

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

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

We assume that $\mu, \xi > 0$ holds as in the discussion in Section 2.1. Let $\epsilon := \frac{1}{3}\xi$. Let $\nu, \eta_{\max}$ 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 $\mathbf{Z}^- = \mathbf{U}^- \mathbf{\Sigma}^- (\mathbf{V}^-)^\top$ 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 $\mathbf{\Sigma}^- \in \mathbb{R}^{1,N}$ is diagonal. The singular decomposition of $\mathbf{L}^-$ is given by

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

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

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

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 $\mathcal{B}_R^\rho$, 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, $[\mathbf{L}^-]_{i,j} = [\tilde{\mathbf{V}}^-]_{i,1} [\mathbf{T}^-]_{1,1} [\tilde{\mathbf{V}}^-]_{i,1}^\top$, for $i, j \in$

$[N]$. In particular, $[\mathbf{L}^-]_{1,1} = ([\tilde{\mathbf{V}}^-]_{1,1})^2 [\mathbf{T}^-]_{1,1}$. 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 $z_n \in \mathbb{R}^{1+D}$ by

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

The Lorentz Gramian of $(z_n)_{n=1}^N$ is given by

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

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

$$\begin{aligned} |z_{0,1}| |z_{0,n}| - z_{0,1} z_{0,n} &> \|z_1^+\|_2 \|z_n^+\|_2 - z_{0,1} z_{0,n} \\ &\geq (z_1^+)^T z_n^+ - z_{0,1} z_{0,n} \\ &= -\langle z_1, z_n \rangle_M \\ &= [L]_{1,n} \\ &= [L^+ - L^-]_{1,n} \\ &\geq 1 \\ &> 0. \end{aligned} \quad (63)$$

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 $(z_n)_{n=1}^N$ as in (58). Then, since (a)-(d) are satisfied, $(z_n)_{n=1}^N \in (\mathbb{L}^D)^N$ and its Lorentz Gramian is $L = L^+ - L^-$. Thus, it suffices to show that condition (e) implies that $\Delta_{\mathbb{L}^D}(z_0, z_N) \leq R$ is valid for all $n \in N$. Since $[L^-]_{n,n} = (z_{0,n})^2 = \cosh^2 \Delta_{\mathbb{L}^D}(z_0, z_N)$, condition (e) implies $\cosh^2 \Delta_{\mathbb{L}^D}(z_0, z_N) \leq \cosh^2 R$, which is equivalent to $\Delta_{\mathbb{L}^D}(z_0, z_N) \leq R$. Hence, we have $(z_n)_{n=1}^N \in \mathcal{B}_R$.

(iii) Assume that conditions (a)-(f) are satisfied. Define $(z_n)_{n=1}^N$ as in (58). Then, since (a)-(e) are satisfied, $(z_n)_{n=1}^N \in \mathcal{B}_R$ and its Lorentz Gramian is $L = L^+ - L^-$. Also, condition (f) implies $\frac{1}{N} \sum_{n=1}^N \cosh^2 \Delta_{\mathbb{L}^D}(z_0, z_n) \leq \cosh^2 \rho$, since the left hand side is given by $\frac{1}{N} \|L^-\|_*$. $\square$

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

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

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

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

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

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

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 $\lambda \in \mathbb{R}_{\geq 0}$, $\mathcal{Q}^\lambda$ 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:

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

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:

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

Hence, we have

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

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

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

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

Lemma 13.

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

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 $\mathcal{Q}_1^\lambda$ by

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

Since $\mathcal{Q}_1^\lambda$ is the convex hull of $\mathcal{Q}^\lambda$ 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. $\square$