Z-duality

basis $\sigma \rightarrow$ poles : $A'_\sigma \lambda = c_\sigma$

basis $\sigma \rightarrow$ poles : $z^{A_\sigma} = e^{c_\sigma}$

a **single** real pole λ in \mathbf{R}^m

$\mu(\sigma)$ complex poles $z = e^\lambda$ in \mathbf{C}^m

$$\lambda = \pi^\sigma$$

$$\lambda = \pi^\sigma + i 2\pi \frac{v}{\mu(\sigma)}; v \in V_\sigma \subset \mathbf{Z}^m$$

$$V_\sigma = \{v \in \mathbf{Z}^m \mid v' A_\sigma = 0 \pmod{\mu(\sigma)}\}$$

So both $\hat{f}_d(b, c)$ and $\hat{f}(b, c)$ are computed with the **same Cauchy's residue technique** but the resulting contribution of each vertex for $\hat{f}_d(b, c)$ is **more complicated to evaluate** because of the additional **complex zeros**.

Brion and Vergne 's discrete formula

Let $\sigma := [A_{\sigma_1} | \cdots | A_{\sigma_m}]$ be a feasible basis of $\{Ax = b, x \geq 0\}$, with $\mu(\sigma) := \det(A_\sigma)$, and $\mu(\sigma)$ "dual" variables $z = e^\lambda e^{2i\pi v/\mu(\sigma)} \in \mathbb{C}^m$ with

$$\lambda = \pi^\sigma + i 2\pi \frac{v}{\mu(\sigma)}; \quad v \in V_\sigma := \{v \in \mathbb{Z}^m \mid v' A_\sigma = 0 \pmod{\mu(\sigma)}\}.$$

Re-interpreted with these data, Brion and Vergne 's original discrete formula reads (Lasserre)

$$\hat{f}_d(b, c) = \sum_{x(\sigma): \text{vertex of } \Omega(b)} e^{c'x(\sigma)} \times \frac{1}{\mu(\sigma)} \left[\sum_{v \in V_\sigma} \frac{e^{2i\pi v'b/\mu(\sigma)}}{\prod_{k \notin \sigma} (1 - e^{-(2i\pi v' A_k/\mu(\sigma))} e^{(c_k - \pi^\sigma A_k)})} \right]$$

Back to **optimization** : $f_d(b, c) = \max\{c'x \mid Ax = b; \quad x \in \mathbb{N}^n\}$.

Theorem (Lasserre). Assume that

$$\max_{\sigma} e^{c'x(\sigma)} \times \lim_{r \rightarrow \infty} \left[\sum_{v \in V_{\sigma}} \frac{e^{2i\pi v'b/\mu(\sigma)}}{\prod_{k \notin \sigma} (1 - e^{-(2i\pi v' A_k/\mu(\sigma))} e^{r(c_k - \pi^{\sigma} A_k)})} \right]^{1/r}$$

is attained at a **unique** basis σ^* . Then :

$$f_d(b, c) = c'x(\sigma^*) + \sum_{k \notin \sigma^*} (c_k - \pi^{\sigma^*} A_k) x_k^* = \sum_j c_j x_j^*.$$

σ^* is an optimal basis of the linear program, and

$x(\sigma^*)$ (resp. x^*) is an optimal solution of the **linear** (resp. **integer**) program.

In this case :

$$f_d(b, c) = c'x(\sigma^*) + \left\{ \begin{array}{l} \max \quad \sum_{k \notin \sigma^*} (c_k - \pi^{\sigma^*} A_k) x_k \\ A_{\sigma^*} u \quad + \quad \sum_{k \notin \sigma^*} A_k x_k \quad = b \\ u \in \mathbf{Z}^m; \quad x_k \in \mathbf{N} \quad \forall k \notin \sigma^* \end{array} \right.$$

$f_d(b, c) = c'x(\sigma^*) +$ **opt. value of GOMORY relaxation!**

As a consequence,

Corollary For $t \in \mathbf{N}$ sufficiently large, the gap $f(tb, c) - f_d(tb, c)$ between the *continuous* and *discrete* optimal values is a (constant) *periodic function* with period $\mu(\sigma^*) = \det(A_{\sigma^*})$, where σ^* is the optimal basis of the continuous LP.

The periodicity is due to the **complex poles** $z = e^{\lambda^*}$ with

$$\lambda^* = \pi^{\sigma^*} + i2\pi \frac{v}{\mu(\sigma^*)}, \quad v \in V_{\sigma^*} \subset \mathbf{Z}^m; \quad v' A_{\sigma^*} = 0 \pmod{\mu(\sigma^*)}.$$

KNAPSACK REVISITED

Let $a \in \mathbf{N}^{n+1}$, $0 \leq c \in \mathbf{R}^{n+1}$, $b \in \mathbf{N}$, and $s := \sum_i a_i$.

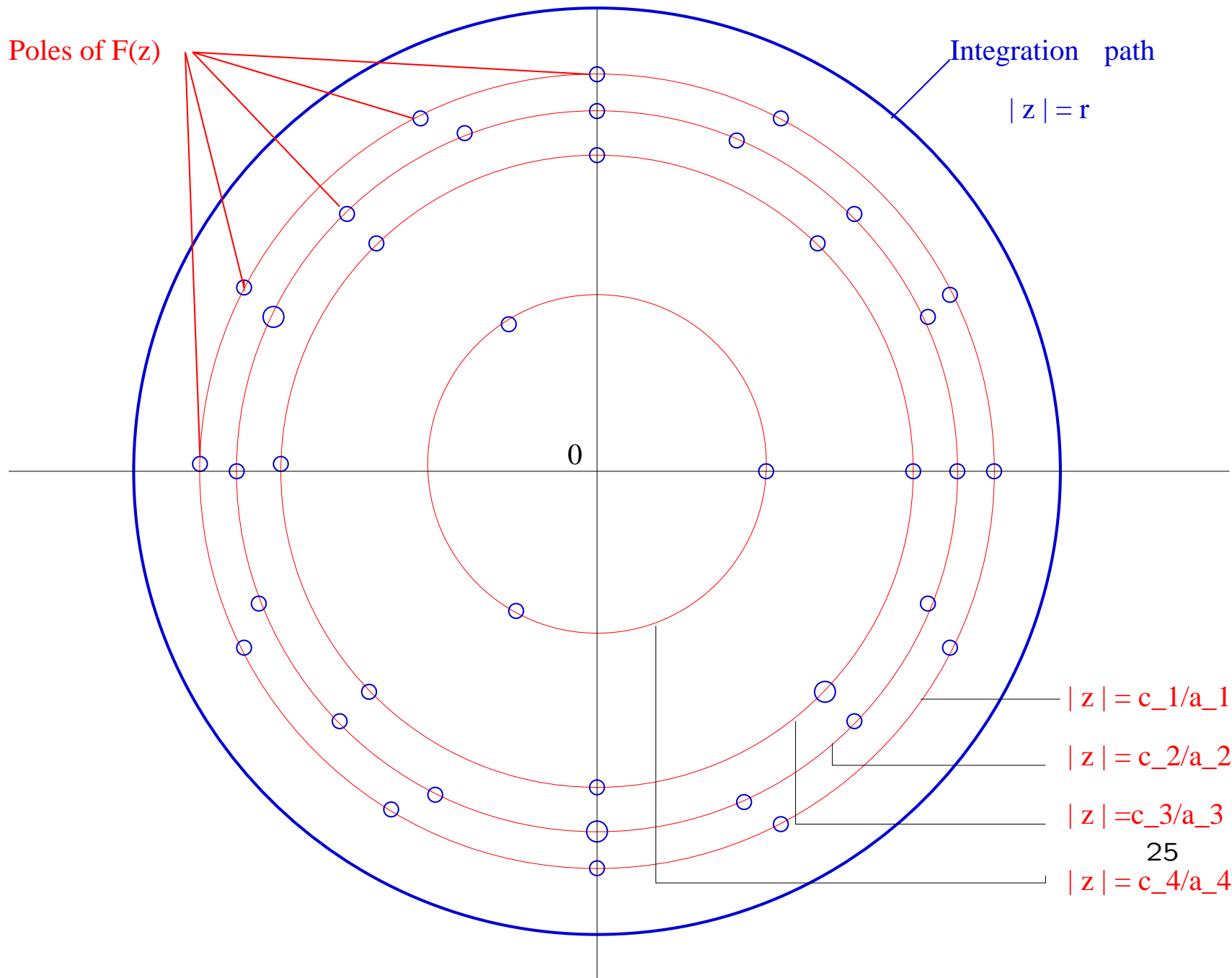
$$f_d(b, c) = \max_{x \in \mathbf{N}^n} \{c'x \mid a'x = b\}; \quad \hat{f}_d(b, c) = \sum_{x \in \mathbf{N}^n} \{e^{c'x} \mid a'x = b\}.$$

$$\hat{F}_d(z, c) = \frac{z^s}{\prod_{j=1}^n (z^{a_j} - e^{c_j})}; \quad \hat{f}_d(b, c) = \frac{1}{2i\pi} \int_{|z|=\gamma} z^{b-1} \hat{F}_d(z, c) dz$$

with $\gamma \in \mathbf{R}^+$ and $\ln \gamma > \max_{i=1}^n c_i/a_i$.

The poles $\{z_{jk}\}$ of \hat{F} satisfy :

$$\ln z_{jk} = c_j/a_j + 2i\pi k/a_j; \quad k = 1, \dots, a_j; \quad j = 1, \dots, n$$



When the ratios $\{c_k/a_k\}$ are close to each other, the integration of $z^{b-1}\hat{F}_d(z)$ on the circle $|z| = \gamma$ becomes more difficult because nearly all the poles of $\hat{F}_d(z, c)$ contribute.

Similarly, it is also known that the corresponding knapsack problems are difficult to solve ...

Let $r \in \mathbf{N}$, $c_1/a_1 < c_2/a_2 < \dots < c_n/a_n$ and $c_j \in \mathbf{N}$ for all j .

$$\frac{\widehat{F}_d(z, rc)}{z} = \frac{z^{s-1}}{\prod_{j=1}^n (z^{a_j} - e^{rc_j})} = \sum_{j=1}^n \frac{Q_j(z)}{(z^{a_j} - e^{rc_j})},$$

for some polynomials

$$z \mapsto Q_j(z) := \sum_{k=0}^{a_j-1} Q_{jk} z^k = \sum_{k=0}^{a_j-1} \left[\frac{M_{jk}(e^r)}{R_{jk}(e^r)} \right] z^k,$$

for some polynomials $\{M_{jk}, R_{jk}\}$ of the variable $y := e^r$.

In principle they can be obtained by symbolic calculation. Therefore, with $b = s_j \bmod a_j$, for all $j = 1, \dots, n$,

$$\widehat{f}(b, rc) = \sum_{j=1}^n Q_{j(a_j-s_j-1)} y^{\frac{(b-s_j)c_j}{a_j}} = \sum_{j=1}^n \frac{M_{j(a_j-s_j-1)}(y)}{R_{j(a_j-s_j-1)}(y)} y^{\frac{(b-s_j)c_j}{a_j}}$$

Hence

$$\begin{aligned}
 e^{f_d(b,c)} &= \lim_{r \rightarrow \infty} \widehat{f}_d(b, rc)^{1/r} \\
 &= \lim_{r \rightarrow \infty} \left[\sum_{j=1}^n \frac{M_j(a_j - s_j - 1)(e^r)}{R_j(a_j - s_j - 1)(e^r)} e^{r \frac{(b-s_j)c_j}{a_j}} \right]^{1/r}
 \end{aligned}$$

Thus, for sufficiently large b ,

$$e^{f_d(b,c)} = \max_{j=1, \dots, n} \lim_{r \rightarrow \infty} \left[\frac{M_j(a_j - s_j - 1)(e^r)}{R_j(a_j - s_j - 1)(e^r)} e^{r \frac{(b-s_j)c_j}{a_j}} \right]^{1/r}$$

$$f_d(b, c) = \frac{c_1 b}{a_1} - \frac{c_1 s_1}{a_1} + \deg M_{1(a_1 - s_1 - 1)}(y) - \deg R_{1(a_1 - s_1 - 1)}(y)$$

Example : Consider the problem

$$\hat{f}(b, c) := \max_{x \in \mathbb{N}^2} \{3x_1 + 5x_2 \mid 2x_1 + 3x_2 \leq b\}.$$

Let $y = e^r$. We obtain

$$\begin{aligned} \frac{F(z)}{z} &= \frac{z^5}{(z-1)(z^2-y^3)(z^3-y^5)} \\ &= \frac{(y^9+y^7+y^6) + z(y^7+y^6+y^4) + z^2(y^6+y^4+y^2)}{(y^5-1)(y-1)(z^3-y^5)} \\ &= \frac{(y^5+y^3) + z(y^3+y^2)}{(y^3-1)(y-1)(z^2-y^3)} + \frac{1}{(y^5-1)(y^3-1)(z-1)} \end{aligned}$$

Hence, $\hat{f}_d(b, c) = f(b, c)$ if $b = 0 \pmod{3}$ and for large b ,

$$\begin{aligned} \hat{f}_d(b, c) &= f_d(b, c) - 5/3 + 1 && \text{whenever } b = 1 \pmod{3} \\ \hat{f}_d(b, c) &= f_d(b, c) - 10/3 + 3 && \text{whenever } b = 2 \pmod{3} \end{aligned}$$

A Discrete Farkas Lemma

Let $A \in \mathbf{N}^{m \times n}$, $b \in \mathbf{N}^m$ and consider the problem deciding **whether or not** $Ax = b$ has a solution $x \in \mathbf{N}^n$.

Theorem: (i) $Ax = b$ has a solution $x \in \mathbf{N}^n$ if and only if the polynomial $b \mapsto z^b - 1$ in $\mathbf{R}[z_1, \dots, z_m]$ can be written

$$z_1^{b_1} \cdots z_m^{b_m} - 1 = \sum_{j=1}^n Q_j(z)(z^{A_j} - 1) = \sum_{j=1}^n Q_j(z)(z_1^{A_{1j}} \cdots z_m^{A_{mj}} - 1)$$

for some polynomials $Q_j(z)$, all with **nonnegative coefficients**.

(ii) The degree of the Q_j 's is bounded by $b^* := \sum_{j=1}^m b_j - \max_k \sum_{j=1}^m A_{jk}$.

A single LP to solve with $n \times \binom{b^*+m}{b^*}$ variables, $\binom{b^*+m}{m}$ constraints and a (sparse) matrix of coefficients in $\{0, \pm 1\}$

One also retrieves the classical **Farkas Lemma** in \mathbf{R}^n , that is,

$$\{x \in \mathbf{R}^n \mid Ax = b; x \geq 0\} \neq \emptyset \quad \Leftrightarrow \quad A'u \geq 0 \Rightarrow b'u \geq 0.$$

Indeed, if $Ax = b$ has a solution $x \in \mathbf{N}^n$, then with $u = \ln z$,

$$e^{b'u} - 1 = \sum_{j=1}^n Q_j(e^{u_1}, \dots, e^{u_m})(e^{(A'u)_j} - 1).$$

Therefore,

$$A'u \geq 0 \Rightarrow e^{(A'u)_j} - 1 \geq 0 \Rightarrow e^{b'u} - 1 \geq 0 \Rightarrow b'u \geq 0,$$

and one retrieves (*), i.e., $Ax = b$ has a solution $x \in \mathbf{R}^n$.

The general case $A \in \mathbf{Z}^{m \times n}, b \in \mathbf{Z}^m$.

Let $\Omega := \{x \in \mathbf{R}^n \mid Ax = b; \quad x \geq 0\}$ be a **polytope**.

Let $\alpha \in \mathbf{N}^n$ be such that for every column A_j of A ,

$A_{kj} + \alpha_j \geq 0 \quad \forall k = 1, \dots, m$; let $\mathbf{N} \ni \beta \geq \rho(\alpha) := \max\{\alpha'x \mid x \in \Omega\}$.

Theorem: (i) $Ax = b$ has a solution $x \in \mathbf{N}^n$ if and only if the polynomial $z \mapsto z^b (zy)^\beta - 1$ in $\mathbf{R}[z_1, \dots, z_m]$ can be written

$$z^b (zy)^\beta - 1 = Q_0(z, y)(zy - 1) + \sum_{j=1}^n Q_j(z, y)(z^{A_j} (zy)^{\alpha_j} - 1),$$

for some polynomials $Q_j(z)$, all with **nonnegative coefficients**.

(ii) The degree of the Q_j 's is bounded by $b^* := (m+1)\beta + \sum_{j=1}^m b_j$.

Back to standard Farkas lemma

$$\{x \in \mathbf{R}^n \mid Ax = b, x \geq 0\} \neq \emptyset \quad \Leftrightarrow \quad [A'\lambda \geq 0] \Rightarrow b'\lambda \geq 0.$$

But, **equivalently** $\{x \in \mathbf{R}^n \mid Ax = b, x \geq 0\} \neq \emptyset$ if and only if the polynomial $\lambda \mapsto b'\lambda$ can be written

$$b'\lambda = \sum_{j=1}^n Q_j(\lambda)(A'\lambda)_j,$$

for some polynomials $\{Q_j\} \subset \mathbf{R}[\lambda_1, \dots, \lambda_m]$, all with **nonnegative** coefficients.

In this case, each Q_j is necessarily a **constant**, that is, $Q_j \equiv Q_j(0) = x_j \geq 0$, and $Ax = b$!

$P = \{x \in \mathbf{R}^n \mid Ax = b, x \geq 0\}$	$P \cap \mathbf{Z}^n$
$x \in P$ $\Leftrightarrow x = Q(0, \dots, 0) \text{ with}$ $Q \in \mathbf{R}[\lambda_1, \dots, \lambda_m]$ $b'\lambda = \langle Q, A'\lambda \rangle$ $Q \succeq 0$	$x \in \text{integer hull}(P)$ $\Leftrightarrow x = Q(1, \dots, 1) \text{ with}$ $Q \in \mathbf{R}[e^{\lambda_1}, \dots, e^{\lambda_m}]$ $e^{b'\lambda} - 1 = \langle Q, e^{A'\lambda} - 1_n \rangle$ $Q \succeq 0$

Comparing continuous and discrete Farkas lemma

An equivalent Linear program

Let $0 \leq q = \{q_{j\alpha}\} \in \mathbf{R}^{ns}$ be the coefficients of the Q_j 's in

$$z^b - 1 = \sum_{j=1}^n Q_j(z)(z^{A_j} - 1)$$

They are solutions of a linear system

$$Mq = r, \quad q \geq 0$$

for some matrix M and vector r , both with $0, \pm 1$ coefficients.

** M and r are easily obtained from A, b with **no** computation

Write $q = (q_1, q_2, \dots, q_n)$ with each $q_j = \{q_{j\alpha}\} \in \mathbf{R}^s$, and let $\hat{c}_{j\alpha} := c_j$ for all α

Theorem : Let $A \in \mathbf{N}^{m \times n}$, $b \in \mathbf{N}^m$, $c \in \mathbf{R}^n$.

(i) The integer program $\mathbf{P} \rightarrow \max\{c'x \mid Ax = b, x \in \mathbf{N}^n\}$ has same value as the linear program

$$\mathbf{Q} \rightarrow \max\left\{\sum_{j=1}^n \tilde{c}'_j q_j \mid Mq = r; q \geq 0\right\}.$$

(ii) Let q^* be an optimal vertex, and let

$$x_j^* := \sum_{\alpha} q_{j\alpha}^* \quad j = 1, \dots, n.$$

Then $x^* \in \mathbf{N}^n$ and x^* is an optimal solution of \mathbf{P} .

The link with superadditive functions

The LP-dual Q^* of the linear program Q reads

$$Q^* \rightarrow \min_{\pi} \{ \pi' r \mid M' \pi \geq \hat{c} \}.$$

More precisely, with $\mathcal{D} := \prod_{j=1}^n \{0, 1, \dots, b_j\} \subset \mathbf{N}^m$,

$$Q^* \rightarrow \begin{cases} \min_{\pi} & \pi(b) - \pi(0) \\ \text{s.t.} & \pi(\alpha + A_j) - \pi(\alpha) \geq c_j, \quad \alpha \in \mathcal{D}, j = 1, \dots, n \end{cases}$$

Let $\Pi := \{ \pi : \mathbf{N}^m \rightarrow \mathbf{R} \cup \{+\infty\} \mid \pi(x) = \infty \text{ if and only if } x \notin \mathcal{D} \}$.

For every $\pi \in \Pi$, let $f_{\pi} : \mathbf{N}^m \rightarrow \mathbf{R} \cup \{+\infty\}$ be the function

$$f_{\pi}(x) := \inf_{\alpha \in \mathcal{D}} \pi(\alpha + x) - \pi(\alpha), \quad x \in \mathbf{N}^m$$

For every $\pi \in \Pi$, the function f_π is **superadditive** and $f_\pi(0) = 0$.

The LP dual Q^* reads

$$Q^* \rightarrow \begin{cases} \min_{\pi \in \Pi} f_\pi(b) \\ \text{s.t.} & f_\pi(A_j) \geq c_j, \quad j = 1, \dots, n. \end{cases}$$

Thus, Q^* is a **simplified and explicit form** of the abstract dual of Jeroslow, Wolsey, stated in terms of **superadditive** functions.

→ In the abstract dual one may restrict to the **subclass of superadditive functions** derived from the **representation**

$$z^b - 1 = \sum_{j=1}^n Q_j(z^{A_j} - 1).$$

And, with $\mathbf{P} = \{x \in \mathbf{R}_+^n \mid Ax = b\}$ the integer hull $\text{co}(\mathbf{P} \cap \mathbf{Z}^n)$ reads

$$\text{co}(\mathbf{P} \cap \mathbf{Z}^n) = \{x \in \mathbf{R}^n \mid \sum_{j=1}^n f_{\pi}(A_j)x_j \leq f_{\pi}(b)\},$$

for finitely many π , generators of the convex cone

$$(\pi, \lambda) : \quad \pi(\alpha + A_j) - \pi(\alpha) + \lambda_j \geq 0, \quad \alpha + A_j \in \mathcal{D}, \quad j = 1, \dots, n$$

CONCLUSION

Generating functions permit to exhibit a natural **duality** for integer programming, an IP-analogue of **LP duality**.

This duality also shows which kind of **superadditive functions** are useful in the abstract dual of Jeroslow, Wolsey.

This might help providing **efficient Gomory cuts** in MIP solvers like CPLEX, or XPRESS-MP.

Another dual problem

Let $A \in \mathbf{Z}^{m \times n}$, $b \in \mathbf{Z}^m$, $c \in \mathbf{R}^n$. Let $y \mapsto f(y, c) = \max\{c'x \mid Ax = y; x \geq 0\}$. The **Fenchel transform** of the convex function $-f(\cdot, c)$ is the convex function

$$\lambda \mapsto (-f)^*(\lambda, c) = \sup_{y \in \mathbf{R}^m} \lambda'y + f(y, c).$$

The **dual** problem of the **linear program** is obtained from **Fenchel duality** as

$$\begin{aligned} f(b, c) &= \inf_{\lambda \in \mathbf{R}^m} b'\lambda + (-f)^*(-\lambda, c) \\ &= \inf_{\lambda \in \mathbf{R}^m} b'\lambda + \sup_{x \in \mathbf{R}^m} (c - A'\lambda)'x = \min \{b'\lambda \mid A'\lambda \geq c\} \end{aligned}$$

Equivalently

$$e^{f(b, c)} = \inf_{\lambda \in \mathbf{R}^m} \sup_{x \in \mathbf{R}^m} e^{(b - Ax)'\lambda} e^{c'x}$$

Define

$$\rho^* := \inf_{z \in \mathbf{C}^m} \sup_{x \in \mathbf{N}^n} \Re \left(z^{b-Ax} e^{c'x} \right) = \inf_{z \in \mathbf{C}^m} f_d^*(z, c).$$

Hence,

$$f_d^*(z, c) = \Re \left(z^b \prod_{j=1}^n (z^{-A_j} e^{c_j})^{x_j} \right) < \infty \quad \text{if } |z^{A_j}| \geq e^{c_j} \quad \forall j$$

(that is, $A' \ln |z| \geq c$). Next, (writing $z \in \mathbf{C}$ as $e^\lambda e^{i\theta}$)

$$\rho^* \leq \inf_{z \in \mathbf{R}^m} \sup_{x \in \mathbf{R}^n} \Re \left(z^{b-Ax} e^{c'x} \right) = \inf_{\lambda \in \mathbf{R}^m} \sup_{x \in \mathbf{R}^n} e^{(b-Ax)' \lambda} e^{c'x} = e^{f(b,c)}$$

Finally, with $z \in \mathbf{C}^m$ arbitrary fixed

$$\sup_{x \in \mathbf{N}^n} \Re \left(z^{b-Ax} e^{c'x} \right) \geq e^{c'x^*} = e^{f_d(b,c)}$$

Hence $f_d(\mathbf{b}, \mathbf{c}) \leq \ln \rho^* \leq \mathbf{f}(\mathbf{b}, \mathbf{c})$.

Let σ^* be an **optimal basis** of the **linear program**. Under **uniqueness** of the “ \max_σ ” in Brion and Vergne ’s formula, and an additional technical condition

$$ef_d(b,c) = \rho^* = \max_{x \in \mathbb{N}^n} \Re \left(\hat{z}^{b-Ax} e^{c'x} \right) = f_d^*(\hat{z}, c)$$

where $\hat{z}^{A_j} = \gamma e^{c_j} \quad \forall j \in \sigma^*$ for some real $\gamma > 1$.

\hat{z} is an optimal solution of the dual problem

$$\inf_{z \in \mathbb{C}^m} f_d^*(z, c)$$