

Asymptotic Approximation of Marginal Likelihood Integrals

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A Statistical Example

132 Schizophrenic Patients

- Evans-Gilula-Guttman(1989) studied schizophrenic patients for connections between recovery time (in years Y) and frequency of visits by relatives.

	$2 \leq Y < 10$	$10 \leq Y < 20$	$20 \leq Y$	<i>Totals</i>
Regularly	43	16	3	62
Rarely	6	11	10	27
Never	9	18	16	43
<i>Totals</i>	58	45	29	132

- Proposed two statistical models to explain the data.

132 Schizophrenic Patients

● Model 1: Independence Model

	$2 \leq Y < 10$	$10 \leq Y < 20$	$20 \leq Y$
Regularly	$a_1 b_1$	$a_1 b_2$	$a_1 b_3$
Rarely	$a_2 b_1$	$a_2 b_2$	$a_2 b_3$
Never	$a_3 b_1$	$a_3 b_2$	$a_3 b_3$

132 Schizophrenic Patients

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● Model 2: Hidden Variable Model

	$2 \leq Y < 10$	$10 \leq Y < 20$	$20 \leq Y$
Regularly	$ta_1 b_1 + (1 - t)c_1 d_1$	$ta_1 b_2 + (1 - t)c_1 d_2$	$ta_1 b_3 + (1 - t)c_1 d_3$
Rarely	$ta_2 b_1 + (1 - t)c_2 d_1$	$ta_2 b_2 + (1 - t)c_2 d_2$	$ta_2 b_3 + (1 - t)c_2 d_3$
Never	$ta_3 b_1 + (1 - t)c_3 d_1$	$ta_3 b_2 + (1 - t)c_3 d_2$	$ta_3 b_3 + (1 - t)c_3 d_3$

Marginal Likelihood Integrals

- In Bayesian statistics, models are selected by comparing *marginal likelihood integrals*.

$$Z = \int_{\Omega} \prod_i p_i(\omega)^{U_i} \varphi(\omega) d\omega$$

U_i the data, Ω parameter space

$p_i(\omega)$ functions parametrizing the model

$\varphi(\omega)$ prior belief about parameter space

Marginal Likelihood Integrals

● e.g. Independence Model

	$2 \leq Y < 10$	$10 \leq Y < 20$	$20 \leq Y$	<i>Totals</i>
Reg.	$a_1 b_1$ (43)	$a_1 b_2$ (16)	$a_1 b_3$ (3)	(62)
Rarely	$a_2 b_1$ (6)	$a_2 b_2$ (11)	$a_2 b_3$ (10)	(27)
Never	$a_3 b_1$ (9)	$a_3 b_2$ (18)	$a_3 b_3$ (16)	(43)
<i>Totals</i>	(58)	(45)	(29)	

$$Z_1 = \int_{\Delta_2} \int_{\Delta_2} a_1^{62} a_2^{27} a_3^{43} b_1^{58} b_2^{45} b_3^{29} da db$$

$$a = (a_1, a_2, a_3), b = (b_1, b_2, b_3)$$

$$\Delta_2 = \{(x_1, x_2, x_3) \in \mathbb{R}^3 : x_i \geq 0, \sum_i x_i = 1\}$$

Asymptotic Approximation

● IMPORTANT PROBLEM IN STATISTICS

Compute
$$Z(n) = \int_{\Omega} \prod_{i=1}^k p_i(\omega)^{nq_i} \varphi(\omega) d\omega$$

where n sample size, $q_i = U_i/n$ relative frequencies.

Assumptions:

Ω compact and semianalytic

i.e. $\Omega = \{\omega \in \mathbb{R}^d : g_1 \geq 0, \dots, g_l \geq 0\}$, g_i real analytic on \mathbb{R}^d

φ nearly analytic

i.e. $\varphi = \varphi_s \varphi_a$, φ_s positive and smooth, φ_a real analytic on Ω

p_i positive real analytic functions on Ω summing to 1

q true distribution lying in $p(\Omega)$

Asymptotic Approximation

- **IMPORTANT PROBLEM IN STATISTICS**

Compute $Z(n) = \int_{\Omega} \prod_{i=1}^k p_i(\omega)^{nq_i} \varphi(\omega) d\omega$

- L.-Sturmfels-Xu(2008) gave efficient algorithms for computing $Z(n)$ *exactly* for small samples n .

Asymptotic Approximation

- **IMPORTANT PROBLEM IN STATISTICS**

Compute
$$Z(n) = \int_{\Omega} \prod_{i=1}^k p_i(\omega)^{nq_i} \varphi(\omega) d\omega$$

- L.-Sturmfels-Xu(2008) gave efficient algorithms for computing $Z(n)$ *exactly* for small samples n .
- *Asymptotically*, as $n \rightarrow \infty$,

$$Z(n) \approx \left(\prod_{i=1}^k q_i^{q_i} \right)^n \cdot C n^{-\lambda} (\log n)^{\theta-1}$$

In this talk, we discuss how to compute (λ, θ) .

In machine learning, λ is called the *learning coefficient* of the statistical model and θ its *multiplicity*.

Statistical Learning Theory and Singularity Theory

Statistical Learning Theory

In statistics, it is customary to write our marginal likelihood integral as the Laplace integral

$$\begin{aligned} Z(n) &= \int_{\Omega} \prod_{i=1}^k p_i(\omega)^{nq_i} \varphi(\omega) d\omega \\ &= \left(\prod_{i=1}^k q_i^{q_i} \right)^n \int_{\Omega} e^{-nK(\omega)} \varphi(\omega) d\omega \end{aligned}$$

where $\prod_{i=1}^k q_i^{q_i}$ is the maximum likelihood and $K(\omega)$ is the Kullback-Leibler distance

$$K(\omega) = \sum_{i=1}^k q_i \log \frac{q_i}{p_i(\omega)}.$$

Statistical Learning Theory

Theorem (Watanabe)

The Laplace integrals defined by functions

$$K(\omega) = \sum_{i=1}^k q_i \log \frac{q_i}{p_i(\omega)}$$
$$Q(\omega) = \sum_{i=1}^k (p_i(\omega) - q_i)^2$$

are asymptotically the same up to a constant, i.e.

$$\int_{\Omega} e^{-nK(\omega)} |\varphi(\omega)| d\omega \approx C_1 n^{-\lambda} (\log n)^{\theta-1}$$
$$\int_{\Omega} e^{-nQ(\omega)} |\varphi(\omega)| d\omega \approx C_2 n^{-\lambda} (\log n)^{\theta-1}$$

for some constants $\lambda, \theta, C_1, C_2$.

Singularity Theory

Theorem (Arnold-Gusein-Zade-Varchenko)

Let f be a real analytic function on Ω with $f(\omega^*) = 0$ for some $\omega^* \in \Omega$. If we have asymptotics

$$Z(n) = \int_{\Omega} e^{-n|f(\omega)|} |\varphi(\omega)| d\omega \approx C n^{-\lambda} (\log n)^{\theta-1},$$

then λ is the smallest pole of the zeta function

$$\zeta(z) = \int_{\Omega} |f(\omega)|^{-z} |\varphi(\omega)| d\omega, \quad z \in \mathbb{C}$$

and θ is the multiplicity of this pole.

Singularity Theory

Theorem (Arnold-Gusein-Zade-Varchenko)

Let f be a real analytic function on Ω with $f(\omega^*) = 0$ for some $\omega^* \in \Omega$.
Then, asymptotically, we have

$$Z(n) = \int_{\Omega} e^{-n|f(\omega)|} |\varphi(\omega)| d\omega \approx \sum_{\alpha} \sum_{i=0}^{d-1} c_{i,\alpha} n^{-\alpha} (\log n)^i.$$

The α in this expansion range over positive rational numbers which are poles of

$$\zeta(z) = \int_{\Omega_{\delta}} |f(\omega)|^{-z} |\varphi(\omega)| d\omega$$

where $\Omega_{\delta} = \{\omega \in \Omega : |f(\omega)| < \delta\}$ for any $\delta > 0$. The coefficients $c_{i,\alpha}$ satisfy

$$c_{i,\alpha} = \frac{(-1)^{i+1}}{i!} \sum_{j=i}^{d-1} \frac{\Gamma^{(j-i)}(\alpha)}{(j-i)!} d_{j+1,\alpha}$$

where $d_{j,\alpha}$ is the coefficient of $(z - \alpha)^{-(j)}$ in the Laurent expansion of $\zeta(z)$.

Example: Monomial Functions

● Let $f = \omega_1^{\kappa_1} \cdots \omega_d^{\kappa_d}$ and $\varphi = \omega_1^{\tau_1} \cdots \omega_d^{\tau_d}$.

$$\int_{[0,\varepsilon]^d} e^{-n\omega^\kappa} \omega^\tau d\omega = C n^{-\lambda} (\log n)^{\theta-1}$$

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• To find (λ, θ) , we study the zeta function

$$\int_{[0,\varepsilon]^d} \omega^{-\kappa z + \tau} d\omega = \left[\frac{\omega_1^{-\kappa_1 z + \tau_1 + 1}}{-\kappa_1 z + \tau_1 + 1} \right]_0^\varepsilon \cdots \left[\frac{\omega_d^{-\kappa_d z + \tau_d + 1}}{-\kappa_d z + \tau_d + 1} \right]_0^\varepsilon$$

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- Thus, $\lambda = \min_i \left\{ \frac{\tau_i + 1}{\kappa_i} \right\}$, $\theta = \# \min_i \left\{ \frac{\tau_i + 1}{\kappa_i} \right\}$

where $\# \min S$ is the number of times the minimum is attained in a set S .

Resolution of Singularities

Theorem (Hironaka)

Let f be a real analytic function at the origin with $f(0) = 0$.

Then, there exists a manifold M , a neighborhood W of the origin and a proper real analytic map $\rho : M \rightarrow W$ such that

- ρ is an isomorphism on $M \setminus (f \circ \rho)^{-1}(0)$
- $f \circ \rho$ and $|\rho'|$ are monomial functions locally at each $y \in (f \circ \rho)^{-1}(0)$

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Thus, we can find the poles of the zeta function of any f , provided we have a resolution of singularities for f .

Finding resolutions is generally a hard problem.

Marginal Likelihood Integral

$$\int_{\Omega} \prod_{i=1}^k p_i(\omega)^{nq_i} |\varphi(\omega)| d\omega$$
$$\approx C(\prod_{i=1}^k q_i^{q_i})^n n^{-\lambda} (\log n)^{\theta-1}$$

Watanabe

Laplace Integral of Sum of Squares

$$\int_{\Omega} e^{-n \sum_{i=1}^k (p_i(\omega) - q_i)^2} |\varphi(\omega)| d\omega$$
$$\approx C n^{-\lambda} (\log n)^{\theta-1}$$

Arnold et al.

Poles of a Monomial Function

$$\int \mu^{-\kappa z + \tau} d\mu$$

Hironaka

Zeta Function of Sum of Squares

(λ, θ) is smallest pole of

$$\zeta(z) = \int_{\Omega} |Q(\omega)|^{-z} |\varphi(\omega)| d\omega$$

Real Log Canonical Thresholds

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- $\Omega \subset \mathbb{R}^d$ compact semianalytic subset
 \mathcal{A}_Ω ring of real analytic functions on Ω
 $I = \langle f_1, \dots, f_r \rangle \subset \mathcal{A}_\Omega$, φ nearly analytic

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$$\zeta(z) = \int_{\Omega} (f_1(\omega)^2 + \dots + f_r(\omega)^2)^{-z/2} |\varphi(\omega)| d\omega$$

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- Consider the zeta function

$$\zeta(z) = \int_{\Omega} (f_1(\omega)^2 + \dots + f_r(\omega)^2)^{-z/2} |\varphi(\omega)| d\omega$$

- Define $\text{RLCT}_\Omega(I; \varphi) = (\lambda, \theta)$ where λ is the smallest pole of $\zeta(z)$ and θ its multiplicity.
If $\zeta(z)$ does not have any poles, set $(\lambda, \theta) = (\infty, \infty)$.

Call λ the *real log canonical threshold* of $(I; \varphi)$ on Ω .

Fundamental Properties

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- RLCT's are local in nature.

$$\text{RLCT}_{\Omega}(I; \varphi) = \min_{x \in \mathcal{V}(I)} \text{RLCT}_{\Omega_x}(I; \varphi)$$

where each Ω_x is a sufficiently small nbhd of x in Ω and $(\lambda_1, \theta_1) < (\lambda_2, \theta_2)$ if $\lambda_1 < \lambda_2$, or $\lambda_1 = \lambda_2$ and $\theta_1 > \theta_2$.

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- RLCT's depend on the boundary structure of Ω_x .

Fundamental Properties

● Formula for disjoint variables

$$\begin{aligned}\text{RLCT}_{X_0 \times Y_0}(I_x + I_y; \varphi_x \varphi_y) &= (\lambda_x + \lambda_y, \theta_x + \theta_y - 1) \\ \text{RLCT}_{X_0 \times Y_0}(I_x I_y; \varphi_x \varphi_y) &= \begin{cases} (\lambda_x, \theta_x) & \text{if } \lambda_x < \lambda_y \\ (\lambda_y, \theta_y) & \text{if } \lambda_x > \lambda_y \\ (\lambda_x, \theta_x + \theta_y) & \text{if } \lambda_x = \lambda_y \end{cases}\end{aligned}$$

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● Formula for change of variables

$$\text{RLCT}_{\Omega_0}(I; \varphi) = \min_{y \in \rho^{-1}(0)} \text{RLCT}_{\rho^{-1}(\Omega_0)_y}(\rho^* I; (\varphi \circ \rho)|\rho'|)$$

Newton Polyhedra

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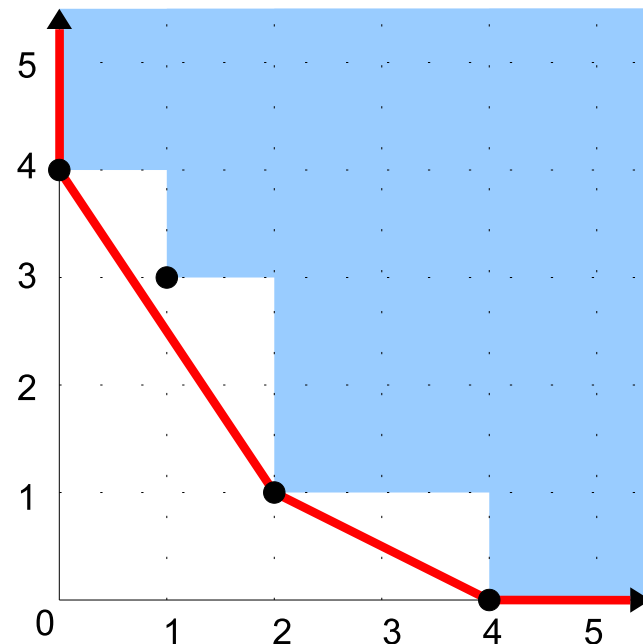
- $\omega_1, \dots, \omega_d$ local coordinates at the origin
 I an ideal of real analytic functions at the origin
Given power series $f = \sum_{\alpha} c_{\alpha} \omega^{\alpha}$, define $[\omega^{\alpha}]f := c_{\alpha}$

Newton Polyhedra

- $\omega_1, \dots, \omega_d$ local coordinates at the origin
 I an ideal of real analytic functions at the origin
Given power series $f = \sum_{\alpha} c_{\alpha} \omega^{\alpha}$, define $[\omega^{\alpha}]f := c_{\alpha}$
- The *Newton polyhedron* of I is the convex hull
$$\mathcal{P}(I) = \text{conv}\{\alpha : f \in I, [\omega^{\alpha}]f \neq 0\}$$

$$I = \langle x^4 + x^2y + xy^3 + y^4 \rangle$$

$$J = \langle x^4, x^2y, xy^3, y^4 \rangle$$



Newton Polyhedra

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Given power series $f = \sum_{\alpha} c_{\alpha} \omega^{\alpha}$, define $[\omega^{\alpha}]f := c_{\alpha}$

- The *Newton polyhedron* of I is the convex hull

$$\mathcal{P}(I) = \text{conv}\{\alpha : f \in I, [\omega^{\alpha}]f \neq 0\}$$

- $\tau = (\tau_1, \dots, \tau_d)$ vector of non-negative integers

The *distance* l_{τ} is the smallest t such that

$$t \cdot (\tau_1 + 1, \dots, \tau_d + 1) \in \mathcal{P}(I)$$

The *multiplicity* θ_{τ} is the codimension of the face of $\mathcal{P}(I)$ at this intersection.

Example: Newton Polyhedra

$$I = \langle x^4 + x^2y + xy^3 + y^4 \rangle$$

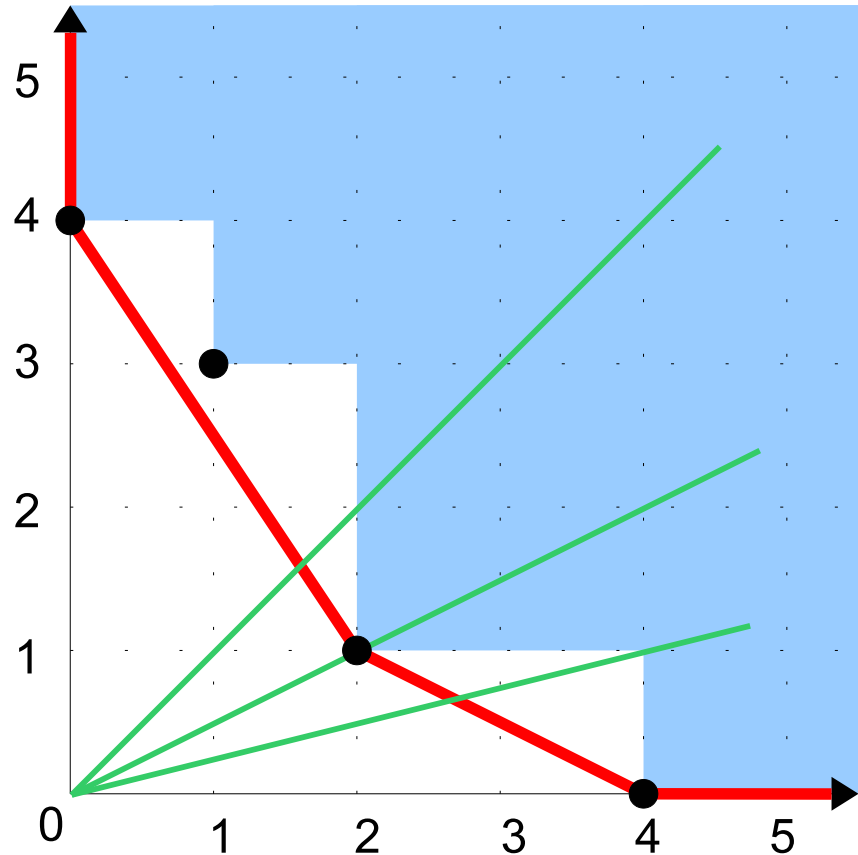
$$J = \langle x^4, x^2y, xy^3, y^4 \rangle$$

Both I, J have the same
Newton polyhedron.

$$l_{(0,0)} = \frac{8}{5}, \theta_{(0,0)} = 1$$

$$l_{(1,0)} = 1, \theta_{(1,0)} = 2$$

$$l_{(3,0)} = \frac{2}{3}, \theta_{(3,0)} = 1$$



Relation to RLCT

Theorem (L.)

Suppose the origin is not on the boundary of Ω .

Then, when φ is a monomial function ω^τ ,

$$\text{RLCT}_{\Omega_0}(I; \omega^\tau) \leq (1/l_\tau, \theta_\tau).$$

Equality holds when I is a monomial ideal.

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Remark

Equality also holds for ideals which are *nondegenerate* (a term due to Varchenko).

Back to Schizophrenic Patients

Learning Coefficients

$P = (p_{ij})$, $S_i = \{\text{rank } i \text{ matrices}\}$

$S_{21} = \{p_{11} = 0; p_{12}, p_{21}, p_{22} \text{ non-zero; up to perm}\} \subset S_2$

$S_{22} = \{p_{11} = p_{22} = 0; p_{12}, p_{21} \text{ non-zero; up to perm}\} \subset S_2$

Theorem (L.)

The learning coefficient (λ, θ) of the model is

$$(\lambda, \theta) = \begin{cases} (5/2, 1) & \text{if } P \in S_1, \\ (7/2, 1) & \text{if } P \in S_2 \setminus (S_{21} \cup S_{22}), \\ (4, 1) & \text{if } P \in S_{21} \setminus S_{22}, \\ (9/2, 1) & \text{if } P \in S_{22}. \end{cases}$$

Learning Coefficients

Proof

Four basic techniques:

1. Changing generators for the ideal
2. Change of variables formula
3. Disjoint variables formula
4. Newton polyhedra method

(systematically peeling an onion)

Complete First Term Asymptotics

Theorem (L.)

If f is nondegenerate, $f(0) = 0$ and $f \geq 0$, then

$$\int_{[0,1]^d} e^{-nf(\omega)} \omega^\tau d\omega \approx C n^{-\lambda} (\log n)^\theta$$

where $(\lambda, \theta) = (1/l_\tau, \theta_\tau)$ and

$$C = \frac{\Gamma(\lambda)}{(\theta - 1)!} \sum_{v \in \mathcal{F}} \prod_{i=1}^{\theta} (v\beta)_i^{-1} \int_{[0,1]^{d-\theta}} g_v(\bar{\mu}) d\bar{\mu}.$$

$\bar{\mu} = (\mu_{\theta+1}, \dots, \mu_d)$, $\bar{e} = e_{\theta+1} + \dots + e_d$,

$g_v(\bar{\mu})$ is $f(\mu^v)^{-\lambda} \mu^{v\tau + \alpha - \bar{e}}$ evaluated at $(0, \dots, 0, \bar{\mu})$,

\mathcal{F} is a unimodular refinement of the normal fan of $\mathcal{P}(f)$,

$\beta(v)$ is the vertex of $\mathcal{P}(f)$ corresponding to v ,

$\alpha(v)$ is the vector of row sums of v .

Complete First Term Asymptotics

Proposition (L.)

If $\langle f_1(\omega), \dots, f_k(\omega) \rangle$ is a monomial ideal,
then $f = \sum_i f_i(\omega)^2$ is nondegenerate.

Proposition (L.)

Let q_1, \dots, q_k be positive real numbers and
 $p_1(\omega), \dots, p_k(\omega)$ be positive real analytic functions
such that $p_i(\omega), q_i \geq 0$ and $\sum_i p_i(\omega) = \sum_i q_i = 1$.

If $\langle p_1(\omega) - q_1, \dots, p_k(\omega) - q_k \rangle$ is a monomial ideal,
then $f = \sum_i q_i \log \frac{q_i}{p_i(\omega)}$ is nondegenerate.

Take Home

1. Compute asymptotics using *zeta functions*.
2. When computing learning coefficients, work with RLCT of *ideals* not *functions*.
3. Newton polyhedra methods can be extended to work with monomial *amplitude functions*.

Open Questions:

1. The RLCT over Ω is the minimum of RLCT's at $x \in \Omega$. How do we identify points with the minimum RLCT?
2. Is there a way to extend Newton polyhedra methods to cases where the origin is on the boundary of Ω ?

Thank you for your kind attention :)

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<http://arxiv.org/abs/1003.5338>

<http://math.berkeley.edu/~shaowei/rlct.html>

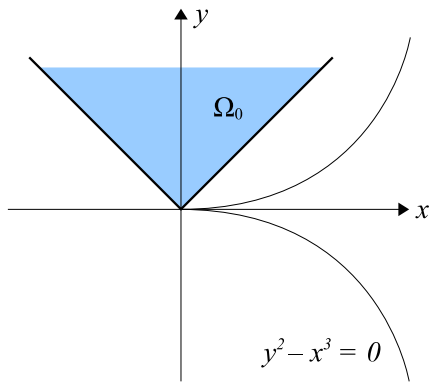
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Example: Boundary Structure

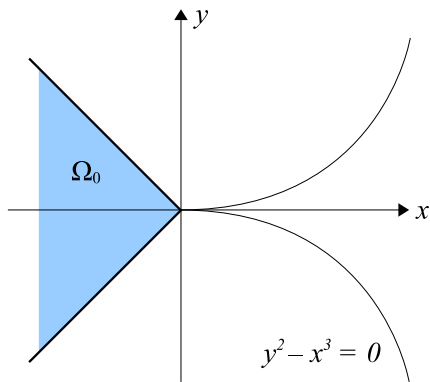
Let $I = \langle y^2 - x^3 \rangle$ and $\varphi = 1$.

● **Case 1:** $\Omega_0 = \{(x, y) \in \mathbb{R}^2 : 0 \leq y \leq \varepsilon, -y \leq x \leq y\}$



$$\text{RLCT}_{\Omega_0}(I; \varphi) = (1, 1)$$

● **Case 2:** $\Omega_0 = \{(x, y) \in \mathbb{R}^2 : -\varepsilon \leq x \leq 0, x \leq y \leq -x\}$



$$\text{RLCT}_{\Omega_0}(I; \varphi) = \left(\frac{5}{6}, 1\right)$$

Disjoint Variables

- Suppose we have disjoint sets of variables

$$x = (x_1, \dots, x_m)$$

$$y = (y_1, \dots, y_n)$$

$$I_x = \langle f_1(x), \dots, f_r(x) \rangle$$

$$I_y = \langle g_1(y), \dots, g_s(y) \rangle$$

$$(\lambda_x, \theta_x) = \text{RLCT}_{X_0}(I_x; \varphi_x) \quad (\lambda_y, \theta_y) = \text{RLCT}_{Y_0}(I_y; \varphi_y)$$

- Recall $I_x + I_y = \langle f_i, g_j \text{ for all } i, j \rangle$, $I_x I_y = \langle f_i g_j \text{ for all } i, j \rangle$

Proposition

$$\text{RLCT}_{X_0 \times Y_0}(I_x + I_y; \varphi_x \varphi_y) = (\lambda_x + \lambda_y, \theta_x + \theta_y - 1)$$

$$\text{RLCT}_{X_0 \times Y_0}(I_x I_y; \varphi_x \varphi_y) = \begin{cases} (\lambda_x, \theta_x) & \text{if } \lambda_x < \lambda_y \\ (\lambda_y, \theta_y) & \text{if } \lambda_x > \lambda_y \\ (\lambda_x, \theta_x + \theta_y) & \text{if } \lambda_x = \lambda_y \end{cases}$$

Change of Variables

- $I = \langle f_1, \dots, f_r \rangle$

- ρ change of variables outside $\mathcal{V}(I)$

i.e. $\rho : M \rightarrow W$ is a proper real analytic map from a manifold M to a neighborhood W of the origin that is an isomorphism on $M \setminus \rho^{-1}(\mathcal{V}(I))$

- $\rho^* I = \langle f_1 \circ \rho, \dots, f_r \circ \rho \rangle, \mathcal{M} = \rho^{-1}(\Omega_0)$

Proposition

$$\text{RLCT}_{\Omega_0}(I; \varphi) = \min_{y \in \rho^{-1}(0)} \text{RLCT}_{\mathcal{M}_y}(\rho^* I; (\varphi \circ \rho)|\rho'|)$$

Computation

Recall $p_{ij}(t, a, b, c, d) = ta_ib_j + (1 - t)c_jd_j$.

Consider $t^* = \frac{1}{2}$ and $a^* = b^* = c^* = d^* = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$.

Denote $\omega = (t, a, b, c, d)$ and $\omega^* = (t^*, a^*, b^*, c^*, d^*)$.

Let $I = \langle p_{ij}(\omega + \omega^*) - p_{ij}(\omega^*) \rangle$ and $\varphi = 1$.

We want to find $\text{RLCT}_{\Omega_{\omega^*}}(I; \varphi)$.

Note that ω^* is not on the boundary of Ω .

Computation

Now, $\varphi = 1$ and I is generated by

$$p_{ij}(\omega + \omega^*) - p_{ij}(\omega^*) \text{ for all } i, j \in \{1, 2, 3\}$$

Computation

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Note that

$$p_{i1} + p_{i2} + p_{i3} = ta_i + tc_i =: p_{i0}$$

$$p_{1j} + p_{2j} + p_{3j} = tb_j + td_j =: p_{0j}$$

Let $g_{ij}(\omega)$ denote $p_{ij}(\omega + \omega^*) - p_{ij}(\omega^*)$.

Computation

Now, $\varphi = 1$ and I is generated by

$$g_{ij}(\omega) \text{ for all } i, j \in \{0, 1, 2\}$$

Computation

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For $i, j \in \{1, 2\}$, we replace $g_{ij}(\omega)$ with

$$g_{ij}(\omega) - (d_j + d_j^*)g_{i0}(\omega) - (a_i + a_i^*)g_{0j}(\omega)$$

Computation

Now, $\varphi = 1$ and I is generated by

$$g_{01}(\omega)$$

$$g_{02}(\omega)$$

$$g_{10}(\omega)$$

$$g_{20}(\omega)$$

$$g_{11}(\omega) - (d_1 + d_1^*)g_{10}(\omega) - (a_1 + a_1^*)g_{01}(\omega)$$

$$g_{12}(\omega) - (d_2 + d_2^*)g_{10}(\omega) - (a_1 + a_1^*)g_{02}(\omega)$$

$$g_{21}(\omega) - (d_1 + d_1^*)g_{20}(\omega) - (a_2 + a_2^*)g_{01}(\omega)$$

$$g_{22}(\omega) - (d_2 + d_2^*)g_{20}(\omega) - (a_2 + a_2^*)g_{02}(\omega)$$

Computation

Now, $\varphi = 1$ and I is generated by

$$g_{01}(\omega)$$

$$g_{02}(\omega)$$

$$g_{10}(\omega)$$

$$g_{20}(\omega)$$

$$g_{11}(\omega) - (d_1 + d_1^*)g_{10}(\omega) - (a_1 + a_1^*)g_{01}(\omega)$$

$$g_{12}(\omega) - (d_2 + d_2^*)g_{10}(\omega) - (a_1 + a_1^*)g_{02}(\omega)$$

$$g_{21}(\omega) - (d_1 + d_1^*)g_{20}(\omega) - (a_2 + a_2^*)g_{01}(\omega)$$

$$g_{22}(\omega) - (d_2 + d_2^*)g_{20}(\omega) - (a_2 + a_2^*)g_{02}(\omega)$$

Expanding these polynomials, we get...

Computation

Now, $\varphi = 1$ and I is generated by

$$c_1\left(\frac{1}{2} - t\right) + a_1\left(t + \frac{1}{2}\right)$$

$$c_2\left(\frac{1}{2} - t\right) + a_2\left(t + \frac{1}{2}\right)$$

$$d_1\left(\frac{1}{2} - t\right) + b_1\left(t + \frac{1}{2}\right)$$

$$d_2\left(\frac{1}{2} - t\right) + b_2\left(t + \frac{1}{2}\right)$$

$$a_1 d_1$$

$$a_1 d_2$$

$$a_2 d_1$$

$$a_2 d_2$$

Computation

Now, $\varphi = 1$ and I is generated by

$$c_1\left(\frac{1}{2} - t\right) + a_1\left(t + \frac{1}{2}\right)$$

$$c_2\left(\frac{1}{2} - t\right) + a_2\left(t + \frac{1}{2}\right)$$

$$d_1\left(\frac{1}{2} - t\right) + b_1\left(t + \frac{1}{2}\right)$$

$$d_2\left(\frac{1}{2} - t\right) + b_2\left(t + \frac{1}{2}\right)$$

$$a_1 d_1$$

$$a_1 d_2$$

$$a_2 d_1$$

$$a_2 d_2$$

Substitute $b_i = \frac{b'_i - d_i(\frac{1}{2} - t)}{t + \frac{1}{2}}$, $c_i = \frac{c'_i - a_i(t + \frac{1}{2})}{\frac{1}{2} - t}$.

The Jacobian determinant of this change of variable is 16.

Computation

Now, $\varphi = 16$ and I is generated by

$$c'_1, c'_2, b'_1, b'_2, a_1d_1, a_1d_2, a_2d_1, a_2d_2$$

Computation

Now, $\varphi = 16$ and I is generated by

$$c'_1, c'_2, b'_1, b'_2, a_1 d_1, a_1 d_2, a_2 d_1, a_2 d_2$$

This is a monomial ideal so we may use the Newton polyhedra method to compute its RLCT.

Alternatively, we can apply the formula for disjoint variables.

$$I = \langle c'_1 \rangle + \langle c'_2 \rangle + \langle b'_1 \rangle + \langle b'_2 \rangle + \left(\langle a_1 \rangle + \langle a_2 \rangle \right) \left(\langle d_1 \rangle + \langle d_2 \rangle \right)$$

Computation

Now, $\varphi = 16$ and I is generated by

$$c'_1, c'_2, b'_1, b'_2, a_1 d_1, a_1 d_2, a_2 d_1, a_2 d_2$$

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Conclusion: $\text{RLCT}_{\Omega_{\omega}^*}(I; \varphi) = (6, 2)$