## SUPPLEMENTARY MATERIAL FOR:

NON-ASYMPTOTIC RATES FOR MANIFOLD, TANGENT SPACE AND CURVATURE ESTIMATION

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## Appendix A: Properties and Stability of the Models

## A.1. Property of the Exponential Map in $\mathcal{C}_{\tau_{\text {min }}}^{2}$

Here we show the following Lemma 1, reproduced as Lemma A.1.
Lemma A.1. If $M \in \mathcal{C}_{\tau_{m i n}}^{2}, \exp _{p}: \mathcal{B}_{T_{p} M}\left(0, \tau_{\min } / 4\right) \rightarrow M$ is one-to-one. Moreover, it can be written as

$$
\begin{aligned}
\exp _{p}: \mathcal{B}_{T_{p} M}\left(0, \tau_{\min } / 4\right) & \longrightarrow M \\
v & \longmapsto p+v+\mathbf{N}_{p}(v)
\end{aligned}
$$

with $\mathbf{N}_{p}$ such that for all $v \in \mathcal{B}_{T_{p} M}\left(0, \tau_{\text {min }} / 4\right)$,

$$
\mathbf{N}_{p}(0)=0, \quad d_{0} \mathbf{N}_{p}=0, \quad\left\|d_{v} \mathbf{N}_{p}\right\|_{o p} \leq L_{\perp}\|v\|
$$

where $L_{\perp}=5 /\left(4 \tau_{\text {min }}\right)$. Furthermore, for all $p, y \in M$,

$$
y-p=\pi_{T_{p} M}(y-p)+R_{2}(y-p),
$$

where $\left\|R_{2}(y-p)\right\| \leq \frac{\|y-p\|^{2}}{2 \tau_{\min }}$.
Proof of Lemma A.1. Proposition 6.1 in [13] states that for all $x \in M$, $\left\|I I_{x}^{M}\right\|_{o p} \leq 1 / \tau_{\min }$. In particular, Gauss equation ([8, Proposition 3.1 (a), p.135]) yields that the sectional curvatures of $M$ satisfy $-2 / \tau_{\text {min }}^{2} \leq \kappa \leq$ $1 / \tau_{\text {min }}^{2}$. Using Corollary 1.4 of [3], we get that the injectivity radius of $M$ is at least $\pi \tau_{\text {min }} \geq \tau_{\text {min }} / 4$. Therefore, $\exp _{p}: \mathcal{B}_{T_{p} M}\left(0, \tau_{\text {min }} / 4\right) \rightarrow M$ is one-toone.

Let us write $\mathbf{N}_{p}(v)=\exp _{p}(v)-p-v$. We clearly have $\mathbf{N}_{p}(0)=0$ and $d_{0} \mathbf{N}_{p}=0$. Let now $v \in \mathcal{B}_{T_{p} M}\left(0, \tau_{\text {min }} / 4\right)$ be fixed. We have $d_{v} \mathbf{N}_{p}=$

[^0]$d_{v} \exp _{p}-I d_{T_{p} M}$. For $0 \leq t \leq\|v\|$, we write $\gamma(t)=\exp _{p}(t v /\|v\|)$ for the arc-length parametrized geodesic from $p$ to $\exp _{p}(v)$, and $P_{t}$ for the parallel translation along $\gamma$. From Lemma 18 of [9],
$$
\left\|d_{t} \frac{v}{\|v\|} \exp _{p}-P_{t}\right\|_{o p} \leq \frac{2}{\tau_{\min }^{2}} \frac{t^{2}}{2} \leq \frac{t}{4 \tau_{\min }}
$$

We now derive an upper bound for $\left\|P_{t}-I d_{T_{p} M}\right\|_{o p}$. For this, fix two unit vectors $u \in \mathbb{R}^{D}$ and $w \in T_{p} M$, and write $g(t)=\left\langle P_{t}(w)-w, u\right\rangle$. Letting $\bar{\nabla}$ denote the ambient derivative in $\mathbb{R}^{D}$, by definition of parallel translation,

$$
\begin{aligned}
\left|g^{\prime}(t)\right| & =\left|\left\langle\bar{\nabla}_{\gamma^{\prime}(t)} P_{t}(w)-w, u\right\rangle\right| \\
& =\left|\left\langle I I_{\gamma(t)}^{M}\left(\gamma^{\prime}(t), P_{t}(w)\right), u\right\rangle\right| \\
& \leq 1 / \tau_{\text {min }} .
\end{aligned}
$$

Since $g(0)=0$, we get $\left\|P_{t}-I d_{T_{p} M}\right\|_{o p} \leq t / \tau_{\text {min }}$. Finally, the triangle inequality leads to

$$
\begin{aligned}
\left\|d_{v} \mathbf{N}_{p}\right\|_{o p} & =\left\|d_{v} \exp -I d_{T_{p} M}\right\|_{o p} \\
& \leq\left\|d_{v} \exp -P_{\|v\|}\right\|_{o p}+\left\|P_{\|v\|}-I d_{T_{p} M}\right\|_{o p} \\
& \leq \frac{5\|v\|}{4 \tau_{\min }}
\end{aligned}
$$

We conclude with the property of the projection $\pi^{*}=\pi_{T_{p} M}$. Indeed, defining $R_{2}(y-p)=(y-p)-\pi^{*}(y-p)$, Lemma 4.7 in [10] gives

$$
\begin{aligned}
\left\|R_{2}(y-p)\right\| & =d\left(y-p, T_{p} M\right) \\
& \leq \frac{\|y-p\|^{2}}{2 \tau_{\min }}
\end{aligned}
$$

## A.2. Geometric Properties of the Models $\mathcal{C}^{k}$

Lemma A.2. For any $M \in \mathcal{C}_{\tau_{m i n}, \mathbf{L}}^{k}$ and $x \in M$, the following holds.
(i) For all $v_{1}, v_{2} \in \mathcal{B}_{T_{x} M}\left(0, \frac{1}{4 L_{\perp}}\right)$,

$$
\frac{3}{4}\left\|v_{2}-v_{1}\right\| \leq\left\|\Psi_{x}\left(v_{2}\right)-\Psi_{x}\left(v_{1}\right)\right\| \leq \frac{5}{4}\left\|v_{2}-v_{1}\right\|
$$

(ii) For all $h \leq \frac{1}{4 L_{\perp}} \wedge \frac{2 \tau_{\text {min }}}{5}$,

$$
M \cap \mathcal{B}\left(x, \frac{3 h}{5}\right) \subset \Psi_{x}\left(\mathcal{B}_{T_{x} M}(x, h)\right) \subset M \cap \mathcal{B}\left(x, \frac{5 h}{4}\right) .
$$

(iii) For all $h \leq \frac{\tau_{\text {min }}}{2}$,

$$
\mathcal{B}_{T_{x} M}\left(0, \frac{7 h}{8}\right) \subset \pi_{T_{x} M}(\mathcal{B}(x, h) \cap M) .
$$

(iv) Denoting by $\pi^{*}=\pi_{T_{x} M}$ the orthogonal projection onto $T_{x} M$, for all $x \in M$, there exist multilinear maps $T_{2}^{*}, \ldots, T_{k-1}^{*}$ from $T_{x} M$ to $\mathbb{R}^{D}$, and $R_{k}$ such that for all $y \in \mathcal{B}\left(x, \frac{\tau_{\min } \wedge L_{\perp}^{-1}}{4}\right) \cap M$,

$$
\begin{aligned}
y-x= & \pi^{*}(y-x)+T_{2}^{*}\left(\pi^{*}(y-x)^{\otimes 2}\right)+\ldots+T_{k-1}^{*}\left(\pi^{*}(y-x)^{\otimes k-1}\right) \\
& +R_{k}(y-x),
\end{aligned}
$$

with

$$
\left\|R_{k}(y-x)\right\| \leq C\|y-x\|^{k} \quad \text { and } \quad\left\|T_{i}^{*}\right\|_{o p} \leq L_{i}^{\prime}, \text { for } 2 \leq i \leq k-1,
$$

where $L_{i}^{\prime}$ depends on $d, k, \tau_{\text {min }}, L_{\perp}, \ldots, L_{i}$, and $C$ on $d, k, \tau_{\text {min }}, L_{\perp}$, $\ldots, L_{k}$. Moreover, for $k \geq 3, T_{2}^{*}=I I_{x}^{M}$.
(v) For all $x \in M,\left\|I I_{x}^{M}\right\|_{o p} \leq 1 / \tau_{\text {min }}$. In particular, the sectional curvatures of $M$ satisfy

$$
\frac{-2}{\tau_{\min }^{2}} \leq \kappa \leq \frac{1}{\tau_{\min }^{2}}
$$

Proof of Lemma A.2. (i) Simply notice that from the reverse triangle inequality,

$$
\left|\frac{\left\|\Psi_{x}\left(v_{2}\right)-\Psi_{x}\left(v_{1}\right)\right\|}{\left\|v_{2}-v_{1}\right\|}-1\right| \leq \frac{\left\|N_{x}\left(v_{2}\right)-N_{x}\left(v_{1}\right)\right\|}{\left\|v_{2}-v_{1}\right\|} \leq L_{\perp}\left(\left\|v_{1}\right\| \vee\left\|v_{2}\right\|\right) \leq \frac{1}{4} .
$$

(ii) The right-hand side inclusion follows straightforwardly from (i). Let us focus on the left-hand side inclusion. For this, consider the map defined by $G=\pi_{T_{x} M} \circ \Psi_{x}$ on the domain $\mathcal{B}_{T_{x} M}(0, h)$. For all $v \in \mathcal{B}_{T_{x} M}(0, h)$, we have

$$
\left\|d_{v} G-I d_{T_{x} M}\right\|_{o p}=\left\|\pi_{T_{x} M} \circ d_{v} \mathbf{N}_{x}\right\|_{o p} \leq\left\|d_{v} \mathbf{N}_{x}\right\|_{o p} \leq L_{\perp}\|v\| \leq \frac{1}{4}<1
$$

Hence, $G$ is a diffeomorphism onto its image and it satisfies $\|G(v)\| \geq$ $3\|v\| / 4$. It follows that

$$
\mathcal{B}_{T_{x} M}\left(0, \frac{3 h}{4}\right) \subset G\left(\mathcal{B}_{T_{x} M}(0, h)\right)=\pi_{T_{x} M}\left(\Psi_{x}\left(\mathcal{B}_{T_{x} M}(0, h)\right)\right) .
$$

Now, according to Lemma A.1, for all $y \in \mathcal{B}\left(x, \frac{3 h}{5}\right) \cap M$,

$$
\left\|\pi_{T_{x} M}(y-x)\right\| \leq\|y-x\|+\frac{\|y-x\|^{2}}{2 \tau_{\min }} \leq\left(1+\frac{1}{4}\right)\|y-x\| \leq \frac{3 h}{4}
$$

from what we deduce $\pi_{T_{x} M}\left(\mathcal{B}\left(x, \frac{3 h}{5}\right) \cap M\right) \subset \mathcal{B}_{T_{x} M}\left(0, \frac{3 h}{4}\right)$. As a consequence,

$$
\pi_{T_{x} M}\left(\mathcal{B}\left(x, \frac{3 h}{5}\right) \cap M\right) \subset \pi_{T_{x} M}\left(\Psi_{x}\left(\mathcal{B}_{T_{x} M}(0, h)\right)\right),
$$

which yields the announced inclusion since $\pi_{T_{x} M}$ is one to one on $\mathcal{B}\left(x, \frac{5 h}{4}\right) \cap M$ from Lemma 3 in [4], and

$$
\left(\mathcal{B}\left(x, \frac{3 h}{5}\right) \cap M\right) \subset \Psi_{x}\left(\mathcal{B}_{T_{x} M}(0, h)\right) \subset \mathcal{B}\left(x, \frac{5 h}{4}\right) \cap M .
$$

(iii) Straightforward application of Lemma 3 in [4].
(iv) Notice that Lemma A. 1 gives the existence of such an expansion for $k=2$. Hence, we can assume $k \geq 3$. Taking $h=\frac{\tau_{\min } \wedge L_{\perp}^{-1}}{4}$, we showed in the proof of (ii) that the map $G$ is a diffeomorphism onto its image, with $\left\|d_{v} G-I d_{T_{x} M}\right\|_{o p} \leq \frac{1}{4}<1$. Additionally, the chain rule yields $\left\|d_{v}^{i} G\right\|_{o p} \leq\left\|d_{v}^{i} \Psi_{x}\right\|_{o p} \leq L_{i}$ for all $2 \leq i \leq k$. Therefore, from Lemma A. 3 , the differentials of $G^{-1}$ up to order $k$ are uniformly bounded. As a consequence, we get the announced expansion writing

$$
y-x=\Psi_{x} \circ G^{-1}\left(\pi^{*}(y-x)\right),
$$

and using the Taylor expansions of order $k$ of $\Psi_{x}$ and $G^{-1}$.
Let us now check that $T_{2}^{*}=I I_{x}^{M}$. Since, by construction, $T_{2}^{*}$ is the second order term of the Taylor expansion of $\Psi_{x} \circ G^{-1}$ at zero, a straightforward computation yields

$$
\begin{aligned}
T_{2}^{*} & =\left(I_{D}-\pi_{T_{x} M}\right) \circ d_{0}^{2} \Psi_{x} \\
& =\pi_{T_{x} M^{\perp}} \circ d_{0}^{2} \Psi_{x} .
\end{aligned}
$$

Let $v \in T_{x} M$ be fixed. Letting $\gamma(t)=\Psi_{x}(t v)$ for $|t|$ small enough, it is clear that $\gamma^{\prime \prime}(0)=d_{0}^{2} \Psi\left(v^{\otimes 2}\right)$. Moreover, by definition of the second fundamental form [8, Proposition 2.1, p.127], since $\gamma(0)=x$ and $\gamma^{\prime}(0)=v$, we have

$$
I I_{x}^{M}\left(v^{\otimes 2}\right)=\pi_{T_{x} M^{\perp}}\left(\gamma^{\prime \prime}(0)\right)
$$

Hence

$$
\begin{aligned}
T_{2}^{*}\left(v^{\otimes 2}\right) & =\pi_{T_{x} M^{\perp}} \circ d_{0}^{2} \Psi_{x}\left(v^{\otimes 2}\right) \\
& =\pi_{T_{x} M^{\perp}}\left(\gamma^{\prime \prime}(0)\right) \\
& =I I_{x}^{M}\left(v^{\otimes 2}\right)
\end{aligned}
$$

which concludes the proof.
(v) The first statement is a rephrasing of Proposition 6.1 in [13]. It yields the bound on sectional curvature, using the Gauss equation [8, Proposition 3.1 (a), p.135].

In the proof of Lemma A. 2 (iv), we used a technical lemma of differential calculus that we now prove. It states quantitatively that if $G$ is $\mathcal{C}^{k}$-close to the identity map, then it is a diffeomorphism onto its image and the differentials of its inverse $G^{-1}$ are controlled.

Lemma A.3. Let $k \geq 2$ and $U$ be an open subset of $\mathbb{R}^{d}$. Let $G: U \rightarrow \mathbb{R}^{d}$ be $\mathcal{C}^{k}$. Assume that $\left\|I_{d}-d G\right\|_{o p} \leq \varepsilon<1$, and that for all $2 \leq i \leq k$, $\left\|d^{i} G\right\|_{o p} \leq L_{i}$ for some $L_{i}>0$. Then $G$ is a $\mathcal{C}^{k}$-diffeomorphism onto its image, and for all $2 \leq i \leq k$,

$$
\left\|I_{d}-d G^{-1}\right\|_{o p} \leq \frac{\varepsilon}{1-\varepsilon} \quad \text { and } \quad\left\|d^{i} G^{-1}\right\|_{o p} \leq L_{i, \varepsilon, L_{2}, \ldots, L_{i}}^{\prime}<\infty, \text { for } 2 \leq i \leq k
$$

Proof of Lemma A.3. For all $x \in U,\left\|d_{x} G-I_{d}\right\|_{o p}<1$, so $G$ is one to one, and for all $y=G(x) \in G(U)$,

$$
\begin{aligned}
\left\|I_{d}-d_{y} G^{-1}\right\|_{o p} & =\left\|I_{d}-\left(d_{x} G\right)^{-1}\right\|_{o p} \\
& \leq\left\|\left(d_{x} G\right)^{-1}\right\|_{o p}\left\|I_{d}-d_{x} G\right\|_{o p} \\
& \leq \frac{\left\|I_{d}-d_{x} G\right\|_{o p}}{1-\left\|I_{d}-d_{x} G\right\|_{o p}} \\
& \leq \frac{\varepsilon}{1-\varepsilon}
\end{aligned}
$$

For $2 \leq i \leq k$ and $1 \leq j \leq i$, write $\Pi_{i}^{(j)}$ for the set of partitions of $\{1, \ldots, i\}$ with $j$ blocks. Differentiating $i$ times the identity $G \circ G^{-1}=I d_{G(U)}$, Faa di Bruno's formula yields that, for all $y=G(x) \in G(U)$ and all unit vectors $h_{1}, \ldots, h_{i} \in \mathbb{R}^{D}$,

$$
0=d_{y}\left(G \circ G^{-1}\right) \cdot\left(h_{\alpha}\right)_{1 \leq \alpha \leq i}=\sum_{j=1}^{i} \sum_{\pi \in \Pi_{i}^{(j)}} d_{x}^{j} G \cdot\left(\left(d_{y}^{I I \mid} G^{-1} \cdot\left(h_{\alpha}\right)_{\alpha \in I}\right)_{I \in \pi}\right) .
$$

Isolating the term for $j=1$ entails

$$
\begin{aligned}
& \left\|d_{x} G \cdot\left(d_{y}^{i} G^{-1} \cdot\left(h_{\alpha}\right)_{1 \leq \alpha \leq i}\right)\right\|_{o p} \\
& =\left\|-\sum_{j=2}^{i} \sum_{\pi \in \Pi_{i}^{(j)}} d_{x}^{j} G \cdot\left(\left(d_{y}^{|I|} G^{-1} \cdot\left(h_{\alpha}\right)_{\alpha \in I}\right)_{I \in \pi}\right)\right\|_{o p} \\
& \leq \sum_{j=2}^{i} \sum_{\pi \in \Pi_{i}^{(j)}}\left\|d^{j} G\right\|_{o p} \prod_{I \in \pi}\left\|d^{|I|} G^{-1}\right\|_{o p}
\end{aligned}
$$

Using the first order Lipschitz bound on $G^{-1}$, we get

$$
\left\|d^{i} G^{-1}\right\|_{o p} \leq \frac{1+\varepsilon}{1-\varepsilon} \sum_{j=2}^{i} L_{j} \sum_{\pi \in \Pi_{i}^{(j)}} \prod_{I \in \pi}\left\|d^{|I|} G^{-1}\right\|_{o p}
$$

The result follows by induction on $i$.

## A.3. Proof of Proposition 1

This section is devoted to prove Proposition 1 (reproduced below as Proposition A.4), that asserts the stability of the model with respect to ambient diffeomorphisms.

Proposition A.4. Let $\Phi: \mathbb{R}^{D} \rightarrow \mathbb{R}^{D}$ be a global $\mathcal{C}^{k}$-diffeomorphism. If $\left\|d \Phi-I_{D}\right\|_{o p},\left\|d^{2} \Phi\right\|_{o p}, \ldots,\left\|d^{k} \Phi\right\|_{o p}$ are small enough, then for all $P$ in $\mathcal{P}_{\tau_{\min }, \mathbf{L}, f_{\min }, f_{\max }}^{k}$, the pushforward distribution $P^{\prime}=\Phi_{*} P$ belongs to $\mathcal{P}_{\tau_{\text {min }} / 2,2 \mathbf{L}, f_{\text {min }} / 2,2 f_{\text {max }}}^{k}$.

Moreover, if $\Phi=\lambda I_{D}(\lambda>0)$ is an homogeneous dilation, then $P^{\prime} \in$ $\mathcal{P}_{\lambda \tau_{\min }, \mathbf{L}_{(\lambda)}, f_{\min } / \lambda^{d}, f_{\max } / \lambda^{d}}^{k}$ where $\mathbf{L}_{(\lambda)}=\left(L_{\perp} / \lambda, L_{3} / \lambda^{2}, \ldots, L_{k} / \lambda^{k-1}\right)$.

Proof of Proposition A.4. The second part is straightforward since the dilation $\lambda M$ has reach $\tau_{\lambda M}=\lambda \tau_{M}$, and can be parametrized locally by $\tilde{\Psi}_{\lambda p}(v)=\lambda \Psi_{p}(v / \lambda)=\lambda p+v+\lambda \mathbf{N}_{p}(v / \lambda)$, yielding the differential bounds $\mathbf{L}_{(\lambda)}$. Bounds on the density follow from homogeneity of the $d$-dimensional Hausdorff measure.

The first part follows combining Proposition A. 5 and Lemma A. 6.
Proposition A. 5 asserts the stability of the geometric model, that is, the reach bound and the existence of a smooth parametrization when a submanifold is perturbed.

Proposition A.5. Let $\Phi: \mathbb{R}^{D} \rightarrow \mathbb{R}^{D}$ be a global $\mathcal{C}^{k}$-diffeomorphism. If $\left\|d \Phi-I_{D}\right\|_{o p},\left\|d^{2} \Phi\right\|_{o p}, \ldots,\left\|d^{k} \Phi\right\|_{o p}$ are small enough, then for all $M$ in $\mathcal{C}_{\tau_{\text {min }}, \mathbf{L}}^{k}$, the image $M^{\prime}=\Phi(M)$ belongs to $\mathcal{C}_{\tau_{\min } / 2,2 L_{\perp}, 2 L_{3}, \ldots, 2 L_{k}}^{k}$.

Proof of Proposition A.5. To bound $\tau_{M^{\prime}}$ from below, we use the stability of the reach with respect to $\mathcal{C}^{2}$ diffeomorphisms. Namely, from Theorem 4.19 in [10],

$$
\begin{aligned}
\tau_{M^{\prime}}=\tau_{\Phi(M)} & \geq \frac{\left(1-\left\|I_{D}-d \Phi\right\|_{o p}\right)^{2}}{\frac{1+\left\|I_{D}-d \Phi\right\|_{o p}}{\tau_{M}}+\left\|d^{2} \Phi\right\|_{o p}} \\
& \geq \tau_{\min } \frac{\left(1-\left\|I_{D}-d \Phi\right\|_{o p}\right)^{2}}{1+\left\|I_{D}-d \Phi\right\|_{o p}+\tau_{\min }\left\|d^{2} \Phi\right\|_{o p}} \geq \frac{\tau_{\min }}{2}
\end{aligned}
$$

for $\left\|I_{D}-d \Phi\right\|_{o p}$ and $\left\|d^{2} \Phi\right\|_{o p}$ small enough. This shows the stability for $k=2$, as well as that of the reach assumption for $k \geq 3$.

By now, take $k \geq 3$. We focus on the existence of a good parametrization of $M^{\prime}$ around a fixed point $p^{\prime}=\Phi(p) \in M^{\prime}$. For $v^{\prime} \in T_{p^{\prime}} M^{\prime}=d_{p} \Phi\left(T_{p} M\right)$, let us define

$$
\begin{aligned}
\Psi_{p^{\prime}}^{\prime}\left(v^{\prime}\right) & =\Phi\left(\Psi_{p}\left(d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}\right)\right) \\
& =p^{\prime}+v^{\prime}+\mathbf{N}_{p^{\prime}}^{\prime}\left(v^{\prime}\right)
\end{aligned}
$$

where $\mathbf{N}_{p^{\prime}}^{\prime}\left(v^{\prime}\right)=\left\{\Phi\left(\Psi_{p}\left(d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}\right)\right)-p^{\prime}-v^{\prime}\right\}$.


The maps $\Psi_{p^{\prime}}^{\prime}\left(v^{\prime}\right)$ and $\mathbf{N}_{p^{\prime}}^{\prime}\left(v^{\prime}\right)$ are well defined whenever $\left\|d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}\right\| \leq \frac{1}{4 L_{\perp}}$, so in particular if $\left\|v^{\prime}\right\| \leq \frac{1}{4\left(2 L_{\perp}\right)} \leq \frac{1-\left\|I_{D}-d \Phi\right\|_{o p}}{4 L_{\perp}}$ and $\left\|I_{D}-d \Phi\right\|_{o p} \leq \frac{1}{2}$. One easily checks that $\mathbf{N}_{p^{\prime}}^{\prime}(0)=0, d_{0} \mathbf{N}_{p^{\prime}}^{\prime}=0$ and writing $c\left(v^{\prime}\right)=p+d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}+$ $\mathbf{N}_{p^{\prime}}\left(d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}\right)$, for all unit vector $w^{\prime} \in T_{p^{\prime}} M^{\prime}$,

$$
\begin{aligned}
& \left\|d_{v^{\prime}}^{2} \mathbf{N}_{p^{\prime}}^{\prime}\left(w^{\prime \otimes 2}\right)\right\|=\| d_{c\left(v^{\prime}\right)}^{2} \Phi\left(\left\{d_{d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}} \Psi_{p} \circ d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\}^{\otimes 2}\right) \\
& +d_{c\left(v^{\prime}\right)} \Phi \circ d_{d_{p^{\prime}} \Phi^{-1} . v^{\prime}}^{2} \Psi_{p}\left(\left\{d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\}^{\otimes 2}\right) \| \\
& =\| d_{c\left(v^{\prime}\right)}^{2} \Phi\left(\left\{d_{d_{p^{\prime}} \Phi^{-1} . v^{\prime}} \Psi_{p} \circ d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\}^{\otimes 2}\right) \\
& +\left(d_{c\left(v^{\prime}\right)} \Phi-I d\right) \circ d_{d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}}^{2} \Psi_{p}\left(\left\{d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\}^{\otimes 2}\right) \\
& +d_{d_{p^{\prime}} \Phi^{-1} . v^{\prime}}^{2} \Psi_{p}\left(\left\{d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\}^{\otimes 2}\right) \| \\
& \leq\left\|d^{2} \Phi\right\|_{o p}\left(1+L_{\perp}\left\|d_{p^{\prime}} \Phi^{-1} \cdot v^{\prime}\right\|\right)^{2}\left\|d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\|^{2} \\
& +\left\|I_{D}-d \Phi\right\|_{o p} L_{\perp}\left\|d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\|^{2} \\
& +L_{\perp}\left\|d_{p^{\prime}} \Phi^{-1} \cdot w^{\prime}\right\|^{2} \\
& \leq\left\|d^{2} \Phi\right\|_{o p}(1+1 / 4)^{2}\left\|d_{p^{\prime}} \Phi^{-1}\right\|_{o p}^{2} \\
& +\left\|I_{D}-d \Phi\right\|_{o p} L_{\perp}\left\|d \Phi^{-1}\right\|_{o p}^{2} \\
& +L_{\perp}\left\|d_{p^{\prime}} \Phi^{-1}\right\|_{o p}^{2} .
\end{aligned}
$$

Writing further $\left\|d \Phi^{-1}\right\|_{o p} \leq\left(1-\left\|I_{D}-d \Phi\right\|_{o p}\right)^{-1} \leq 1+2\left\|I_{D}-\Phi\right\|_{o p}$ for $\left\|I_{D}-d \Phi\right\|_{o p}$ small enough depending only on $L_{\perp}$, it is clear that the righthand side of the latter inequality goes below $2 L_{\perp}$ for $\left\|I_{D}-d \Phi\right\|_{o p}$ and $\left\|d^{2} \Phi\right\|_{o p}$ small enough. Hence, for $\left\|I_{D}-d \Phi\right\|_{o p}$ and $\left\|d^{2} \Phi\right\|_{o p}$ small enough depending only on $L_{\perp},\left\|d_{v^{\prime}}^{2} \mathbf{N}_{p^{\prime}}^{\prime}\right\|_{o p} \leq 2 L_{\perp}$ for all $\left\|v^{\prime}\right\| \leq \frac{1}{4\left(2 L_{\perp}\right)}$. From the chain rule, the same argument applies for the order $3 \leq i \leq k$ differential of $\mathbf{N}_{p^{\prime}}^{\prime}$.

Lemma A. 6 deals with the condition on the density in the models $\mathcal{P}^{k}$. It gives a change of variable formula for pushforward of measure on submanifolds, ensuring a control on densities with respect to intrinsic volume measure.

Lemma A. 6 (Change of variable for the Hausdorff measure). Let $P$ be a probability distribution on $M \subset \mathbb{R}^{D}$ with density $f$ with respect to the
d-dimensional Hausdorff measure $\mathcal{H}^{d}$. Let $\Phi: \mathbb{R}^{D} \rightarrow \mathbb{R}^{D}$ be a global diffeomorphism such that $\left\|I_{D}-d \Phi\right\|_{\mathrm{op}}<1 / 3$. Let $P^{\prime}=\Phi_{*} P$ be the pushforward of $P$ by $\Phi$. Then $P^{\prime}$ has a density $g$ with respect to $\mathcal{H}^{d}$. This density can be chosen to be, for all $z \in \Phi(M)$,

$$
g(z)=\frac{f\left(\Phi^{-1}(z)\right)}{\sqrt{\operatorname{det}\left(\left.\pi_{T_{\Phi^{-1}(z)} M} \circ d_{\Phi^{-1}(z)} \Phi^{T} \circ d_{\Phi^{-1}(z)} \Phi\right|_{T_{\Phi^{-1}(z)} M}\right)}}
$$

In particular, if $f_{\min } \leq f \leq f_{\max }$ on $M$, then for all $z \in \Phi(M)$,

$$
\left(1-3 d / 2\left\|I_{D}-d \Phi\right\|_{\mathrm{op}}\right) f_{\min } \leq g(z) \leq f_{\max }\left(1+3\left(2^{d / 2}-1\right)\left\|I_{D}-d \Phi\right\|_{\mathrm{op}}\right)
$$

Proof of Lemma A.6. Let $p \in M$ be fixed and $A \subset \mathcal{B}(p, r) \cap M$ for $r$ small enough. For a differentiable map $h: \mathbb{R}^{d} \rightarrow \mathbb{R}^{D}$ and for all $x \in \mathbb{R}^{d}$, we let $J_{h}(x)$ denote the $d$-dimensional Jacobian $J_{h}(x)=\sqrt{\operatorname{det}\left(d_{x} h^{T} d_{x} h\right)}$. The area formula ([11, Theorem 3.2.5]) states that if $h$ is one-to-one,

$$
\int_{A} u(h(x)) J_{h}(x) \lambda^{d}(d x)=\int_{h(A)} u(y) \mathcal{H}^{d}(d y)
$$

whenever $u: \mathbb{R}^{D} \rightarrow \mathbb{R}$ is Borel, where $\lambda^{d}$ is the Lebesgue measure on $\mathbb{R}^{d}$. By definition of the pushforward, and since $d P=f d \mathcal{H}^{d}$,

$$
\int_{\Phi(A)} d P^{\prime}(z)=\int_{A} f(y) \mathcal{H}^{d}(d y)
$$

Writing $\Psi_{p}=\exp _{p}: T_{p} M \rightarrow \mathbb{R}^{D}$ for the exponential map of $M$ at $p$, we have

$$
\int_{A} f(y) \mathcal{H}^{d}(d y)=\int_{\Psi_{p}^{-1}(A)} f\left(\Psi_{p}(x)\right) J_{\Psi_{p}}(x) \lambda^{d}(d x)
$$

Rewriting the right hand term, we apply the area formula again with $h=$ $\Phi \circ \Psi_{p}$,

$$
\begin{aligned}
\int_{\Psi_{p}-1}(A) & f\left(\Psi_{p}(x)\right) J_{\Psi_{p}}(x) \lambda^{d}(d y) \\
& =\int_{\Psi_{p}-1} f(A) \\
& =\int_{\Phi(A)} f\left(\Phi^{-1}(h(x))\right) \frac{J_{\Psi_{p}}\left(h^{-1}(h(x))\right)}{J_{\Phi \circ \Psi_{p}}\left(h^{-1}(h(x))\right)} J_{\Phi \circ \Psi_{p}}(x) \lambda^{d}(d x) \\
J_{\Phi \circ \Psi_{p}}\left(h^{-1}(z)\right) & \mathcal{H}^{d}(d z)
\end{aligned}
$$

Since this is true for all $A \subset \mathcal{B}(p, r) \cap M, P^{\prime}$ has a density $g$ with respect to $\mathcal{H}^{d}$, with

$$
g(z)=f\left(\Phi^{-1}(z)\right) \frac{J_{\Psi_{\Phi^{-1}(z)}}\left(\Psi_{\Phi^{-1}(z)}^{-1} \circ \Phi^{-1}(z)\right)}{J_{\Phi \circ \Psi^{-1}(z)}\left(\Psi_{\Phi^{-1}(z)}^{-1} \circ \Phi^{-1}(z)\right)} .
$$

Writing $p=\Phi^{-1}(z)$, it is clear that $\Psi_{\Phi^{-1}(z)}^{-1} \circ \Phi^{-1}(z)=\Psi_{p}^{-1}(p)=0 \in T_{p} M$. Since $d_{0} \exp _{p}: T_{p} M \rightarrow \mathbb{R}^{D}$ is the inclusion map, we get the first statement.

We now let $B$ and $\pi_{T}$ denote $d_{p} \Phi$ and $\pi_{T_{p} M}$ respectively. For any unit vector $v \in T_{p} M$,

$$
\begin{aligned}
\mid\left\|\pi_{T} B^{T} B v\right\|-\|v\| \| & \leq\left\|\pi_{T}\left(B^{T} B-I_{D}\right) v\right\| \\
& \leq\left\|B^{T} B-I_{D}\right\|_{\mathrm{op}} \\
& \leq\left(2+\left\|I_{D}-B\right\|_{\mathrm{op}}\right)\left\|I_{D}-B\right\|_{\mathrm{op}} \\
& \leq 3\left\|I_{D}-B\right\|_{\mathrm{op}}
\end{aligned}
$$

Therefore, $1-3\left\|I_{D}-B\right\|_{\mathrm{op}} \leq\left\|\pi_{T} B^{T} B \mid T_{p} M\right\|_{\mathrm{op}} \leq 1+3\left\|I_{D}-B\right\|_{\mathrm{op}}$. Hence,

$$
\sqrt{\operatorname{det}\left(\pi_{T} B^{T} B{\mid T_{p} M}\right)} \leq\left(1+3\left\|I_{D}-B\right\|_{\mathrm{op}}\right)^{d / 2} \leq \frac{1}{1-\frac{3 d}{2}\left\|I_{D}-B\right\|_{\mathrm{op}}}
$$

and

$$
\sqrt{\operatorname{det}\left(\left.\pi_{T} B^{T} B\right|_{T_{p} M}\right)} \geq\left(1-3\left\|I_{D}-B\right\|_{\mathrm{op}}\right)^{d / 2} \geq \frac{1}{1+3\left(2^{d / 2}-1\right)\left\|I_{D}-B\right\|_{\mathrm{op}}},
$$

which yields the result.

## Appendix B: Some Probabilistic Tools

## B.1. Volume and Covering Rate

The first lemma of this section gives some details about the covering rate of a manifold with bounded reach.

Lemma B.7. Let $P_{0} \in \mathcal{P}^{k}$ have support $M \subset \mathbb{R}^{D}$. Then for all $r \leq$ $\tau_{\text {min }} / 4$ and $x$ in $M$,

$$
c_{d} f_{\min } r^{d} \leq p_{x}(r) \leq C_{d} f_{\max } r^{d}
$$

for some $c_{d}, C_{d}>0$, with $p_{x}(r)=P_{0}(\mathcal{B}(x, r))$.

Moreover, letting $h=\left(\frac{C_{d}^{\prime} k}{f_{\min }} \frac{\log n}{n}\right)^{1 / d}$ with $C_{d}^{\prime}$ large enough, the following holds. For $n$ large enough so that $h \leq \tau_{\min } / 4$, with probability at least $1-$ $\left(\frac{1}{n}\right)^{k / d}$,

$$
d_{H}\left(M, \mathbb{Y}_{n}\right) \leq h / 2
$$

Proof of Lemma B.7. Denoting by $\mathcal{B}_{M}(x, r)$ the geodesic ball of radius $r$ centered at $x$, Proposition 25 of [1] yields

$$
\mathcal{B}_{M}(x, r) \subset \mathcal{B}(x, r) \cap M \subset \mathcal{B}_{M}(x, 6 r / 5) .
$$

Hence, the bounds on the Jacobian of the exponential map given by Proposition 27 of [1] yield

$$
c_{d} r^{d} \leq \operatorname{Vol}(\mathcal{B}(x, r) \cap M) \leq C_{d} r^{d}
$$

for some $c_{d}, C_{d}>0$. Now, since $P$ has a density $f_{\min } \leq f \leq f_{\max }$ with respect to the volume measure of $M$, we get the first result.

Now we notice that since $p_{x}(r) \geq c_{d} f_{\min } r^{d}$, Theorem 3.3 in [7] entails, for $s \leq \tau_{\text {min }} / 8$,

$$
\mathbb{P}\left(d_{H}\left(M, \mathbb{X}_{n}\right) \geq s\right) \leq \frac{4^{d}}{c_{d} f_{\min } s^{d}} \exp \left(-\frac{c_{d} f_{\min }}{2^{d}} n s^{d}\right)
$$

Hence, taking $s=h / 2$, and $h=\left(\frac{C_{d}^{\prime} k}{f_{m i n}} \frac{\log n}{n}\right)^{1 / d}$ with $C_{d}^{\prime}$ so that $C_{d}^{\prime} \geq$ $\frac{8^{d}}{c_{d} k} \vee \frac{2^{d}(1+k / d)}{c_{d} k}$ yields the result. Since $k \geq 1$, taking $C_{d}^{\prime}=\frac{8^{d}}{c_{d}}$ is sufficient.

## B.2. Concentration Bounds for Local Polynomials

This section is devoted to the proof of the following proposition.
Proposition B.8. Set $h=\left(K \frac{\log n}{n-1}\right)^{\frac{1}{d}}$. There exist constants $\kappa_{k, d}, c_{k, d}$ and $C_{d}$ such that, if $K \geq\left(\kappa_{k, d} f_{\text {max }}^{2} / f_{\text {min }}^{3}\right)$ and $n$ is large enough so that $3 h / 2 \leq h_{0} \leq \tau_{\text {min }} / 4$, then with probability at least $1-\left(\frac{1}{n}\right)^{\frac{k}{d}+1}$, we have

$$
\begin{array}{cc}
P_{0, n-1}\left[S^{2}\left(\pi^{*}(x)\right) \mathbb{1}_{\mathcal{B}(h / 2)}(x)\right] & \geq c_{k, d} h^{d} f_{\min }\left\|S_{h}\right\|_{2}^{2}, \\
N(3 h / 2) & \leq C_{d} f_{\max }(n-1) h^{d},
\end{array}
$$

for every $S \in \mathbb{R}^{k}\left[x_{1: d}\right]$, where $N(h)=\sum_{j=2}^{n} \mathbb{1}_{\mathcal{B}(0, h)}\left(Y_{j}\right)$.

A first step is to ensure that empirical expectations of order $k$ polynomials are close to their deterministic counterparts.

Proposition B.9. Let $b \leq \tau_{\text {min }} / 8$. For any $y_{0} \in M$, we have

$$
\begin{aligned}
& \mathbb{P}\left[\sup _{u_{1}, \ldots, u_{k}, \varepsilon \in\{0,1\}^{k}}\left|\left(P_{0}-P_{0, n-1}\right) \prod_{j=1}^{p}\left(\frac{\left\langle u_{j}, y\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}\left(y_{0}, b\right)}(y)\right|\right. \\
& \left.\geq p_{y_{0}}(b)\left(\frac{4 k \sqrt{2 \pi}}{\sqrt{(n-1) p_{y_{0}}(b)}}+\sqrt{\frac{2 t}{(n-1) p_{y_{0}}(b)}}+\frac{2}{3(n-1) p_{y_{0}}(b)}\right)\right] \leq e^{-t}
\end{aligned}
$$

where $P_{0, n-1}$ denotes the empirical distribution of $n-1$ i.i.d. random variables $Y_{i}$ drawn from $P_{0}$.

Proof of Proposition B.9. Without loss of generality we choose $y_{0}=$ 0 and shorten notation to $\mathcal{B}(b)$ and $p(b)$. Let $\mathcal{Z}$ denote the empirical process on the left-hand side of Proposition B.9. Denote also by $f_{u, \varepsilon}$ the map $\prod_{j=1}^{k}\left(\frac{\left\langle u_{j}, y\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}(b)}(y)$, and let $\mathcal{F}$ denote the set of such maps, for $u_{j}$ in $\mathcal{B}(1)$ and $\varepsilon$ in $\{0,1\}^{k}$.

Since $\left\|f_{u, \varepsilon}\right\|_{\infty} \leq 1$ and $P f_{u, \varepsilon}^{2} \leq p(b)$, the Talagrand-Bousquet inequality ([6, Theorem 2.3]) yields

$$
\mathcal{Z} \leq 4 \mathbb{E} \mathcal{Z}+\sqrt{\frac{2 p(b) t}{n-1}}+\frac{2 t}{3(n-1)},
$$

with probability larger than $1-e^{-t}$. It remains to bound $\mathbb{E Z}$ from above.
Lemma B.10. We may write

$$
\mathbb{E} \mathcal{Z} \leq \frac{\sqrt{2 \pi p(b)}}{\sqrt{n-1}} k
$$

Proof of Lemma B.10. Let $\sigma_{i}$ and $g_{i}$ denote some independent Rademacher and Gaussian variables. For convenience, we denote by $\mathbb{E}_{A}$ the expectation with respect to the random variable $A$. Using symmetrization
inequalities we may write

$$
\begin{aligned}
\mathbb{E} \mathcal{Z} & =\mathbb{E}_{Y} \sup _{u, \varepsilon}\left|\left(P_{0}-P_{0, n-1}\right) \prod_{j=1}^{k}\left(\frac{\left\langle u_{j}, y\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}(b)}(y)\right| \\
& \leq \frac{2}{n-1} \mathbb{E}_{Y} \mathbb{E}_{\sigma} \sup _{u, \varepsilon} \sum_{i=1}^{n-1} \sigma_{i} \prod_{j=1}^{k}\left(\frac{\left\langle u_{j}, Y_{i}\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) \\
& \leq \frac{\sqrt{2 \pi}}{n-1} \mathbb{E}_{Y} \mathbb{E}_{g} \sup _{u, \varepsilon} \sum_{i=1}^{n-1} g_{i} \prod_{j=1}^{k}\left(\frac{\left\langle u_{j}, Y_{i}\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) .
\end{aligned}
$$

Now let $\mathcal{Y}_{u, \varepsilon}$ denote the Gaussian process $\sum_{i=1}^{n-1} g_{i} \prod_{j=1}^{k}\left(\frac{\left\langle u_{j}, Y_{i}\right\rangle}{b}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right)$. Since, for any $y$ in $\mathcal{B}(b), u, v$ in $\mathcal{B}(1)^{k}$, and $\varepsilon, \varepsilon^{\prime}$ in $\{0,1\}^{k}$, we have

$$
\begin{aligned}
& \left|\prod_{j=1}^{k}\left(\frac{\left\langle y, u_{j}\right\rangle}{b}\right)^{\varepsilon_{j}}-\prod_{j=1}^{k}\left(\frac{\left\langle y, v_{j}\right\rangle}{b}\right)^{\varepsilon_{j}^{\prime}}\right| \\
& \leq \left\lvert\, \sum_{r=1}^{k}\left(\prod_{j=1}^{k+1-r}\left(\frac{\left\langle y, u_{j}\right\rangle}{b}\right)^{\varepsilon_{j}} \prod_{j=k+2-r}^{k}\left(\frac{\left\langle y, v_{j}\right\rangle}{b}\right)^{\varepsilon_{j}^{\prime}}\right.\right. \\
& \left.\quad-\prod_{j=1}^{k-r}\left(\frac{\left\langle y, u_{j}\right\rangle}{b}\right)^{\varepsilon_{j}} \prod_{j=k+1-r}^{k}\left(\frac{\left\langle y, v_{j}\right\rangle}{b}\right)^{\varepsilon_{j}^{\prime}}\right) \mid \\
& \leq \sum_{r=1}^{k} \left\lvert\, \prod_{j=1}^{k-r}\left(\frac{\left\langle y, u_{j}\right\rangle}{b}\right)^{\varepsilon_{j}} \prod_{j=k+2-r}^{k}\left(\frac{\left\langle y, v_{j}\right\rangle}{b}\right)^{\varepsilon_{j}^{\prime}}\left[\left(\frac{\left\langle u_{k+1-r}, y\right\rangle}{b}\right)^{\varepsilon_{k+1-r}}\right.\right. \\
& \quad \leq \sum_{r=1}^{k}\left|\frac{\left.\left\langle\left(\frac{\left\langle v_{k+1-r}, y\right\rangle}{b}\right)^{\varepsilon_{k+1-r}}\right] \right\rvert\,}{b}\right|
\end{aligned}
$$

we deduce that

$$
\begin{aligned}
\mathbb{E}_{g}\left(\mathcal{Y}_{u, \varepsilon}-\mathcal{Y}_{v, \varepsilon^{\prime}}\right)^{2} & \leq k \sum_{i=1}^{n-1} \sum_{r=1}^{k}\left(\frac{\left\langle\varepsilon_{r} u_{r}, Y_{i}\right\rangle}{b}-\frac{\left\langle\varepsilon_{r}^{\prime} v_{r}, Y_{i}\right\rangle}{b}\right)^{2} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) \\
& \leq \mathbb{E}_{g}\left(\Theta_{u, \varepsilon}-\Theta_{v, \varepsilon^{\prime}}\right)^{2},
\end{aligned}
$$

where $\Theta_{u, \varepsilon}=\sqrt{k} \sum_{i=1}^{n-1} \sum_{r=1}^{k} g_{i, r} \frac{\left\langle\varepsilon_{r} u_{r}, Y_{i}\right\rangle}{\frac{b}{13}} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right)$. According to Slepian's

Lemma [5, Theorem 13.3], it follows that

$$
\begin{aligned}
\mathbb{E}_{g} \sup _{u, \varepsilon} \mathcal{Y}_{g} & \leq \mathbb{E}_{g} \sup _{u, \varepsilon} \Theta_{u, \varepsilon} \\
& \leq \sqrt{k} \mathbb{E}_{g} \sup _{u, \varepsilon} \sum_{r=1}^{k} \frac{\left\langle\varepsilon_{r} u_{r}, \sum_{i=1}^{n-1} g_{i, r} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) Y_{i}\right\rangle}{b} \\
& \leq \sqrt{k} \mathbb{E}_{g} \sup _{u, \varepsilon} \sqrt{k \sum_{r=1}^{k} \frac{\left\langle\varepsilon_{r} u_{r}, \sum_{i=1}^{n-1} g_{i, r} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) Y_{i}\right\rangle^{2}}{b^{2}}} .
\end{aligned}
$$

We deduce that

$$
\begin{aligned}
\mathbb{E}_{g} \sup _{u, \varepsilon} Y_{g} & \leq \mathbb{E}_{g} \sup _{u, \varepsilon} \Theta_{g} \\
& \leq k \sqrt{\mathbb{E}_{g} \sup _{\|u\|=1, \varepsilon \in\{0,1\}} \frac{\left\langle\varepsilon u, \sum_{i=1}^{n-1} g_{i} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right) Y_{i}\right\rangle^{2}}{b^{2}}} \\
& \leq k \sqrt{\mathbb{E}_{g}\left\|\sum_{i=1}^{n-1} \frac{g_{i} Y_{i}}{b} \mathbb{1}_{\mathcal{B}(b)}\left(Y_{i}\right)\right\|^{2}} \\
& \leq k \sqrt{N(b)} .
\end{aligned}
$$

Then we can deduce that $\mathbb{E}_{X} \mathbb{E}_{g} \sup _{u, \varepsilon} Y_{g} \leq k \sqrt{p(b)}$.
Combining Lemma B. 10 with Talagrand-Bousquet's inequality gives the result of Proposition B.9.

We are now in position to prove Proposition B.8.
Proof of Proposition B.8. If $h / 2 \leq \tau_{\text {min }} / 4$, then, according to Lemma
B. $7, p(h / 2) \geq c_{d} f_{\text {min }} h^{d}$, hence, if $h=\left(K \frac{\log (n)}{n-1}\right)^{\frac{1}{d}},(n-1) p(h / 2) \geq K c_{d} f_{\text {min }} \log (n)$.

Choosing $b=h / 2$ and $t=(k / d+1) \log (n)+\log (2)$ in Proposition B. 9 and $K=K^{\prime} / f_{\text {min }}$, with $K^{\prime}>1$ leads to

$$
\begin{array}{r}
\mathbb{P}\left[\sup _{u_{1}, \ldots, u_{k}, \varepsilon \in\{0,1\}^{k}}\left|\left(P_{0}-P_{0, n-1}\right) \prod_{j=1}^{k}\left(2 \frac{\left\langle u_{j}, y\right\rangle}{h}\right)^{\varepsilon_{j}} \mathbb{1}_{\mathcal{B}\left(y_{0}, h / 2\right)}(y)\right|\right. \\
\left.\geq \frac{c_{d, k} f_{\text {max }}}{\sqrt{K^{\prime}}} h^{d}\right] \leq \frac{1}{2}\left(\frac{1}{n}\right)^{\frac{k}{d}+1} .
\end{array}
$$

On the complement of the probability event mentioned just above, for a polynomial $S=\sum_{\alpha \in[0, k]^{d}| | \alpha \mid \leq k} a_{\alpha} y_{1: d}^{\alpha}$, we have

$$
\begin{aligned}
\left(P_{0, n-1}-P_{0}\right) S^{2}\left(y_{1: d}\right) \mathbb{1}_{\mathcal{B}(h / 2)}(y) & \geq-\sum_{\alpha, \beta} \frac{c_{d, k} f_{\max }}{\sqrt{K^{\prime}}}\left|a_{\alpha} a_{\beta}\right| h^{d+|\alpha|+|\beta|} \\
& \geq-\frac{c_{d, k} f_{\max }}{\sqrt{K^{\prime}}} h^{d}\left\|S_{h}\right\|_{2}^{2}
\end{aligned}
$$

On the other hand, we may write, for all $r>0$,

$$
\int_{\mathcal{B}(0, r)} S^{2}\left(y_{1: d}\right) d y_{1} \ldots d y_{d} \geq C_{d, k} r^{d}\left\|S_{r}\right\|_{2}^{2}
$$

for some constant $C_{d, k}$. It follows that

$$
P_{0} S^{2}\left(y_{1: d}\right) \mathbb{1}_{\mathcal{B}(h / 2)}(y) \geq P_{0} S^{2}\left(y_{1: d}\right) \mathbb{1}_{B(7 h / 16)}\left(y_{1: d}\right) \geq c_{k, d} h^{d} f_{\text {min }}\left\|S_{h}\right\|_{2}^{2}
$$

according to Lemma A.2. Then we may choose $K^{\prime}=\kappa_{k, d}\left(f_{\max } / f_{\min }\right)^{2}$, with $\kappa_{k, d}$ large enough so that

$$
P_{0, n-1} S^{2}\left(x_{1: d}\right) \mathbb{1}_{\mathcal{B}(h / 2)}(y) \geq c_{k, d} f_{\min } h^{d}\left\|S_{h}\right\|_{2}^{2}
$$

The second inequality of Proposition B. 8 is derived the same way from Proposition B.9, choosing $\varepsilon=(0, \ldots, 0), b=3 h / 2$ and $h \leq \tau_{\min } / 8$ so that $b \leq \tau_{\min } / 4$.

## Appendix C: Minimax Lower Bounds

## C.1. Conditional Assouad's Lemma

This section is dedicated to the proof of Lemma 7, reproduced below as Lemma C. 11.

Lemma C. 11 (Conditional Assouad). Let $m \geq 1$ be an integer and let $\left\{\mathcal{Q}_{\tau}\right\}_{\tau \in\{0,1\}^{m}}$ be a family of $2^{m}$ submodels $\mathcal{Q}_{\tau} \subset \mathcal{Q}$. Let $\left\{U_{k} \times U_{k}^{\prime}\right\}_{1 \leq k \leq m}$ be a family of pairwise disjoint subsets of $\mathcal{X} \times \mathcal{X}^{\prime}$, and $\mathcal{D}_{\tau, k}$ be subsets of $\mathcal{D}$. Assume that for all $\tau \in\{0,1\}^{m}$ and $1 \leq k \leq m$,

- for all $Q_{\tau} \in \mathcal{Q}_{\tau}, \theta_{X}\left(Q_{\tau}\right) \in \mathcal{D}_{\tau, k}$ on the event $\left\{X \in U_{k}\right\}$;
- for all $\theta \in \mathcal{D}_{\tau, k}$ and $\theta^{\prime} \in \mathcal{D}_{\tau^{k}, k}, d\left(\theta, \theta^{\prime}\right) \geq \Delta$.

For all $\tau \in\{0,1\}^{m}$, let $\bar{Q}_{\tau} \in \overline{\operatorname{Conv}}\left(\mathcal{Q}_{\tau}\right)$, and write $\bar{\mu}_{\tau}$ and $\bar{\nu}_{\tau}$ for the marginal distributions of $\bar{Q}_{\tau}$ on $\mathcal{X}$ and $\mathcal{X}^{\prime}$ respectively. Assume that if $\left(X, X^{\prime}\right)$
has distribution $\bar{Q}_{\tau}, X$ and $X^{\prime}$ are independent conditionally on the event $\left\{\left(X, X^{\prime}\right) \in U_{k} \times U_{k}^{\prime}\right\}$, and that

$$
\min _{\substack{\tau \in\{0,1\}^{m} \\ 1 \leq k \leq m}}\left\{\left(\int_{U_{k}} d \bar{\mu}_{\tau} \wedge d \bar{\mu}_{\tau^{k}}\right)\left(\int_{U_{k}^{\prime}} d \bar{\nu}_{\tau} \wedge d \bar{\nu}_{\tau^{k}}\right)\right\} \geq 1-\alpha .
$$

Then,

$$
\inf _{\hat{\theta}} \sup _{Q \in \mathcal{Q}} \mathbb{E}_{Q}\left[d\left(\theta_{X}(Q), \hat{\theta}\left(X, X^{\prime}\right)\right)\right] \geq m \frac{\Delta}{2}(1-\alpha)
$$

where the infimum is taken over all the estimators $\hat{\theta}: \mathcal{X} \times \mathcal{X}^{\prime} \rightarrow \mathcal{D}$.
Proof of Lemma C.11. The proof follows that of Lemma 2 in [14]. Let $\hat{\theta}=\hat{\theta}\left(X, X^{\prime}\right)$ be fixed. For any family of $2^{m}$ distributions $\left\{Q_{\tau}\right\}_{\tau} \in\left\{\mathcal{Q}_{\tau}\right\}_{\tau}$, since the $U_{k} \times U_{k}^{\prime}$ 's are pairwise disjoint,

$$
\begin{aligned}
& \sup _{Q \in \mathcal{Q}} \mathbb{E}_{Q} {\left[d\left(\theta_{X}(Q), \hat{\theta}\left(X, X^{\prime}\right)\right)\right] } \\
& \geq \max _{\tau} \mathbb{E}_{Q_{\tau}} d\left(\hat{\theta}, \theta_{X}\left(Q_{\tau}\right)\right) \\
& \geq \max _{\tau} \mathbb{E}_{Q_{\tau}} \sum_{k=1}^{m} d\left(\hat{\theta}, \theta_{X}\left(Q_{\tau}\right)\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right) \\
& \geq \geq 2^{-m} \sum_{\tau} \sum_{k=1}^{m} \mathbb{E}_{Q_{\tau}} d\left(\hat{\theta}, \theta_{X}\left(Q_{\tau}\right)\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right) \\
& \geq \geq 2^{-m} \sum_{\tau} \sum_{k=1}^{m} \mathbb{E}_{Q_{\tau}} d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right) \\
&=\sum_{k=1}^{m} 2^{-(m+1)} \sum_{\tau}\left(\mathbb{E}_{Q_{\tau}} d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right)\right. \\
&\left.\quad+\mathbb{E}_{Q_{\tau k}} d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right)\right) .
\end{aligned}
$$

Since the previous inequality holds for all $Q_{\tau} \in \mathcal{Q}_{\tau}$, it extends to $\bar{Q}_{\tau} \in$ $\overline{\operatorname{Conv}}\left(\mathcal{Q}_{\tau}\right)$ by linearity. Let us now lower bound each of the terms of the sum for fixed $\tau \in\{0,1\}^{m}$ and $1 \leq k \leq m$. By assumption, if ( $X, X^{\prime}$ ) has distribution $\bar{Q}_{\tau}$, then conditionally on $\left\{\left(X, X^{\prime}\right) \in U_{k} \times U_{k}^{\prime}\right\}, X$ and $X^{\prime}$ are
independent. Therefore,

$$
\begin{aligned}
\mathbb{E}_{\bar{Q}_{\tau}} d & \left.d \hat{\theta}, \mathcal{D}_{\tau, k}\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right)+\mathbb{E}_{\bar{Q}_{\tau^{k}}} d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) \mathbb{1}_{U_{k} \times U_{k}^{\prime}}\left(X, X^{\prime}\right) \\
\geq & \mathbb{E}_{\bar{Q}_{\tau}} d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right) \mathbb{1}_{U_{k}}(X) \mathbb{1}_{U_{k}^{\prime}}\left(X^{\prime}\right)+\mathbb{E}_{\bar{Q}_{\tau^{k}}} d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) \mathbb{1}_{U_{k}}(X) \mathbb{1}_{U_{k}^{\prime}}\left(X^{\prime}\right) \\
= & \mathbb{E}_{\bar{\nu}_{\tau}}\left[\mathbb{E}_{\bar{\mu}_{\tau}}\left(d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right) \mathbb{1}_{U_{k}}(X)\right) \mathbb{1}_{U_{k}^{\prime}}\left(X^{\prime}\right)\right] \\
& \quad+\mathbb{E}_{\bar{\nu}_{\tau^{k}}}\left[\mathbb{E}_{\bar{\mu}_{\tau^{k}}}\left(d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) \mathbb{1}_{U_{k}}(X)\right) \mathbb{1}_{U_{k}^{\prime}}\left(X^{\prime}\right)\right] \\
= & \int_{U_{k}} \int_{U_{k}^{\prime}} d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right) d \bar{\mu}_{\tau}(x) d \bar{\nu}_{\tau}\left(x^{\prime}\right)+\int_{U_{k}} \int_{U_{k}^{\prime}} d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) d \bar{\mu}_{\tau^{k}}(x) d \bar{\nu}_{\tau^{k}}\left(x^{\prime}\right) \\
\geq & \int_{U_{k}} \int_{U_{k}^{\prime}}\left(d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right)+d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right)\right) d \bar{\mu}_{\tau} \wedge d \bar{\mu}_{\tau^{k}}(x) d \bar{\nu}_{\tau} \wedge d \bar{\nu}_{\tau^{k}}\left(x^{\prime}\right) \\
\geq & \Delta\left(\int_{U_{k}} d \bar{\mu}_{\tau} \wedge d \bar{\mu}_{\tau^{k}}\right)\left(\int_{U_{k}^{\prime}} d \bar{\nu}_{\tau} \wedge d \bar{\nu}_{\tau^{k}}\right) \\
\geq & \Delta(1-\alpha)
\end{aligned}
$$

where we used that $d\left(\hat{\theta}, \mathcal{D}_{\tau, k}\right)+d\left(\hat{\theta}, \mathcal{D}_{\tau^{k}, k}\right) \geq \Delta$. The result follows by summing the above bound $\left|\{1, \ldots, m\} \times\{0,1\}^{m}\right|=m 2^{m}$ times.

## C.2. Construction of Generic Hypotheses

Let $M_{0}^{(0)}$ be a $d$-dimensional $\mathcal{C}^{\infty}$-submanifold of $\mathbb{R}^{D}$ with reach greater than 1 and such that it contains $\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}(0,1 / 2) . M_{0}^{(0)}$ can be built for example by flattening smoothly a unit $d$-sphere in $\mathbb{R}^{d+1} \times\{0\}^{D-d-1}$. Since $M_{0}^{(0)}$ is $\mathcal{C}^{\infty}$, the uniform probability distribution $P_{0}^{(0)}$ on $M_{0}^{(0)}$ belongs to $\mathcal{P}_{1, \mathbf{L}^{(0)}, 1 / V_{0}^{(0)}, 1 / V_{0}^{(0)}}^{k}$, for some $\mathbf{L}^{(0)}$ and $V_{0}^{(0)}=\operatorname{Vol}\left(M_{0}^{(0)}\right)$.

Let now $M_{0}=\left(2 \tau_{\min }\right) M_{0}^{(0)}$ be the submanifold obtained from $M_{0}^{(0)}$ by homothecy. By construction, and from Proposition A.4, we have

$$
\tau_{M_{0}} \geq 2 \tau_{\text {min }}, \quad \mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(0, \tau_{\text {min }}\right) \subset M_{0}, \quad \operatorname{Vol}\left(M_{0}\right)=C_{d} \tau_{\text {min }}^{d}
$$

and the uniform probability distribution $P_{0}$ on $M_{0}$ satisfies

$$
P_{0} \in \mathcal{P}_{2 \tau_{\min }, \mathbf{L} / 2,2 f_{\text {min }}, f_{\max } / 2}^{k}
$$

whenever $L_{\perp} / 2 \geq L_{\perp}^{(0)} /\left(2 \tau_{\text {min }}\right), \ldots, L_{k} / 2 \geq L_{k}^{(0)} /\left(2 \tau_{\text {min }}\right)^{k-1}$, and provided that $2 f_{\text {min }} \leq\left(\left(2 \tau_{\text {min }}\right)^{d} V_{0}^{(0)}\right)^{-1} \leq f_{\max } / 2$. Note that $L_{\perp}^{(0)}, \ldots, L_{k}^{(0)}$, $\operatorname{Vol}\left(M_{0}^{(0)}\right)$ depend only on $d$ and $k$. For this reason, all the lower bounds will
be valid for $\tau_{\text {min }} L_{\perp}, \ldots, \tau_{\text {min }}^{k-1} L_{k},\left(\tau_{\text {min }}^{d} f_{\text {min }}\right)^{-1}$ and $\tau_{\text {min }}^{d} f_{\text {max }}$ large enough to exceed the thresholds $L_{\perp}^{(0)} / 2, \ldots, L_{k}^{(0)} / 2^{k-1}, 2^{d} V_{0}^{(0)}$ and $\left(2^{d} V_{0}^{(0)}\right)^{-1}$ respectively.

For $0<\delta \leq \tau_{\min } / 4$, let $x_{1}, \ldots, x_{m} \in M_{0} \cap \mathcal{B}\left(0, \tau_{\text {min }} / 4\right)$ be a family of points such that

$$
\text { for } \quad 1 \leq k \neq k^{\prime} \leq m, \quad\left\|x_{k}-x_{k^{\prime}}\right\| \geq \delta .
$$

For instance, considering the family $\left\{\left(l_{1} \delta, \ldots, l_{d} \delta, 0, \ldots, 0\right)\right\}_{l_{i} \in \mathbb{Z},\left|l_{i}\right| \leq\left\lfloor\tau_{\text {min }} /(4 \delta)\right]}$,

$$
m \geq c_{d}\left(\frac{\tau_{\min }}{\delta}\right)^{d}
$$

for some $c_{d}>0$.
We let $e \in \mathbb{R}^{D}$ denote the $(d+1)$ th vector of the canonical basis. In particular, we have the orthogonal decomposition of the ambient space

$$
\mathbb{R}^{D}=\left(\mathbb{R}^{d} \times\{0\}^{D-d}\right)+\operatorname{span}(e)+\left(\{0\}^{d+1} \times \mathbb{R}^{D-d-1}\right) .
$$

Let $\phi: \mathbb{R}^{D} \rightarrow[0,1]$ be a smooth scalar map such that $\left.\phi\right|_{\mathcal{B}\left(0, \frac{1}{2}\right)}=$ 1 and $\left.\phi\right|_{\mathcal{B}(0,1)^{c}}=0$.

Let $\Lambda_{+}>0$ and $1 \geq A_{+}>A_{-}>0$ be real numbers to be chosen later. Let $\boldsymbol{\Lambda}=\left(\Lambda_{1}, \ldots, \Lambda_{m}\right)$ with entries $-\Lambda_{+} \leq \Lambda_{k} \leq \Lambda_{+}$, and $\mathbf{A}=\left(A_{1}, \ldots, A_{m}\right)$ with entries $A_{-} \leq A_{k} \leq A_{+}$. For $z \in \mathbb{R}^{D}$, we write $z=\left(z_{1}, \ldots, z_{D}\right)$ for its coordinates in the canonical basis. For all $\tau=\left(\tau_{1}, \ldots, \tau_{m}\right) \in\{0,1\}^{m}$, define the bump map as

$$
\begin{equation*}
\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A}, i}(x)=x+\sum_{k=1}^{m} \phi\left(\frac{x-x_{k}}{\delta}\right)\left\{\tau_{k} A_{k}\left(x-x_{k}\right)_{1}^{i}+\left(1-\tau_{k}\right) \Lambda_{k}\right\} e . \tag{1}
\end{equation*}
$$

An analogous deformation map was considered in [1]. We let $P_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}$ denote the pushforward distribution of $P_{0}$ by $\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}$, and write $M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}$ for its support. Roughly speaking, $M_{\tau}^{\Lambda, \mathbf{A}, i}$ consists of $m$ bumps at the $x_{k}$ 's having different shapes (Figure 1). If $\tau_{k}=0$, the bump at $x_{k}$ is a symmetric plateau function and has height $\Lambda_{k}$. If $\tau_{k}=1$, it fits the graph of the polynomial $A_{k}\left(x-x_{k}\right)_{1}^{i}$ locally. The following Lemma C. 12 gives differential bounds and geometric properties of $\Phi_{\tau}^{\Lambda, \mathbf{A}, i}$.

Lemma C.12. There exists $c_{\phi, i}<1$ such that if $A_{+} \leq c_{\phi, i} \delta^{i-1}$ and $\Lambda_{+} \leq c_{\phi, i} \delta$, then $\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A}, i}$ is a global $\mathcal{C}^{\infty}$-diffeomorphism of $\mathbb{R}^{D}$ such that for


Figure 1: The three shapes of the bump map $\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A}, i}$ around $x_{k}$.
all $1 \leq k \leq m, \Phi_{\tau}^{\Lambda, \mathbf{A}, i}\left(\mathcal{B}\left(x_{k}, \delta\right)\right)=\mathcal{B}\left(x_{k}, \delta\right)$. Moreover,

$$
\left\|I_{D}-d \Phi_{\tau}^{\Lambda, \mathbf{A}, i}\right\|_{o p} \leq C_{i}\left\{\frac{A_{+}}{\delta^{1-i}}\right\} \vee\left\{\frac{\Lambda_{+}}{\delta}\right\}
$$

and for $j \geq 2$,

$$
\left\|d^{j} \Phi_{\tau}^{\Lambda, \mathbf{A}, i}\right\|_{o p} \leq C_{i, j}\left\{\frac{A_{+}}{\delta^{j-i}}\right\} \vee\left\{\frac{\Lambda_{+}}{\delta^{j}}\right\} .
$$

Proof of Lemma C.12. Follows straightforwardly from chain rule, similarly to Lemma 11 in [1].

Lemma C.13. If $\tau_{\min } L_{\perp}, \ldots, \tau_{\min }^{k-1} L_{k},\left(\tau_{\text {min }}^{d} f_{\text {min }}\right)^{-1}$ and $\tau_{\text {min }}^{d} f_{\text {max }}$ are large enough (depending only on $d$ and $k$ ), then provided that $\Lambda_{+} \vee A_{+} \delta^{i} \leq$ $c_{k, d, \tau_{m i n}} \delta^{k}$, for all $\tau \in\{0,1\}^{m}, P_{\tau}^{\mathbf{\Lambda}, \mathbf{A}, i} \in \mathcal{P}_{\tau_{m i n}, \mathbf{L}, f_{\min }, f_{\max }}^{k}$

Proof of Lemma C.13. Follows using the stability of the model Lemma A. 4 applied to the distribution $P_{0} \in \mathcal{P}_{2 \tau_{m i n}, \mathbf{L} / 2,2 f_{\min }, f_{\max } / 2}^{k}$ and the map $\Phi_{\tau}^{\Lambda, \mathbf{A}, i}$, of which differential bounds are asserted by Lemma C.12.

## C.3. Hypotheses for Tangent Space and Curvature

## C.3.1. Proof of Lemma 8

This section is devoted to the proof of Lemma 8, for which we first derive two slightly more general results, with parameters to be tuned later. The proof is split into two intermediate results Lemma C. 14 and Lemma C.15.

Let us write $\bar{Q}_{\tau, n}^{(i)}$ for the mixture distribution on $\left(\mathbb{R}^{D}\right)^{n}$ defined by

$$
\begin{equation*}
\bar{Q}_{\tau, n}^{(i)}=\int_{\left[-\Lambda_{+}, \Lambda_{+}\right]^{m}} \int_{\left[A_{-}, A_{+}\right]^{m}}\left(P_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}\right)^{\otimes n} \frac{d \mathbf{A}}{\left(A_{+}-A_{-}\right)^{m}} \frac{d \boldsymbol{\Lambda}}{\left(2 \Lambda_{+}\right)^{m}} . \tag{2}
\end{equation*}
$$

Although the probability distribution $\bar{Q}_{\tau, n}^{(i)}$ depends on $A_{-}, A_{+}$and $\Lambda_{+}$, we omit this dependency for the sake of compactness. Another way to define $\bar{Q}_{\tau, n}^{(i)}$ is the following: draw uniformly $\boldsymbol{\Lambda}$ in $\left[-\Lambda_{+}, \Lambda_{+}\right]^{m}$ and $\mathbf{A}$ in $\left[A_{-}, A_{+}\right]^{m}$, and given $(\boldsymbol{\Lambda}, \mathbf{A})$, take $Z_{i}=\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A}, i}\left(Y_{i}\right)$, where $Y_{1}, \ldots, Y_{n}$ is an i.i.d. $n$-sample with common distribution $P_{0}$ on $M_{0}$. Then $\left(Z_{1}, \ldots, Z_{n}\right)$ has distribution $\bar{Q}_{\tau, n}^{(i)}$.

Lemma C.14. Assume that the conditions of Lemma C.12 hold, and let

$$
U_{k}=\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta / 2\right)+\mathcal{B}_{\text {span }(e)}\left(0, \tau_{\text {min }} / 2\right),
$$

and

$$
U_{k}^{\prime}=\left(\mathbb{R}^{D} \backslash\left\{\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta\right)+\mathcal{B}_{\text {span }(e)}\left(0, \tau_{\min } / 2\right)\right\}\right)^{n-1}
$$

Then the sets $U_{k} \times U_{k}^{\prime}$ are pairwise disjoint, $\bar{Q}_{\tau, n}^{(i)} \in \overline{\operatorname{Conv}}\left(\left(\mathcal{P}_{\tau}^{(i)}\right)^{\otimes n}\right)$, and if $\left(Z_{1}, \ldots, Z_{n}\right)=\left(Z_{1}, Z_{2: n}\right)$ has distribution $\bar{Q}_{\tau, n}^{(i)}, Z_{1}$ and $Z_{2: d}$ are independent conditionally on the event $\left\{\left(Z_{1}, Z_{2: n}\right) \in U_{k} \times U_{k}^{\prime}\right\}$.

Moreover, if $\left(X_{1}, \ldots, X_{n}\right)$ has distribution $\left(P_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}\right)^{\otimes n}$ (with fixed $\mathbf{A}$ and $\boldsymbol{\Lambda})$, then on the event $\left\{X_{1} \in U_{k}\right\}$, we have:

- if $\tau_{k}=0$,

$$
T_{X_{1}} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}=\mathbb{R}^{d} \times\{0\}^{D-d} \quad, \quad\left\|I I_{X_{1}}^{M_{\tau}^{\Lambda, \mathbf{A},(i)}} \circ \pi_{T_{X_{1}} M_{\tau}^{\Lambda}, \mathbf{A},(i)}\right\|_{o p}=0
$$

and $d_{H}\left(M_{0}, M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}\right) \geq\left|\Lambda_{k}\right|$.

- if $\tau_{k}=1$,
- for $i=1: \angle\left(T_{X_{1}} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(1)}, \mathbb{R}^{d} \times\{0\}^{D-d}\right) \geq A_{-} / 2$.
- for $i=2:\left\|I I_{X_{1}}^{M_{\tau}^{\Lambda, \mathbf{A},(2)}} \circ \pi_{T_{X_{1}} M_{\tau}^{\Lambda, \mathbf{A},(2)}}\right\|_{o p} \geq A_{-} / 2$.

Proof of Lemma C.14. It is clear from the definition (2) that $\bar{Q}_{\tau, n}^{(i)} \in$ $\overline{\operatorname{Conv}}\left(\left(\mathcal{P}_{\tau}^{(i)}\right)^{\otimes n}\right)$. By construction of the $\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A}, i}$, s , these maps leave the sets

$$
\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta\right)+\mathcal{B}_{\text {span }(e)}\left(0, \tau_{\min } / 2\right)
$$

unchanged for all $\boldsymbol{\Lambda}, \mathbf{L}$. Therefore, on the event $\left\{\left(Z_{1}, Z_{2: n}\right) \in U_{k} \times U_{k}^{\prime}\right\}$, one can write $Z_{1}$ only as a function of $X_{1}, \Lambda_{k}, A_{k}$, and $Z_{2: n}$ as a function of the rest of the $X_{j}$ 's, $\Lambda_{k}$ 's and $A_{k}$ 's. Therefore, $Z_{1}$ and $Z_{2: n}$ are independent.

We now focus on the geometric statements. For this, we fix a deterministic point $z=\Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}\left(x_{0}\right) \in U_{k} \cap M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}$. By construction, one necessarily has $x_{0} \in M_{0} \cap \mathcal{B}\left(x_{k}, \delta / 2\right)$.

- If $\tau_{k}=0$, locally around $x_{0}, \Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(1)}$ is the translation of vector $\Lambda_{k} e$. Therefore, since $M_{0}$ satisfies $T_{x_{0}} M_{0}=\mathbb{R}^{d} \times\{0\}^{D-d}$ and $I I_{x_{0}}^{M_{0}}=0$, we have

$$
T_{z} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}=\mathbb{R}^{d} \times\{0\}^{D-d} \quad \text { and } \quad\left\|I I_{z}^{M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}} \circ \pi_{T_{z} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}}\right\|_{o p}=0
$$

- if $\tau_{k}=1$,
- for $i=1$ : locally around $x_{0}, \Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(1)}$ can be written as $x \mapsto x+$ $A_{k}\left(x-x_{k}\right)_{1} e$. Hence, $T_{z} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(i)}$ contains the direction $\left(1, A_{k}\right)$ in the plane $\operatorname{span}\left(e_{1}, e\right)$ spanned by the first vector of the canonical basis and $e$. As a consequence, since $e$ is orthogonal to $\mathbb{R}^{d} \times$ $\{0\}^{D-d}$,

$$
\angle\left(T_{z} M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(1)}, \mathbb{R}^{d} \times\{0\}^{D-d}\right) \geq\left(1+1 / A_{k}^{2}\right)^{-1 / 2} \geq A_{k} / 2 \geq A_{-} / 2
$$

- for $i=2$ : locally around $x_{0}, \Phi_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(2)}$ can be written as $x \mapsto$ $x+A_{k}\left(x-x_{k}\right)_{1}^{2} e$. Hence, $M_{\tau}^{\boldsymbol{\Lambda}, \mathbf{A},(2)}$ contains an arc of parabola of equation $y=A_{k}\left(x-x_{k}\right)_{1}^{2}$ in the plane $\operatorname{span}\left(e_{1}, e\right)$. As a consequence,

$$
\left\|I I_{z}^{M_{\tau}^{\Lambda, \mathbf{A},(2)}} \circ \pi_{T_{z} M_{\tau}^{\Lambda, \mathbf{A},(2)}}\right\|_{o p} \geq A_{k} / 2 \geq A_{-} / 2
$$

Lemma C.15. Assume that the conditions of Lemma C. 12 and Lemma C. 14 hold. If in addition, $c A_{+}(\delta / 4)^{i} \leq \Lambda_{+} \leq C A_{+}(\delta / 4)^{i}$ for some absolute constants $C \geq c>3 / 4$, and $A_{-}=A_{+} / 2$, then,

$$
\int_{U_{k}} d \bar{Q}_{\tau, 1}^{(i)} \wedge d \bar{Q}_{\tau^{k}, 1}^{(i)} \geq \frac{c_{d, i}}{C}\left(\frac{\delta}{\tau_{\min }}\right)^{d}
$$

and

$$
\int_{U_{k}^{\prime}} d \bar{Q}_{\tau, n-1}^{(i)} \wedge d \bar{Q}_{\tau^{k}, n-1}^{(i)}=\left(1-c_{d}^{\prime}\left(\frac{\delta}{\tau_{\min }}\right)^{d}\right)^{n-1}
$$

Proof of Lemma C.15. First note that all the involved distributions have support in $\mathbb{R}^{d} \times \operatorname{span}(e) \times\{0\}^{D-(d+1)}$. Therefore, we use the canonical coordinate system of $\mathbb{R}^{d} \times \operatorname{span}(e)$, centered at $x_{k}$, and we denote the components by $\left(x_{1}, x_{2}, \ldots, x_{d}, y\right)=\left(x_{1}, x_{2: d}, y\right)$. Without loss of generality, assume that $\tau_{k}=0$ (if not, flip $\tau$ and $\tau^{k}$ ). Recall that $\phi$ has been chosen to be constant and equal to 1 on the ball $\mathcal{B}(0,1 / 2)$.

By definition (2), on the event $\left\{Z \in U_{k}\right\}$, a random variable $Z$ having distribution $\bar{Q}_{\tau, 1}^{(i)}$ can be represented by $Z=X+\phi\left(\frac{X-x_{k}}{\delta}\right) \Lambda_{k} e=X+\Lambda_{k} e$ where $X$ and $\Lambda_{k}$ are independent and have respective distributions $P_{0}$ (the uniform distribution on $M_{0}$ ) and the uniform distribution on $\left[-\Lambda_{+}, \Lambda_{+}\right]$. Therefore, on $U_{k}, \bar{Q}_{\tau, 1}^{(i)}$ has a density with respect to the Lebesgue measure $\lambda_{d+1}$ on $\mathbb{R}^{d} \times \operatorname{span}(e)$ that can be written as

$$
\bar{q}_{\tau, 1}^{(i)}\left(x_{1}, x_{2: d}, y\right)=\frac{\mathbb{1}_{\left[-\Lambda_{+}, \Lambda_{+}\right]}(y)}{2 \operatorname{Vol}\left(M_{0}\right) \Lambda_{+}} .
$$

Analogously, nearby $x_{k}$ a random variable $Z$ having distribution $\bar{Q}_{\tau^{k}, 1}^{(i)}$ can be represented by $Z=X+A_{k}\left(X-x_{k}\right)_{1}^{i} e$ where $A_{k}$ has uniform distribution on $\left[A_{-}, A_{+}\right]$. Therefore, a straightforward change of variable yields the density

$$
\bar{q}_{\tau^{k}, 1}^{(i)}\left(x_{1}, x_{2: d}, y\right)=\frac{\mathbb{1}_{\left[A_{-} x_{1}^{i}, A_{+} x_{1}^{i}\right]}(y)}{\operatorname{Vol}\left(M_{0}\right)\left(A_{+}-A_{-}\right) x_{1}^{i}} .
$$

We recall that $\operatorname{Vol}\left(M_{0}\right)=\left(2 \tau_{\text {min }}\right)^{d} \operatorname{Vol}\left(M_{0}^{(0)}\right)=c_{d}^{\prime} \tau_{\text {min }}^{d}$. Let us now tackle the right-hand side inequality, writing

$$
\begin{array}{rl}
\int_{U_{k}} & d \bar{Q}_{\tau, 1}^{(i)} \wedge d \bar{Q}_{\tau^{k}, 1}^{(i)} \\
& =\int_{\mathcal{B}\left(x_{k}, \delta / 2\right)}\left(\frac{\mathbb{1}_{\left[-\Lambda_{+}, \Lambda_{+}\right]}(y)}{2 \operatorname{Vol}\left(M_{0}\right) \Lambda_{+}}\right) \wedge\left(\frac{\mathbb{1}_{\left[A_{-} x_{1}^{i}, A_{+} x_{1}^{i}\right]}(y)}{\operatorname{Vol}\left(M_{0}\right)\left(A_{+}-A_{-}\right) x_{1}^{i}}\right) d y d x_{1} d x_{2: d} \\
& \geq \int_{\mathcal{B}_{\mathbb{R}^{d-1}}\left(0, \frac{\delta}{4}\right)} \int_{-\delta / 4}^{\delta / 4} \int_{\mathbb{R}}\left(\frac{\mathbb{1}_{\left[-\Lambda_{+}, \Lambda_{+}\right]}(y)}{2 \Lambda_{+}}\right) \wedge\left(\frac{\mathbb{1}_{\left[A_{-} x_{1}^{i}, A_{+} x_{1}^{i}\right]}(y)}{A_{+} x_{1}^{i} / 2}\right) \frac{d y d x_{1} d x_{2: d}}{\operatorname{Vol}\left(M_{0}\right)} .
\end{array}
$$

It follows that

$$
\begin{aligned}
& \int_{U_{k}} d \bar{Q}_{\tau, 1}^{(i)} \wedge d \bar{Q}_{\tau^{k}, 1}^{(i)} \\
& \quad \geq \frac{c_{d}}{\tau_{\text {min }}^{d}} \delta^{d-1} \int_{0}^{\delta / 4} \int_{A_{+} x_{1}^{i} / 2}^{\Lambda_{+} \wedge\left(A_{+} x_{1}^{i}\right)} \frac{1}{2 \Lambda_{+}} \wedge \frac{2}{A_{+} x_{1}^{i}} d y d x_{1} \\
& \quad \geq \frac{c_{d}}{\tau_{\text {min }}^{d}} \delta^{d-1} \int_{0}^{\delta / 4} \int_{A_{+} x_{1}^{i} / 2}^{(c \wedge 1)\left(A_{+} x_{1}^{i}\right)} \frac{(2 c \wedge 1 / 2)}{2 \Lambda_{+}} d y d x_{1} \\
& \quad=\frac{c_{d}}{\tau_{\text {min }}^{d}} \delta^{d-1}(2 c \wedge 1 / 2)(c \wedge 1-1 / 2) \frac{A_{+}}{\Lambda_{+}} \frac{(\delta / 4)^{i+1}}{i+1} \\
& \quad \geq \frac{c_{d, i}}{C}\left(\frac{\delta}{\tau_{\text {min }}}\right)^{d}
\end{aligned}
$$

For the integral on $U_{k}^{\prime}$, notice that by definition, $\bar{Q}_{\tau, n-1}^{(i)}$ and $\bar{Q}_{\tau^{k}, n-1}^{(i)}$ coincide on $U_{k}^{\prime}$ since they are respectively the image distributions of $P_{0}$ by functions that are equal on that set. Moreover, these two functions leave $\mathbb{R}^{D} \backslash\left\{\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta\right)+\mathcal{B}_{\text {span }(e)}\left(0, \tau_{\text {min }} / 2\right)\right\}$ unchanged. Therefore,

$$
\begin{aligned}
\int_{U_{k}^{\prime}} d \bar{Q}_{\tau, n-1}^{(i)} \wedge & d \bar{Q}_{\tau^{k}, n-1}^{(i)} \\
& =P_{0}^{\otimes n-1}\left(U_{k}^{\prime}\right) \\
& =\left(1-P_{0}\left(\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta\right)+\mathcal{B}_{\text {span }(e)}\left(0, \tau_{\min } / 2\right)\right)\right)^{n-1} \\
& =\left(1-\omega_{d} \delta^{d} / \operatorname{Vol}\left(M_{0}\right)\right)^{n-1}
\end{aligned}
$$

hence the result.
Proof of Lemma 8. The properties of $\left\{\bar{Q}_{\tau, n}^{(i)}\right\}_{\tau}$ and $\left\{U_{k} \times U_{k}^{\prime}\right\}_{k}$ given by Lemma C. 14 and Lemma C. 15 yield the result, setting $\Lambda_{+}=A_{+} \delta^{i} / 4$, $A_{+}=2 A_{-}=\varepsilon \delta^{k-i}$ for $\varepsilon=\varepsilon_{k, d, \tau_{m i n}}$, and $\delta$ such that $c_{d}^{\prime}\left(\frac{\delta}{\tau_{\text {min }}}\right)^{d}=\frac{1}{n-1}$.

## C.3.2. Proof of Lemma 9

This section details the construction leading to Lemma 9 that we restate in Lemma C.16.

Lemma C.16. Assume that $\tau_{\min } L_{\perp}, \ldots, \tau_{\min }^{k-1} L_{k},\left(\tau_{\min }^{d} f_{\min }\right)^{-1}, \tau_{\min }^{d} f_{\max }$ are large enough (depending only on $d$ and $k$ ), and $\sigma \geq C_{k, d, \tau_{m i n}}(1 /(n-1))^{k / d}$ for $C_{k, d, \tau_{\text {min }}}>0$ large enough. Given $i \in\{1,2\}$, there exists a collection of $2^{m}$ distributions $\left\{\mathbf{P}_{\tau}^{(i), \sigma}\right\}_{\tau \in\{0,1\}^{m}} \subset \mathcal{P}^{k}(\sigma)$ with associated submanifolds $\left\{M_{\tau}^{(i), \sigma}\right\}_{\tau \in\{0,1\}^{m}}$, together with pairwise disjoint subsets $\left\{U_{k}^{\sigma}\right\}_{1 \leq k \leq m}$ of $\mathbb{R}^{D}$ such that the following holds for all $\tau \in\{0,1\}^{m}$ and $1 \leq k \leq m$.

If $x \in U_{k}^{\sigma}$ and $y=\pi_{M_{\tau}^{(i), \sigma}}(x)$, we have

- if $\tau_{k}=0$,

$$
T_{y} M_{\tau}^{(i), \sigma}=\mathbb{R}^{d} \times\{0\}^{D-d} \quad, \quad\left\|I I_{y}^{M_{\tau}^{(i), \sigma}} \circ \pi_{T_{y} M_{\tau}^{(i), \sigma}}\right\|_{o p}=0,
$$

- if $\tau_{k}=1$,

$$
\begin{aligned}
& - \text { for } i=1: \angle\left(T_{y} M_{\tau}^{(1), \sigma}, \mathbb{R}^{d} \times\{0\}^{D-d}\right) \geq c_{k, d, \tau_{\min }}\left(\frac{\sigma}{n-1}\right)^{\frac{k-1}{k+d}}, \\
& - \text { for } i=2:\left\|I I_{y}^{M_{\tau}^{(2), \sigma}} \circ \pi_{T_{y} M_{\tau}^{(2), \sigma}}\right\|_{o p} \geq c_{k, d, \tau_{\min }}^{\prime}\left(\frac{\sigma}{n-1}\right)^{\frac{k-2}{k+d}} .
\end{aligned}
$$

Furthermore,

$$
\int_{\left(\mathbb{R}^{D}\right)^{n-1}}\left(\mathbf{P}_{\tau}^{(i), \sigma}\right)^{\otimes n-1} \wedge\left(\mathbf{P}_{\tau^{k}}^{(i), \sigma}\right)^{\otimes n-1} \geq c_{0}, \text { and } \quad m \cdot \int_{U_{k}^{\sigma}} \mathbf{P}_{\tau}^{(i), \sigma} \wedge \mathbf{P}_{\tau^{k}}^{(i), \sigma} \geq c_{d}
$$

Proof of Lemma C.16. Following the notation of Section C.2, for $i \in$ $\{1,2\}, \tau \in\{0,1\}^{m}, \delta \leq \tau_{\text {min }} / 4$ and $A>0$, consider

$$
\begin{equation*}
\Phi_{\tau}^{A, i}(x)=x+\sum_{k=1}^{m} \phi\left(\frac{x-x_{k}}{\delta}\right)\left\{\tau_{k} A\left(x-x_{k}\right)_{1}^{i}\right\} e . \tag{3}
\end{equation*}
$$

Note that (3) is a particular case of (1). Clearly from the definition, $\Phi_{\tau}^{A, i}$ and $\Phi_{\tau^{k}}^{A, i}$ coincide outside $\mathcal{B}\left(x_{k}, \delta\right),(\Phi(x)-x) \in \operatorname{span}(e)$ for all $x \in \mathbb{R}^{D}$, and $\left\|I_{D}-\Phi\right\|_{\infty} \leq A \delta^{i}$. Let us define $M_{\tau}^{A, i}=\Phi_{\tau}^{A, i}\left(M_{0}\right)$. From Lemma C.13, we have $M_{\tau}^{A, i} \in \mathcal{C}_{\tau_{\text {min }}, \mathbf{L}}^{k}$ provided that $\tau_{\min } L_{\perp}, \ldots, \tau_{\min }^{k-1} L_{k}$ are large enough, and that $\delta \leq \tau_{\min } / 2$, with $A / \delta^{k-i} \leq \varepsilon$ for $\varepsilon=\varepsilon_{k, d, \tau_{\min }, i}$ small enough.

Furthermore, let us write

$$
U_{k}^{\sigma}=\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(x_{k}, \delta / 2\right)+\mathcal{B}_{\{0\}^{d} \times \mathbb{R}^{D-d}}\left(x_{k}, \sigma / 2\right)
$$

Then the family $\left\{U_{k}^{\sigma}\right\}_{1 \leq k \leq m}$ is pairwise disjoint. Also, since $\tau_{k}=0$ implies that $M_{\tau}^{A, i}$ coincides with $M_{0}$ on $\mathcal{B}\left(x_{k}, \delta\right)$, we get that if $x \in U_{k}^{\sigma}$ and $y=$ $\pi_{M_{\tau}^{A, i}}(x)$,

$$
T_{y} M_{\tau}^{A, i}=\mathbb{R}^{d} \times\{0\}^{D-d} \quad, \quad\left\|I I_{y}^{M_{\tau}^{A, i}} \circ \pi_{T_{y} M_{\tau}^{A, i}}\right\|_{o p}=0 .
$$

Furthermore, by construction of the bump function $\Phi_{\tau}^{A, i}$, if $x \in U_{k}^{\sigma}$ and $\tau_{k}=1$, then

$$
\angle\left(T_{y} M_{\tau}^{A, i}, \mathbb{R}^{d} \times\{0\}^{D-d}\right) \geq \frac{A}{2}
$$

and

$$
\left\|I I_{y}^{M_{\tau}^{A, i}} \circ \pi_{T_{y} M_{\tau}^{A, i}}\right\|_{o p} \geq \frac{A}{2} .
$$

Now, let us write

$$
\mathcal{O}_{\tau}^{A, i}=\left\{y+\xi \mid y \in M_{\tau}^{A, i}, \xi \in\left(T_{y} M_{\tau}^{A, i}\right)^{\perp},\|\xi\| \leq \sigma / 2\right\}
$$

for the offset of $M_{\tau}^{\Lambda, A, i}$ of radius $\sigma / 2$. The sets $\left\{\mathcal{O}_{\tau}^{A, i}\right\}_{\tau}$ are closed subsets of $\mathbb{R}^{D}$ with non-empty interiors. Let $\mathbf{P}_{\tau}^{A, i}$ denote the uniform distribution on $\mathcal{O}_{\tau}^{A, i}$. Finally, let us denote by $P_{\tau}^{A, i}=\left(\pi_{M_{\tau}^{A, i}}\right)_{*} \mathbf{P}_{\tau}^{A, i}$ the pushforward distributions of $\mathbf{P}_{\tau}^{A, i}$ by the projection maps $\pi_{M_{\tau}^{A, i}}$. From Lemma 19 in [12], $P_{\tau}^{A, i}$ has a density $f_{\tau}^{A, i}$ with respect to the volume measure on $M_{\tau}^{A, i}$, and this density satisfies

$$
\operatorname{Vol}\left(M_{\tau}^{A, i}\right) f_{\tau}^{A, i} \leq\left(\frac{\tau_{\min }+\sigma / 2}{\tau_{\min }-\sigma / 2}\right)^{d} \leq\left(\frac{5}{3}\right)^{d}
$$

and

$$
\operatorname{Vol}\left(M_{\tau}^{A, i}\right) f_{\tau}^{A, i} \geq\left(\frac{\tau_{\min }-\sigma / 2}{\tau_{\min }+\sigma / 2}\right)^{d} \geq\left(\frac{3}{5}\right)^{d}
$$

Since, by construction, $\operatorname{Vol}\left(M_{0}\right)=c_{d} \tau_{\text {min }}^{d}$, and $c_{d}^{\prime} \leq \operatorname{Vol}\left(M_{\tau}^{\Lambda, A, i}\right) / \operatorname{Vol}\left(M_{0}\right) \leq$ $C_{d}^{\prime}$ whenever $A / \delta^{i-1} \leq \varepsilon_{d, \tau_{m i n}, i}^{\prime}$, we get that $P_{\tau}^{A, i}$ belongs to the model $\mathcal{P}^{k}$ provided that $\left(\tau_{\text {min }}^{d} f_{\text {min }}\right)^{-1}$ and $\tau_{\text {min }}^{d} f_{\text {max }}$ are large enough. This proves that under these conditions, the family $\left\{\mathbf{P}_{\tau}^{A, i}\right\}_{\tau \in\{0,1\}^{m}}$ is included in the model $\mathcal{P}^{k}(\sigma)$.

Let us now focus on the bounds on the $L^{1}$ test affinities. Let $\tau \in\{0,1\}^{m}$ and $1 \leq k \leq m$ be fixed, and assume, without loss of generality, that $\tau_{k}=0$ (if not, flip the role of $\tau$ and $\tau^{k}$ ). First, note that

$$
\int_{\left(\mathbb{R}^{D}\right)^{n-1}}\left(\mathbf{P}_{\tau}^{A, i}\right)^{\otimes n-1} \wedge\left(\mathbf{P}_{\tau^{k}}^{A, i}\right)^{\otimes n-1} \geq\left(\int_{\mathbb{R}^{D}} \mathbf{P}_{\tau}^{A, i} \wedge \mathbf{P}_{\tau^{k}}^{A, i}\right)^{n-1}
$$

Furthermore, since $\mathbf{P}_{\tau}^{A, i}$ and $\mathbf{P}_{\tau^{k}}^{A, i}$ are the uniform distributions on $\mathcal{O}_{\tau}^{A, i}$ and $\mathcal{O}_{\tau}^{A, i}$,

$$
\begin{aligned}
& \int_{\mathbb{R}^{D}} \mathbf{P}_{\tau}^{A, i} \wedge \mathbf{P}_{\tau^{k}}^{A, i}=1-\frac{1}{2} \int_{\mathbb{R}^{D}}\left|\mathbf{P}_{\tau}^{A, i}-\mathbf{P}_{\tau^{k}}^{A, i}\right| \\
& =1-\frac{1}{2} \int_{\mathbb{R}^{D}}\left|\frac{\mathbb{1}_{\mathcal{O}_{\tau}^{A, i}}(a)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right)}-\frac{\mathbb{1}_{\mathcal{O}_{\tau}^{A, i}}(a)}{\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)}\right| d \mathcal{H}^{D}(a) \text {. }
\end{aligned}
$$

Furthermore,

$$
\begin{aligned}
\frac{1}{2} \int_{\mathbb{R}^{D}} \left\lvert\, \frac{\mathbb{1}_{\mathcal{O}_{\tau}^{A, i}}(a)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right)}\right. & \left.-\frac{\mathbb{O}_{\mathcal{O}^{k}}^{A, i}(a)}{\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)} \right\rvert\, d \mathcal{H}^{D}(a) \\
= & \frac{1}{2} \operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i} \cap \mathcal{O}_{\tau^{k}}^{A, i}\right)\left|\frac{1}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right)}-\frac{1}{\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)}\right| \\
& +\frac{1}{2}\left(\frac{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right)}+\frac{\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i} \backslash \mathcal{O}_{\tau^{\prime}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)}\right) \\
\leq & \frac{3}{2} \frac{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}\right) \vee \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i} \backslash \mathcal{O}_{\tau^{\prime}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right) \wedge \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)}
\end{aligned}
$$

To get a lower bound on the denominator, note that for $\delta \leq \tau_{\min } / 2, M_{\tau}^{A, i}$ and $M_{\tau^{k}}^{A, i}$ both contain

$$
\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(0, \tau_{\min }\right) \backslash \mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(0, \tau_{\min } / 4\right),
$$

so that $\mathcal{O}_{\tau}^{A, i}$ and $\mathcal{O}_{\tau^{k}}^{A, i}$ both contain

$$
\left(\mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(0, \tau_{\text {min }}\right) \backslash \mathcal{B}_{\mathbb{R}^{d} \times\{0\}^{D-d}}\left(0, \tau_{\text {min }} / 4\right)\right)+\mathcal{B}_{\{0\}^{d} \times \mathbb{R}^{D-d}}(0, \sigma / 2) .
$$

As a consequence, $\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right) \wedge \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right) \geq c_{d} \omega_{d} \tau_{\text {min }}^{d} \omega_{D-d}(\sigma / 2)^{D-d}$, where $\omega_{\ell}$ denote the volume of a $\ell$-dimensional unit Euclidean ball.
 consider $a_{0}=y+\xi \in \mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}$, with $y \in M_{\tau}^{A, i}$ and $\xi \in\left(T_{y} M_{\tau}^{A, i}\right)^{\perp}$. Since $\Phi_{\tau}^{A, i}$ and $\Phi_{\tau^{k}}^{A, i}$ coincide outside $\mathcal{B}\left(x_{k}, \delta\right)$, so do $M_{\tau}^{A, i}$ and $M_{\tau^{k}}^{A, i}$. Hence, one necessarily has $y \in \mathcal{B}\left(x_{k}, \delta\right)$. Thus, $\left(T_{y} M_{\tau}^{A, i}\right)^{\perp}=T_{y} M_{0}^{\perp}=\operatorname{span}(e)+$ $\{0\}^{d+1} \times \mathbb{R}^{D-d-1}$, so we can write $\xi=s e+z$ with $s \in \mathbb{R}$ and $z \in\{0\}^{d+1} \times$ $\mathbb{R}^{D-d-1}$. By definition of $\mathcal{O}_{\tau}^{A, i},\|\xi\|=\sqrt{s^{2}+\|z\|^{2}} \leq \sigma / 2$, which yields $\|z\| \leq$ $\sigma / 2$ and $|s| \leq \sqrt{(\sigma / 2)^{2}-\|z\|^{2}}$. Furthermore, $y_{0}$ does not belong to $\mathcal{O}_{\tau^{k}}^{A, i}$, which translates to

$$
\begin{aligned}
\sigma / 2<d\left(a_{0}, M_{\tau^{k}}^{A, i}\right) & \leq\left\|y_{0}+s e+z-\Phi_{\tau^{k}}^{A, i}\left(y_{0}\right)\right\| \\
& =\sqrt{\left|s+\left\langle e, y_{0}-\Phi_{\tau^{k}}^{A, i}\left(y_{0}\right)\right\rangle\right|^{2}+\|z\|^{2}}
\end{aligned}
$$

from what we get $|s| \geq \sqrt{(\sigma / 2)^{2}-\|z\|^{2}}-\left\|I_{D}-\Phi_{\tau^{k}}^{A, i}\right\|_{\infty}$. We just proved that $\mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}$ is a subset of

$$
\begin{aligned}
\mathcal{B}_{d}\left(x_{k}, \delta\right)+ & \left\{s e+z \mid(s, z) \in \mathbb{R} \times \mathbb{R}^{D-d-1},\|z\| \leq \sigma / 2\right. \text { and } \\
& \left.\sqrt{(\sigma / 2)^{2}-\|z\|^{2}}-\left\|I_{D}-\Phi_{\tau^{k}}^{A, i}\right\|_{\infty} \leq|s| \leq \sqrt{(\sigma / 2)^{2}-\|z\|^{2}}\right\} .
\end{aligned}
$$

Hence,

$$
\begin{equation*}
\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}\right) \leq \omega_{d} \delta^{d} \times 2\left\|I_{D}-\Phi_{\tau^{k}}^{A, i}\right\|_{\infty} \times \omega_{D-d-1}(\sigma / 2)^{D-d-1} \tag{4}
\end{equation*}
$$

Similar arguments lead to

$$
\begin{equation*}
\operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i} \backslash \mathcal{O}_{\tau}^{A, i}\right) \leq \omega_{d} \delta^{d} \times 2\left\|I_{D}-\Phi_{\tau}^{A, i}\right\|_{\infty} \times \omega_{D-d-1}(\sigma / 2)^{D-d-1} \tag{5}
\end{equation*}
$$

Since $\left\|I_{D}-\Phi_{\tau}^{A, i}\right\|_{\infty} \vee\left\|I_{D}-\Phi_{\tau^{k}}^{A, i}\right\|_{\infty} \leq A \delta^{i}$, summing up bounds (4) and (5) yields

$$
\begin{aligned}
\int_{\mathbb{R}^{D}} \mathbf{P}_{\tau}^{A, i} \wedge \mathbf{P}_{\tau^{k}}^{A, i} & \geq 1-3 \frac{\omega_{d} \omega_{D-d-1} A \delta^{i} \cdot \delta^{d}(\sigma / 2)^{D-d-1}}{\omega_{d} \tau_{\min }^{d} \omega_{D-d}(\sigma / 2)^{D-d}} \\
& \geq 1-3 \frac{A \delta^{i}}{\sigma}\left(\frac{\delta}{\tau_{\min }}\right)^{d}
\end{aligned}
$$

To derive the last bound, we notice that since $U_{k}^{\sigma} \subset \mathcal{O}_{\tau}^{A, i}=\operatorname{Supp}\left(\mathbf{P}_{\tau}^{A, i}\right)$, we have

$$
\begin{aligned}
\int_{U_{k}^{\sigma}} \mathbf{P}_{\tau}^{A, i} \wedge \mathbf{P}_{\tau^{k}}^{A, i} & \geq \frac{\operatorname{Vol}\left(U_{k}^{\sigma} \cap \mathcal{O}_{\tau^{k}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right) \wedge \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)} \\
& \geq \frac{\operatorname{Vol}\left(U_{k}^{\sigma}\right)-\operatorname{Vol}\left(U_{k}^{\sigma} \backslash \mathcal{O}_{\tau^{k}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right) \wedge \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)} \\
& \geq \frac{\operatorname{Vol}\left(U_{k}^{\sigma}\right)-\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i} \backslash \mathcal{O}_{\tau^{k}}^{A, i}\right)}{\operatorname{Vol}\left(\mathcal{O}_{\tau}^{A, i}\right) \wedge \operatorname{Vol}\left(\mathcal{O}_{\tau^{k}}^{A, i}\right)} \\
& \geq \frac{\omega_{d}(\delta / 2)^{d} \omega_{D-d}(\sigma / 2)^{D-d}-\omega_{d} \delta^{d} A \delta^{i} \omega_{D-d-1}(\sigma / 2)^{D-d-1}}{\omega_{d} \tau_{\min }^{d} \omega_{D-d}(\sigma / 2)^{D-d}}
\end{aligned}
$$

Hence, whenever $A \delta^{i} \leq c_{d} \sigma$ for $c_{d}$ small enough, we get

$$
\int_{U_{k}^{\sigma}} \mathbf{P}_{\tau}^{A, i} \wedge \mathbf{P}_{\tau^{k}}^{A, i} \geq c_{d}^{\prime}\left(\frac{\delta}{\tau_{\min }}\right)^{d}
$$

Since $m$ can be chosen such that $m \geq c_{d}\left(\tau_{\min } / \delta\right)^{d}$, we get the last bound.
Eventually, writting $\mathbf{P}_{\tau}^{(i), \sigma}=\mathbf{P}_{\tau}^{A, i}$ for the particular parameters $A=$ $\varepsilon \delta^{k-i}$, for $\varepsilon=\varepsilon_{k, d, \tau_{m i n}}$ small enough, and $\delta$ such that $\frac{3 A \delta^{i}}{\sigma}\left(\frac{\delta}{\tau_{m i n}}\right)^{d}=\frac{1}{n-1}$ yields the result. Such a choice of parameter $\delta$ does meet the condition $A \delta^{i}=\varepsilon \delta^{k} \leq c_{d} \sigma$, provided that $\sigma \geq \frac{c_{d}}{\varepsilon}\left(\frac{1}{n-1}\right)^{k / d}$.

## C.4. Hypotheses for Manifold Estimation

## C.4.1. Proof of Lemma 5

Let us prove Lemma 5, stated here as Lemma C.17.
Lemma C.17. If $\tau_{\min } L_{\perp}, \ldots, \tau_{\min }^{k-1} L_{k},\left(\tau_{\min }^{d} f_{\min }\right)^{-1}$ and $\tau_{\min }^{d} f_{\max }$ are large enough (depending only on $d$ and $k$ ), there exist $P_{0}, P_{1} \in \mathcal{P}^{k}$ with associated submanifolds $M_{0}, M_{1}$ such that

$$
d_{H}\left(M_{0}, M_{1}\right) \geq c_{k, d, \tau_{m i n}}\left(\frac{1}{n}\right)^{\frac{k}{d}}, \text { and } \quad\left\|P_{0} \wedge P_{1}\right\|_{1}^{n} \geq c_{0} .
$$

Proof of Lemma C.17. Following the notation of Section C.2, for $\delta \leq$ $\tau_{\text {min }} / 4$ and $\Lambda>0$, consider

$$
\Phi_{\tau}^{\Lambda}(x)=x+\phi\left(\frac{x}{\delta}\right) \Lambda \cdot e
$$

which is a particular case of (1). Define $M^{\Lambda}=\Phi^{\Lambda}\left(M_{0}\right)$, and $P^{\Lambda}=\Phi_{*}^{\Lambda} P_{0}$. Under the conditions of Lemma C.13, $P_{0}$ and $P^{\Lambda}$ belong to $\mathcal{P}^{k}$, and by construction, $d_{H}\left(M_{0}, M^{\Lambda}\right)=\Lambda$. In addition, since $P_{0}$ and $P^{\Lambda}$ coincide outside $\mathcal{B}(0, \delta)$,

$$
\int_{\mathbb{R}^{D}} d P_{0} \wedge d P^{\Lambda}=P_{0}(\mathcal{B}(0, \delta))=\omega_{d}\left(\frac{\delta}{\tau_{\min }}\right)^{d}
$$

Setting $P_{1}=P^{\Lambda}$ with $\omega_{d}\left(\frac{\delta}{\tau_{\min }}\right)^{d}=\frac{1}{n}$ and $\Lambda=c_{k, d, \tau_{\min }} \delta^{k}$ for $c_{k, d, \tau_{m i n}}>0$ small enough yields the result.

## C.4.2. Proof of Lemma 6

Here comes the proof of Lemma 6, stated here as Lemma C.17.
LEMMA C.18. If $\tau_{\min } L_{\perp}, \ldots, \tau_{\min }^{k-1} L_{k},\left(\tau_{\min }^{d} f_{\min }\right)^{-1}$ and $\tau_{\min }^{d} f_{\max }$ are large enough (depending only on $d$ and $k$ ), there exist $P_{0}^{\sigma}, P_{1}^{\sigma} \in \mathcal{P}^{k}(\sigma)$ with associated submanifolds $M_{0}^{\sigma}, M_{1}^{\sigma}$ such that

$$
d_{H}\left(M_{0}^{\sigma}, M_{1}^{\sigma}\right) \geq c_{k, d, \tau_{\min }}\left(\frac{\sigma}{n}\right)^{\frac{k}{d+k}}, \text { and }\left\|P_{0}^{\sigma} \wedge P_{1}^{\sigma}\right\|_{1}^{n} \geq c_{0}
$$

Proof of Lemma C.18. The proof follows the lines of that of Lemma C.16. Indeed, with the notation of Section C.2, for $\delta \leq \tau_{\text {min }} / 4$ and $0<\Lambda \leq$ $c_{k, d, \tau_{\text {min }}} \delta^{k}$ for $c_{k, d, \tau_{\text {min }}}>0$ small enough, consider

$$
\Phi_{\tau}^{\Lambda}(x)=x+\phi\left(\frac{x}{\delta}\right) \Lambda \cdot e .
$$

Define $M^{\Lambda}=\Phi^{\Lambda}\left(M_{0}\right)$. Write $\mathcal{O}_{0}, \mathcal{O}^{\Lambda}$ for the offsets of radii $\sigma / 2$ of $M_{0}, M^{\Lambda}$, and and $\mathbf{P}_{0}, \mathbf{P}^{\Lambda}$ for the uniform distributions on these sets.

By construction, we have $d_{H}\left(M_{0}, M^{\Lambda}\right)=\Lambda$, and as in the proof of Lemma C.16, we get

$$
\int_{\mathbb{R}^{D}} \mathbf{P}_{0} \wedge \mathbf{P}^{\Lambda} \geq 1-3 \frac{\Lambda}{\sigma}\left(\frac{\delta}{\tau_{\min }}\right)^{d}
$$

Denoting $P_{0}^{\sigma}=\mathbf{P}_{0}$ and $P_{1}^{\sigma}=\mathbf{P}^{\Lambda}$ with $\Lambda=\varepsilon_{k, d, \tau_{\text {min }}} \delta^{k}$ and $\delta$ such that $3 \frac{\Lambda}{\sigma}\left(\frac{\delta}{\tau_{\text {min }}}\right)^{d}$ yields the result.

## C.5. Minimax Inconsistency Results

This section is devoted to the proof of Theorem 1, reproduced here as Theorem C.19.

Theorem C.19. Assume that $\tau_{\text {min }}=0$. If $D \geq d+3$, then, for all $k \geq 2$ and $L_{\perp}>0$, provided that $L_{3} / L_{\perp}^{2}, \ldots, L_{k} / L_{\perp}^{k-1}, L_{\perp}^{d} / f_{\min }$ and $f_{\max } / L_{\perp}^{d}$ are large enough (depending only on $d$ and $k$ ), for all $n \geq 1$,

$$
\inf _{\hat{T}} \sup _{P \in \mathcal{P}_{(x)}^{k}} \mathbb{E}_{P \otimes_{n}} \angle\left(T_{x} M, \hat{T}\right) \geq \frac{1}{2}>0
$$

where the infimum is taken over all the estimators $\hat{T}=\hat{T}\left(X_{1}, \ldots, X_{n}\right)$.
Moreover, for any $D \geq d+1$, provided that $L_{3} / L_{\perp}^{2}, \ldots, L_{k} / L_{\perp}^{k-1}, L_{\perp}^{d} / f_{\text {min }}$ and $f_{\max } / L_{\perp}^{d}$ are large enough (depending only on $d$ and $k$ ), for all $n \geq 1$,

$$
\inf _{\widehat{I I}} \sup _{P \in \mathcal{P}_{(x)}^{k}} \mathbb{E}_{P \otimes n}\left\|I I_{x}^{M} \circ \pi_{T_{x} M}-\widehat{I I}\right\|_{o p} \geq \frac{L_{\perp}}{4}>0
$$

where the infimum is taken over all the estimators $\widehat{I I}=\widehat{I I}\left(X_{1}, \ldots, X_{n}\right)$.
We will make use of Le Cam's Lemma, which we recall here.
Theorem C. 20 (Le Cam's Lemma [14]). For all pairs $P, P^{\prime}$ in $\mathcal{P}$,

$$
\inf _{\hat{\theta}} \sup _{P \in \mathcal{P}} \mathbb{E}_{P \otimes n} d(\theta(P), \hat{\theta}) \geq \frac{1}{2} d\left(\theta(P), \theta\left(P^{\prime}\right)\right)\left\|P \wedge P^{\prime}\right\|_{1}^{n}
$$

where the infimum is taken over all the estimators $\hat{\theta}=\hat{\theta}\left(X_{1}, \ldots, X_{n}\right)$.
Proof of Theorem C.19. For $\delta \geq \Lambda>0$, let $\mathcal{C}, \mathcal{C}^{\prime} \subset \mathbb{R}^{3}$ be closed curves of the Euclidean space as in Figure 2, and such that outside the figure, $\mathcal{C}$ and $\mathcal{C}^{\prime}$ coincide and are $\mathcal{C}^{\infty}$. The bumped parts are obtained with a smooth diffeomorphism similar to (1) and centered at $x$. Here, $\delta$ and $\Lambda$ can be chosen arbitrarily small.

Let $\mathcal{S}^{d-1} \subset \mathbb{R}^{d}$ be a $d-1$-sphere of radius $1 / L_{\perp}$. Consider the Cartesian products $M_{1}=\mathcal{C} \times \mathcal{S}^{d-1}$ and $M_{1}^{\prime}=\mathcal{C}^{\prime} \times \mathcal{S}^{d-1} . M_{1}$ and $M_{1}^{\prime}$ are subsets of $\mathbb{R}^{d+3} \subset \mathbb{R}^{D}$. Finally, let $P_{1}$ and $P_{1}^{\prime}$ denote the uniform distributions on $M$ and $M^{\prime}$. Note that $M, M^{\prime}$ can be built by homothecy of ratio $\lambda=$ $1 / L_{\perp}$ from some unitary scaled $M_{1}^{(0)}, M^{\prime(0)}$, similarly to Section 5.3.2 in [2], yielding, from Proposition A.4, that $P_{1}, P_{1}^{\prime}$ belong to $\mathcal{P}_{(x)}^{k}$ provided that $L_{3} / L_{\perp}^{2}, \ldots, L_{k} / L_{\perp}^{k-1}, L_{\perp}^{d} / f_{\text {min }}$ and $f_{\max } / L_{\perp}^{d}$ are large enough (depending


Figure 2: Hypotheses for minimax lower bound on tangent space estimation with $\tau_{\text {min }}=0$.
only on $d$ and $k$ ), and that $\Lambda, \delta$ and $\Lambda^{k} / \delta$ are small enough. From Le Cam's Lemma C.20, we have for all $n \geq 1$,

$$
\inf _{\hat{T}} \sup _{P \in \mathcal{P}_{(x)}^{k}} \mathbb{E}_{P \otimes n} \angle\left(T_{x} M, \hat{T}\right) \geq \frac{1}{2} \angle\left(T_{x} M_{1}, T_{x} M_{1}^{\prime}\right)\left\|P_{1} \wedge P_{1}^{\prime}\right\|_{1}^{n}
$$

By construction, $\angle\left(T_{x} M_{1}, T_{x} M_{1}^{\prime}\right)=1$, and since $\mathcal{C}$ and $\mathcal{C}^{\prime}$ coincide outside $\mathcal{B}_{\mathbb{R}^{3}}(0, \delta)$,

$$
\begin{aligned}
\left\|P_{1} \wedge P_{1}^{\prime}\right\|_{1} & =1-\operatorname{Vol}\left(\left(\mathcal{B}_{\mathbb{R}^{3}}(0, \delta) \cap \mathcal{C}\right) \times \mathcal{S}^{d-1}\right) / \operatorname{Vol}\left(\mathcal{C} \times \mathcal{S}^{d-1}\right) \\
& =1-\operatorname{Length}\left(\mathcal{B}_{\mathbb{R}^{3}}(0, \delta) \cap \mathcal{C}\right) / \operatorname{Length}(\mathcal{C}) \\
& \geq 1-c_{L_{\perp}} \delta .
\end{aligned}
$$

Hence, at fixed $n \geq 1$, letting $\Lambda, \delta$ go to 0 with $\Lambda^{k} / \delta$ small enough, we get the announced bound.

We now tackle the lower bound on curvature estimation with the same strategy. Let $M_{2}, M_{2}^{\prime} \subset \mathbb{R}^{D}$ be $d$-dimensional submanifolds as in Figure 3: they both contain $x$, the part on the top of $M_{2}$ is a half $d$-sphere of radius $2 / L_{\perp}$, the bottom part of $M_{2}^{\prime}$ is a piece of a $d$-plane, and the bumped parts are obtained with a smooth diffeomorphism similar to (1), centered at $x$. Outside $\mathcal{B}(x, \delta), M_{2}$ and $M_{2}^{\prime}$ coincide and connect smoothly the upper and lower parts. Let $P_{2}, P_{2}^{\prime}$ be the probability distributions obtained by the pushforward given by the bump maps. Under the same conditions on the parameters as previously, $P_{2}$ and $P_{2}^{\prime}$ belong to $\mathcal{P}_{(x)}^{k}$ according to Proposition


Figure 3: Hypotheses for minimax lower bound on curvature estimation with $\tau_{\text {min }}=0$.
A.4. Hence from Le Cam's Lemma C. 20 we deduce

$$
\begin{aligned}
\underset{\widehat{I I}}{\inf } \sup _{P \in \mathcal{P}_{(x)}^{k}} \mathbb{E}_{P \otimes n} & \left\|I I_{x}^{M} \circ \pi_{T_{x} M}-\widehat{I I}\right\|_{o p} \\
& \geq \frac{1}{2}\left\|I I_{x}^{M_{2}} \circ \pi_{T_{x} M_{2}}-I I_{x}^{M_{2}^{\prime}} \circ \pi_{T_{x} M_{2}^{\prime}}\right\|_{o p}\left\|P_{2} \wedge P_{2}^{\prime}\right\|_{1}^{n} .
\end{aligned}
$$

But by construction, $\left\|I I_{x}^{M_{2}} \circ \pi_{T_{x} M_{2}}\right\|_{o p}=0$, and since $M_{2}^{\prime}$ is a part of a sphere of radius $2 / L_{\perp}$ nearby $x,\left\|I I_{x}^{M_{2}^{\prime}} \circ \pi_{T_{x} M_{2}^{\prime}}\right\|_{o p}=L_{\perp} / 2$. Hence,

$$
\left\|I I_{x}^{M_{2}} \circ \pi_{T_{x} M_{2}}-I I_{x}^{M_{2}^{\prime}} \circ \pi_{T_{x} M_{2}^{\prime}}\right\|_{o p} \geq L_{\perp} / 2 .
$$

Moreover, since $P_{2}$ and $P_{2}^{\prime}$ coincide on $\mathbb{R}^{D} \backslash \mathcal{B}(x, \delta)$,

$$
\left\|P_{2} \wedge P_{2}^{\prime}\right\|_{1}=1-P_{2}(\mathcal{B}(x, \delta)) \geq 1-c_{d, L_{\perp}} \delta^{d} .
$$

At $n \geq 1$ fixed, letting $\Lambda, \delta$ go to 0 with $\Lambda^{k} / \delta$ small enough, we get the desired result.

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