

Supplementary Materials for Generalization Error Bound for Hyperbolic Ordinal Embedding

A. Proof of Lemma 4

Proof. (The first equation) Let $\mathbf{Z} := [\mathbf{z}_1 \ \mathbf{z}_2 \ \dots \ \mathbf{z}_N] \in \mathbb{R}^{D,N}$, and $\mathbf{Z} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^\top$ be the singular value decomposition of \mathbf{Z} . Regarding the Gramian matrix, we can obtain the singular value decomposition of \mathbf{G} as follows.

$$\begin{aligned} \mathbf{G} &= \mathbf{Z}^\top \mathbf{Z} \\ &= \mathbf{V}\mathbf{\Sigma}^\top \mathbf{U}^\top \mathbf{U}\mathbf{\Sigma}\mathbf{V}^\top \\ &= \mathbf{V}\mathbf{\Sigma}^2 \mathbf{V}^\top. \end{aligned} \quad (37)$$

Hence, we have $\|\mathbf{G}\|_* = \text{Tr}(\mathbf{\Sigma}^2)$. Therefore, we have that

$$\begin{aligned} \|\mathbf{G}\|_* &= \text{Tr}(\mathbf{\Sigma}^2) \\ &= \text{Tr}(\mathbf{\Sigma}^\top \mathbf{\Sigma} \mathbf{V}^\top \mathbf{V}) \\ &= \text{Tr}(\mathbf{V}\mathbf{\Sigma}\mathbf{U}^\top \mathbf{U}\mathbf{\Sigma}\mathbf{V}^\top) \\ &= \text{Tr}(\mathbf{Z}^\top \mathbf{Z}) \\ &= \text{Tr}(\mathbf{G}) \\ &= \sum_{n=1}^N [\mathbf{G}]_{n,n} \\ &= \sum_{n=1}^N \mathbf{z}_n^\top \mathbf{z}_n \\ &= \sum_{n=1}^N (\Delta_{\mathbb{R}^D}(\mathbf{0}, \mathbf{z}_n))^2, \end{aligned} \quad (38)$$

which completes the proof of the first equation.

(The second equation) For any $i, j \in [N]$, we have that

$$\begin{aligned} [\mathbf{G}]_{i,j} &= \mathbf{z}_i^\top \mathbf{z}_j \\ &\leq \|\mathbf{z}_i\|_2 \|\mathbf{z}_j\|_2 \\ &\leq \max \left\{ \|\mathbf{z}_i\|_2^2, \|\mathbf{z}_j\|_2^2 \right\} \\ &\leq \max_{n \in [N]} \|\mathbf{z}_n\|_2^2 \\ &= \max_{n \in [N]} [\mathbf{G}]_{n,n} \\ &\leq \max_{n \in [N]} (\Delta_{\mathbb{R}^D}(\mathbf{0}, \mathbf{z}_n))^2, \end{aligned} \quad (39)$$

where the first inequality holds from the Cauchy Schwartz inequality. Hence, we have $\max_{i,j \in [N]} [\mathbf{G}]_{i,j} \leq \max_{n \in [N]} [\mathbf{G}]_{n,n}$. Conversely, obviously, $\max_{i,j \in [N]} [\mathbf{G}]_{i,j} \geq \max_{n \in [N]} [\mathbf{G}]_{n,n}$ is valid. Therefore, we have that $\max_{i,j \in [N]} [\mathbf{G}]_{i,j} = \max_{n \in [N]} [\mathbf{G}]_{n,n}$. Since the left hand side equals $\|\mathbf{G}\|_{\max}$,

and right hand side equals $\max_{n \in [N]} (\Delta_{\mathbb{R}^D}(\mathbf{0}, \mathbf{z}_n))^2$, we have that $\|\mathbf{G}\|_{\max} = \max_{n \in [N]} (\Delta_{\mathbb{R}^D}(\mathbf{0}, \mathbf{z}_n))^2$, which completes the proof of the second equation. \square

B. Proof of Lemma 6

Proof. Define

$$\begin{aligned} \mu &:= \min \{ \Delta^*(i, j) \mid i \neq j \}, \\ \xi &:= \min \left\{ \left| 1 - \frac{\Delta^*(i, j)}{\Delta^*(i', j')} \right| \mid \begin{array}{l} (i, j), (i', j') \in [N] \times [N], \\ i \neq j, i' \neq j', (i, j) \neq (i', j') \end{array} \right\}. \end{aligned} \quad (40)$$

We assume that $\mu, \xi > 0$ holds as in the discussion in Section 2.1. Let $\epsilon := \frac{1}{3}\xi$. Let ν, η_{\max} be the constants determined on the weighted graph defined by (Sarkar, 2011). Let $\tau := \max \left\{ \eta_{\max}, \frac{\nu(1+\epsilon)}{\mu\epsilon}, \frac{1}{\mu\epsilon} \right\}$. Then by the $(1+\epsilon)$ -distortion algorithm, we can obtain representations $\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_N \in \mathbb{L}^2$ such that

$$(1-\epsilon)\tau\Delta^*(i, j) < \Delta_{\mathbb{L}^2}(\mathbf{z}_i, \mathbf{z}_j) \leq (1+\epsilon)\tau\Delta^*(i, j), \quad (41)$$

for any $i, j \in [N]$. Here, the following is valid for $i, j, i', j' \in [N]$: if $\Delta^*(i, j) > \Delta^*(i', j')$, then

$$\begin{aligned} &\Delta_{\mathbb{L}^2}(\mathbf{z}_i, \mathbf{z}_j) - \Delta_{\mathbb{L}^2}(\mathbf{z}_{i'}, \mathbf{z}_{j'}) \\ &\geq \tau(1-\epsilon)\Delta^*(i, j) - (1+\epsilon)\Delta^*(i', j') \\ &\geq \tau\Delta^*(i, j) \left[(1-\epsilon) - (1+\epsilon) \frac{\Delta^*(i', j')}{\Delta^*(i, j)} \right] \\ &\geq \tau\Delta^*(i, j) \left[\left(1 - \frac{1}{3}\xi\right) - \left(1 + \frac{1}{3}\xi\right)(1-\xi) \right] \\ &> \tau\Delta^*(i, j) \frac{1}{3}\xi \\ &\geq \tau\mu\epsilon \\ &\geq 1. \end{aligned} \quad (42)$$

\square

C. Proof of Lemma 7

Proof. Let c be the center of the 6-star subgraph and n_1, n_2, \dots, n_6 be its neighborhood. Define $\Delta_m^* := \Delta^*(c, n_m)$. Let $B_\Delta(\mathbf{z})$ be an open ball of radius Δ centered at \mathbf{z} in \mathbb{R}^2 . Assume that $\mathbf{z}_c, \mathbf{z}_{n_1}, \dots, \mathbf{z}_{n_6} \in \mathbb{R}^2$ satisfies (2) and define $\Delta_m := \Delta_{\mathbb{R}^2}(\mathbf{z}_c, \mathbf{z}_{n_m})$. For m, m' such that $\Delta_m^* < \Delta_{m'}^*$, $\mathbf{z}_{n_m} \in B_{\Delta_{m'}}(\mathbf{z}_c)$ and $\mathbf{z}_{n_{m'}} \notin B_{\Delta_m}(\mathbf{z}_{n_m})$,

are necessary. Hence, $\angle z_{n_m} z_c z_{n_m'} > 60^\circ$. Thus, segments $z_c z_{n_1}, z_c z_{n_2}, \dots, z_c z_{n_6}$ partition 360° into 6 angles larger than 60° , which is contradiction. \square

D. Proof of Proposition 8

Proof. If $a, b, c > 0$ and assume $x > 0$, then we have that

$$ax^2 + bx < c \Leftrightarrow 0 < x < -b + \sqrt{b^2 + 4ac}. \quad (43)$$

We are going to find S that satisfies

$$\begin{aligned} & O\left(L_\phi(\exp R)^2 \left(\sqrt{\frac{N \ln N}{S}} + \frac{N \ln N}{S} + \sqrt{\frac{\ln \frac{2}{\delta}}{S}}\right)\right) \\ & < 2\alpha \frac{V_{\min}^{\mathbb{R}^2}}{|\mathcal{T}|}, \end{aligned} \quad (44)$$

where $L_\phi = 1$. This corresponds to $x = \frac{1}{\sqrt{S}}$, $a = O(N \ln N)$, $b = O\left(\sqrt{N \ln N} + \sqrt{\ln \frac{2}{\delta}}\right)$, $c = O\left(\frac{2\alpha V_{\min}^{\mathbb{R}^2}}{(\exp R)^2 |\mathcal{T}|}\right)$. We can formulate the condition as follows:

$$\begin{aligned} 0 < \frac{1}{\sqrt{S}} < -b + \sqrt{b^2 + 4ac} \\ \Leftrightarrow S > \frac{1}{(-b + \sqrt{b^2 + 4ac})^2}. \end{aligned} \quad (45)$$

Here, we have that

$$\begin{aligned} & \frac{1}{(-b + \sqrt{b^2 + 4ac})^2} \\ & = \frac{(b + \sqrt{b^2 + 4ac})^2}{(4ac)^2} \\ & \leq \frac{(b + b(1 + \frac{2ac}{b^2}))^2}{(4ac)^2} \\ & = \left(\frac{b(2 + \frac{2ac}{b^2})}{4ac}\right)^2 \\ & = \left(\frac{b}{2ac} + \frac{1}{2b}\right)^2 \\ & \leq \left(\frac{b}{2ac} + \frac{1}{2b}\right)^2 \\ & = O\left(\frac{(\exp R)^2 |\mathcal{T}|}{\alpha V_{\min}^{\mathbb{R}^2} N \ln N} \left(\sqrt{N \ln N} + \sqrt{\ln \frac{2}{\delta}}\right) \right. \\ & \quad \left. + \frac{1}{\sqrt{N \ln N} + \sqrt{\ln \frac{2}{\delta}}}\right)^2, \end{aligned} \quad (46)$$

which completes the proof. Here, the first inequality holds since $\sqrt{1+y} \leq 1 + \frac{1}{2}y$. \square

E. Proof of Lemma 9

Proof. The statement (iv) follows from (ii) and (iii). Therefore, we prove (i)-(iii) in the following.

(Sufficiency) (i) Assume that $(z_n)_{n=1}^N \in (\mathbb{L}^D)^N$ is valid. For $d = 0, 1, \dots, D$ and $n = 1, 2, \dots, N$, we denote the d -th element of $z_n \in \mathbb{L}^D$ by $z_{d,n}$, and for $n = 1, 2, \dots, N$, we define $z_n^- \in \mathbb{R}^1$ and $z_n^+ \in \mathbb{R}^1$ by

$$z_n^- := [z_{0,n}], z_n^+ := [z_{1,n} \ z_{2,n} \ \dots \ z_{D,n}]^\top. \quad (47)$$

Also, we define $Z^- \in \mathbb{R}^{1,N}$ and $Z^+ \in \mathbb{R}^{D-1,N}$ by

$$\begin{aligned} Z^- &:= [z_1^- \ z_2^- \ \dots \ z_N^-], \\ Z^+ &:= [z_1^+ \ z_2^+ \ \dots \ z_N^+], \end{aligned} \quad (48)$$

respectively. Define $L^-, L^+ \in \mathbb{R}^{N,N}$ by $L^- := (Z^-)^\top Z^-$ and $L^+ := (Z^+)^\top Z^+$, respectively. For all $x \in \mathbb{R}^N$, $x^\top L^- x = (L^- x)^\top L^- x \geq 0$. Therefore, we have $L^- \succeq \mathbf{O}$. Likewise, $L^+ \succeq \mathbf{O}$ is valid, and thus we obtain (a). Because $Z^- \in \mathbb{R}^{1,N}$ and $Z^+ \in \mathbb{R}^{D-1,N}$, we have $\text{rank } L^- = 1$ and $\text{rank } L^+ \leq D$, respectively. As $z_n \in \mathbb{L}^D$, $z_{0,n} \geq 1$ is valid, $\text{rank } L^- \neq 0$, and therefore $\text{rank } L^- = 1$. Thus, we have (b). If $i, j \in [N]$, then the following inequality holds:

$$\begin{aligned} & [L^+ - L^-]_{i,j} \\ & = (z_i^+)^\top z_j^+ - (z_i^-)^\top z_j^- \\ & = (z_i^+)^\top z_j^+ - \sqrt{1 + (z_i^+)^\top (z_i^+)} \sqrt{1 + (z_j^+)^\top (z_j^+)} \\ & = (z_i^+)^\top z_j^+ - \left\| \begin{bmatrix} 1 & (z_i^+)^\top \end{bmatrix} \right\|_2 \left\| \begin{bmatrix} 1 & (z_j^+)^\top \end{bmatrix} \right\|_2 \\ & \leq (z_i^+)^\top z_j^+ - \begin{bmatrix} 1 & (z_i^+)^\top \end{bmatrix} \begin{bmatrix} 1 & (z_j^+)^\top \end{bmatrix}^\top \\ & = 1, \end{aligned} \quad (49)$$

where the inequality comes from the Cauchy Schwarz inequality, and the equality holds if $i = j$. These imply (c) and (d), which completes the proof of the sufficiency in (i).

(ii) Assume $(z_n)_{n=1}^N \in \mathcal{B}_R$, that is, for all $n \in [N]$, $\Delta_{\mathbb{L}^D}(z_0, z_n) \leq R$ is valid. Since $\mathcal{B}_R \subset \mathbb{L}^D$, conditions (a)-(d) holds true from the above discussion. Since $\Delta_{\mathbb{L}^D}(z_0, z_n) = \text{arcosh}(-\langle z_0, z_n \rangle_M) = \text{arcosh}(z_{0,n}) = \text{arcosh}\left(\sqrt{1 + (z_n^+)^\top z_n^+}\right)$, the followings are valid:

$$\begin{aligned} (z_n^-)^\top z_n^- &= (z_{0,n})^2 = \cosh^2 \Delta_{\mathbb{L}^D}(z_0, z_n), \\ (z_n^+)^\top z_n^+ &= 1 + \cosh^2 \Delta_{\mathbb{L}^D}(z_0, z_n) = \sinh^2 \Delta_{\mathbb{L}^D}(z_0, z_n). \end{aligned} \quad (50)$$

Therefore, we have

$$\begin{aligned} [\mathbf{L}^-]_{n,n} &= (\mathbf{z}_n^-)^\top \mathbf{z}_n^- \leq \cosh^2 R \\ [\mathbf{L}^+]_{n,n} &= (\mathbf{z}_n^+)^\top \mathbf{z}_n^+ \leq \sinh^2 R. \end{aligned} \quad (51)$$

For all $i, j \in [N]$,

$$\begin{aligned} [\mathbf{L}^-]_{i,j} &= (\mathbf{z}_i^-)^\top \mathbf{z}_j^- \\ &\leq \|\mathbf{z}_i^-\|_2 \|\mathbf{z}_j^-\|_2 \\ &\leq \max_{n=i,j} \left\{ (\mathbf{z}_n^-)^\top \mathbf{z}_n^- \right\} \\ &\leq \max_{n \in [N]} \left\{ (\mathbf{z}_n^-)^\top \mathbf{z}_n^- \right\} \\ &\leq \cosh^2 R \end{aligned} \quad (52)$$

is valid, where the first inequality is from the Cauchy Schwarz inequality. Thus, we have $\|\mathbf{L}^-\|_{\max} \leq \cosh^2 R$, and likewise, we also have $\|\mathbf{L}^+\|_{\max} \leq \sinh^2 R$, which imply condition (e). We have proved the sufficiency in (ii).

(iii) If $(\mathbf{z}_n)_{n=1}^N \in \mathcal{B}_R^\rho$, since $\mathcal{B}_R^\rho \subset \mathcal{B}_R$, conditions (a)-(e) follows from the above discussion. Let $\mathbf{Z}^- = \mathbf{U}^- \boldsymbol{\Sigma}^- (\mathbf{V}^-)^\top$ be the singular value decomposition of \mathbf{Z}^- , where $\mathbf{U}^- \in \mathbb{R}^{1,1}$, $\mathbf{V}^- \in \mathbb{R}^{N,N}$ are orthogonal and $\boldsymbol{\Sigma}^- \in \mathbb{R}^{1,N}$ is diagonal. The singular decomposition of \mathbf{L}^- is given by

$$\begin{aligned} \mathbf{L}^- &= \mathbf{V}^- (\boldsymbol{\Sigma}^-)^\top (\mathbf{U}^-)^\top \mathbf{U}^- \boldsymbol{\Sigma}^- (\mathbf{V}^-)^\top \\ &= \mathbf{V}^- (\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^- (\mathbf{V}^-)^\top, \end{aligned} \quad (53)$$

where the diagonal elements of $(\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^-$ indicate the singular values of \mathbf{L}^- . Hence, $\|\mathbf{L}^-\|_* = \text{Tr}((\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^-)$. As \mathbf{V}^- is orthogonal, we have

$$\begin{aligned} \text{Tr}((\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^-) &= \text{Tr}((\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^- (\mathbf{V}^-)^\top \mathbf{V}^-) \\ &= \text{Tr}(\mathbf{V}^- (\boldsymbol{\Sigma}^-)^\top \boldsymbol{\Sigma}^- (\mathbf{V}^-)^\top) \\ &= \text{Tr}(\mathbf{L}^-). \end{aligned} \quad (54)$$

Hence, we get

$$\begin{aligned} \|\mathbf{L}^-\|_* &= \text{Tr}(\mathbf{L}^-) \\ &= \sum_{n=1}^N (\mathbf{z}_n^-)^\top \mathbf{z}_n^- \\ &= \sum_{n=1}^N \cosh^2 \Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_n). \end{aligned} \quad (55)$$

Likewise, we have

$$\begin{aligned} \|\mathbf{L}^+\|_* &= \text{Tr}(\mathbf{L}^+) \\ &= \sum_{n=1}^N (\mathbf{z}_n^+)^\top \mathbf{z}_n^+ \\ &= \sum_{n=1}^N \sinh^2 \Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_n). \end{aligned} \quad (56)$$

By the definition of \mathcal{B}_R^ρ , we have $\|\mathbf{L}^+\|_* \leq N \sinh^2 \rho$ and $\|\mathbf{L}^-\|_* \leq N \cosh^2 \rho$, which imply condition (f). This completes the proof of the sufficiency in (iii).

(Necessity) (i) Assume that conditions (a)-(d) are satisfied.

Noting that $\mathbf{L}^- \succeq \mathbf{O}$, let $\mathbf{L}^- = \tilde{\mathbf{V}}^- \mathbf{T}^- (\tilde{\mathbf{V}}^-)^\top$ be a singular value decomposition of \mathbf{L}^- , where $\tilde{\mathbf{V}}^- \in \mathbb{R}^{N,N}$ is orthogonal and \mathbf{T}^- is diagonal, that is, $[\mathbf{T}^-]_{i,j} = 0$ if $i \neq j$.

Since $\text{rank } \mathbf{L}^- = 1$ and $\mathbf{L}^- \succeq \mathbf{O}$, we can assume that $[\mathbf{T}^-]_{1,1} > 0$ and $[\mathbf{T}^-]_{n,n} = 0$ for all $n = 2, 3, \dots, N$.

Therefore, $[\mathbf{L}^-]_{i,j} = [\tilde{\mathbf{V}}^-]_{i,1} [\mathbf{T}^-]_{1,1} [\tilde{\mathbf{V}}^-]_{i,1}$, for $i, j \in$

$[N]$. In particular, $[\mathbf{L}^-]_{1,1} = \left([\tilde{\mathbf{V}}^-]_{1,1} \right)^2 [\mathbf{T}^-]_{1,1}$. As

\mathbf{L}^+ is positive semi-definite, its diagonal entries are all non-negative. In particular, $[\mathbf{L}^+]_{1,1} \geq 0$. Since $[\mathbf{L}^+]_{1,1} - [\mathbf{L}^-]_{1,1} = -1$ from (c), we have $[\mathbf{L}^-]_{1,1} \geq 1$. Hence, we have $[\tilde{\mathbf{V}}^-]_{1,1} \neq 0$. Define \mathbf{V}^- by

$$\mathbf{V}^- := \begin{cases} \mathbf{V}^- & \text{if } [\tilde{\mathbf{V}}^-]_{1,1} > 0, \\ -\mathbf{V}^- & \text{if } [\tilde{\mathbf{V}}^-]_{1,1} < 0. \end{cases} \quad (57)$$

Then $[\mathbf{V}^-]_{1,1} > 0$ and $\mathbf{L}^- = \mathbf{V}^- \mathbf{T}^- (\mathbf{V}^-)^\top$ is valid. Let

$\mathbf{L}^+ = \mathbf{V}^+ \mathbf{T}^+ (\mathbf{V}^+)^\top$ be a singular value decomposition of \mathbf{L}^+ , where $\tilde{\mathbf{V}}^+ \in \mathbb{R}^{N,N}$ is orthogonal and \mathbf{T}^+ is diagonal.

Since $\text{rank } \mathbf{L}^+ \leq D$ and $\mathbf{L}^+ \succeq \mathbf{O}$, we can assume that $[\mathbf{T}^+]_{n,n} = 0$ for all $n = D+1, D+2, \dots, N$. Define $\mathbf{z}_n^- \in \mathbb{R}^1$ and $\mathbf{z}_n^+ \in \mathbb{R}^D$ by

$$\begin{aligned} \mathbf{z}_n^- &:= [z_{0,n}], \\ \mathbf{z}_n^+ &:= [z_{1,n} \quad z_{2,n} \quad \dots \quad z_{D,n}]^\top, \end{aligned} \quad (58)$$

respectively, where

$$z_{d,n} = \begin{cases} \sqrt{[\mathbf{T}^-]_{1,1} [\mathbf{V}^-]_{n,1}} & \text{if } d = 0, \\ \sqrt{[\mathbf{T}^+]_{d,d} [\mathbf{V}^+]_{n,d}} & \text{if } d = 1, 2, \dots, D. \end{cases} \quad (59)$$

respectively, and define $\mathbf{Z}^- \in \mathbb{R}^{1,N}$ and $\mathbf{Z}^+ \in \mathbb{R}^{D,N}$ by

$$\begin{aligned} \mathbf{Z}^- &= [\mathbf{z}_1^- \quad \mathbf{z}_2^- \quad \dots \quad \mathbf{z}_N^-], \\ \mathbf{Z}^+ &= [\mathbf{z}_1^+ \quad \mathbf{z}_2^+ \quad \dots \quad \mathbf{z}_N^+], \end{aligned} \quad (60)$$

respectively. Now, for $n \in [N]$, we define $\mathbf{z}_n \in \mathbb{R}^{1+D}$ by

$$\mathbf{z}_n = \begin{bmatrix} \mathbf{z}_n^- \\ \mathbf{z}_n^+ \end{bmatrix}. \quad (61)$$

The Lorentz Gramian of $(\mathbf{z}_n)_{n=1}^N$ is given by

$$\begin{aligned} (\mathbf{Z}^+)^{\top} \mathbf{Z}^+ - (\mathbf{Z}^-)^{\top} \mathbf{Z}^- &= \mathbf{V}^+ \mathbf{T}^+ (\mathbf{V}^+)^{\top} - \mathbf{V}^- \mathbf{T}^- (\mathbf{V}^-)^{\top} \\ &= \mathbf{L}^+ - \mathbf{L}^-. \end{aligned} \quad (62)$$

In the following, we prove $\mathbf{z}_n \in \mathbb{L}^D$ for all $n \in [N]$. Since $\langle \mathbf{z}_n, \mathbf{z}_n \rangle_{\mathbb{M}} = [\mathbf{L}]_{n,n} = [\mathbf{L}^+ - \mathbf{L}^-]_{n,n} = -1$, it is sufficient to prove that $z_{0,n} > 0$. From $[\mathbf{V}^-]_{1,1} > 0$ and $z_{0,n} = \sqrt{[\mathbf{T}^-]_{1,1}} [\mathbf{V}^-]_{n,1}$, we have $z_{0,1} > 0$. For general $n \in [N]$, the following is valid:

$$\begin{aligned} |z_{0,1}| |z_{0,n}| - z_{0,1} z_{0,n} &> \|\mathbf{z}_1^+\|_2 \|\mathbf{z}_n^+\|_2 - z_{0,1} z_{0,n} \\ &\geq (\mathbf{z}_1^+)^{\top} \mathbf{z}_n^+ - z_{0,1} z_{0,n} \\ &= -\langle \mathbf{z}_1, \mathbf{z}_n \rangle_{\mathbb{M}} \\ &= [\mathbf{L}]_{1,n} \\ &= [\mathbf{L}^+ - \mathbf{L}^-]_{1,n} \\ &\geq 1 \\ &> 0. \end{aligned} \quad (63)$$

Therefore, $z_{0,1}$ and $z_{0,n}$ must have the same sign. Hence, $z_{0,n} > 0$, which completes the proof of the necessity in (i).

(ii) Assume that conditions **(a)-(e)** are satisfied. Define $(\mathbf{z}_n)_{n=1}^N$ as in (58). Then, since **(a)-(d)** are satisfied, $(\mathbf{z}_n)_{n=1}^N \in (\mathbb{L}^D)^N$ and its Lorentz Gramian is $\mathbf{L} = \mathbf{L}^+ - \mathbf{L}^-$. Thus, it suffices to show that condition **(e)** implies that $\Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_N) \leq R$ is valid for all $n \in N$. Since $[\mathbf{L}^-]_{n,n} = (z_{0,n})^2 = \cosh^2 \Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_N)$, condition **(e)** implies $\cosh^2 \Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_N) \leq \cosh^2 R$, which is equivalent to $\Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_N) \leq R$. Hence, we have $(\mathbf{z}_n)_{n=1}^N \in \mathcal{B}_R$.

(iii) Assume that conditions **(a)-(f)** are satisfied. Define $(\mathbf{z}_n)_{n=1}^N$ as in (58). Then, since **(a)-(e)** are satisfied, $(\mathbf{z}_n)_{n=1}^N \in \mathcal{B}_R$ and its Lorentz Gramian is $\mathbf{L} = \mathbf{L}^+ - \mathbf{L}^-$. Also, condition **(f)** implies $\frac{1}{N} \sum_{n=1}^N \cosh^2 \Delta_{\mathbb{L}^D}(\mathbf{z}_0, \mathbf{z}_n) \leq \cosh^2 \rho$, since the left hand side is given by $\frac{1}{N} \|\mathbf{L}^-\|_*$. \square

Remark 3. The statement of Lemma 9 (i) is equivalent to that of Proposition 1 in (Tabaghi & Dokmanic, 2020). However, the proof there does not consider the case where the representations given by the decomposition of the Lorentz Gramian are all in $-\mathbb{L}^D$ instead of \mathbb{L}^D . The above construction of representations solves this problem by forcing $z_{0,n}$ to be positive.

F. Rademacher Complexity and Proof of Theorem 1

First, we define Rademacher complexity (Koltchinskii, 2001; Koltchinskii & Panchenko, 2000; Bartlett et al., 2002), the key tool to obtain HOE's generalization error bound. Let \mathcal{T} be our input space and $\mathcal{H} \subset \{h|h: \mathcal{T} \rightarrow \mathbb{R}\}$ be our hypothesis space. Let $S \in \mathbb{Z}_{\geq 0}$ be the number of data points, and suppose that data points $(t_1, y_1), (t_2, y_2), \dots, (t_S, y_S) \in \mathcal{T} \times \{-1, +1\}$ are independently distributed according to some unknown fixed distribution μ . The Rademacher complexity of h is defined as follows:

Definition 3. Let $\sigma_1, \sigma_2, \dots, \sigma_S$ be random values such that $\sigma_1, \sigma_2, \dots, \sigma_S, (t_1, y_1), (t_2, y_2), \dots, (t_S, y_S)$ are mutually independent and each of $\sigma_1, \sigma_2, \dots, \sigma_S$ takes values $\{-1, +1\}$ with equal probability. The Rademacher complexity $\mathfrak{R}_S(\mathcal{H})$ is defined by

$$\mathfrak{R}_S(\mathcal{H}) := \mathbb{E}_{(t_s)_{s=1}^S} \mathbb{E}_{(\sigma_s)_{s=1}^S} \left[\frac{1}{S} \sup_{h \in \mathcal{H}} \sum_{s=1}^S \sigma_s h(t_s) \right]. \quad (64)$$

We use the following theorem provided by Bartlett & Mendelson (2002) and arranged by Kakade et al. (2008).

Theorem 10 ((Bartlett & Mendelson, 2002; Kakade et al., 2008)). *Let $\phi: \mathcal{T} \times \{-1, +1\} \rightarrow \mathbb{R}$ be a loss function. Define the empirical risk function $\hat{\mathcal{R}}_S(h)$ and expected risk function $\mathcal{R}(h)$ by*

$$\begin{aligned} \hat{\mathcal{R}}_S(h) &:= \frac{1}{S} \sum_{s=1}^S \phi(h(t_s), y_s), \\ \mathcal{R}(h) &:= \mathbb{E}_{t,y} \phi(h(t), y). \end{aligned} \quad (65)$$

Assume that $\phi(\cdot, -1)$ and $\phi(\cdot, +1)$ are L_ϕ -Lipschitz and bounded. Define

$$c_\phi := \sup_{\substack{t \in \mathcal{T}, \\ y \in \{-1, +1\}, \\ h \in \mathcal{H}}} \phi(h(t), y) - \inf_{\substack{t \in \mathcal{T}, \\ y \in \{-1, +1\}, \\ h \in \mathcal{H}}} \phi(h(t), y). \quad (66)$$

Then for any $\delta \in \mathbb{R}_{>0}$ and with probability at least $1 - \delta$ simultaneously for all $h \in \mathcal{H}$ we have that

$$\mathcal{R}(h) - \hat{\mathcal{R}}_S(h) \leq 2\mathfrak{R}_S(\mathcal{H}) + c_\phi \sqrt{\frac{\ln(1/\delta)}{2S}}. \quad (67)$$

From this theorem, we can easily derive an upper bound for the excess risk of the empirical risk minimizer as follows.

Corollary 11. *Define the empirical risk minimizer $\hat{h} \in \mathcal{H}$ and expected loss minimizer by $h^* \in \mathcal{H}$ by*

$$\hat{h} := \operatorname{argmin}_{h \in \mathcal{H}} \hat{\mathcal{R}}_S(h), \quad h^* := \operatorname{argmin}_{h \in \mathcal{H}} \mathcal{R}(h), \quad (68)$$

and we call $\mathcal{R}(\hat{h}) - \mathcal{R}(h^*)$ the excess risk of \hat{h} . Then for any $\delta \in \mathbb{R}_{>0}$ and with probability at least $1 - \delta$ we have that

$$\mathcal{R}(\hat{h}) - \mathcal{R}(h^*) \leq 2\mathfrak{R}_S(\mathcal{H}) + 2c_\phi \sqrt{\frac{\ln(2/\delta)}{2S}}. \quad (69)$$

Proof. We have that

$$\begin{aligned} & \mathcal{R}(\hat{h}) - \mathcal{R}(h^*) \\ &= \left(\mathcal{R}(\hat{h}) - \hat{\mathcal{R}}_S(\hat{h}) \right) + \left(\hat{\mathcal{R}}_S(\hat{h}) - \hat{\mathcal{R}}_S(h^*) \right) \\ & \quad + \left(\hat{\mathcal{R}}_S(h^*) - \mathcal{R}(h^*) \right) \\ & \leq \left(\mathcal{R}(\hat{h}) - \hat{\mathcal{R}}_S(\hat{h}) \right) + \left(\hat{\mathcal{R}}_S(h^*) - \mathcal{R}(h^*) \right), \end{aligned} \quad (70)$$

where the last inequality holds from the definition of \hat{h} . We complete the proof by evaluating the first and second term by Theorem 10 and Hoeffding's inequality, respectively. \square

Therefore, it suffices to obtain the upper bound of HOE model's Rademacher complexity, which is given below. Let $\mathcal{B} \subset \mathbb{L}^D$. We define a hypothesis function class $h(\cdot; \mathcal{B})$ by

$$h(\cdot; \mathcal{B}) := \left\{ h(\cdot; (\mathbf{z}_n)_{n=1}^N) \mid (\mathbf{z}_n)_{n=1}^N \in \mathcal{B} \right\}. \quad (71)$$

We evaluate the Rademacher complexity of the hypothesis function class $h(\cdot; \mathcal{B}^\rho)$ defined by

$$\begin{aligned} & \mathfrak{R}_S(h(\cdot; \mathcal{B}^\rho)) \\ &:= \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{(\mathbf{z}_n)_{n=1}^N \in \mathcal{B}^\rho} \frac{1}{S} \sum_{s=1}^S \sigma_s h(i_s, j_s, k_s; (\mathbf{z}_n)_{n=1}^N) \right]. \end{aligned} \quad (72)$$

The following evaluates the complexity.

Lemma 12. *The Rademacher complexity of the hypothesis function class $h(\cdot; \mathcal{B}^\rho)$ satisfies the following inequality:*

$$\begin{aligned} & \mathfrak{R}_S(h(\cdot; \mathcal{B}^\rho)) \\ & \leq (\cosh^2 \rho + \sinh^2 \rho) \left(\sqrt{\frac{2N \ln N}{S}} + \frac{N \ln N}{6S} \right). \end{aligned} \quad (73)$$

Proof. Define $\mathcal{G}_R^-, \mathcal{G}_R^+, \bar{\mathcal{G}}_R^-, \bar{\mathcal{G}}_R^+ \subset \mathbb{S}^{N,N}$ by

$$\begin{aligned} \mathcal{G}^\rho &:= \{ \mathbf{L}^+ - \mathbf{L}^- \mid \mathbf{L}^-, \mathbf{L}^+ \in \mathbb{S}^{N,N} \text{ satisfy (a)-(d), (f)} \}, \\ \bar{\mathcal{G}}^\rho &:= \{ \mathbf{L}^+ - \mathbf{L}^- \mid \mathbf{L}^- \in \mathcal{Q}^{N \cosh^2 \rho}, \mathbf{L}^+ \in \mathcal{Q}^{N \sinh^2 \rho} \}, \end{aligned} \quad (74)$$

where (a)-(f) are the conditions defined in Lemma 9, and for $\lambda \in \mathbb{R}_{\geq 0}$, \mathcal{Q}^λ is defined by

$$\mathcal{Q}^\lambda := \{ \mathbf{L} \in \mathbb{S}^{N,N} \mid \mathbf{L} \succeq \mathbf{O}, \|\mathbf{L}\|_* \leq \lambda \}. \quad (75)$$

Note that $\mathcal{G}_R^\rho \subset \bar{\mathcal{G}}^\rho$. Let $\sigma_1, \sigma_2, \dots, \sigma_S$ be i.i.d Rademacher random variables and $\boldsymbol{\sigma} = [\sigma_1 \ \sigma_2 \ \dots \ \sigma_S]^\top$. The Rademacher complexity is calculated as follows:

$$\begin{aligned} & \mathfrak{R}_S(h(\cdot; \mathcal{B}^\rho)) \\ &:= \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{(\mathbf{z}_n)_{n=1}^N \in \mathcal{B}^\rho} \frac{1}{S} \sum_{s=1}^S \sigma_s h(i_s, j_s, k_s; (\mathbf{z}_n)_{n=1}^N) \right] \\ &= \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\mathbf{L} \in \mathcal{G}^\rho} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \right] \\ &\leq \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\mathbf{L} \in \bar{\mathcal{G}}^\rho} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \right], \end{aligned} \quad (76)$$

where the last inequality results from $\mathcal{G}_R^\rho \subset \bar{\mathcal{G}}^\rho$. We can decompose the integrand of the above expectation operator as follows:

$$\begin{aligned} & \sup_{\mathbf{L} \in \bar{\mathcal{G}}^\rho} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \\ &= \sup_{\substack{\mathbf{L}^- \in \mathcal{Q}^{N \sinh^2 \rho}, \\ \mathbf{L}^+ \in \mathcal{Q}^{N \cosh^2 \rho}}} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}^+ - \mathbf{L}^-, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \\ &\leq \sup_{\mathbf{L}^+ \in \mathcal{Q}^{N \sinh^2 \rho}} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}^+, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \\ & \quad + \sup_{\mathbf{L}^- \in \mathcal{Q}^{N \cosh^2 \rho}} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}^-, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}}. \end{aligned} \quad (77)$$

Hence, we have

$$\begin{aligned} & \mathfrak{R}_S(h(\cdot; \mathcal{B}^\rho)) \\ & \leq \mathfrak{R}_S^{\mathcal{G}}(\langle \mathcal{Q}^{N \sinh^2 \rho}, \cdot \rangle_{\mathbb{F}}) + \mathfrak{R}_S^{\mathcal{G}}(\langle \mathcal{Q}^{N \cosh^2 \rho}, \cdot \rangle_{\mathbb{F}}), \end{aligned} \quad (78)$$

where $\mathfrak{R}_S^{\mathcal{G}}(\langle \mathcal{Q}^\lambda, \cdot \rangle_{\mathbb{F}})$ is defined as

$$\mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\mathbf{L} \in \mathcal{Q}^\lambda} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \mathbf{L}, \mathbf{T}_{i_s, j_s, k_s} \rangle_{\mathbb{F}} \right]. \quad (79)$$

We can bound $\mathfrak{R}_S^{\mathcal{G}}(\langle \mathcal{Q}^\lambda, \cdot \rangle_{\mathbb{F}})$ by the following lemma.

Lemma 13.

$$\mathfrak{R}_S^{\mathcal{G}}(\langle \mathcal{Q}^\lambda, \cdot \rangle_{\mathbb{F}}) \leq \frac{\lambda}{N} \left(\sqrt{\frac{2(N+1) \ln N}{S}} + \frac{N \ln N}{\sqrt{12}S} \right). \quad (80)$$

We prove Lemma 13 later. By Lemma 13, we obtain

$$\begin{aligned} & \mathfrak{R}_S(h(\cdot; \mathcal{B}^\rho)) \\ & \leq (\cosh^2 \rho + \sinh^2 \rho) \left(\sqrt{\frac{2(N+1) \ln N}{S}} + \frac{N \ln N}{\sqrt{12S}} \right), \end{aligned} \quad (81)$$

which completes the proof. \square

Proof of Lemma 13. For $\lambda \in \mathbb{R}_{\geq 0}$, define \mathcal{Q}_1^λ by

$$\{\lambda \mathbf{u} \mathbf{u}^\top \mid \mathbf{u} \in \mathbb{R}^N, \|\mathbf{u}\|_2 = 1\}. \quad (82)$$

Since \mathcal{Q}_1^λ is the convex hull of \mathcal{Q}^λ and the Rademacher complexity of a function class equals that of its convex hull (Bartlett & Mendelson, 2002, Theorem 12-2), we have that

$$\begin{aligned} & \mathfrak{R}_S^G(\langle \mathcal{Q}^\lambda, \cdot \rangle_F) \\ & = \mathfrak{R}_S^G(\langle \mathcal{Q}_1^\lambda, \cdot \rangle_F) \\ & = \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\|\mathbf{u}\|_2 \leq 1} \frac{1}{S} \sum_{s=1}^S \sigma_s \langle \lambda \mathbf{u} \mathbf{u}^\top, \mathbf{T}_{i_s, j_s, k_s} \rangle_F \right] \\ & = \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\|\mathbf{u}\|_2 \leq 1} \frac{1}{S} \sum_{s=1}^S \sigma_s \lambda \text{Tr}(\mathbf{u} \mathbf{u}^\top \mathbf{T}_{i_s, j_s, k_s}) \right] \\ & = \frac{\lambda}{S} \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\|\mathbf{u}\|_2 \leq 1} \sum_{s=1}^S \sigma_s \text{Tr}(\mathbf{u}^\top \mathbf{T}_{i_s, j_s, k_s} \mathbf{u}) \right] \\ & = \frac{\lambda}{S} \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\sup_{\|\mathbf{u}\|_2 \leq 1} \text{Tr} \left(\mathbf{u}^\top \left(\sum_{s=1}^S \sigma_s \mathbf{T}_{i_s, j_s, k_s} \right) \mathbf{u} \right) \right] \\ & = \frac{\lambda}{S} \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\left\| \sum_{s=1}^S \sigma_s \mathbf{T}_{i_s, j_s, k_s} \right\|_{\text{op},2} \right], \end{aligned} \quad (83)$$

where $\|\cdot\|_{\text{op},2}$ denotes the operator norm with respect to the 2-norm defined by

$$\|\mathbf{A}\|_{\text{op},2} := \max_{\|\mathbf{u}\|_2 \leq 1} \|\mathbf{A}\mathbf{u}\|_2 \quad (84)$$

To evaluate this, we can apply the following the matrix Bernstein inequality.

Theorem 14 ((Tropp, 2015) Theorem 6.6.1). *Let $\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_S \in \mathbb{S}^{N,N}$ be independent random matrices that satisfies*

$$\mathbb{E} \mathbf{A}_s = \mathbf{O}, \quad \|\mathbf{A}_s\|_{\text{op},2} \leq \sigma. \quad (85)$$

Then

$$\mathbb{E} \left\| \sum_{s=1}^S \mathbf{A}_s \right\|_{\text{op},2} \leq \sqrt{2v \left(\sum_{s=1}^S \mathbf{A}_s \right) \ln N} + \frac{1}{3} \sigma \ln N, \quad (86)$$

where v is the matrix variance statistics defined by

$$v(\mathbf{A}) := \|\mathbb{E} \mathbf{A}^2\|_{\text{op},2}. \quad (87)$$

Note that $v \left(\sum_{s=1}^S \mathbf{A}_s \right) = \left\| \sum_{s=1}^S \mathbb{E} \mathbf{A}_s^2 \right\|_{\text{op},2}$ is valid since

$\mathbf{A}_1, \mathbf{A}_2, \dots, \mathbf{A}_S \in \mathbb{S}^{N,N}$ are independent. We apply Theorem 14 to the right hand side of (83) by substituting \mathbf{A}_s by $\sigma_s \mathbf{T}_{i_s, j_s, k_s}$. Here, $\mathbb{E} \sigma_s \mathbf{T}_{i_s, j_s, k_s} = \mathbf{O}$ is valid because $\mathbb{E} \sigma_s = 0$. The singular values of $\sigma_s \mathbf{T}_{i_s, j_s, k_s}$ is equal to those of $\sigma_s \tilde{\mathbf{T}}$, where

$$\tilde{\mathbf{T}} := \begin{bmatrix} 0 & -\frac{1}{2} & +\frac{1}{2} \\ -\frac{1}{2} & 0 & 0 \\ +\frac{1}{2} & 0 & 0 \end{bmatrix}. \quad (88)$$

Since

$$\left(\sigma_s \tilde{\mathbf{T}} \right)^2 = \left(\sigma_s \tilde{\mathbf{T}} \right)^\top \left(\sigma_s \tilde{\mathbf{T}} \right) = \begin{bmatrix} +\frac{1}{2} & 0 & 0 \\ 0 & +\frac{1}{4} & -\frac{1}{4} \\ 0 & -\frac{1}{4} & +\frac{1}{4} \end{bmatrix}, \quad (89)$$

and its eigenvalues are $0, +\frac{1}{2}, +\frac{1}{2}$, the singular values of $\tilde{\mathbf{T}}$ are $0, +\frac{1}{\sqrt{2}}, +\frac{1}{\sqrt{2}}$. Hence, we have that $\|\sigma_s \mathbf{T}_{i_s, j_s, k_s}\|_{\text{op},2} \leq \frac{1}{\sqrt{2}}$. Lastly, we evaluate $\left\| \sum_{s=1}^S \mathbb{E} (\sigma_s \mathbf{T}_{i_s, j_s, k_s})^2 \right\|_{\text{op},2} = S \|\mathbb{E} \mathbf{T}_{i_s, j_s, k_s}^2\|_{\text{op},2}$. In general, since $\|\cdot\|_{\text{op},2}$ is a convex function, we have that $\|\mathbb{E} \mathbf{T}_{i_s, j_s, k_s}^2\|_{\text{op},2} \leq \mathbb{E} \|\mathbf{T}_{i_s, j_s, k_s}^2\|_{\text{op},2} = \frac{1}{2}$.

The diagonal elements and off-diagonal elements of $\mathbb{E} \mathbf{T}_{i_s, j_s, k_s}^2$ are all $\frac{1}{N}$ and all $-\frac{1}{2} \frac{1}{N(N-1)}$, respectively, because the sum of the diagonal elements and that of the off-diagonal elements in $\mathbf{T}_{i_s, j_s, k_s}$ are always 1 and $-\frac{1}{2}$, respectively, and from the symmetry among the N diagonal elements and $N(N-1)$ off-diagonal elements. Hence, we have $\mathbb{E} \mathbf{T}_{i_s, j_s, k_s}^2 = \left(\frac{1}{N} + \frac{1}{2} \frac{1}{N(N-1)} \right) \mathbf{I} - \frac{1}{2} \frac{1}{N(N-1)} \mathbf{1} \mathbf{1}^\top$, whose eigenvalues (singular values) are $\frac{1}{N} + \frac{1}{2} \frac{1}{N(N-1)}$ (multiplicity $N-1$) and $\frac{1}{N}$ (multiplicity 1). Thus, we have

$$\begin{aligned} v \left(\sum_{s=1}^S \sigma_s \mathbf{T}_{i_s, j_s, k_s} \right) & = S \|\mathbb{E} \mathbf{T}_{i_s, j_s, k_s}^2\|_{\text{op},2} \\ & = S \left(\frac{1}{N} + \frac{1}{2} \frac{1}{N(N-1)} \right) \\ & = \frac{S}{2} \left(\frac{1}{N-1} + \frac{1}{N} \right) \\ & = \frac{S}{2} \frac{1}{N^2} \left(\frac{N^2}{N-1} + N \right) \\ & \leq \frac{S}{N^2} (N+1) \end{aligned} \quad (90)$$

. Therefore we have

$$\begin{aligned} & \mathbb{E}_{(i,j,k)} \mathbb{E}_\sigma \left[\left\| \sum_{s=1}^S \sigma_s \mathbf{T}_{i_s, j_s, k_s} \right\|_{\text{op},2} \right] \\ & \leq \frac{1}{N} \sqrt{2S(N+1) \ln N} + \frac{1}{3} \cdot \frac{1}{\sqrt{2}} \ln N, \end{aligned} \quad (91)$$

which completes the proof. \square

Proof of Theorem 1. We complete the proof by applying Corollary 11 to Lemma 12. \square