



Consideration on singularities in learning theory and real log canonical threshold

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log canonical threshold

Definition

Y : a smooth variety

f : a nonzero regular function on Y .

$Z \subset Y$: a closed subscheme

The log canonical threshold $c_Z(Y, f)$

$$c_Z(Y_{\mathbb{C}}, f) = \sup\{c : |f|^{-c} \text{ is locally } L^2 \text{ on n.b.d. of } Z\}.$$

$$c_Z(Y_{\mathbb{R}}, f) = \sup\{c : |f|^{-c} \text{ is locally } L^1 \text{ on n.b.d. of } Z\}.$$

$$\left(\int |z^m|^{-2c} dz d\bar{z} < +\infty, \int |x^m|^{-c} dx < +\infty \Leftrightarrow c < 1/m\right)$$

$$\pi : Y' \rightarrow Y$$

$$K_{Y'} = \pi^*(K_Y) + \sum a_i E_i, \quad \pi^*(f = 0) = \sum b_i E_i \text{ normal crossing,}$$

$$c_Z(Y, f) = \min_{\pi(E_i) \cap Z \neq \emptyset} \left\{ \frac{a_i + 1}{b_i} \right\} \quad 0 < c_Z(Y_{\mathbb{C}}, f) \leq 1, 0 < c_Z(Y_{\mathbb{R}}, f)$$

Bernstein-Sato polynomial

Theorem (Bernstein 71, Björk79)

$$I := \langle b(s) \in \mathbb{C}[s] : b(s)f^s = Pf^{s+1}, \\ P : \text{a linear differential operator} \rangle.$$

Bernstein-Sato polynomial $b_f(s)$:

$$I = \langle b_f(s) \rangle \text{ with leading coefficient 1}$$

Theorem (Kashiwara76, Lichtin89)

$$c_0(\mathbb{C}^d, f) = \text{the largest root of } b_f(s)$$

$$(c_0(\mathbb{R}^d, f) = \text{the root of } b_f(s) \bmod 1)$$

Theorem Atiyah, Bernstein, Sato & Shintani, Björk

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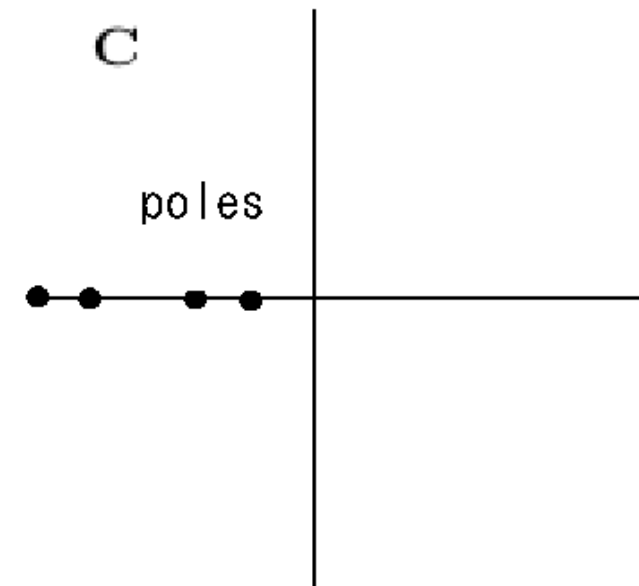
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Theorem Atiyah, Bernstein, Sato & Shintani, Björk

$\psi : C^\infty$ – function with compact support.

$$\zeta(z) = \int_{\text{n.b.d. of } z} |f_{\mathbb{C}}(w)|^z \psi dw \wedge d\bar{w}$$
$$\zeta(z) = \int_{\text{n.b.d. of } z} |f_{\mathbb{R}}(w)|^z \psi dw$$

are holomorphic in the right-half plane.



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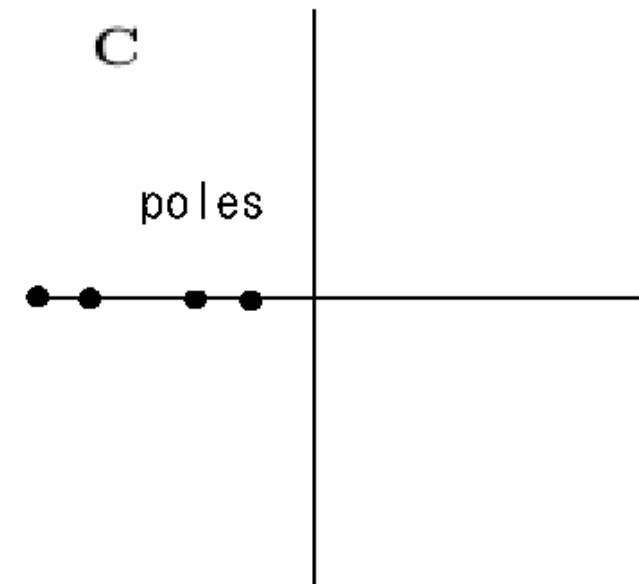
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Its poles are negative rational numbers.



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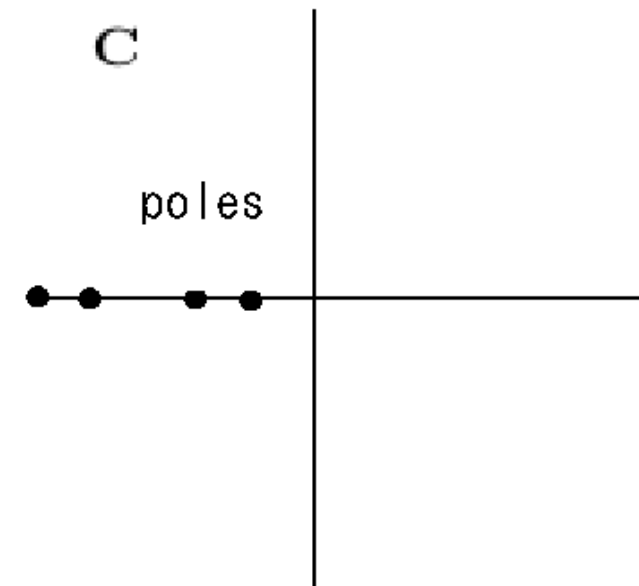
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$-c_Z(\mathbb{C}^n, f)$ = the largest pole of $\int_{\text{n.b.d. of } z} |f|^{2z} \psi dw d\bar{w}$

$-c_Z(\mathbb{R}^n, f)$ = the largest pole of $\int_{\text{n.b.d. of } z} |f|^z \psi dw$

Hironaka Theorem

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f : analytic in a neighborhood of $w \in \mathbf{R}^d$ with
 $f(w) = 0$.

Hironaka Theorem

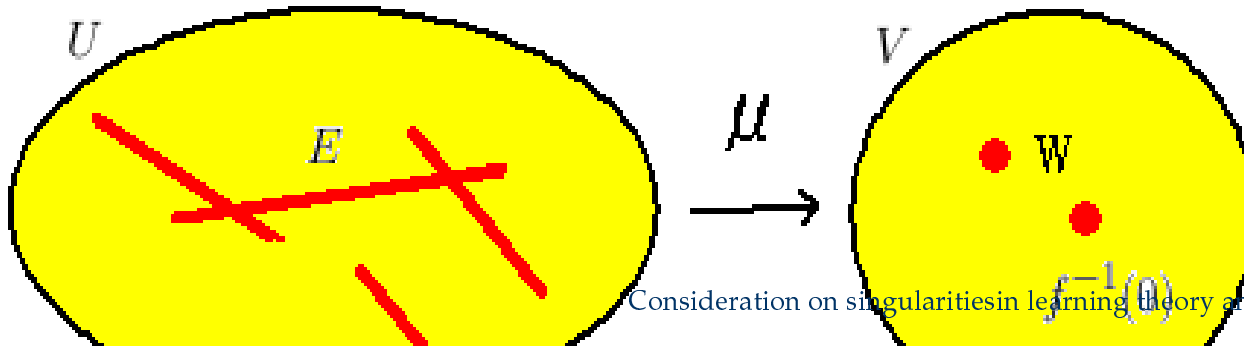
f : analytic in a neighborhood of $w \in \mathbf{R}^d$ with $f(w) = 0$.

Then there exist $\left\{ \begin{array}{l} \text{an open set } V \ni w, \\ \text{a analytic manifold } U, \\ \text{a proper analytic map } \mu \end{array} \right.$

such that

$$\mu : U - E \xrightarrow{\sim} V - f^{-1}(0), \quad E = \mu^{-1}(f^{-1}(0))$$

$$f(\mu(u_1, u_2, \dots, u_d)) = \pm u_1^{s_1} u_2^{s_2} \cdots u_n^{s_d}, \quad s_i \in \mathbf{Z}_{\geq 0}.$$



Apply Hironaka's Theorem

$$\begin{aligned}\zeta(z) &= \int_V |f(w)|^z \psi(w) dw \\ &= \sum_{U_u} \int_{U_u} |u_1^{s_1} u_2^{s_2} \cdots u_d^{s_d}|^z u_1^{k_1} u_2^{k_2} \cdots u_d^{k_d} du.\end{aligned}$$

We have poles $-\frac{k_1+1}{s_1}, -\frac{k_2+1}{s_2}, \dots, -\frac{k_d+1}{s_d}$

Fact

- $I = \langle f_1, \dots, f_m \rangle$
 $c_Z(Y_{\mathbf{R}}, I) := c_Z(Y, f_1^2 + \dots + f_m^2)$
- A : a matrix, P, Q : regular matrices
 $c_Z(Y_{\mathbf{R}}, \|A\|^2) = c_Z(Y, \|PAQ\|^2)$

Definition (Vandermonde matrix type singularities)

a^*, b^* : constants

Fix $Q \in \mathbb{N}$ and $m \in \mathbb{N} \cup \{0\}$.

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1H} & a_{1,H+1}^* & \cdots & a_{1,H+r}^* \\ a_{21} & \cdots & a_{2H} & a_{2,H+1}^* & \cdots & a_{2,H+r}^* \\ \vdots & & & & & \\ a_{M1} & \cdots & a_{MH} & a_{M,H+1}^* & \cdots & a_{M,H+r}^* \end{pmatrix},$$

$$I = (\ell_1, \dots, \ell_N) \in (\mathbb{N} \cup \{0\})^N,$$

$$B_I =$$

$$\left(\prod_{j=1}^N b_{1j}^{\ell_j}, \prod_{j=1}^N b_{2j}^{\ell_j}, \dots, \prod_{j=1}^N b_{Hj}^{\ell_j}, \prod_{j=1}^N b_{H+1,j}^{\ell_j}, \dots, \prod_{j=1}^N b_{H+r,j}^{\ell_j} \right)^t \text{ and}$$

$$B = (B_I)_{\ell_1 + \dots + \ell_N = Qn + m, 0 \leq n \leq H+r}$$

We call $\|AB\|^2 = 0$ **Vandermonde matrix type singularities**

Example

$\langle C \rangle$: ideal generated by $\{c_{ij}\}$.

Remark $\langle AB \rangle = \langle AB' \rangle$ where

$B' = (B_I)_{\ell_1 + \dots + \ell_N = Qn + m, 0 \leq n \leq H'}$ and $H' \geq H + r$.

Example

$N = 1, m = 0, Q = 1$ and $r = 0$,

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1H} \\ & \vdots & \\ a_{M1} & \cdots & a_{MH} \end{pmatrix}, B = \begin{pmatrix} 1 & b_{11} & b_{11}^2 & \cdots & b_{11}^{H-1} \\ 1 & b_{21} & b_{21}^2 & \cdots & b_{21}^{H-1} \\ & \vdots & & & \\ 1 & b_{H1} & b_{H1}^2 & \cdots & b_{H1}^{H-1} \end{pmatrix}$$

Example

$N = 2, Q = 2$ and $m = r = H = 1,$

$$A = \begin{pmatrix} a_{11} & a_{12}^* \\ a_{21} & a_{22}^* \\ \vdots & \vdots \\ a_{M1} & a_{M,2}^* \end{pmatrix}$$

$$B = \begin{pmatrix} b_{11} & b_{11}^3 & b_{12} & b_{12}^3 & b_{11}b_{12}^2 & b_{11}^2b_{12} \\ b_{21}^* & b_{21}^{*3} & b_{22}^* & b_{22}^{*3} & b_{21}^*b_{22}^{*2} & b_{21}^{*2}b_{22}^* \end{pmatrix}.$$

Example

$$N = 3, m = Q = r = 1, H = 2$$

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13}^* \\ a_{21} & a_{12} & a_{23}^* \\ \vdots & \vdots & \\ a_{M1} & a_{12} & a_{M3}^* \end{pmatrix}$$

$$B = \begin{pmatrix} b_{11} & b_{11}^2 & b_{12} & b_{12}^2 & b_{13} & b_{13}^2 & b_{11}b_{12} & b_{11}b_{13} & b_{12}b_{13} & b_{11}^3 & \cdots \\ b_{21} & b_{21}^2 & b_{22} & b_{22}^2 & b_{23} & b_{23}^2 & b_{21}b_{22} & b_{11}b_{23} & b_{22}b_{23} & b_{21}^3 & \cdots \\ b_{31}^* & b_{31}^{*2} & b_{32}^* & b_{32}^{*2} & b_{33}^* & b_{33}^{*2} & b_{31}^*b_{32}^* & b_{11}^*b_{33}^* & b_{32}^*b_{33}^* & b_{31}^{*3} & \cdots \end{pmatrix}$$

Remark Their singularities are not isolated. Degenerate with respect to their Newton polyhedrons

Three layered neural network

Three layered neural network :

N input, H hidden and M output

$$p(y|x, w) = \frac{1}{(2\pi)^{M/2}} \exp\left(-\frac{1}{2} \sum_{k=1}^M \left(y_k - \sum_{i=1}^H a_{ki} \tanh\left(\sum_{j=1}^N b_{ij} x_j\right)\right)^2\right).$$

True distribution with r hidden units:

$$p(y|x, w^*) = \frac{1}{(2\pi)^{M/2}} \exp\left(-\frac{1}{2} \sum_{k=1}^M \left(y_k - \sum_{i=1}^r a_{k,H+i}^* \tanh\left(\sum_{j=1}^N b_{H+i,j}^* x_j\right)\right)^2\right).$$

$$Q = 2, m = 1. \quad (\tanh(x) = c_1 x + c_3 x^3 + \dots)$$

Normal mixture model

Normal mixture model :

$$p(x|w) = \frac{1}{(2\pi)^{N/2}} \sum_{i=1}^H a_{1i} \exp\left(-\frac{\sum_{j=1}^N (x_j - b_{ij})^2}{2}\right),$$

$$\sum_{i=1}^H a_{1i} = 1.$$

True distribution with r peaks:

$$p(x|w^*) = \frac{1}{(2\pi)^{N/2}} \sum_{i=H+1}^{H+r} a_{1i}^* \exp\left(-\frac{\sum_{j=1}^N (x_j - b_{ij}^*)^2}{2}\right),$$

$$\sum_{i=H+1}^{H+r} a_{1i}^* = 1.$$

$$Q = 1, m = 1 \text{ and } M = 1.$$

Theorem 1

Consider a neighborhood of $w^* = \{a_{ki}^*, b_{ij}^*\}$. Assume

$$\left. \begin{array}{l} (b_{11}^*, \dots, b_{1N}^*) \\ \vdots \\ (b_{H_0 1}^*, \dots, b_{H_0 N}^*) \end{array} \right\} = 0, \quad \left. \begin{array}{l} (b_{H_0+1,1}^*, \dots, b_{H_0+1,N}^*) \\ \vdots \\ (b_{H_0+H_1,1}^*, \dots, b_{H_0+H_1,N}^*) \end{array} \right\} = w_1^*,$$

$$\dots, \quad \left. \begin{array}{l} (b_{H_0+\dots+H_{r'-1}+1,1}^*, \dots, b_{H_0+\dots+H_{r'-1}+1,N}^*) \\ \vdots \end{array} \right\} = w_{r'}^*.$$

$$\left. (b_{H_0+\dots+H_{r'-1}+H_{r'},1}^*, \dots, b_{H_0+\dots+H_{r'-1}+H_{r'},N}^*) \right\}$$

and $H_0 + \dots + H_{r'} = H$.

Then $c_{w^*}(\|AB\|^2) = \sum_{i=1}^{r'} c_{w_i^*}(\|A_i B_i\|^2)$

$A_i B_i$ correspond to w_i^*

Theorem 2

$$\text{bound}_i = \begin{cases} \frac{MH}{2}, & \text{if } mM \leq N - 1, \text{ and } i = 1, \\ \frac{mM(H - 1) + N}{2m}, & \text{if } mM \leq N - 1, \text{ and } i = 2, \\ \frac{NH}{2m}, & \text{if } N \leq mM \leq m(N - 1), \\ \frac{NH}{2m}, & \text{if } M \geq N, (N - 1)(m - 1) \geq 1, \\ \frac{2HN + Q(M(1 + k) + (N - 1)(2H - k - 1))k}{4Qk + 4m}, & \text{if } M \geq N, (N - 1)(m - 1) = 0, \end{cases}$$

for $i = 1, 2$, where $k = \max\{i \in \mathbb{Z}; 2H \geq (Qi(i - 1) + 2mi)(M - N + 1)\}$.

$$\text{bound}_3 = \frac{NH + \sum_{i=0}^{k'-1} MQ(k' - i) \binom{N+m+Qi-1}{N-1}}{2m + 2Qk'},$$

where $k' = \max\{i \in \mathbb{Z}; NH \geq M \sum_{i'=0}^{i-1} (m + Qi') \binom{N+m+Qi'-1}{N-1}\}$.

Theorem 2

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1H} \\ & \vdots & \\ a_{M1} & \cdots & a_{MH} \end{pmatrix} \quad \begin{aligned} B_I &= (\prod_{j=1}^N b_{1j}^{\ell_j}, \prod_{j=1}^N b_{2j}^{\ell_j}, \cdots, \prod_{j=1}^N b_{Hj}^{\ell_j})^t, \\ B &= (B_I)_{\ell_1 + \cdots + \ell_N = Qn + m, 0 \leq n \leq H+r} \end{aligned}$$

$$c_0(\|AB\|^2) \leq \min\{\mathbf{bound}_1, \mathbf{bound}_3\}.$$

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1H-1} & a_{1,H}^* \\ & \vdots & \\ a_{M1} & \cdots & a_{MH-1} & a_{M,H}^* \end{pmatrix} \quad \begin{aligned} B_I &= (\prod_{j=1}^N b_{1j}^{\ell_j}, \prod_{j=1}^N b_{2j}^{\ell_j}, \cdots, \prod_{j=1}^N b_{Hj}^{\ell_j}) \\ \hat{B} &= (B_I)_{\ell_1 + \cdots + \ell_N = Qn + m, 0 \leq n \leq H+r} \end{aligned}$$

$$c_0(\|AB\|^2) \leq \min\{\mathbf{bound}_2, \mathbf{bound}_3\}.$$

Theorem 3: Exact Values 1

If $N = 1$, we have

$$c_{w^*}(\|AB\|^2) = \frac{MQk_0(k_0 + 1) + 2H_0}{4(m + k_0Q)} + \frac{Mr'}{2} \\ + \sum_{\alpha=1}^r \frac{Mk_\alpha(k_\alpha + 1) + 2H_\alpha}{4(1 + k_\alpha)} + \sum_{\alpha=r+1}^{r'} \frac{Mk_\alpha(k_\alpha + 1) + 2(H_\alpha - 1)}{4(1 + k_\alpha)}.$$

$$k_0 = \max\{i \in \mathbb{Z}; 2H_0 \geq M(i(i - 1)Q + 2mi)\},$$

$$k_\alpha = \max\{i \in \mathbb{Z}; 2H_\alpha \geq M(i^2 + i)\},$$

$$k_\alpha = \max\{i \in \mathbb{Z}; 2(H_\alpha - 1) \geq M(i^2 + i)\}.$$

Theorem 3: Exact Values 2

Case 1 $H = 1$.

$$\lambda(\|A_{M1}B_{1N}\|^2) = \min\left\{\frac{M}{2}, \frac{N}{2m}\right\}, \theta = \begin{cases} 1, & \text{if } mM \neq N, \\ 2, & \text{if } mM = N, \end{cases}$$

$$\lambda(\|\mathbf{a}^* B_{1N}\|^2) = \frac{N}{2m}, \theta = 1.$$

Case 2 $H = 2$. $\lambda = c_0(\|A_{M2}B_{2N}\|^2)$.

1. If $mM \leq N - 1$ then $\lambda = M$ and $\theta = 1$.
2. If $m = 1, M = N$, then $\lambda = \frac{2N+Q(2N-1)}{2(Q+1)}$ and $\theta = 1$.
3. If $m = 1, N = M - 1$ then $\lambda = N$ and $\theta = 2$.
4. If $m = 1, N < M - 1$ then $\lambda = N$ and $\theta = 1$.
5. If $m = 2, N = 1, M = 1$, then $\lambda = \frac{1}{2}$ and $\theta = 2$.
6. If $m = 2, N < mM, M > 1$ then $\lambda = \frac{N}{m}$ and $\theta = 1$.
7. If $m \geq 2, N = mM$ then $\lambda = \frac{N}{m}$ and $\theta = 3$.
8. If $m > 2, N < mM$ then $\lambda = \frac{N}{m}$ and $\theta = 1$.

Theorem 3: Exact Values 3

$$\lambda = c_0(\|(A_{M1}, \mathbf{a}^*)B_{2N}\|^2),$$

1. If $m \geq 2, mM \leq N - 1$ then $\lambda = \frac{mM+N}{2^m}$ and $\theta = 1$.
2. If $m = 1, N \geq M + Q + 1$ then $\lambda = \frac{N+M}{2}$ and $\theta = 1$.
3. If $m = 1, N = M + Q$ then $\lambda = \frac{N+M}{2}$ and $\theta = 2$.
4. If $m = 1, M + 1 \leq N \leq M + Q - 1$ then $\lambda = \frac{2N+Q(2N-1)}{2(Q+1)}$ and $\theta = 1$.
5. If $m = 1, N = M$ then $\lambda = \frac{2N+Q(2N-1)}{2(Q+1)}$ and $\theta = 1$.
6. If $m = 1, N = M - 1$ then $\lambda = N$ and $\theta = 2$.
7. If $m = 1, N < M - 1$ then $\lambda = N$ and $\theta = 1$.
8. If $m = 2, N = 1, M = 1$, then $\lambda = \frac{1}{2}$ and $\theta = 2$.
9. If $m = 2, N < mM, M > 1$ then $\lambda = \frac{N}{m}$ and $\theta = 1$.
10. If $m \geq 2, N = mM$ then $\lambda = \frac{N}{m}$ and $\theta = 2$.
11. If $m > 2, N < mM$ then $\lambda = \frac{N}{m}$ and $\theta = 1$.

Three layered Neural Network

True distribution : r hidden units

bound₀

$$= \frac{Mr}{2} + \frac{H(N-1)}{2} + \min \left\{ \frac{r}{2} + \frac{H-r+(M-N+1)(k'+k'^2)}{4k'+2}, \right. \\ \left. \frac{r-1}{2} + \frac{2(H-r+1)+(M-N+1)(k''+k''^2)}{4k''+4} \right\},$$

where $k' = \max\{i \in \mathbb{Z}; H-r \geq i^2(M-N+1)\}$ and

$k'' = \max\{i \in \mathbb{Z}; 2(H-r-1) \geq (i^2+i)(M-N+1)\}$.

Also let **bound₃** = $\frac{N(H-r) + \sum_{i=0}^{k'-1} 2M(k'-i) \binom{N+2i}{N-1}}{2+4k'}$,

where $k' = \max\{i \in \mathbb{Z}; N(H-r) \geq M \sum_{i'=0}^{i-1} (1+2i') \binom{N+2i'}{N-1}\}$.

If $M < N$, then $\lambda \leq \min\left\{\frac{MH+Nr}{2}, \frac{(M+N)r}{2} + \text{bound}_3\right\}$.

If $M \geq N$, then $\lambda \leq \min\left\{\text{bound}_0, \frac{(M+N)r}{2} + \text{bound}_3\right\}$.

Three Layered Neural Network

1. $H - r = 0 : \lambda = r \left(\frac{M+N}{2} \right), \theta = 1.$
2. $H = 1, r = 0 : \lambda = \min \left\{ \frac{M}{2}, \frac{N}{2} \right\}, \theta = \begin{cases} 1, & \text{if } M \neq N, \\ 2, & \text{if } M = N. \end{cases}$
3. $H - r = 1, r \geq 1 :$
 - (a) If $N > M + 1$ then $\lambda = (r - 1) \left(\frac{M+N}{2} \right) + \frac{2M+N}{2}$ and $\theta = 1.$
 - (b) If $N = M + 1$ then $\lambda = (r - 1) \left(\frac{M+N}{2} \right) + \frac{2M+N}{2}$ and $\theta = 2.$
 - (c) If $N = M$ then $\lambda = (r - 1) \left(\frac{M+N}{2} \right) + \frac{3M+3N-1}{4}$ and $\theta = 1.$
 - (d) If $N = M - 1$ then $\lambda = (r - 1) \left(\frac{M+N}{2} \right) + \frac{M+2N}{2}$ and $\theta = 2.$
 - (e) If $N < M - 1$ then $\lambda = (r - 1) \left(\frac{M+N}{2} \right) + \frac{M+2N}{2}$ and $\theta = 1.$
4. $H = 2, r = 0 :$
 - (a) If $N \geq M + 1$ then $\lambda = M$ and $\theta = 1.$
 - (b) If $N = M$ then $\lambda = \frac{3M-1}{3}$ and $\theta = 1.$
 - (c) If $N = M - 1$ then $\lambda = N$ and $\theta = 2.$
 - (d) If $N < M - 1$ then $\lambda = N$ and $\theta = 1.$

Normal Mixture Model

$$\text{bound}_3 = \frac{N(H - r + 1) + \sum_{i=0}^{k'-1} (k' - i) \binom{N+i}{N-1}}{2 + 2k'},$$

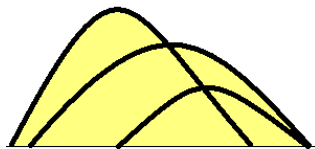
where $k' = \max\{i \in \mathbb{Z}; N(H - r + 1) \geq \sum_{i'=0}^{i-1} (1 + i') \binom{N+i'}{N-1}\}$.

If $1 < N$, then $\lambda \leq \min\left\{\frac{H - 1 + Nr}{2}, \frac{(N + 1)(r - 1)}{2} + \text{bound}_3\right\}$.

If $N = 1$, then

$$\lambda = r - 1 + \frac{i + i^2 + 2(H - (r - 1))}{4(i + 1)}, \theta = \begin{cases} 1, & \text{if } i^2 + i < 2(H - (r - 1)) \\ 2, & \text{if } i^2 + i = 2(H - (r - 1)) \end{cases}$$

where $i = \max\{j \in \mathbb{Z}; j^2 + j \leq 2(H - (r - 1))\}$.



1. $H - r = 0: \lambda = \frac{r-1+rN}{2}, \theta = 1.$

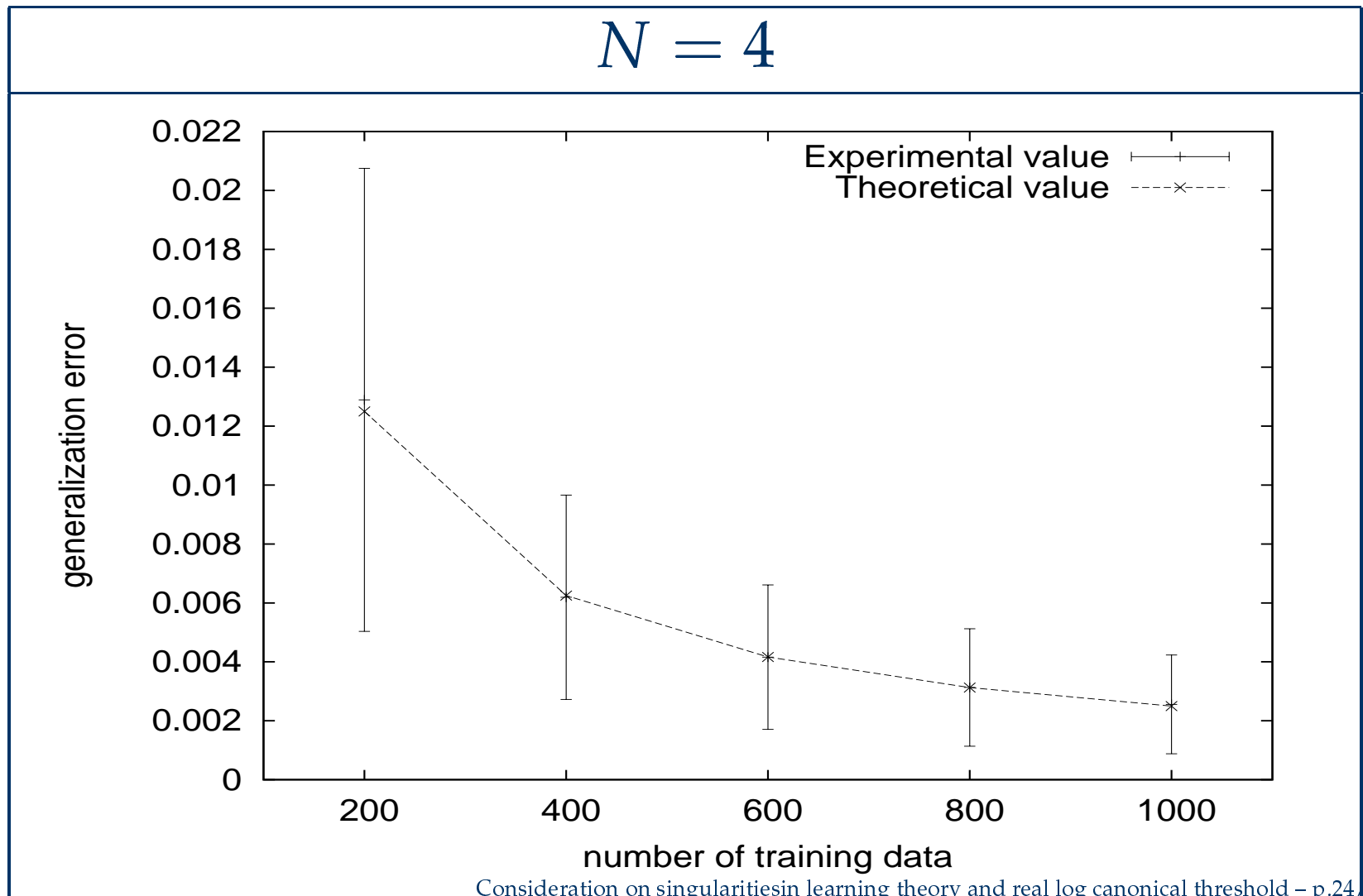
2. $H - r = 1:$

- (a) If $N > 2, \lambda = \frac{r(N+1)}{2}, \theta = 1.$

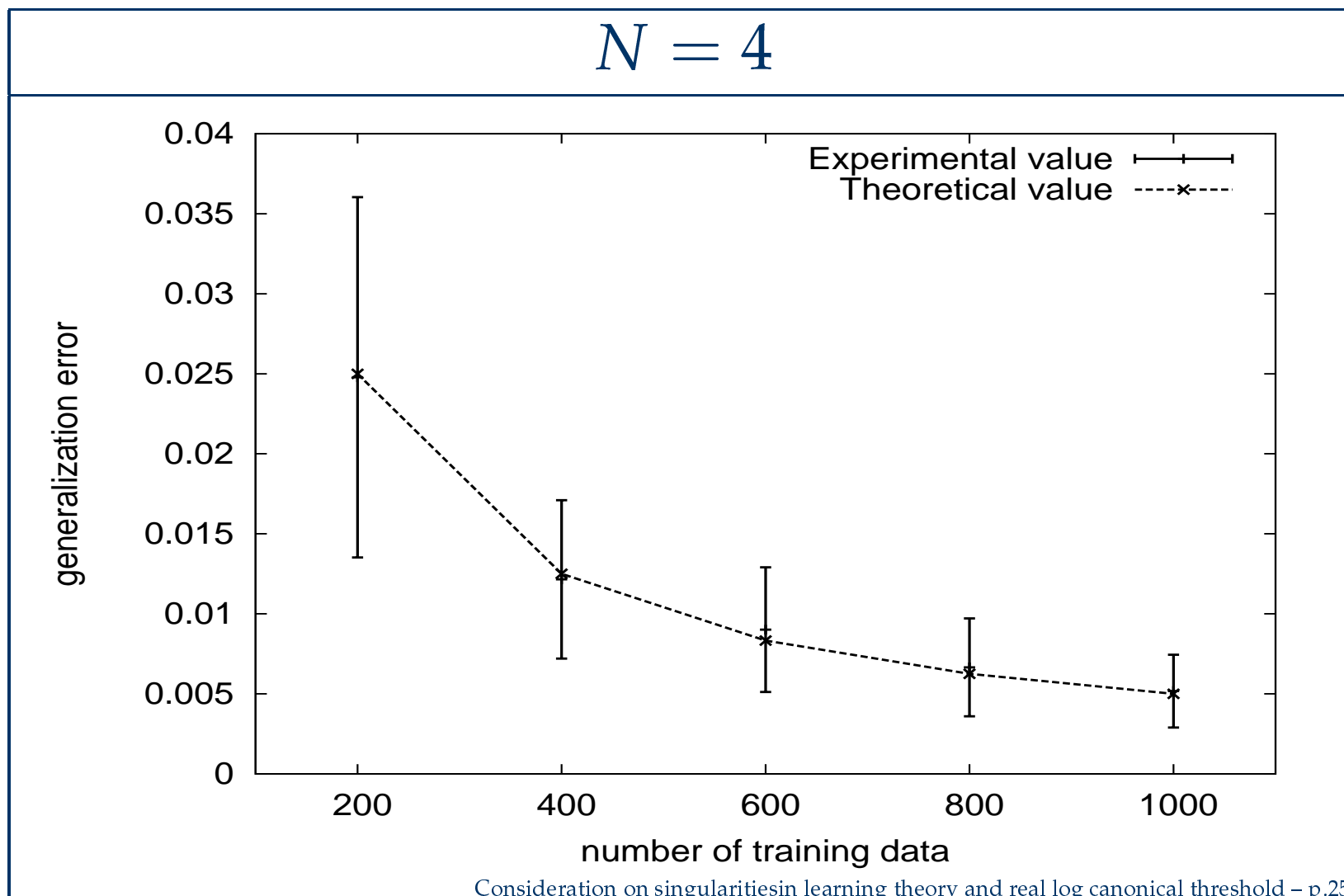
- (b) If $N = 2, \lambda = \frac{3r}{2}, \theta = 2.$

- (c) If $N = 1, \lambda = \frac{3}{4} + r - 1, \theta = 1.$

Numerical Results (By K.Nagata)



Numerical Results (By K.Nagata)



Lemma 4

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1H} \\ a_{21} & \cdots & a_{2H} \\ \vdots & \vdots & \vdots \\ a_{M1} & \cdots & a_{MH} \end{pmatrix}, Y = \begin{pmatrix} f_{11}(y) & \cdots & f_{1N}(y) \\ f_{21}(y) & \cdots & f_{2N}(y) \\ \vdots & & \vdots \\ f_{H1}(y) & \cdots & f_{HN}(y) \end{pmatrix},$$

where $y = (y_1, \dots, y_n)$. Then we have the blowing up process of $\|AY\|^2$ as

$$\sum_{j=1}^k \sum_{i=1}^M a_{ij}^2 \frac{Y_{jj}^{(j)}}{Y_{j-1,j-1}^{(j)}} + \left\| \frac{1}{Y_{j-1,j-1}^{(j)}} A^{(k)} Y^{(k)} \right\|^2,$$

where

Lemma 4

$$A^{(k)} = \begin{pmatrix} a_{1,k+1} & \cdots & a_{1H} \\ & & \vdots \\ a_{M,k+1} & \cdots & a_{MH} \end{pmatrix}, Y^{(k)} = \begin{pmatrix} Y_{k+1,k+1}^{(k)} & \cdots & Y_{k+1,N}^{(k)} \\ \vdots & & \vdots \\ Y_{H,k+1}^{(k)} & \cdots & Y_{HN}^{(k)} \end{pmatrix},$$

$$Y_{ij}^{(k)} = \begin{vmatrix} f_{11}(y) & \cdots & f_{1k}(y) & f_{1j}(y) \\ \vdots & \vdots & \vdots & \vdots \\ f_{k1}(y) & \cdots & f_{kk}(y) & f_{kj}(y) \\ f_{i1}(y) & \cdots & f_{ik}(y) & f_{ij}(y) \end{vmatrix},$$

and local analytic coordinate system $(y_1^{(k)}, \dots, y_n^{(k)})$ such that $Y_{ij}^{(k)}(y^{(k)})$ are normal crossing.

Lemma 5

$f_1(w_1, \dots, w_d), \dots, f_m(w_1, \dots, w_d)$
: homogeneous functions of w_1, \dots, w_d
with $\deg f_i(w_1, \dots, w_d) = n_i$.
 $\psi : C^\infty$ function with $\psi_{(0, \dots, 0)} \geq \psi_{(w_1^*, \dots, w_d^*)}$
normal crossing at $(0, \dots, 0), (w_1^*, \dots, w_d^*)$.

Then we have

$$\begin{aligned} & c_{(0, \dots, 0)}(\mathbb{R}^d, f_1^2 + \dots + f_m^2, \psi) \\ & \leq c_{(w_1^*, \dots, w_d^*)}(\mathbb{R}^d, f_1^2 + \dots + f_m^2, \psi), \end{aligned}$$

Remark

In general, it is not true that

$$C_{w_0}(\mathbb{R}^d, f_1^2 + \cdots + f_m^2) \leq C_{w^*}(\mathbb{R}^d, f_1^2 + \cdots + f_m^2),$$

where $w_0 \in \mathbb{R}^d$ satisfies

$$f_i(w_0) = \frac{\partial f_i}{\partial w_j}(w_0) = 0, 1 \leq i \leq m, 1 \leq j \leq d.$$

Example

$$\text{Let } f_1 = x(x - 1)^2,$$

$$f_2 = ((y - 1)^6 + x)(y^2 + (x - 1)^2)$$

$$f_3 = ((z - 1)^6 + x)(z^2 + (x - 1)^2).$$

Then we have

$$f_1 = f_2 = f_3 = \frac{\partial f_1}{\partial x} = \frac{\partial f_2}{\partial y} = \frac{\partial f_2}{\partial x} = \frac{\partial f_3}{\partial z} = \frac{\partial f_3}{\partial x} = 0$$

if and only if $x = 1, y = 0, z = 0$.

In this case, we have

$$c_{(1,0,0)}(f_1^2 + f_2^2 + f_3^2) = 1/4 + 1/4 + 1/4 >$$

$$c_{(0,1,1)}(f_1^2 + f_2^2 + f_3^2) = 1/2 + 1/12 + 1/12.$$

Lemma 6 Toric Variety

$f(x)$: non-degenerate with respect to its Newton polyhedron Γ_+
If $c = \min\{c' \geq 0 : c'\mathbf{e} \in \Gamma_+\} > 1$ then we have $c_0(f) = 1/c$.

$$f_1 = u_1^{s_{11}} \cdots u_d^{s_{1d}}, \dots, f_p = u_1^{s_{p1}} \cdots u_d^{s_{pd}},$$

$$g = u_1^{t_1} u_2^{t_2} \cdots u_d^{t_d} du$$

Γ_+ : the Newton diagram of $f_1^2 + \cdots + f_p^2$.

$$\mathbf{e} = (1, \dots, 1)^t, \mathbf{t} = (t_1, \dots, t_d)^t$$

$$c = \min\{c' \geq 0 : c'(\mathbf{t} + \mathbf{e}) \in \Gamma_+\}$$

θ' : the number of faces $T \ni c(\mathbf{t} + \mathbf{e})$ with $\dim. d - 1$ of Γ_+ .

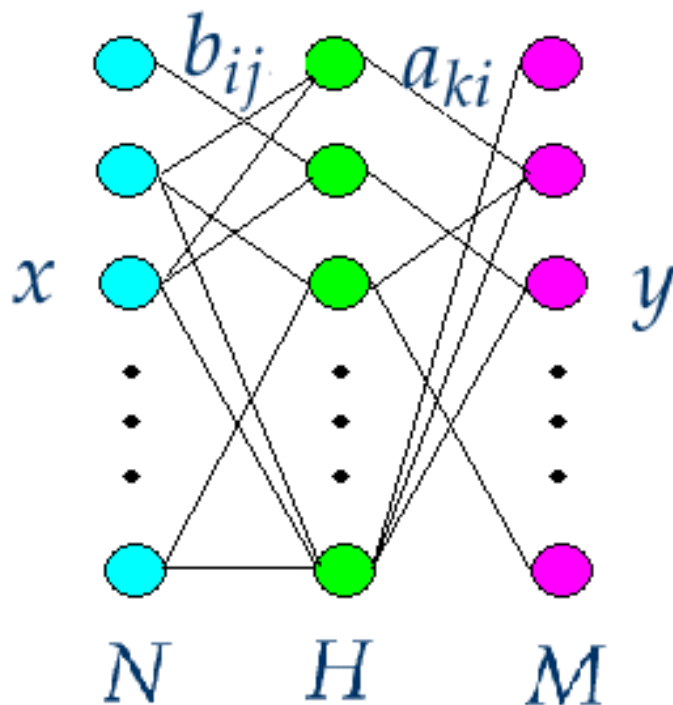
$$\theta = \min\{d, \theta'\},$$

Then, the largest pole of $\int_{\text{near } 0} (f_1^2 + \cdots + f_p^2)^z g$ is $1/c$ and its order is θ .

In this case, the condition $c > 1$ is not necessary.

Reduced rank regression

Reduced rank regression. (linear hidden units.)



(1) $N + r \leq M + H'$, $M + r \leq N + H'$, $H' + r \leq M + N$.

(a) If $M + H' + N + r$ is even, then $\theta = 1$ and $\lambda = \frac{-(H'+r)^2 - M^2 - N^2 + 2(H'+r)M + 2(H'+r)N + 2MN}{8}$.

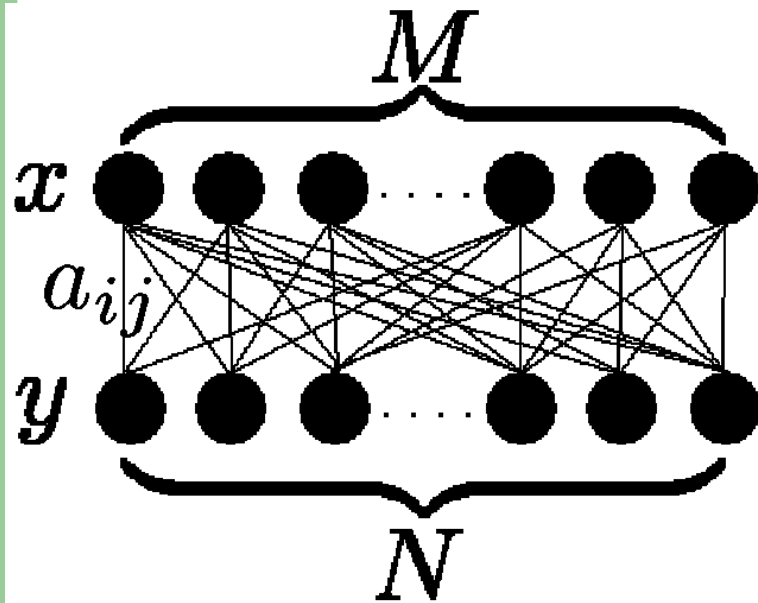
(b) If $M + H' + N + r$ is odd, then $\theta = 2$ and $\lambda = \frac{-(H'+r)^2 - M^2 - N^2 + 2(H'+r)M + 2(H'+r)N + 2MN + 1}{8}$.

(2) $M + H' < N + r$ then $\theta = 1$ and $\lambda = \frac{H'M - H'r + Nr}{2}$.

(3) $N + H' < M + r$ then $\theta = 1$ and $\lambda = \frac{H'M - H'r + Nr}{2}$.

(4) $M + N < H' + r$ then $\theta = 1$ and $\lambda = \frac{MN}{2}$.

Restricted Boltzmann Machine



$x = (x_j) \in \{1, -1\}^M$ inputs
 $y = (y_j) \in \{1, -1\}^N$ hidden units

$$p(x, y | a) = \frac{\exp(\sum_{i=1}^M \sum_{j=1}^N a_{ij} x_i y_j)}{Z(a)}$$

$$p(x | a) = \sum_{y_i = \pm 1} p(x, y | a) = \frac{\prod_{j=1}^N \prod_{i=1}^M \cosh(a_{ij})}{Z(a)}$$

$$\times \prod_{j=1}^N (2 \sum_{0 \leq p \leq M/2} \sum_{i_1 < \dots < i_{2p}} x_{i_1} \cdots x_{i_{2p}} \tanh(a_{i_1 j}) \cdots \tanh(a_{i_{2p} j}))$$

Restricted Boltzmann Machine

$$\prod_{j=1}^N \left(\sum_{0 \leq p \leq M/2} \sum_{i_1 < \dots < i_{2p}} x_{i_1} \cdots x_{i_{2p}} \tanh(a_{i_1 j}) \cdots \tanh(a_{i_{2p} j}) \right)$$

$$= \sum_{I \in \mathcal{I}} B_N^I x^I, \quad B = (b_{ij}) = (\tanh(a_{ij})) : M \times N \text{ matrix.}$$

Let $B = (b_{ij}) = (\tanh(a_{ij}))$. Denote $B^J = \prod_{i=1}^M \prod_{j=1}^N b_{ij}^{J_{ij}}$ where $J = (J_{ij})$ is an $M \times N$ matrix with $J_{ij} \in \{0, 1\}$.

$$\mathcal{I} = \{I \in \{0, 1\}^M \mid \sum_{i=1}^M I_i \text{ is even}\},$$

$$B^I = \sum_{\substack{\sum_{i=1}^M J_{ij} = 0 \pmod{2} \\ \sum_{j=1}^N J_{ij} = 0 \pmod{I_i}}} B^J \text{ for } I \in \mathcal{I}.$$

Consider $c_{B^*} \left(\sum_{I \in \mathcal{I}} \left(\frac{B_N^I}{B_N^{*I}} - \frac{B_N^0}{B_N^{*0}} \right)^2 \right)$

Example

Example (M=3), $I = (0,0,0), (1,1,0), (1,0,1)$ or $(0,1,1)$.

$$C_i = \begin{pmatrix} 1 & b_{1i}b_{2i} & b_{1i}b_{3i} & b_{2i}b_{3i} \\ b_{1i}b_{2i} & 1 & b_{2i}b_{3i} & b_{1i}b_{3i} \\ b_{1i}b_{3i} & b_{2i}b_{3i} & 1 & b_{1i}b_{2i} \\ b_{2i}b_{3i} & b_{1i}b_{3i} & b_{1i}b_{2i} & 1 \end{pmatrix}, \begin{pmatrix} B_N^{(0,0,0)} \\ B_N^{(1,1,0)} \\ B_N^{(1,0,1)} \\ B_N^{(0,1,1)} \end{pmatrix} = \left(\prod_{i=1}^N C_i \right) \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

Theorem

Consider the matrix $C_N = (c_N^{I,I'})$ where $c_N^{I,I'} = b_N^{I''}$ with $I' + I'' = I \pmod{2}$.
Note that $B_N = C_N B_{N-1}$.

Theorem

Let $K_1 \cap K_2 = \emptyset$ and $K_1 \cup K_2 = \{1, \dots, M\}$.

Set $\ell_I = \begin{cases} -1, & \text{if } \#\{i \in K_1 : I_i = 1\} \text{ is odd,} \\ 1, & \text{otherwise.} \end{cases}$

Then $\ell = (\ell_I)$ is an **eigenvector** of C_N and its **eigenvalue** is $\sum_{I \in \mathcal{I}} \ell_I b_N^I$.

Example ($M = 3$)

$$C_j = \begin{pmatrix} 1 & b_{1j}b_{2j} & b_{1j}b_{3j} & b_{2j}b_{3j} \\ b_{1j}b_{2j} & 1 & b_{2j}b_{3j} & b_{1j}b_{3j} \\ b_{1j}b_{3j} & b_{2j}b_{3j} & 1 & b_{1j}b_{2j} \\ b_{2j}b_{3j} & b_{1j}b_{3j} & b_{1j}b_{2j} & 1 \end{pmatrix}, D = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & 1 & -1 \\ 1 & -1 & -1 & 1 \end{pmatrix},$$

$$B_N = \begin{pmatrix} B_N^{(0,0,0)} \\ B_N^{(1,1,0)} \\ B_N^{(1,0,1)} \\ B_N^{(0,1,1)} \end{pmatrix} = C_N C_{N-1} C_1 \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, DC_j D^{-1} = \begin{pmatrix} s_j^0 & 0 & 0 & 0 \\ 0 & s_j^1 & 0 & 0 \\ 0 & 0 & s_j^2 & 0 \\ 0 & 0 & 0 & s_j^3 \end{pmatrix}$$

$$\begin{cases} s_j^0 = 1 + b_{1j}b_{2j} + b_{1j}b_{3j} + b_{2j}b_{3j}, & s_j^1 = 1 + b_{1j}b_{2j} - b_{1j}b_{3j} - b_{2j}b_{3j}, \\ s_j^2 = 1 - b_{1j}b_{2j} + b_{1j}b_{3j} - b_{2j}b_{3j}, & s_j^3 = 1 - b_{1j}b_{2j} - b_{1j}b_{3j} + b_{2j}b_{3j}, \end{cases}$$

Main Theorem

(1) If $M = 2$ then

$$\lambda = \frac{1}{2} \text{ and } \theta = \begin{cases} 2, & \text{if } N = 1, b^* = 0 \\ 1, & \text{otherwise.} \end{cases}$$

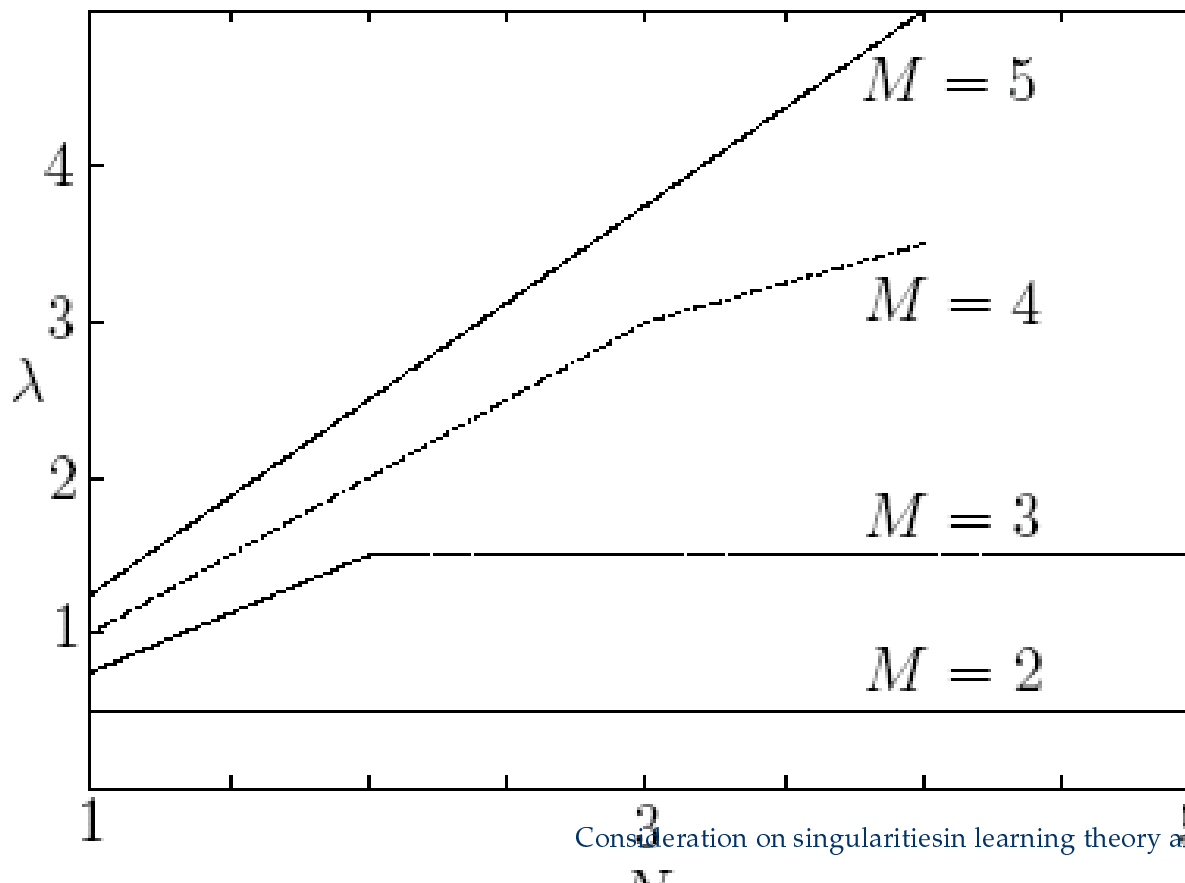
(2) If $M = 3$ then

$$\lambda = \begin{cases} 3/4, & \text{if } N = 1, b^* = 0 \\ 1/2, & \text{if } N = 1, b^* \neq 0, \prod_{i=1}^3 b_{i1}^* = 0 \\ 3/2, & \text{if } N = 1, \prod_{i=1}^3 b_{i1}^* \neq 0 \\ 3/2, & \text{if } N \geq 2, \end{cases} \quad \theta = \begin{cases} 3, & \text{if } N = 2, b^* = 0, \\ 2, & \text{if } N = 2, b^* \neq 0, \\ & b_{i_0 j}^* = b_{i_1 j}^* = 0 \\ 2, & \text{if } N = 2, b_{i_0 j_0}^* b_{i_1 j_0}^* \neq 0, \\ & b_{i_2 j_0}^* = b_{ij}^* = 0 \text{ for } 1 \leq i \leq 3 \\ 1, & \text{otherwise,} \end{cases}$$

$i_0, i_1, i_2 \in \{1, 2, 3\}$ are different from each other, $1 \leq j_0 \leq N$.

Main Theorem 2

If $M > N$ and $b^* = 0$ then $\lambda = \frac{MN}{4}$ and $\theta = \begin{cases} 1, & \text{if } M > N + 1, \\ M, & \text{if } M = N + 1. \end{cases}$



Main Theorem 2

$(b_{1j}, b_{2j}, \dots, b_{Mj}) \neq 0$ for $j = 1, \dots, N_0$

$(b_{1j}, b_{2j}, \dots, b_{Mj}) = 0$ for $j = N_0 + 1, \dots, N$ in V ,

where V is a sufficiently small neighborhood of b^* .

Then

$$\frac{M(N - N_0)}{4} \leq \lambda \leq \frac{M(N - N_0)}{4} + \frac{MN_0}{2},$$

if $M > N - N_0$

$$\frac{M(M - 1)}{4} + \frac{MN_0}{2} \leq \lambda \leq \frac{2N_0 + (M - 1)(M - 2)}{4} + \frac{MN_0}{2}$$
$$\left(< \frac{MN_0}{2} + \frac{M(N - N_0)}{4} \right),$$

if $M \leq N - N_0$.

Conclusion

Conclusion

- We show new bound values of learning coefficients for Vandermonde matrix type singularities,
- and the explicit values in some conditions.
- Our future research aims to improve our methods and to apply them to obtain exact values.

Proof

$$\left\langle \begin{pmatrix} b_1^m \\ \vdots \\ b_H^m \end{pmatrix}, \begin{pmatrix} 0 \\ b_2^m (b_1^Q - b_2^Q) \\ \vdots \\ b_H^m (b_1^Q - b_H^Q) \end{pmatrix}, \dots, \begin{pmatrix} 0 \\ \vdots \\ 0 \\ b_1^m (b_1^Q - b_H^Q) \cdots (b_{H-1}^Q - b_H^Q) \end{pmatrix} \right\rangle$$

$$= \langle \mathbf{b}'_1, \mathbf{b}'_2, \dots, \mathbf{b}'_H \rangle.$$

We have $\langle \mathbf{b}'_1, \mathbf{b}'_2, \dots, \mathbf{b}'_H \rangle = \langle \mathbf{b}'_1, \mathbf{b}'_2, \dots, \mathbf{b}'_{H-1}, \mathbf{b}''_H \rangle$.

Let $g_{j,j}(x), \dots, g_{H,j}(x)$ s.t. $g_{j',j}(x\gamma_{j'}) = g_{j'',j}(x\gamma_{j''})$ if $|b_{j'}^*| = |b_{j''}^*| \neq 0$ and $g_{j',j}(x) - g_{j'',j}(x')$ can be divided by $x^Q - x'^Q$ if $b_{j'}^* = b_{j''}^* = 0$

Assume that $(0, \dots, 0, g_{j,j}(b_j)\mathbf{b}''_{jj}, \dots, g_{H,j}(b_H)\mathbf{b}''_{Hj}) \in \langle \mathbf{b}''_j, \dots, \mathbf{b}''_H \rangle$

Proof

Since

$$\begin{pmatrix} 0 \\ \vdots \\ 0 \\ b_{j-1}^m (b_1^Q - b_{j-1}^Q) \cdots (b_{j-2}^Q - b_{j-1}^Q) \\ \vdots \\ b_H^m (b_1^Q - b_H^Q) \cdots (b_{j-2}^Q - b_H^Q) \end{pmatrix} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ g_{j-1,j-1}(b_{j-1}) \mathbf{b}_{j-1,j-1}'' \\ \vdots \\ g_{H,j-1}(b_H) \mathbf{b}_{H,j-1}'' \end{pmatrix},$$

Proof

we have $\mathbf{b}'_{j-1} =$

$$\mathbf{b}''_{j-1} g_{j-1,j-1}(b_{j-1}) + \begin{pmatrix} 0 \\ \vdots \\ 0 \\ (g_{j,j-1}(b_j) - g_{j-1,j-1}(b_{j-1})) \mathbf{b}''_{j,j-1} \\ \vdots \\ (g_{H,j-1}(b_H) - g_{j-1,j-1}(b_{j-1})) \mathbf{b}''_{H,j-1} \end{pmatrix}$$

Proof

$$= \mathbf{b}_{j-1}'' g_{j-1,j-1}(b_{j-1}) + (0, \dots, 0, g_{j,j}(b_j) \mathbf{b}_{j,j}'', \dots, g_{H,j}(b_H) \mathbf{b}_{H,j}'')^t,$$

where

$$\left\{ \begin{array}{l} g_{k,j}(b_k) = g_{k,j-1}(b_k) - g_{j-1,j-1}(b_{j-1}), \\ \quad \text{if } |b_k^*| \neq |b_{j-1}^*|, \\ g_{k,j}(b_k) = (g_{k,j-1}(b_k) - g_{j-1,j-1}(b_{j-1})) / (b_{j-1}/\gamma_{j-1} - b_k/\gamma_k), \\ \quad \text{if } |b_k^*| = |b_{j-1}^*| \neq 0, \\ g_{k,j}(b_k) = (g_{k,j-1}(b_k) - g_{j-1,j-1}(b_{j-1})) / (b_{j-1}^Q - b_k^Q) \\ \quad \text{if } b_k^* = b_{j-1}^* = 0. \end{array} \right.$$

By the inductive assumption,

$$(0, \dots, 0, g_{j,j}(b_j) \mathbf{b}_{j,j}'', \dots, g_{H,j}(b_H) \mathbf{b}_{H,j}'')^t \in \langle \mathbf{b}_j'', \dots, \mathbf{b}_H'' \rangle.$$

$$\langle \mathbf{b}'_1, \dots, \mathbf{b}'_{j-1}, \mathbf{b}''_j, \dots, \mathbf{b}''_H \rangle = \langle \mathbf{b}'_1, \dots, \mathbf{b}'_{j-2}, \mathbf{b}''_{j-1}, \mathbf{b}''_j, \dots, \mathbf{b}''_H \rangle.$$

Lemma 例

$$b_1^* = 1, b_2^* = 1, b_3^* = 2, b_4^* = 2$$

$$\left(\begin{array}{cccc} b_1 & 0 & 0 & 0 \\ b_2 & b_2(b_1^2 - b_2^2) & 0 & 0 \\ b_3 & b_3(b_1^2 - b_3^2) & b_3(b_1^2 - b_3^2)(b_2^2 - b_3^2) & 0 \\ b_4 & b_4(b_1^2 - b_4^2) & b_4(b_1^2 - b_4^2)(b_2^2 - b_4^2) & b_4(b_1^2 - b_4^2)(b_2^2 - b_4^2)(b_3^2 - b_4^2) \end{array} \right)$$

$$\left(\begin{array}{cccc} b_1 & 0 & 0 & 0 \\ b_2 & b_2(b_1^2 - b_2^2) & 0 & 0 \\ b_3 & b_3(b_1^2 - b_3^2) & b_3(b_1^2 - b_3^2)(b_2^2 - b_3^2) & 0 \\ b_4 & b_4(b_1^2 - b_4^2) & b_4(b_1^2 - b_4^2)(b_2^2 - b_4^2) & (b_3 - b_4) \end{array} \right) R_1$$

Lemma 例

$$\begin{pmatrix}
 & b_3(b_2^2 - b_3^2)(b_1^2 - b_3^2) & & 0 \\
 b_3(b_2^2 - b_3^2)(b_1^2 - b_3^2) + b_4(b_1^2 - b_4^2)(b_2^2 - b_4^2) - b_3(b_1^2 - b_3^2)(b_2^2 - b_3^2) & & & (b_3 - b_4) \\
 & & & \\
 \left(\begin{array}{cccc}
 b_1 & 0 & 0 & 0 \\
 b_2 & b_2(b_1^2 - b_2^2) & 0 & 0 \\
 b_3 & b_3(b_1^2 - b_3^2) & b_3(b_2^2 - b_3^2)(b_1^2 - b_3^2) & 0 \\
 b_4 & b_4(b_1^2 - b_4^2) & b_3(b_2^2 - b_3^2)(b_1^2 - b_3^2) & (b_3 - b_4)
 \end{array} \right) & R_3 \\
 & & & \\
 \left(\begin{array}{cccc}
 b_1 & 0 & 0 & 0 \\
 b_2 & b_2(b_1 + b_2)(b_1 - b_2) & 0 & 0 \\
 b_3 & b_3(b_1^2 - b_3^2) & 1 & 0 \\
 b_4 & b_4(b_1^2 - b_4^2) & 1 & (b_3 - b_4)
 \end{array} \right) & R_4
 \end{pmatrix}$$

Lemma 例

$$\left(\begin{array}{cccc} b_1 & & 0 & & 0 & 0 \\ b_2 & & (b_1 - b_2) & & 0 & 0 \\ b_3 & (b_1 - b_2) + b_3(b_1^2 - b_3^2) - (b_1 - b_2) & & & 1 & 0 \\ b_4 & (b_1 - b_2) + b_4(b_1^2 - b_4^2) - (b_1 - b_2) & & & 1 & (b_3 - b_4) \end{array} \right) R_5$$

$$\left(\begin{array}{cccc} b_1 & 0 & 0 & 0 \\ b_2 & (b_1 - b_2) & 0 & 0 \\ b_3 & (b_1 - b_2) & 1 & 0 \\ b_4 & (b_1 - b_2) & 1 & (b_3 - b_4) \end{array} \right) R_6$$

Lemma 例

$$\begin{pmatrix} b_1 & 0 & 0 & 0 \\ b_2 & b_1 - b_2 & 0 & 0 \\ b_3 & 0 & 1 & 0 \\ b_3 & 0 & 1 & b_3 - b_4 \end{pmatrix} R_7 = \begin{pmatrix} b_1 & 0 & 0 & 0 \\ b_1 & b_1 - b_2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & b_3 - b_4 \end{pmatrix} R_8$$
$$= \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & b_1 - b_2 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & b_3 - b_4 \end{pmatrix} R_9$$

Lemma

Let $\{|b_0^{**}|, \dots, |b_r^{**}|; |b_i^*|, i = 1, \dots, H\}$.

Assume that $b_1^* = \dots = b_{H_0}^* = b_0^{**}$, $|b_{H_0+1}^*| = \dots = |b_{H_0+H_1}^*| = |b_1^{**}|$,
 \dots , $|b_{H_0+\dots+H_{r-1}+1}^*| = \dots = |b_{H_0+\dots+H_r}^*| = |b_r^{**}|$.

$$\exists R \text{ s.t. } B'R = \begin{pmatrix} B^{(0)} & 0 & 0 & \dots & 0 \\ 0 & B^{(1)} & 0 & \dots & 0 \\ \vdots & & & \ddots & \\ 0 & 0 & 0 & \dots & B^{(r)} \end{pmatrix}, \text{ where}$$

$$B^{(0)} = \begin{pmatrix} b_1^{(0)m} & \dots & b_1^{(0)Q(H_0-1)+m} \\ \vdots & & \vdots \\ b_{H_0}^{(0)m} & \dots & b_{H_0}^{(0)Q(H_0-1)+m} \end{pmatrix}, B^{(\alpha)} = \begin{pmatrix} 1 & b_1^{(\alpha)} & \dots & b_1^{(\alpha)H_\alpha-1} \\ \vdots & & \vdots & \\ 1 & b_{H_\alpha}^{(\alpha)} & \dots & b_{H_\alpha}^{(\alpha)H_\alpha-1} \end{pmatrix}$$

Proof

Since

$$\begin{pmatrix} \prod_{j=1}^N b_{1j}^{\ell_j} \\ \prod_{j=1}^N b_{2j}^{\ell_j} \\ \vdots \\ \prod_{j=1}^N b_{Hj}^{\ell_j} \end{pmatrix} = \begin{pmatrix} b_{11}^{\ell'_1} \prod_{j=2}^N b_{1j}^{\ell_j} & 0 & \cdots & 0 \\ 0 & b_{21}^{\ell'_1} \prod_{j=2}^N b_{2j}^{\ell_j} & \cdots & 0 \\ \vdots & \ddots & \ddots & 0 \\ 0 & 0 & \cdots & b_{H1}^{\ell'_1} \prod_{j=2}^N b_{Hj}^{\ell_j} \end{pmatrix} \times \begin{pmatrix} b_{11}^{\ell_1 - \ell'_1} \\ b_{21}^{\ell_1 - \ell'_1} \\ \vdots \\ b_{H1}^{\ell_1 - \ell'_1} \end{pmatrix}, \text{ we have Theorem 1.}$$

Theorem 2 の証明

$$\Psi' dz = \left\{ \prod_{\tau=0}^r v_{\tau}^{t_{\tau}} \left(d_1^2 + d_2^2 + \cdots + d_K^2 \right) + \sum_{i=K+1}^{p'} C_i^2 \right\} \prod_{\tau=0}^r v_{\tau}^{q_{\tau}} ddvdade_m$$

$$C_i = \prod_{\tau=0}^r v_{\tau}^{t(i,0,\tau)} \sum_{\substack{i \leq m \leq p' \\ J_m^{(\alpha)} = 0}} g(i, m) a_m e_m \prod_{\substack{k+1 \leq m' < i \\ J_{m'}^{(\alpha)} = 0}} (e_m^Q - e_{m'}^Q)$$

$$+ \sum_{J \in RJ^{(\alpha)}} \prod_{\tau=0}^r v_{\tau}^{t(i,J,\tau)} \sum_{\substack{i \leq m \leq p' \\ J_m^{(\alpha)} = J}} g(i, m) a_m e_m \prod_{\substack{k+1 \leq m' < i \\ J_{m'}^{(\alpha)} = J}} (e_m - e_{m'})$$

$$+ \sum_{J \notin RJ^{(\alpha)}, J \neq 0} \prod_{\tau=0}^r v_{\tau}^{t(i,J,\tau)} \sum_{\substack{i \leq m \leq p' \\ J_m^{(\alpha)} = J}} g(i, m) a_m \prod_{\substack{k+1 \leq m' < i \\ J_{m'}^{(\alpha)} = J}} (e_m - e_{m'}).$$

Theorem 2 の証明

$$\text{Let } A = \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1H} \\ a_{21} & a_{22} & \cdots & a_{2H} \\ \vdots & \vdots & \ddots & \vdots \\ a_{M1} & a_{M2} & \cdots & a_{MH} \end{pmatrix} \text{ and}$$

$$B = \begin{pmatrix} b_1^m & 0 & 0 & \cdots & 0 \\ 0 & b_2^m (b_2^Q - b_1^Q) & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & b_H^m (b_H^Q - b_1^Q)(b_H^Q - b_2^Q) \cdots (b_H^Q - b_{H-1}^Q) \end{pmatrix}$$

の log canonical threshold と同じ

Normal mixture model

$$\begin{pmatrix} a_1 \\ \vdots \\ a_H \\ a_{H+1}^* \\ \vdots \\ a_{H+r}^* \end{pmatrix}^t \begin{pmatrix} \prod_{j=1}^N b_{1j}^{\ell_j} \\ \vdots \\ \prod_{j=1}^N b_{H-1,j}^{\ell_j} \\ \prod_{j=1}^N b_{Hj}^{\ell_j} \\ \prod_{j=1}^N b_{H+1,j}^{*\ell_j} \\ \vdots \\ \prod_{j=1}^N b_{H+r,j}^{*\ell_j} \end{pmatrix} = \begin{pmatrix} a_1 \\ \vdots \\ a_{H-1} \\ a_{H+1}^* \\ \vdots \\ a_{H+r}^* \end{pmatrix}^t \begin{pmatrix} \prod_{j=1}^N b_{1j}^{\ell_j} - \prod_{j=1}^N b_{Hj}^{\ell_j} \\ \prod_{j=1}^N b_{2j}^{\ell_j} - \prod_{j=1}^N b_{Hj}^{\ell_j} \\ \vdots \\ \prod_{j=1}^N b_{H-1,j}^{\ell_j} - \prod_{j=1}^N b_{Hj}^{\ell_j} \\ \prod_{j=1}^N b_{H+1,j}^{*\ell_j} - \prod_{j=1}^N b_{Hj}^{\ell_j} \\ \vdots \\ \prod_{j=1}^N b_{H+r,j}^{*\ell_j} - \prod_{j=1}^N b_{Hj}^{\ell_j} \end{pmatrix}$$

by using $\sum_{i=1}^H a_i = 1, \sum_{i=1}^r a_{H+i}^* = -1$.

$$\left(\prod_{j=1}^N (b_{1j} - b_{Hj})^{\ell_j}, \dots, \prod_{j=1}^N (b_{H-1,j} - b_{Hj})^{\ell_j}, \prod_{j=1}^N (b_{H+1,j}^* - b_{Hj})^{\ell_j}, \dots, \prod_{j=1}^N (b_{H+r,j}^* - b_{Hj})^{\ell_j} \right)$$