

AATRN Applied Topology Seminar – 11/06/2025

Detection of representation orbits of compact Lie groups from point clouds

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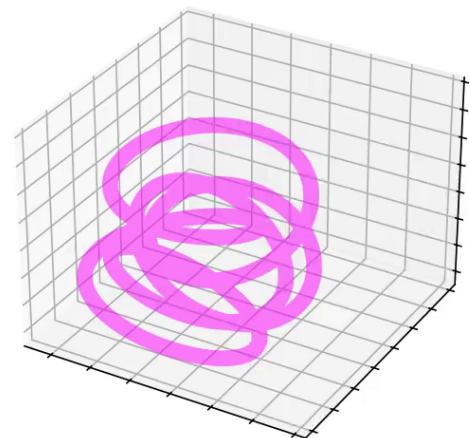
Henrique Ennes – INRIA Sophia



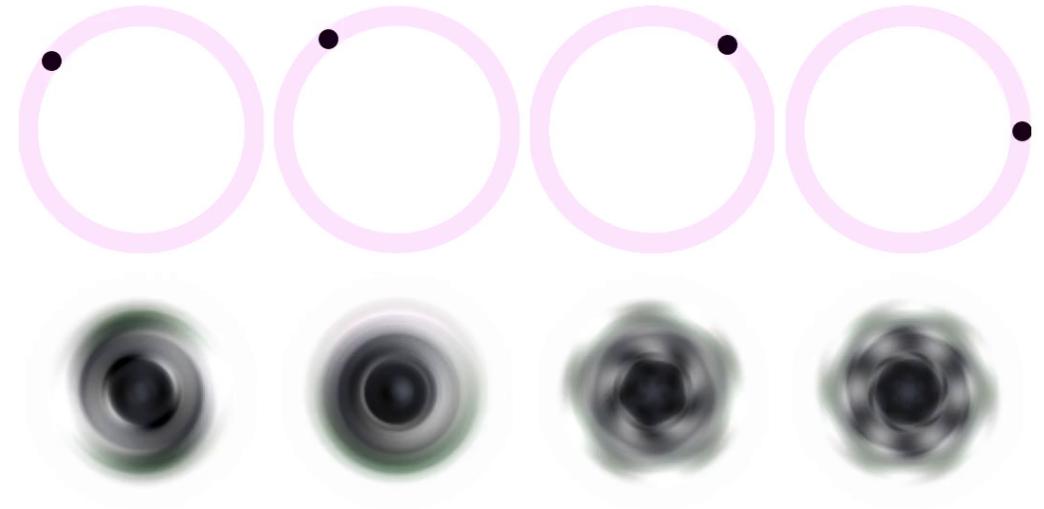
Rotations of $m \times m$ RGB image



Embedding in $\mathbb{R}^{m \times m \times 3}$



Projection in eigenplanes



Eigenvalues of the point cloud's covariance matrix:

311.2, 311.2, 221.3, 221.3, 82.3, 82.3, 79.4, 79.4, ...

In these eigenplanes, the orbit is close to

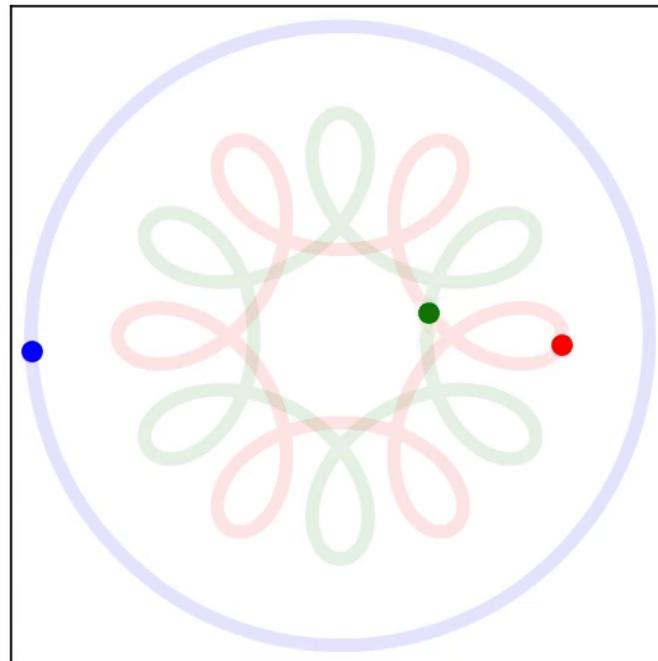
$$\theta \mapsto \begin{pmatrix} \mu_1 \cos \omega_1 \theta \\ \mu_1 \sin \omega_1 \theta \\ \mu_2 \cos \omega_2 \theta \\ \mu_2 \sin \omega_2 \theta \\ \vdots \\ \mu_k \cos \omega_k \theta \\ \mu_k \sin \omega_k \theta \end{pmatrix} = \begin{pmatrix} \cos \omega_1 \theta & -\sin \omega_1 \theta \\ \sin \omega_1 \theta & \cos \omega_1 \theta \\ & & \cos \omega_2 \theta & -\sin \omega_2 \theta \\ & & \sin \omega_2 \theta & \cos \omega_2 \theta \\ & & & & \ddots \\ & & & & \cos \omega_k \theta & -\sin \omega_k \theta \\ & & & & \sin \omega_k \theta & \cos \omega_k \theta \end{pmatrix} \begin{pmatrix} \mu_1 \\ 0 \\ \mu_2 \\ 0 \\ \vdots \\ \mu_k \\ 0 \end{pmatrix}$$

In 1975, Roger Broucke found several periodic orbits.

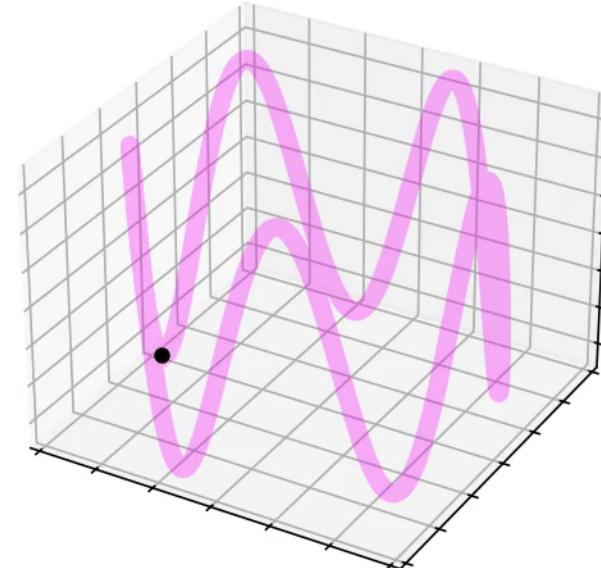
Let $x_1(t)$, $x_2(t)$, $x_3(t)$ be the three bodies, and define $z(t) = (x_1(t), x_2(t), x_3(t)) \in \mathbb{R}^6$.

Orbit A3

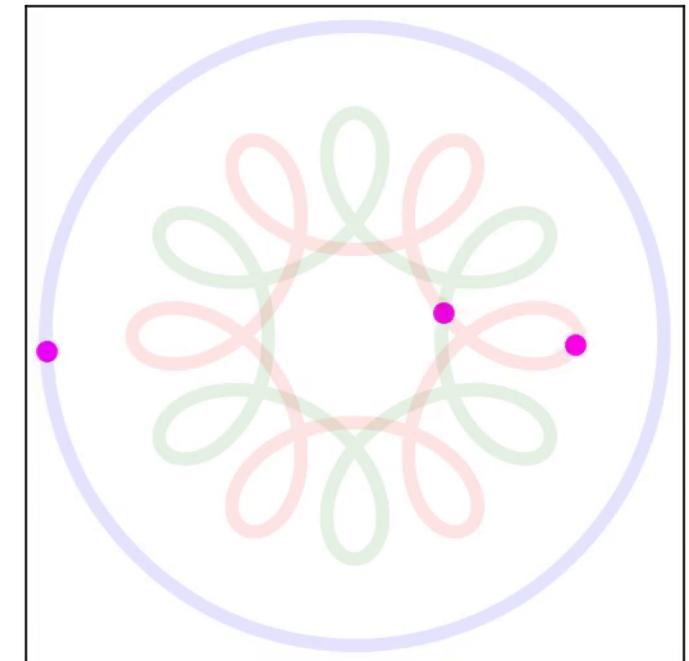
Trajectory of x_1, x_2, x_3
(found by integration)



Trajectory of z



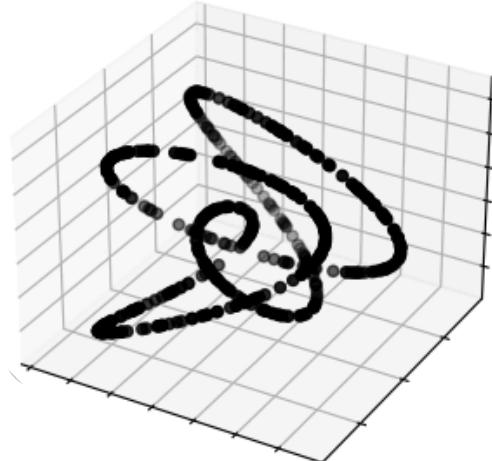
Reconstructed orbit of $\text{SO}(2)$



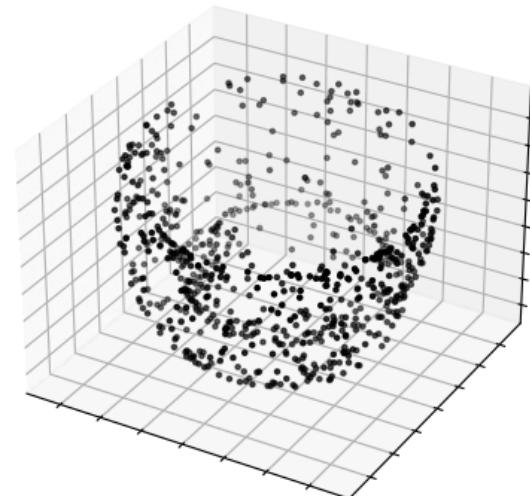
Input: A point cloud $X = \{x_1, \dots, x_N\} \subset \mathbb{R}^n$.

Output: A compact Lie group G , a representation ϕ in \mathbb{R}^n , and an orbit \mathcal{O} close to X .

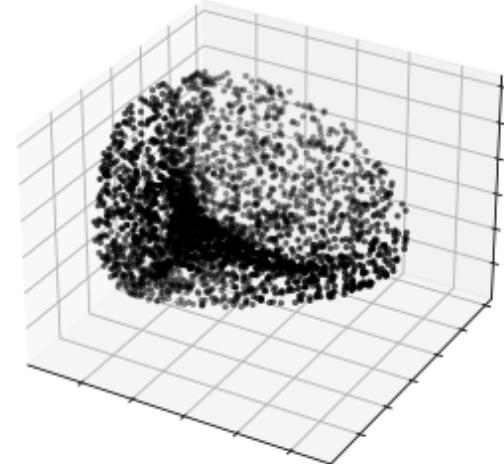
Orbit of $\text{SO}(2)$ in \mathbb{R}^6



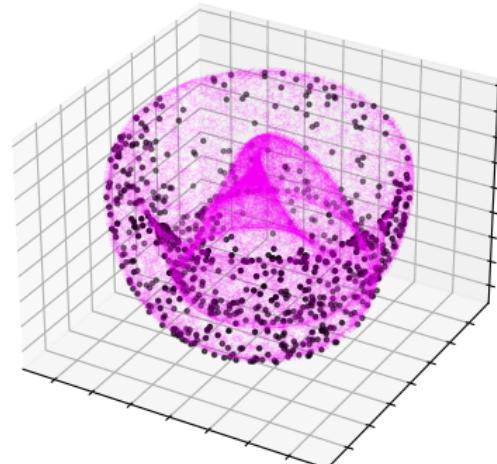
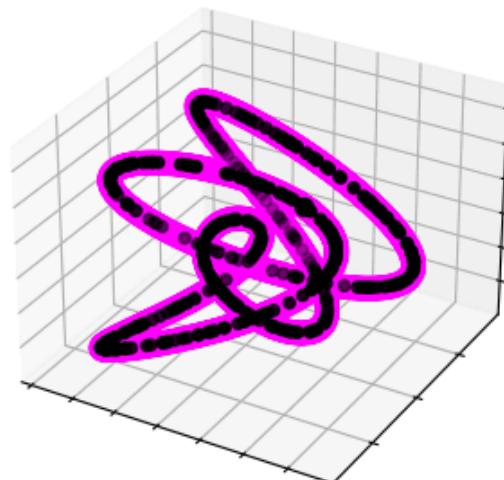
Orbit of T^2 in \mathbb{R}^6



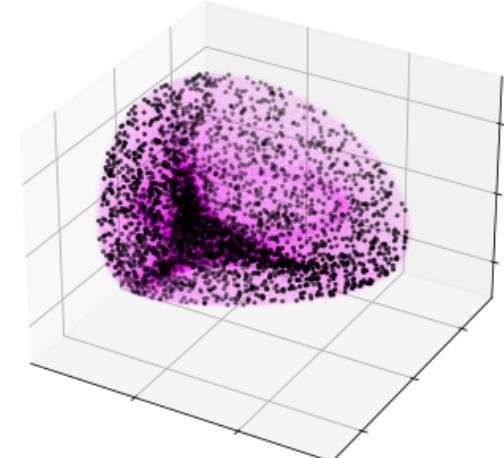
Orbit of $\text{SO}(3)$ in \mathbb{R}^9



Input:



Output:



A **Lie group** is a smooth manifold endowed with a group operation, and such that $(g, h) \mapsto gh^{-1}$ is smooth.

- $\mathrm{GL}_n(\mathbb{R})$ general linear group: the $n \times n$ invertible matrices.
- $\mathrm{O}(n)$ orthogonal group: the $n \times n$ orthogonal matrices ($A^\top = A^{-1}$).
- $\mathrm{SO}(n)$ special orthogonal group: the $n \times n$ orthogonal matrices with determinant +1.
- $\mathrm{U}(n)$ unitary group: $n \times n$ (complex) unitary matrices ($A^* = A^{-1}$).
- $\mathrm{SU}(n)$ special unitary group: $n \times n$ (complex) unitary matrices with determinant +1.
- T^d d -torus: the product $\mathrm{SO}(2) \times \cdots \times \mathrm{SO}(2)$.

A (real) **representation** of dimension n is a smooth homomorphism $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$.

- If G is a matrix group, the natural embedding $G \rightarrow \mathrm{GL}_n(\mathbb{R})$ is a representation.
- For $G = \mathrm{SO}(2)$ and $\omega \in \mathbb{Z}$, one has $\theta \mapsto \begin{pmatrix} \cos \omega\theta & -\sin \omega\theta \\ \sin \omega\theta & \cos \omega\theta \end{pmatrix}$.

A representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$ is **irreducible** (irrep) if no non-trivial subspace $V \subset \mathbb{R}^n$ is stabilized.

Fact: Every representation ϕ is equivalent to a sum of irreps. That is, one has a decomposition $\mathbb{R}^n = V_1 \oplus \cdots \oplus V_k$, irreps $\phi_i: G \rightarrow V_i$ and a change of basis $A \in \mathrm{GL}_n(\mathbb{R})$ such that

$$A\phi A^{-1} = \phi_1 \oplus \cdots \oplus \phi_k.$$

Irreps can be explicitly enumerated:

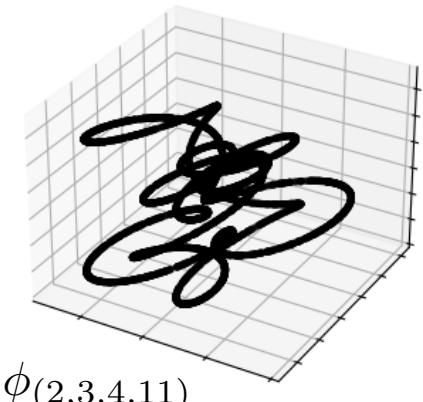
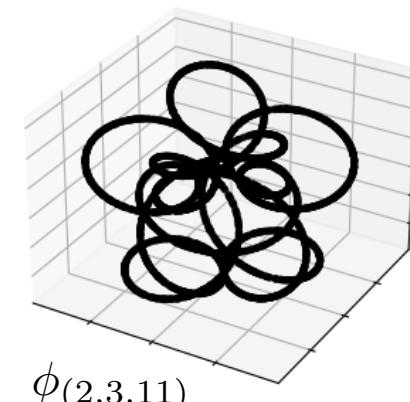
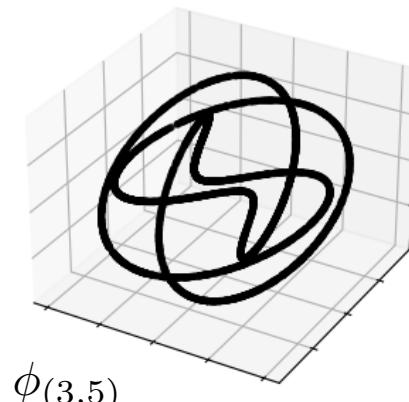
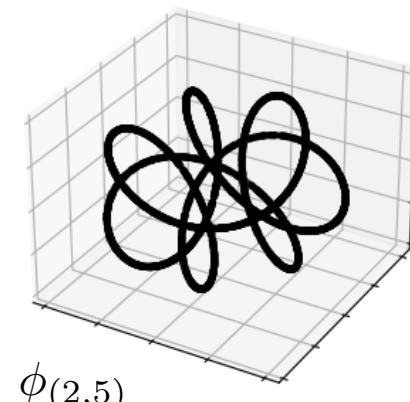
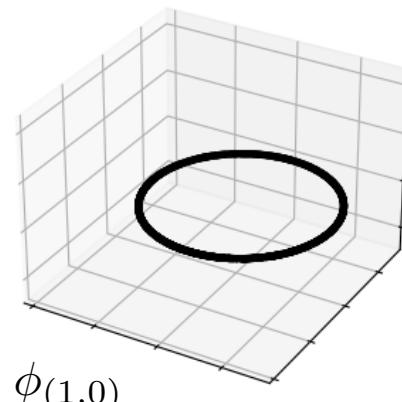
- $\mathrm{SO}(2)$ the $\theta \mapsto R(\omega\theta)$ for $\omega \in \mathbb{Z} \setminus \{0\}$, where $R(\theta) \mapsto \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$.
- T^d the $(\theta_i)_{i=1}^d \mapsto R(\sum_{i=1}^d \omega_i \theta_i)$ for $(\omega_i)_{i=1}^d \in \mathbb{Z}^d \setminus \{0\}$.
- $\mathrm{SO}(3)$ one irrep in \mathbb{R}^n for n odd.
- $\mathrm{SU}(2)$ one irrep in \mathbb{R}^n for n odd or $n \equiv 0 \pmod{4}$.

The **orbit** of $x_0 \in \mathbb{R}^n$ under a representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$ is $\mathcal{O} = \{\phi(g)x_0 \mid g \in G\}$.

Example: Orbits of $\mathrm{SO}(2)$ in \mathbb{R}^{2k} .

Let us write $\phi \simeq \phi_{\omega_1} \oplus \cdots \oplus \phi_{\omega_k}$. The orbit is made of the points

$$\phi(\theta)x_0 = \begin{pmatrix} \cos \omega_1 \theta & -\sin \omega_1 \theta \\ \sin \omega_1 \theta & \cos \omega_1 \theta \\ & \ddots \\ & & \cos \omega_2 \theta & -\sin \omega_2 \theta \\ & & \sin \omega_2 \theta & \cos \omega_2 \theta \\ & & & \ddots \\ & & & & \cos \omega_k \theta & -\sin \omega_k \theta \\ & & & & \sin \omega_k \theta & \cos \omega_k \theta \end{pmatrix} x_0$$



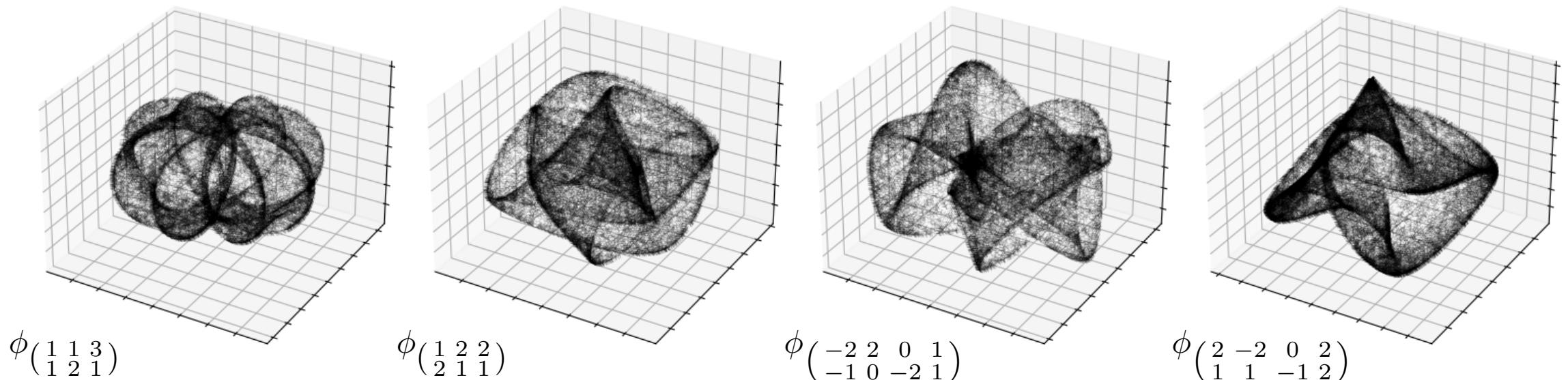
The **orbit** of $x_0 \in \mathbb{R}^n$ under a representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$ is $\mathcal{O} = \{\phi(g)x_0 \mid g \in G\}$.

Example: Orbits of T^2 in \mathbb{R}^{2k} .

Let us write

$$\phi \simeq \phi \begin{pmatrix} \omega_1^{(1)} \\ \omega_1^{(2)} \end{pmatrix} \oplus \cdots \oplus \phi \begin{pmatrix} \omega_k^{(1)} \\ \omega_k^{(2)} \end{pmatrix}$$

One builds the integer matrix of weights $\begin{pmatrix} \omega_1^{(1)} & \dots & \omega_k^{(1)} \\ \omega_1^{(2)} & \dots & \omega_k^{(2)} \end{pmatrix}$.



The **orbit** of $x_0 \in \mathbb{R}^n$ under a representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$ is $\mathcal{O} = \{\phi(g)x_0 \mid g \in G\}$.

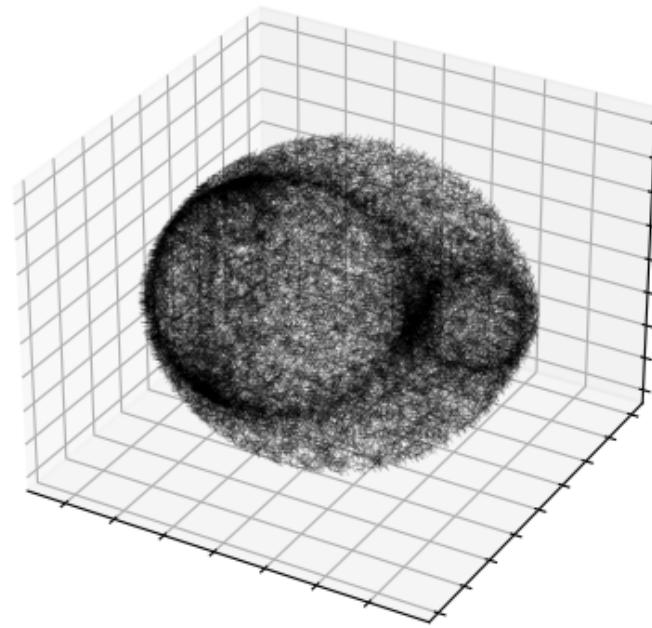
Example: Orbits of $\mathrm{SO}(3)$ and $\mathrm{SU}(2)$ in \mathbb{R}^n .

$\mathrm{SO}(3)$ has a finite number of (equivalence classes) of representations in \mathbb{R}^n : one for each decomposition

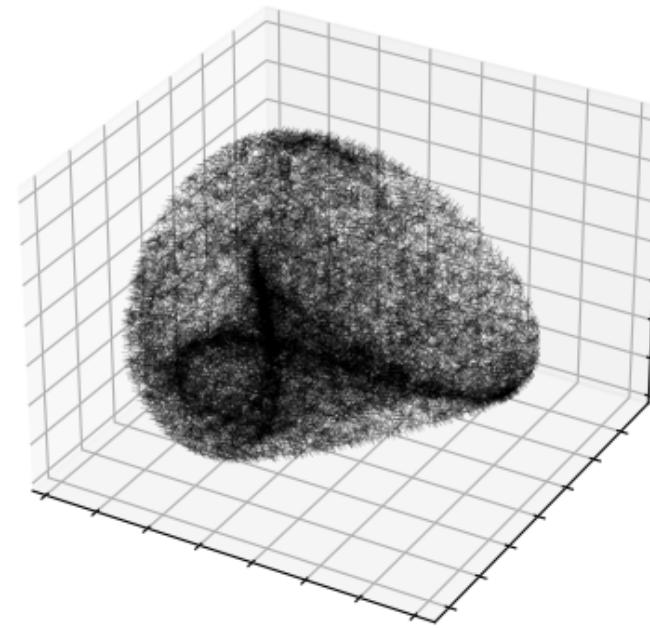
$$n = \omega_1 + \cdots + \omega_k$$

where the ω_i are odd.

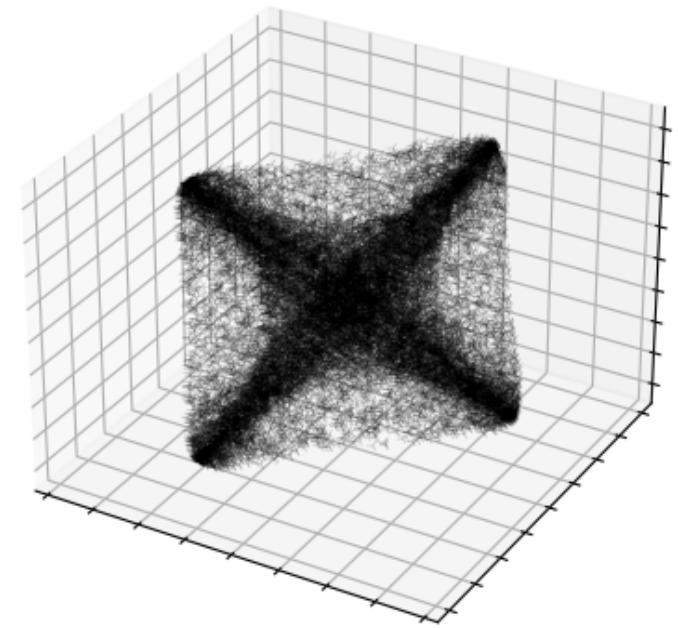
For $\mathrm{SU}(2)$, one can also use multiples of 4.



$\phi_{(5)}$ in \mathbb{R}^5



$\phi_{(3,4)}$ in \mathbb{R}^7



$\phi_{(8)}$ in \mathbb{R}^8

Say we observe an orbit \mathcal{O} of a representation $\phi_1: G \rightarrow \mathrm{GL}_n(\mathbb{R})$, and we want to find ϕ_1 .

Identifiability problem: another representation ϕ_2 may generate \mathcal{O} .

$$\begin{aligned} \phi_1: \theta &\mapsto \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} & \phi_2: \theta &\mapsto \begin{pmatrix} \cos 2\theta & -\sin 2\theta \\ \sin 2\theta & \cos 2\theta \end{pmatrix} \end{aligned}$$

The representations are said **orbit-equivalent** if there exists a $A \in \mathrm{GL}_n(\mathbb{R})$ such that for all $x_0 \in \mathbb{R}^n$,

$$\underbrace{\{A\phi_1(g)A^{-1}x_0 \mid g \in G\}}_{\text{orbit of } x_0 \text{ under } A\phi_1 A^{-1}} = \underbrace{\{\phi_2(g)x_0 \mid g \in G\}}_{\text{orbit of } x_0 \text{ under } \phi_2}.$$

The orbit-equivalence classes of representations of:

- $\mathrm{SO}(2)$ in \mathbb{R}^{2k} are the (increasing and positive) primitive k -tuples of integers.
- T^d in \mathbb{R}^{2k} are the primitive d -dimensional lattices in \mathbb{Z}^k .

For $\mathrm{SO}(3)$ and $\mathrm{SU}(2)$, equivalence and orbit-equivalence coincide.

A Lie group G admits a **Lie algebra**, denoted \mathfrak{g} . It is a vector space endowed with a **Lie bracket** $\mathfrak{g} \times \mathfrak{g} \rightarrow \mathfrak{g}$.

For $G \subset \mathrm{M}_n(\mathbb{R})$, \mathfrak{g} is the tangent space of G at identity, and the bracket is the commutator $[A, B] = AB - BA$.

- $\mathrm{GL}_n(\mathbb{R}) \quad \mathfrak{gl}(n)$ is the set of $n \times n$ matrices.
- $\mathrm{SO}(n) \quad \mathfrak{so}(n)$ is the set of $n \times n$ skew-symmetric matrices.
- $T^n \quad t^n$ is the set of $2n \times 2n$ skew-symmetric matrices that are 2×2 block-diagonal.

One has an **exponential map** $\exp: \mathfrak{g} \rightarrow G$. Is it surjective when G is connected and compact.

Given a representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$, one builds the **derived homomorphism** $d\phi$:

$$\begin{array}{ccc} G & \xrightarrow{\phi} & \mathrm{GL}_n(\mathbb{R}) \\ \exp \uparrow & & \uparrow \exp \\ \mathfrak{g} & \xrightarrow{d\phi} & \mathfrak{gl}(n) \end{array}$$

We call $d\phi(\mathfrak{g})$ the **pushforward** Lie algebra. It is a subalgebra of $\mathfrak{gl}(n)$.

Fact: Two representations ϕ_1, ϕ_2 are orbit-equivalent iff there exists $A \in \mathrm{GL}_n(\mathbb{R})$ such that

$$\mathrm{Ad}\phi_1(\mathfrak{g})A^{-1} = d\phi_2(\mathfrak{g}).$$

Moduli space of Lie algebras: This is an invitation to work in

$$\mathcal{G}^{\mathrm{Lie}}(d, \mathfrak{gl}(n)) \diagup \mathrm{GL}_n(\mathbb{R})$$

where $\mathcal{G}^{\mathrm{Lie}}(d, \mathfrak{gl}(n))$ is the Grassmannian of d -dimensional Lie subalgebras of $\mathfrak{gl}(n)$, acted upon by $\mathrm{GL}_n(\mathbb{R})$.

Denote $\mathfrak{h} = \mathbf{d}\phi(\mathfrak{g})$. There exists an intermediate space between $\mathbf{d}\phi(\mathfrak{g}) \subset \mathfrak{gl}(n)$.

$$\begin{array}{ccccccc}
 G & \xrightarrow{\phi} & \phi(G) & \subset & \text{Sym}(\mathcal{O}) & \subset & \text{GL}_n(\mathbb{R}) \\
 \exp \uparrow & & \exp \uparrow & & \exp \uparrow & & \exp \uparrow \\
 \mathfrak{g} & \xrightarrow{\mathbf{d}\phi} & \mathbf{d}\phi(\mathfrak{g}) & \subset & \mathfrak{sym}(\mathcal{O}) & \subset & \mathfrak{gl}(n)
 \end{array}$$

Symmetry group:

$$\text{Sym}(\mathcal{O}) = \{P \in \text{GL}_n(\mathbb{R}) \mid P\mathcal{O} = \mathcal{O}\}$$

Symmetry algebra:

$$\mathfrak{sym}(\mathcal{O}) = \{P \in \mathfrak{gl}(n) \mid \exp(P) \in \text{Sym}(\mathcal{O})\}$$

Temporary hypothesis: We will suppose that $\mathbf{d}\phi(\mathfrak{g}) = \mathfrak{sym}(\mathcal{O})$.

Good news: $\mathfrak{sym}(\mathcal{O})$ can be estimated from \mathcal{O} .

[Cahill, Mixon, Parshall, [Lie PCA: Density estimation for symmetric manifolds](#), 2023]

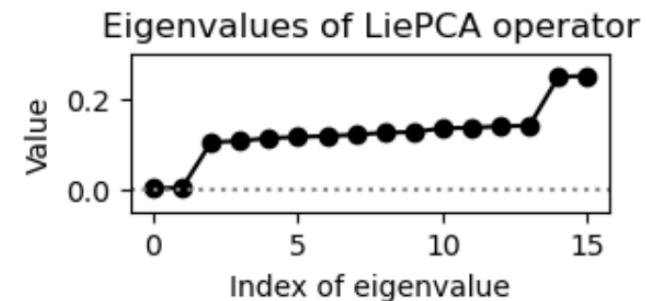
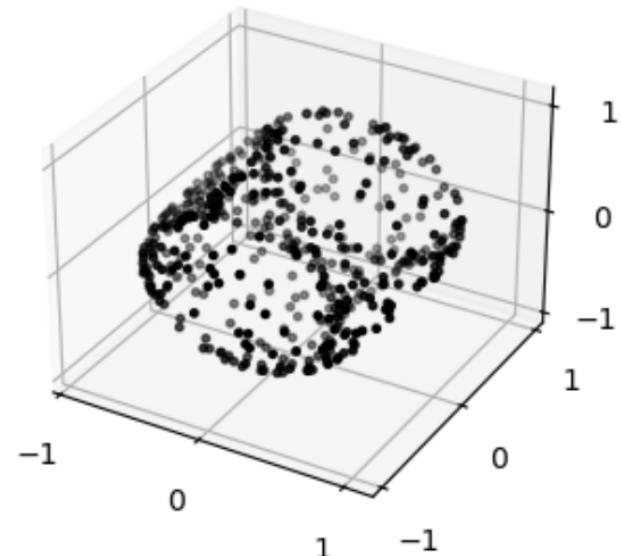
LiePCA operator: Say we observe $X = \{x_1, \dots, x_N\} \subset \mathbb{R}^n$, assumed close to \mathcal{O} .

Define $\Lambda: M_n(\mathbb{R}) \rightarrow M_n(\mathbb{R})$ as
$$\Lambda(A) = \frac{1}{N} \sum_{1 \leq i \leq N} \widehat{\Pi}[N_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle]$$

where

- $\widehat{\Pi}[N_{x_i} X]$ are estimations of projection matrices onto the normal spaces $N_{x_i} \mathcal{O}$,
- $\Pi[\langle x_i \rangle]$ are projection matrices on the lines $\langle x_i \rangle$.

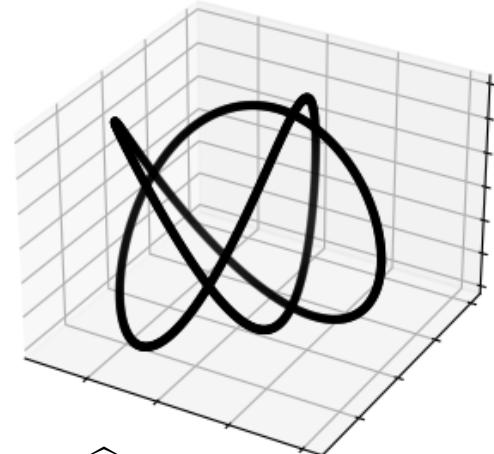
Lemma: $\ker \Lambda \approx \mathfrak{sym}(\mathcal{O})$.



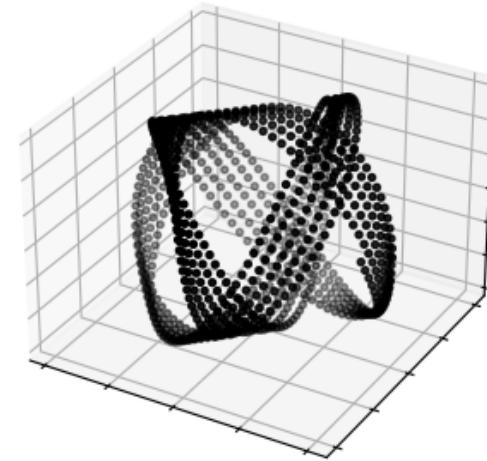
Define $\widehat{\mathfrak{h}}$ as the subspace of $\mathfrak{gl}(n)$ spanned by the d bottom eigenvectors of Λ .

What can go wrong: $\widehat{\mathfrak{h}}$ is estimated as if it were a vector subspace.

- It may not be a Lie algebra ($A, B \in \widehat{\mathfrak{h}} \implies AB - BA \in \widehat{\mathfrak{h}}$).
- It may not come from a compact Lie group.
- We still do not know what is the representation.



exact $\widehat{\mathfrak{h}}$



inexact $\widehat{\mathfrak{h}}$

We wish to find the Lie algebra closest to $\widehat{\mathfrak{h}}$. The problem reads

$$\min \left\{ d(\widehat{\mathfrak{h}}, V) \mid V \in \mathcal{G}^{\text{Lie}}(d, \mathfrak{gl}(n)) \right\}.$$

Remember that $d\phi(\mathfrak{g}) \subset \mathfrak{sym}(\mathcal{O})$ and $\ker \Lambda \approx \mathfrak{sym}(\mathcal{O})$ (LiePCA operator).

Case $d\phi(\mathfrak{g}) = \mathfrak{sym}(\mathcal{O})$: We compute the span $\widehat{\mathfrak{h}}$ of bottom eigenvectors of Λ , and solve

$$\min \left\{ d(\widehat{\mathfrak{h}}, V) \mid V \in \mathcal{G}^{\text{Lie}}(d, \mathfrak{gl}(n)) \right\}.$$

Case $d\phi(\mathfrak{g}) \subsetneq \mathfrak{sym}(\mathcal{O})$: We consider instead

$$\min \left\{ \sum_{i=1}^d \|\Lambda(A_i)\|^2 \mid \langle A_1, \dots, A_d \rangle = V \in \mathcal{G}^{\text{Lie}}(d, \mathfrak{gl}(n)) \right\}$$

Tentative implementation: Let us embed $\mathcal{G}^{\text{Lie}}(d, \mathfrak{gl}(n)) \hookrightarrow M_{n^2}(\mathbb{R})$, the $n^2 \times n^2$ matrices, via $V \mapsto \text{proj}[V]$.

$$\min \text{tr}(\Lambda^2 P) \quad \text{such that} \quad \begin{cases} P \text{ is a } n^2 \times n^2 \text{ matrix,} \\ P^2 = P, \\ P^\top = P, \\ \text{rank}(P) = d, \\ \forall i, j \in [1 \dots, d], \quad P(Pe_i \cdot Pe_j - Pe_j \cdot Pe_i) = Pe_i \cdot Pe_j - Pe_j \cdot Pe_i. \end{cases}$$

Fix G and let $\mathcal{G}(G, \mathfrak{gl}(n))$ be the d -dimensional Lie subalgebras of $\mathfrak{gl}(n)$ that are pushforward of \mathfrak{g} .

The set $\mathcal{G}(G, \mathfrak{gl}(n)) / \mathrm{GL}_n(\mathbb{R})$ is in correspondence with the orbit-equivalence classes of reps of G in \mathbb{R}^n .

Let $\mathfrak{orb}(G, n)$ denote a choice of representatives.

Lemma: The optimization problem is equivalent to

$$\min \sum_{i=1}^d \left\| \Lambda(A \mathrm{diag}(B_i^k)_{k=1}^p A^{-1}) \right\|^2 \quad \text{such that} \quad \begin{cases} (B^1, \dots, B^p) \in \mathfrak{orb}(G, n), \\ A \in \mathrm{GL}_n(\mathbb{R}), \end{cases}$$

Any representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$, up to a change of basis, decomposes as $\phi = \phi_1 \oplus \dots \oplus \phi_p$.

By denoting $B^i = d\phi_i(\mathfrak{g})$, the element $d\phi(\mathfrak{g})$ of $\mathfrak{orb}(G, n)$ is associated to (B^1, \dots, B^p) .

Fix G and let $\mathcal{G}(G, \mathfrak{gl}(n))$ be the d -dimensional Lie subalgebras of $\mathfrak{gl}(n)$ that are pushforward of \mathfrak{g} .

The set $\mathcal{G}(G, \mathfrak{gl}(n)) / \mathrm{GL}_n(\mathbb{R})$ is in correspondence with the orbit-equivalence classes of reps of G in \mathbb{R}^n .

Let $\mathfrak{orb}(G, n)$ denote a choice of representatives.

Lemma: The optimization problem is equivalent to

$$\min \sum_{i=1}^d \left\| \Lambda(A \operatorname{Adiag}(B_i^k)_{k=1}^p A^{-1}) \right\|^2 \quad \text{such that} \quad \begin{cases} (B^1, \dots, B^p) \in \mathfrak{orb}(G, n), \\ \cancel{A \in \mathrm{GL}_n(\mathbb{R})}, \\ A \in \mathrm{O}(n) \end{cases}$$

Any representation $\phi: G \rightarrow \mathrm{GL}_n(\mathbb{R})$, up to a change of basis, decomposes as $\phi = \phi_1 \oplus \dots \oplus \phi_p$.

By denoting $B^i = d\phi_i(\mathfrak{g})$, the element $d\phi(\mathfrak{g})$ of $\mathfrak{orb}(G, n)$ is associated to (B^1, \dots, B^p) .

Orthonormalization trick: After a pre-processing step, we can reduce the program to $A \in \mathrm{O}(n)$.

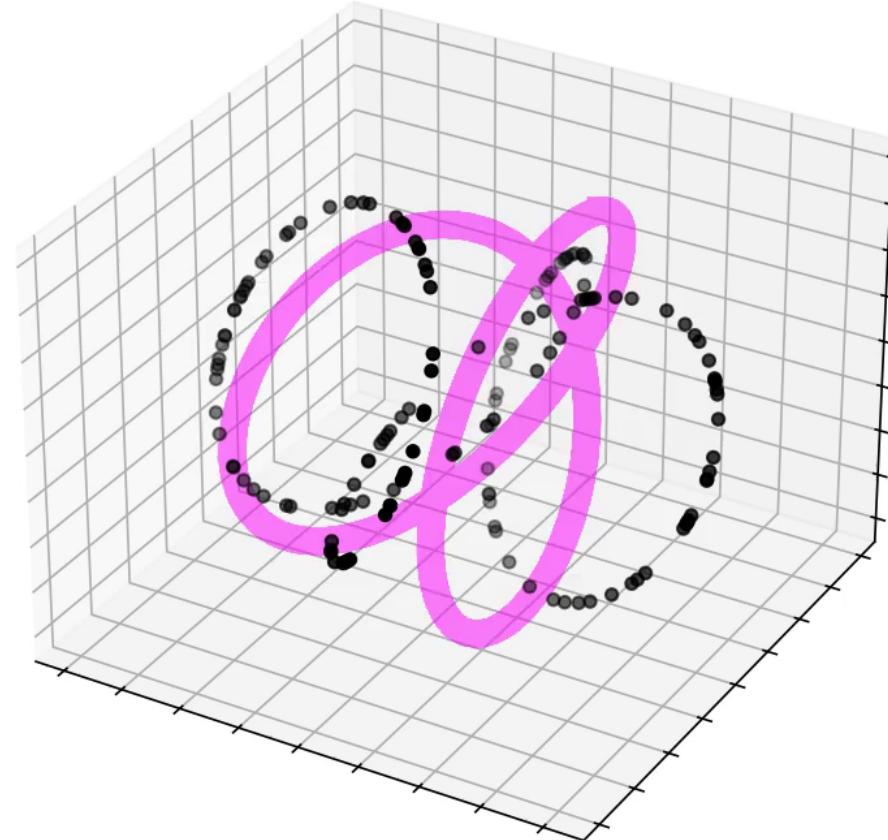
Closest Lie algebra

9/16 (4/6)

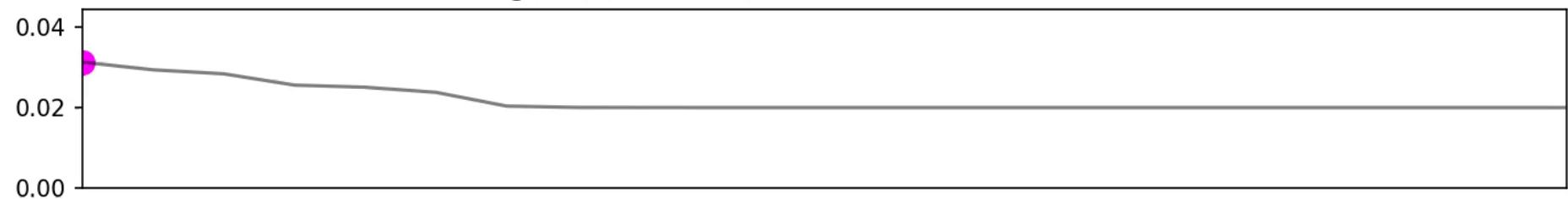
$\text{SO}(2)$ -orbit in \mathbb{R}^4

Rep	Score
(1,2)	+1 : 0.020 -1 : 0.001
(1,3)	+1 : 0.017 -1 : 1×10^{-5}
(1,4)	+1 : 0.014 -1 : 4×10^{-4}
(2,3)	+1 : 0.020 -1 : 0.004
(3,4)	+1 : 0.022 -1 : 0.005

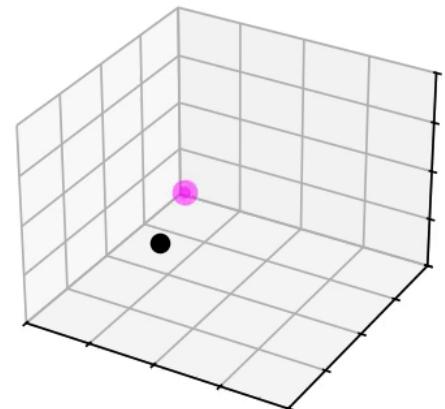
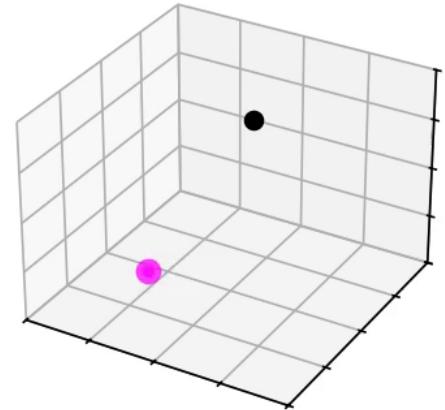
Generated orbit - $d_H(X | \widehat{\mathcal{O}}_x) = 1.164$



Weights (1, 2) - Determinant +1 - Final cost 2.00e-02



Lie algebra



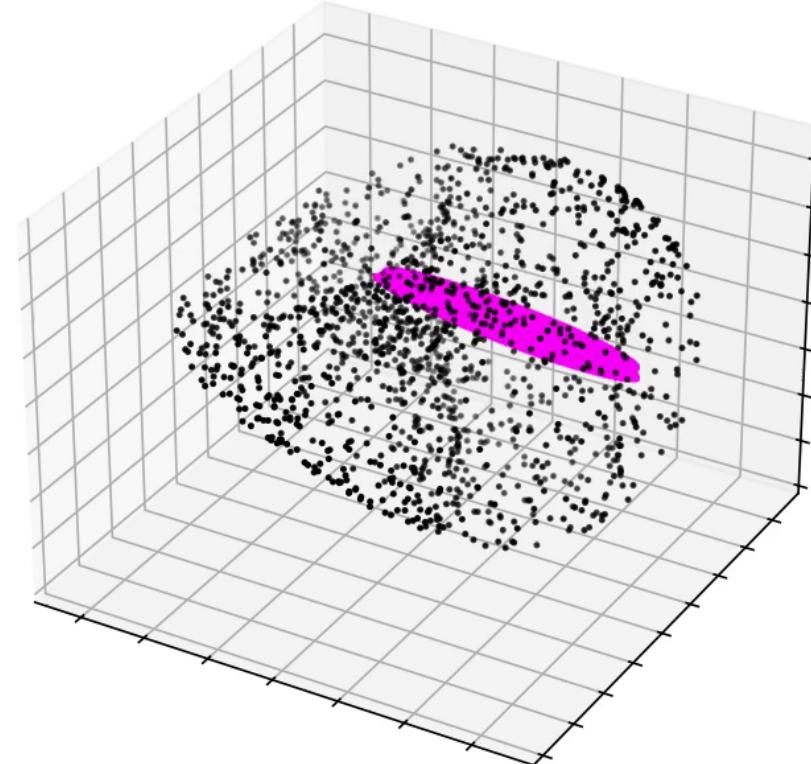
Closest Lie algebra

9/16 (5/6)

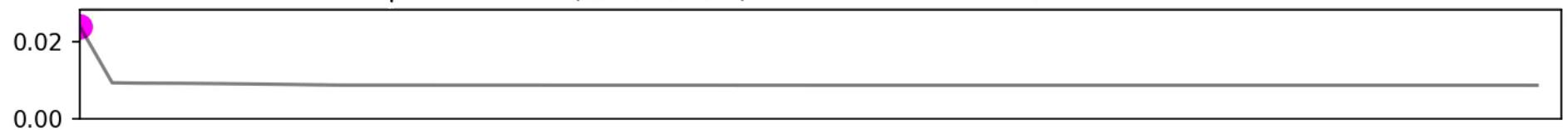
$SU(2)$ -orbit in \mathbb{R}^7

Rep	Score
(3)	+1 : 0.008
(4)	+1 : 0.013
(5)	+1 : 0.003
(3,3)	+1 : 0.003
(3,4)	+1 : 3×10^{-5}
(7)	+1 : 0.005

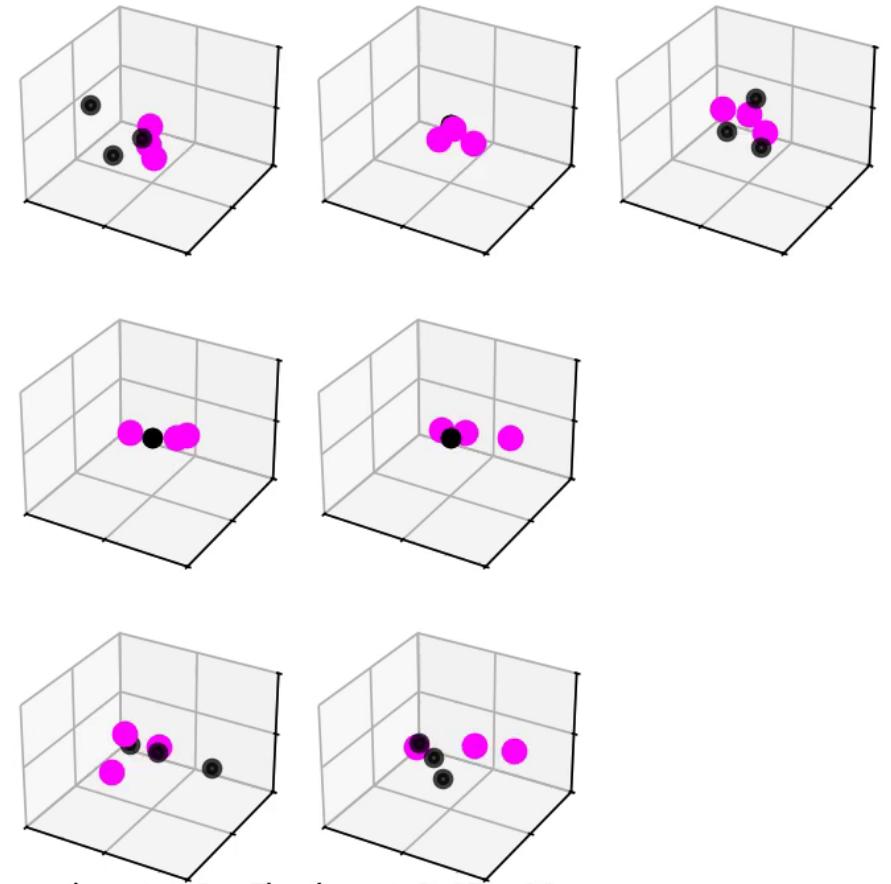
Generated orbit - $d_H(X | \widehat{\mathcal{O}}_x) = 1.739$



Representation (1, 1, 1, 1, 3) - Determinant +1 - Final cost 8.67e-03



Lie algebra



Reformulation of the optimization program:

$$\min \left\{ d(\hat{\mathfrak{h}}, V) \mid V \in \mathcal{G}^{\text{Lie}}(d, \mathfrak{gl}(n)) \right\}.$$

reduces to:

- $\text{SO}(2)$ two-sided orthogonal Procrustes problem \longrightarrow reduction of skew-symmetric matrix
- T^d simultaneous reduction of d skew-symmetric matrices \longrightarrow optimization over $\text{O}(n)$
- $\text{SO}(3), \text{SU}(2)$ no reduction found

Fact: If G is compact, for every representation $G \rightarrow \mathrm{GL}_n(\mathbb{R})$, there exists M positive-definite such that

$$\forall g \in G, M\phi(g)M^{-1} \in \mathrm{O}(n).$$

Given an orbit $\mathcal{O} = G \cdot x_0$, consider the Haar measure μ_G , and define the **covariance matrix**

$$\Sigma[\mathcal{O}] = \int_G (\phi(g)x_0)(\phi(g)x_0)^\top d\mu_G(g).$$

M is found as the square root of the Moore-Penrose pseudo-inverse:

$$M[\mathcal{O}] = \sqrt{\Sigma[\mathcal{O}]^+}.$$

Given a sample X , we build $\Sigma[X] = \frac{1}{N} \sum_{i=1}^N x_i x_i^\top$ and $M[X] = \sqrt{\Sigma[X]^+}$.

Example: With $M = \frac{1}{\sqrt{2}} \mathrm{diag}(1, 1/2, 1, 1)$,

$$\phi: t \mapsto \mathrm{diag}\left(\begin{pmatrix} \cos t & -(1/2)\sin t \\ 2\sin t & \cos t \end{pmatrix}, \begin{pmatrix} \cos 4t & -\sin 4t \\ \sin 4t & \cos 4t \end{pmatrix}\right), \quad \mathcal{O} = \{(\cos t, 2\sin t, \cos 4t, \sin 4t) \mid t \in [0, 2\pi]\}.$$

$$M\phi M^{-1}: t \mapsto \mathrm{diag}\left(\begin{pmatrix} \cos t & \sin t \\ \sin t & \cos t \end{pmatrix}, \begin{pmatrix} \cos 4t & -\sin 4t \\ \sin 4t & \cos 4t \end{pmatrix}\right), \quad M\mathcal{O} = \left\{\frac{1}{\sqrt{2}}(\cos t, \sin t, \cos 4t, \sin 4t) \mid t \in [0, 2\pi]\right\}.$$

Input: A point cloud $X = \{x_1, \dots, x_N\} \subset \mathbb{R}^n$ and a candidate Lie group G .

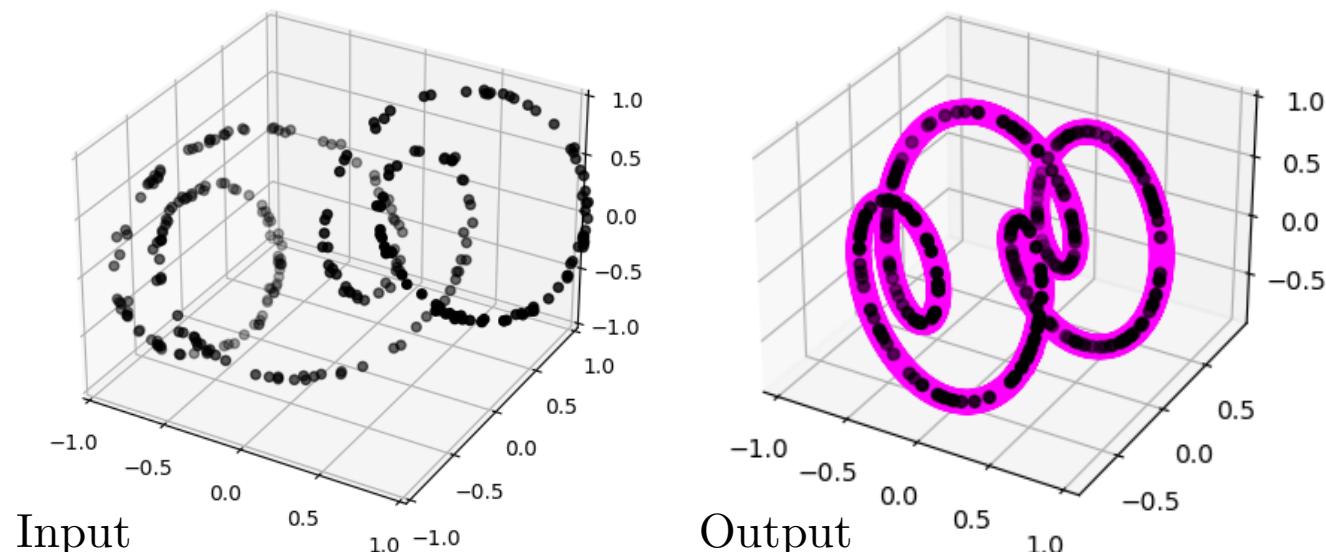
Output: A representation ϕ of G in \mathbb{R}^n , and an orbit \mathcal{O} close to X .

Step 1 (Orthonormalization): Reduce the dimension and orthonormalize the orbit.

Step 2 (LiePCA): Diagonalize the operator $\Lambda: M_n(\mathbb{R}) \rightarrow M_n(\mathbb{R})$.

Step 3 (Closest Lie algebra): Estimate $\widehat{\mathfrak{h}}$ through an optimization over $O(n)$.

Step 4 (Distance to orbit): Choose a $x \in X$, generate $\widehat{\mathcal{O}}_x = \exp(\widehat{\mathfrak{h}}) \cdot x$ and verify that it is close to X .



In **Step 4**, we compute the (non-symmetric) Hausdorff distance $d_H(X | \widehat{\mathcal{O}}_x)$.

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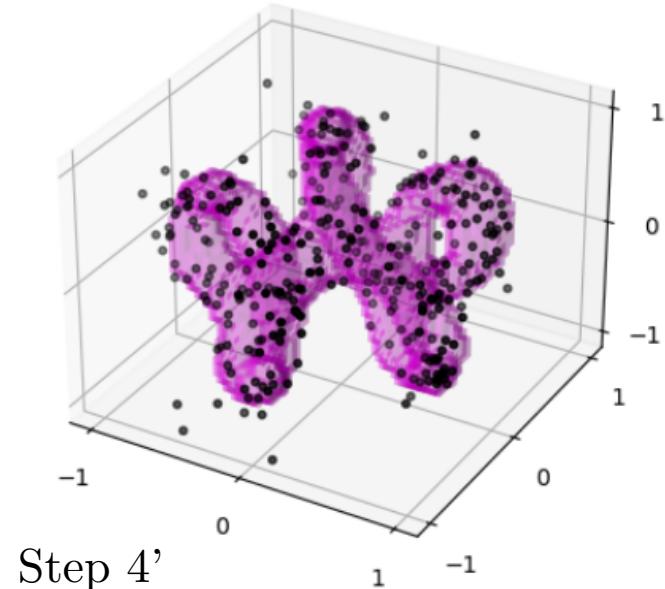
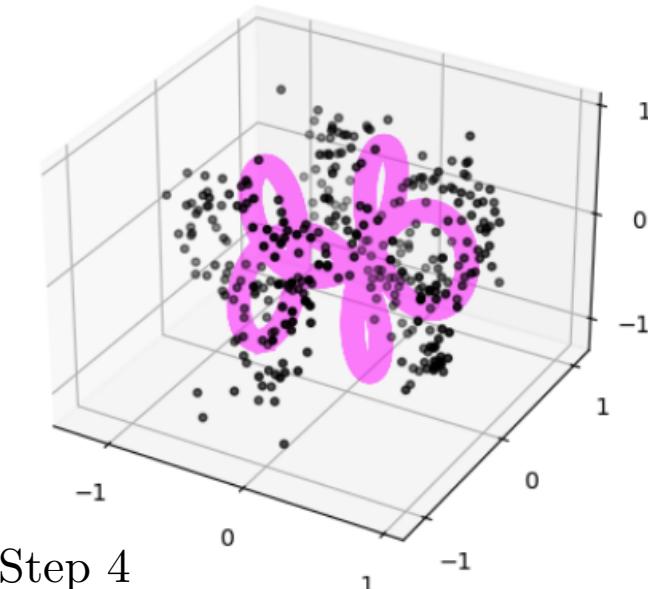
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Step 4' (Distance to noisy orbit): Build the measure $\mu_{\widehat{\mathcal{O}}} = \frac{1}{N} \sum_{x \in X} \mu_{\widehat{\mathcal{O}}_x}$ and verify that it is close to μ_X .



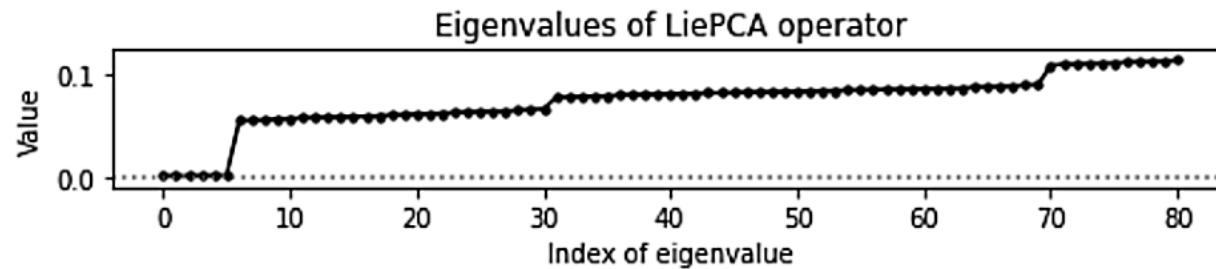
In **Step 4**, we compute the (non-symmetric) Hausdorff distance $d_H(X | \widehat{\mathcal{O}}_x)$.

In **Step 4'**, we compute the Wasserstein distance $W_2(\mu_X, \mu_{\widehat{\mathcal{O}}})$.

Example: Embed $\text{SO}(3) \hookrightarrow \mathbb{R}^9$ and sample 3000 points on it.

LiePCA shows a kernel of dimension 6.

This is consistent with $\text{Isom}(\text{SO}(3)) \simeq \text{SO}(3) \times \text{SO}(3) \times \{\pm 1\}$



We look for an action of $\text{SO}(3)$ or $\text{SU}(2)$. **Step 3** yields

Representation	(3, 5)	(3, 3, 3)	(4, 5)	(8)	(5)	(7)
Cost	2×10^{-5}	4×10^{-5}	0.001	0.001	0.03	0.004
Representation	(9)	(3, 3)	(3, 4)	(4, 4)	(3)	(4)
Cost	0.004	0.006	0.007	0.009	0.011	0.013

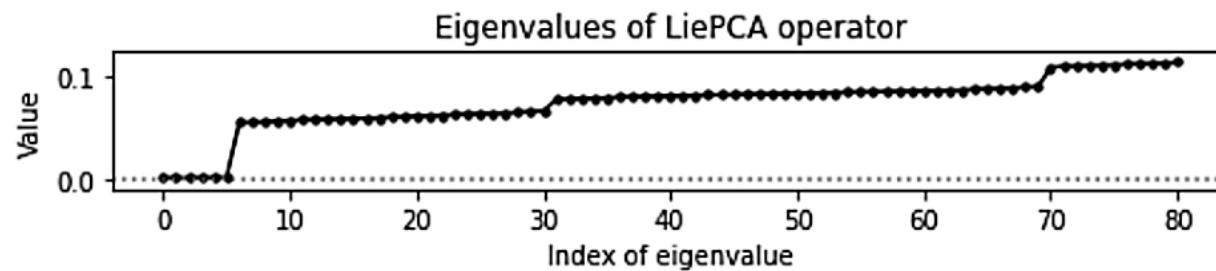
Representation (3, 5): $d_H(X|\hat{\mathcal{O}}_x) \approx 2.658$. However, $d_H(\hat{\mathcal{O}}_x|X) \approx 0.543$.

Representation (3, 3, 3): $d_H(X|\hat{\mathcal{O}}_x) \approx 0.061$.

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Representation (3, 5): $d_H(X|\widehat{\mathcal{O}}_x) \approx 2.658$. However, $d_H(\widehat{\mathcal{O}}_x|X) \approx 0.543$.

action $\text{SO}(3) \curvearrowright \text{SO}(3)$ by conjugation (not transitive)

Representation (3, 3, 3): $d_H(X|\widehat{\mathcal{O}}_x) \approx 0.061$.

action $\text{SO}(3) \curvearrowright \text{SO}(3)$ by translation (transitive)

Input: $X = \{x_1 \dots, x_N\} \subset \mathbb{R}^n$ and G compact.

Model: X sampled close to an orbit \mathcal{O} of a representation $\phi: G \rightarrow \mathbb{R}^n$

Step 1: Orthonormalization via
$$X \leftarrow \sqrt{\Sigma[X]^+} \cdot \Pi_{\Sigma[X]}^{>\epsilon} \cdot X$$

Step 2: Diagonalize the operator
$$\Lambda: A \mapsto \frac{1}{N} \sum_{i=1}^N \widehat{\Pi}[\mathbf{N}_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle]$$

Step 3: Solve $\arg \min \sum_{i=1}^d \|\Lambda(A_i)\|^2$
with $(A_i)_{i=1}^d \in \mathcal{V}^{\text{Lie}}(G, \mathfrak{so}(n))$

Step 4: Output $\widehat{\mathcal{O}}_x = \{ \exp(A)x \mid A \in \widehat{\mathfrak{h}} \}$

Goal: Show that $\widehat{\mathcal{O}}_x$ is close to \mathcal{O}

Input: $X = \{x_1 \dots, x_N\} \subset \mathbb{R}^n$ and G compact.

μ measure on \mathbb{R}^n . E.g., μ_X empirical measure on X

Model: X sampled close to an orbit \mathcal{O} of a representation $\phi: G \rightarrow \mathbb{R}^n$

$\mu_{\mathcal{O}}$ uniform measure on \mathcal{O}

Step 1: Orthonormalization via

$$X \leftarrow \sqrt{\Sigma[X]^+} \cdot \Pi_{\Sigma[X]}^{>\epsilon} \cdot X$$

$$\mu \leftarrow \sqrt{\Sigma[\mu]^+} \cdot \Pi_{\Sigma[\mu]}^{>\epsilon} \cdot \mu$$

Step 2: Diagonalize the operator

$$\Lambda: A \mapsto \frac{1}{N} \sum_{i=1}^N \widehat{\Pi}[\mathbf{N}_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle]$$

$$\Lambda[\mu]: A \mapsto \int_{i=1}^N \widehat{\Pi}[\mathbf{N}_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle] d\mu$$

Step 3: Solve $\arg \min \sum_{i=1}^d \|\Lambda(A_i)\|^2$
with $(A_i)_{i=1}^d \in \mathcal{V}^{\text{Lie}}(G, \mathfrak{so}(n))$

$$\arg \min \sum_{i=1}^d \|\Lambda[\mu](A_i)\|^2$$

with $(A_i)_{i=1}^d \in \mathcal{V}^{\text{Lie}}(G, \mathfrak{so}(n))$

Step 4: Output $\widehat{\mathcal{O}}_x = \{ \exp(A)x \mid A \in \widehat{\mathfrak{h}} \}$

$$\mu_{\widehat{\mathcal{O}}_x} = \exp(\widehat{\mathfrak{h}}) \cdot \mu$$

Goal: Show that $\widehat{\mathcal{O}}_x$ is close to \mathcal{O}

Show that $W_2(\mu_{\widehat{\mathcal{O}}_x}, \mu_{\mathcal{O}})$ is small

Theorem: Under technical assumptions (sufficiently small $W_2(\mu_X, \mu_{\mathcal{O}})$), for a certain choice of parameters, the algorithm outputs a representation $\hat{\phi}$ that is orbit-equivalent to ϕ .

Let $l = \dim \mathcal{O}$. The output measure $\mu_{\hat{\mathcal{O}}}$ satisfies

$$W_2(\mu_{\hat{\mathcal{O}}}, \mu_{\mathcal{O}}) \leq \text{constant} \cdot W_2(\mu_X, \mu_{\mathcal{O}})^{1/4(l+3)}.$$

In addition, for all $x \in X$, the output orbit $\hat{\mathcal{O}}_x$ satisfies

$$d_H(\hat{\mathcal{O}}_x, \mathcal{O}) \leq \text{constant} \cdot d(x, \mathcal{O}) + \text{constant} \cdot W_2(\mu_X, \mu_{\mathcal{O}})^{1/4(l+3)}.$$

Theorem: Let G be a compact Lie group of dimension d , \mathcal{O} an orbit of an almost-faithful representation $\phi: G \rightarrow \mathbb{R}^n$, potentially non-orthogonal, and l its dimension. Let $\mu_{\mathcal{O}}$ be the uniform measure on \mathcal{O} , and $\mu_{\tilde{\mathcal{O}}}$ that on the orthonormalized orbit. Let $X \subset \mathbb{R}^n$ be a finite point cloud and μ_X its empirical measure.

Let $\hat{\phi}, \hat{\mathfrak{h}}, \hat{\mathcal{O}}_x$ and $\mu_{\hat{\mathcal{O}}}$ be the output of the algorithm. Under technical assumptions, $\hat{\phi}$ is equivalent to ϕ , and

$$\begin{aligned}\|\Pi[\hat{h}] - \Pi[\mathfrak{sym}(\mathcal{O})]\|_F &\leq 9d \frac{\rho}{\lambda} \left(r + 4 \left(\frac{\tilde{\omega}}{r^{l+1}} \right)^{1/2} \right) \\ d_H(\hat{\mathcal{O}}_x, \mathcal{O}) &\leq \sqrt{2} \frac{d(x, \mathcal{O})}{\sigma_{\min}} + 3\sqrt{dn} \left(\frac{\rho}{\lambda} \right)^{1/2} \left(r + 4 \left(\frac{\tilde{\omega}}{r^{l+1}} \right)^{1/2} \right)^{1/2} \\ W_2(\mu_{\hat{\mathcal{O}}}, \mu_{\tilde{\mathcal{O}}}) &\leq \frac{1}{\sqrt{2}} \frac{W_2(\mu_X, \mu_{\mathcal{O}})}{\sigma_{\min}} + 3\sqrt{dn} \left(\frac{\rho}{\lambda} \right)^{1/2} \left(r + 4 \left(\frac{\tilde{\omega}}{r^{l+1}} \right)^{1/2} \right)^{1/2}\end{aligned}$$

where

- $\rho = 16l(l+2)6^l \max(\text{vol}(\tilde{\mathcal{O}}), \text{vol}(\tilde{\mathcal{O}})^{-1}) / \min(1, \text{reach}(\tilde{\mathcal{O}}))$,
- $\sigma_{\max}^2, \sigma_{\min}^2$ the top and bottom nonzero eigenvalues of the covariance matrix $\Sigma[\mu_{\mathcal{O}}]$,
- $\tilde{\omega} = 4(n+1)^{3/2} \left(\frac{\sigma_{\max}^3}{\sigma_{\min}^3} \right) \left(\omega(v+\omega) \right)^{1/2}$ with $\omega = \frac{W_2(\mu_{\mathcal{O}}, \mu_X)}{\sigma_{\min}}$ and $v = \left(\frac{\mathbb{V}[\|\mu_{\mathcal{O}}\|]}{\sigma_{\min}^2} \right)^{1/2}$,
- r is the radius of local PCA (estimation of tangent spaces),
- λ the bottom nonzero eigenvalue of the ideal Lie-PCA operator $\Lambda_{\mathcal{O}}$.

Technical assumptions: Define the quantities

$$\omega = \frac{W_2(\mu_{\mathcal{O}}, \mu_X)}{\sigma_{\min}}, \quad v = \left(\frac{\mathbb{V}[\|\mu_{\mathcal{O}}\|]}{\sigma_{\min}^2} \right)^{1/2},$$

$$\tilde{\omega} = 4(n+1)^{3/2} \left(\frac{\sigma_{\max}^3}{\sigma_{\min}^3} \right) \left(\omega(v + \omega) \right)^{1/2}, \quad \rho = \left(16l(l+2)6^l \right) \frac{\max(\text{vol}(\tilde{\mathcal{O}}), \text{vol}(\tilde{\mathcal{O}})^{-1})}{\min(1, \text{reach}(\tilde{\mathcal{O}}))},$$

$$\gamma = (4(2d+1)\sqrt{2})^{-1} \cdot \lambda \cdot \Gamma(G, n, \omega_{\max}) \quad (\text{rigidity constant of Lie subalgebras})$$

Suppose that ω is small enough, so as to satisfy

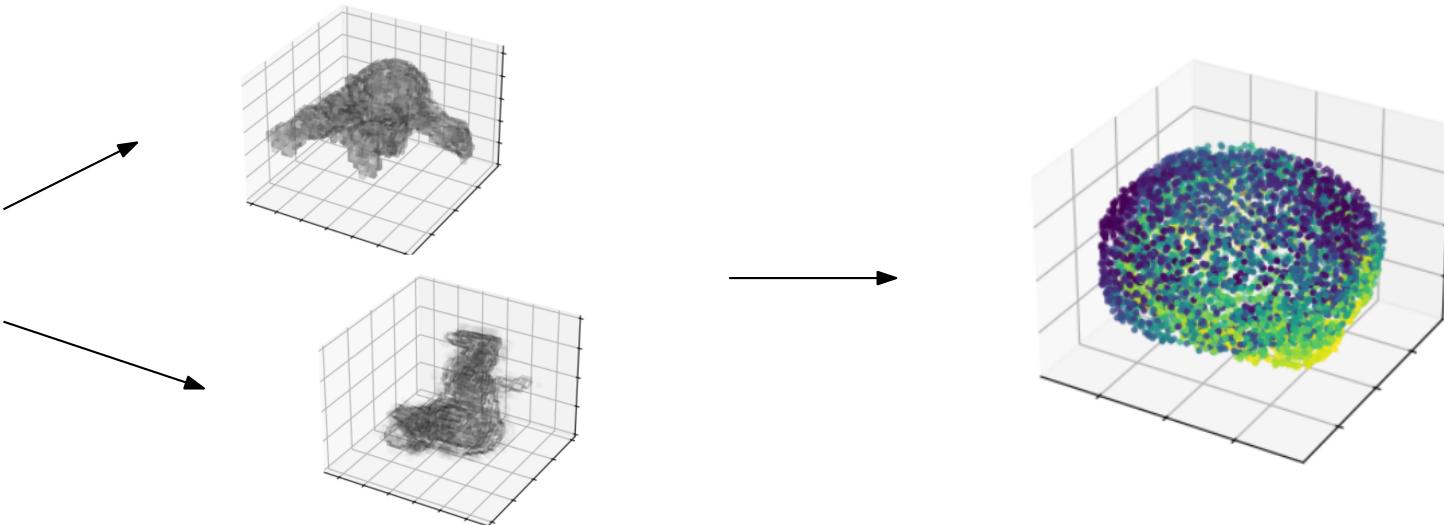
$$\omega < \left(\left(v^2 + \frac{1}{2} \right)^{1/2} - v \right) / \left(3(n+1) \frac{\sigma_{\max}^2}{\sigma_{\min}^2} \right), \quad \tilde{\omega} \leq \min \left\{ \left(\frac{1}{6\rho} \right)^{3(l+1)}, \frac{\gamma^{l+3}}{16}, \left(\frac{\gamma}{(6\rho)^2} \right)^{l+1} \right\}.$$

Choose two parameters ϵ and r in the following nonempty sets:

$$\epsilon \in \left((2v + \omega)\omega\sigma_{\min}^2, \frac{1}{2}\sigma_{\min}^2 \right], \quad r \in \left[(6\rho)^2 \cdot \tilde{\omega}^{1/(l+1)}, (6\rho)^{-1} \right] \cap \left[(4/\gamma)^{2/(l+1)} \cdot \tilde{\omega}^{1/(l+1)}, \gamma \right].$$

Moreover, we suppose that

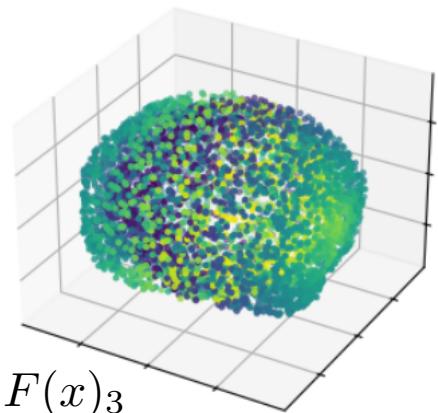
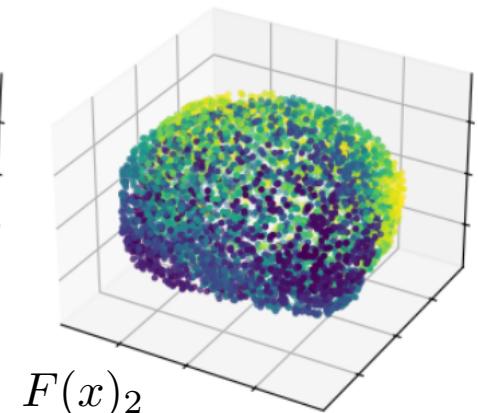
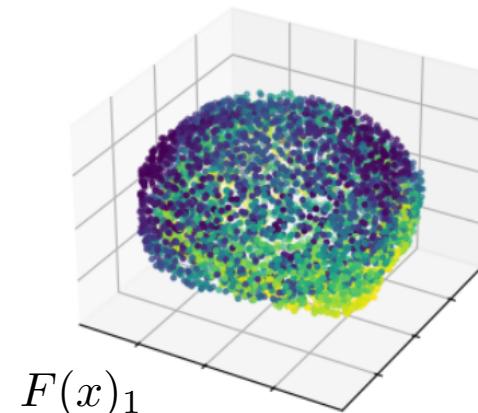
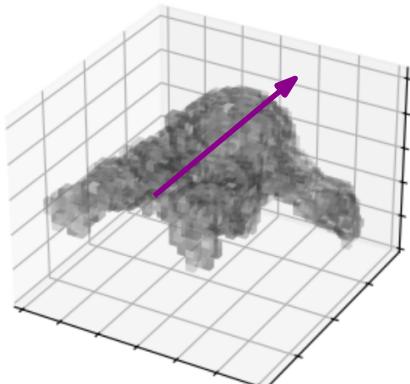
- the minimization problems are computed exactly,
- $\mathfrak{sym}(\mathcal{O})$ is spanned by matrices whose spectra come from primitive vectors of coordinates at most ω_{\max} ,
- the candidate Lie group has Lie algebra $\simeq \mathfrak{sym}(\mathcal{O})$.



- (1) Take a $m \times m \times m$ image.
- (2) Generate several rotations to get a point cloud $X \subset \mathbb{R}^{m \times m \times m}$.
- (3) Project X in \mathbb{R}^n via PCA.

Problem: given $x \in X$, estimate the unit vector $F(x) \in \mathbb{R}^3$ that points toward the armadillo's head.

We define train/test sets of 90%/10%.



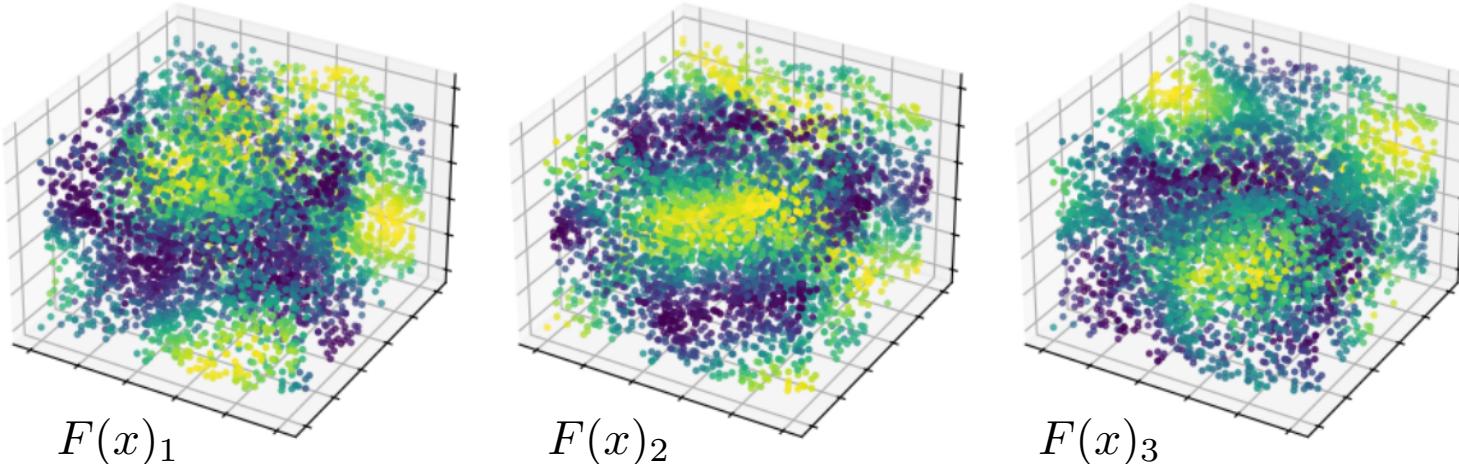
Conventional solution: Train a SVM.

Orthogonal coordinates: Our algorithm detect a $\text{SO}(3)$ -orbit in \mathbb{R}^8 that is close to X : $d_H(X, \mathcal{O}) \simeq 0.1909$.

$$\begin{array}{ccc} \text{SO}(3) & \xrightarrow{\phi} & \text{GL}_n(\mathbb{R}) \\ \exp \uparrow & & \uparrow \exp \\ \mathfrak{so}(3) & \xrightarrow{d\phi} & \mathfrak{gl}(n) \end{array}$$

The orbit is $\mathcal{O} = \{\phi(g) \cdot x_0 \mid g \in G\}$. Every $x \in X$ can be pulled back to $\mathfrak{so}(3)$ via

$$\min_{c \in \mathfrak{so}(3)} \|x - \phi(\exp(c)) \cdot x_0\|.$$



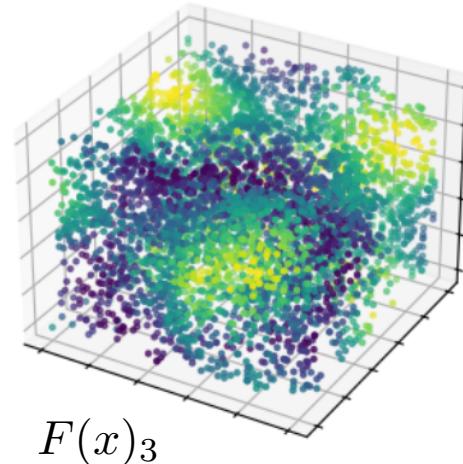
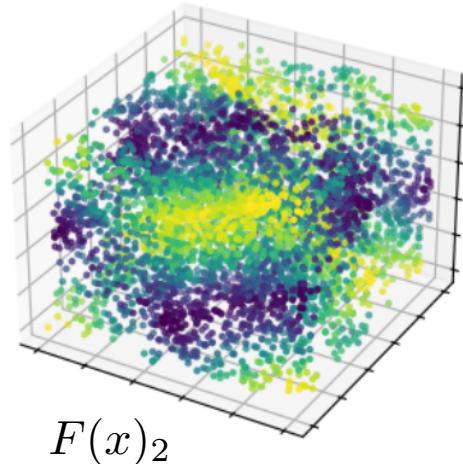
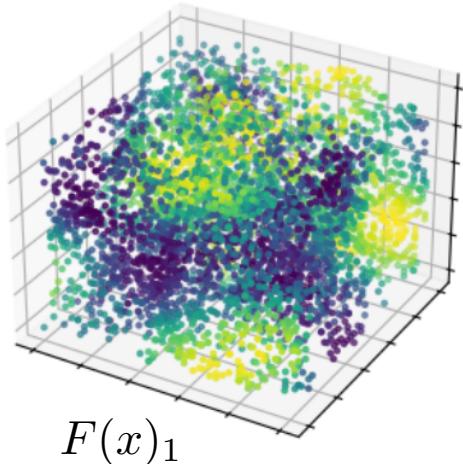
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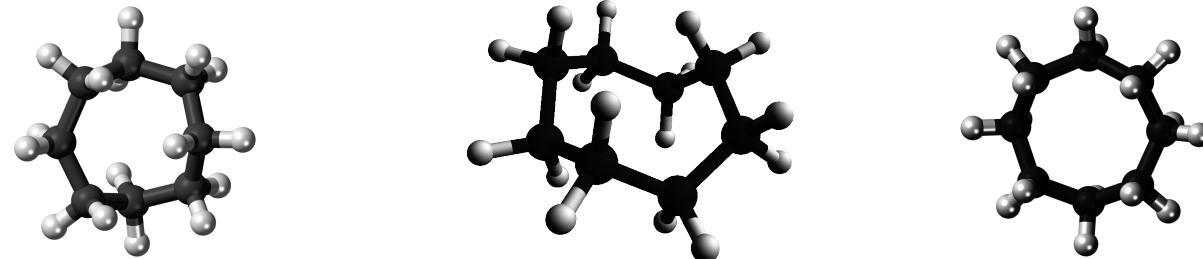
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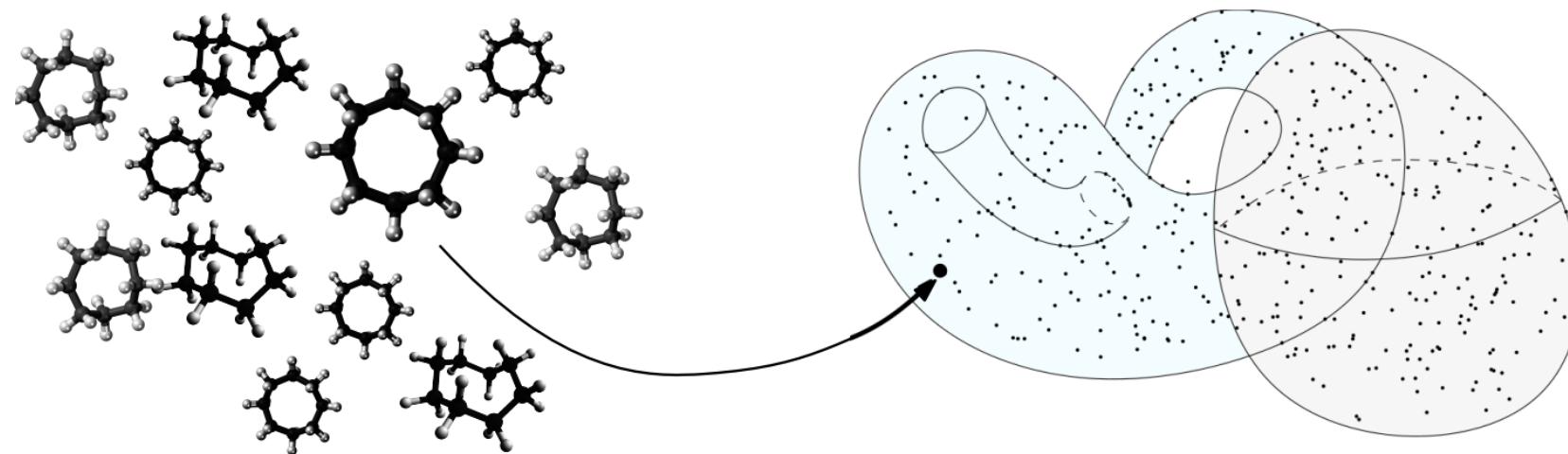


Model	MSE on test data
SVM in dimension 3	0.4003
SVM in dimension 4	0.2496
SVM in dimension 5	0.1295
SVM in dimension 6	0.0380
SVM in dimension 7	0.0148
SVM in dimension 8	0.0119
SVM in dimension 9	0.0114
SVM in dimension 10	0.0122
SVM on orthogonal coordinates	0.0066



A conformer of cyclooctane can be seen as a point in \mathbb{R}^{72} ($3 \times 24 = 72$).

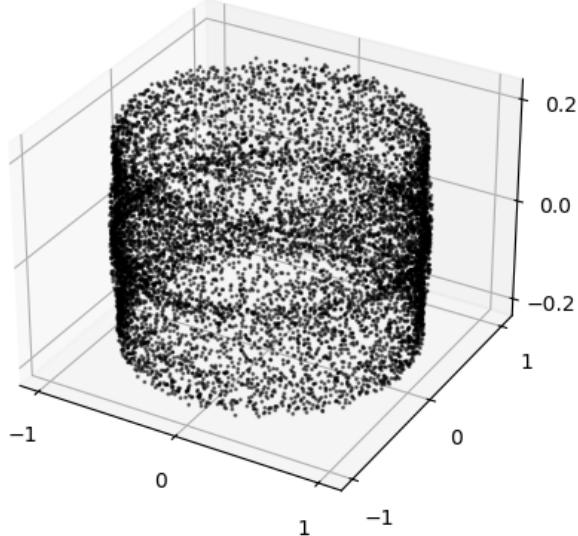
A collection of conformers yield a point cloud $X \subset \mathbb{R}^{72}$.



[Martin, Thompson, Coutsias & Watson, Topology of cyclo-octane energy landscape, 2010]

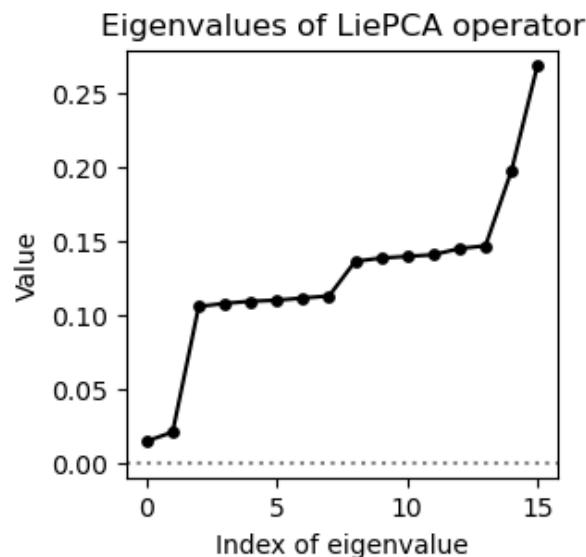
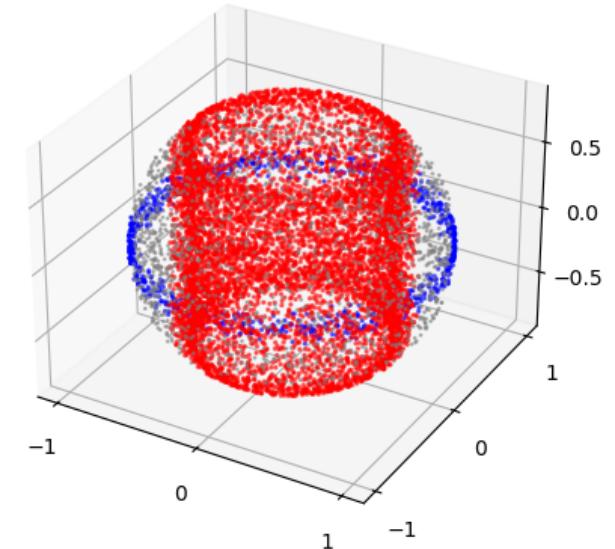
Idea: check whether X lies close to a linear orbit of a Lie group.

Unaligned conformers: We generate 10,000 cyclooctane conformers without aligning them.



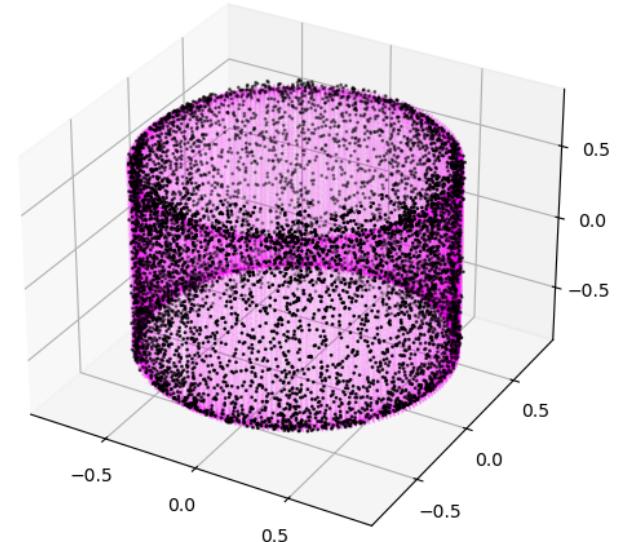
Projected in dimension 3, we see a cylinder surrounded by a circle.

X is projected onto \mathbb{R}^4 and orthonormalized. After discarding 15% of the outliers (gray), two clusters appear. We take the red one.

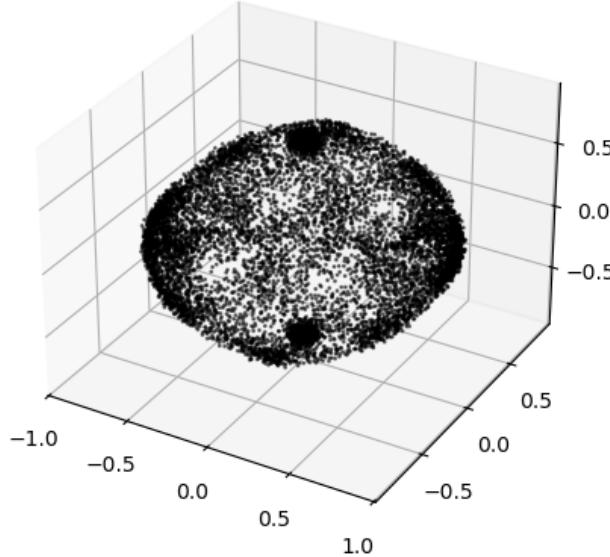


LiePCA has two small eigenvalues, suggesting a symmetry group of dim 2.

We find a T^2 -orbit in \mathbb{R}^4 close X : $d_H(X, \mathcal{O}) \simeq 0.2$.

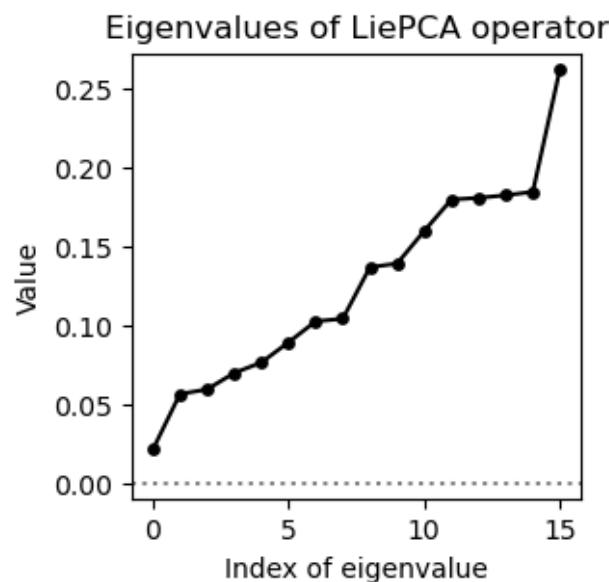


Aligned conformers: We now generate 10,000 aligned conformers (`AlignMolConformers` in RDKit).



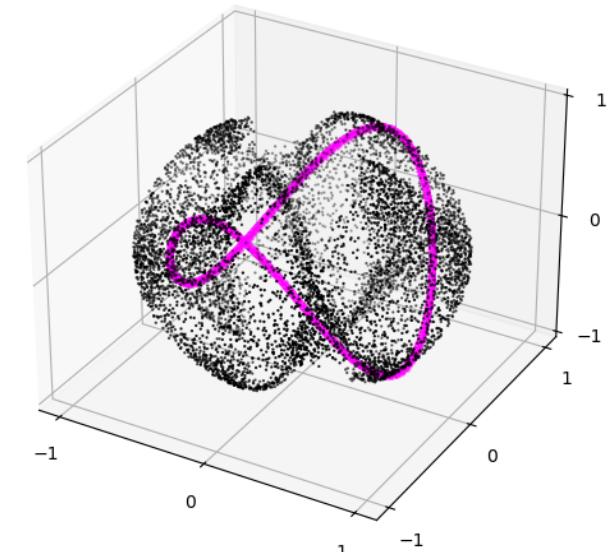
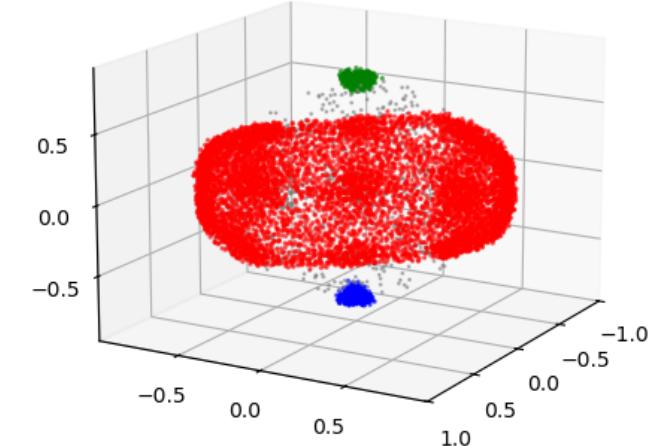
We see three components:
a surface and two clusters.

After discarding 10% of the outliers (gray),
the points are grouped into three classes.
We keep the red class.



LiePCA has one small eigenvalue,
suggesting a symmetry group of dim 1.

We find a SO(2)-action that stabilizes X .
Average distance: $d_H(\widehat{\mathcal{O}}_x | X) \simeq 0.1$.



Consider a neural network

$$V = V_1 \xrightarrow{f_1} V_2 \xrightarrow{f_2} V_3 \xrightarrow{f_3} \dots \xrightarrow{f_{p-1}} V_p = W.$$

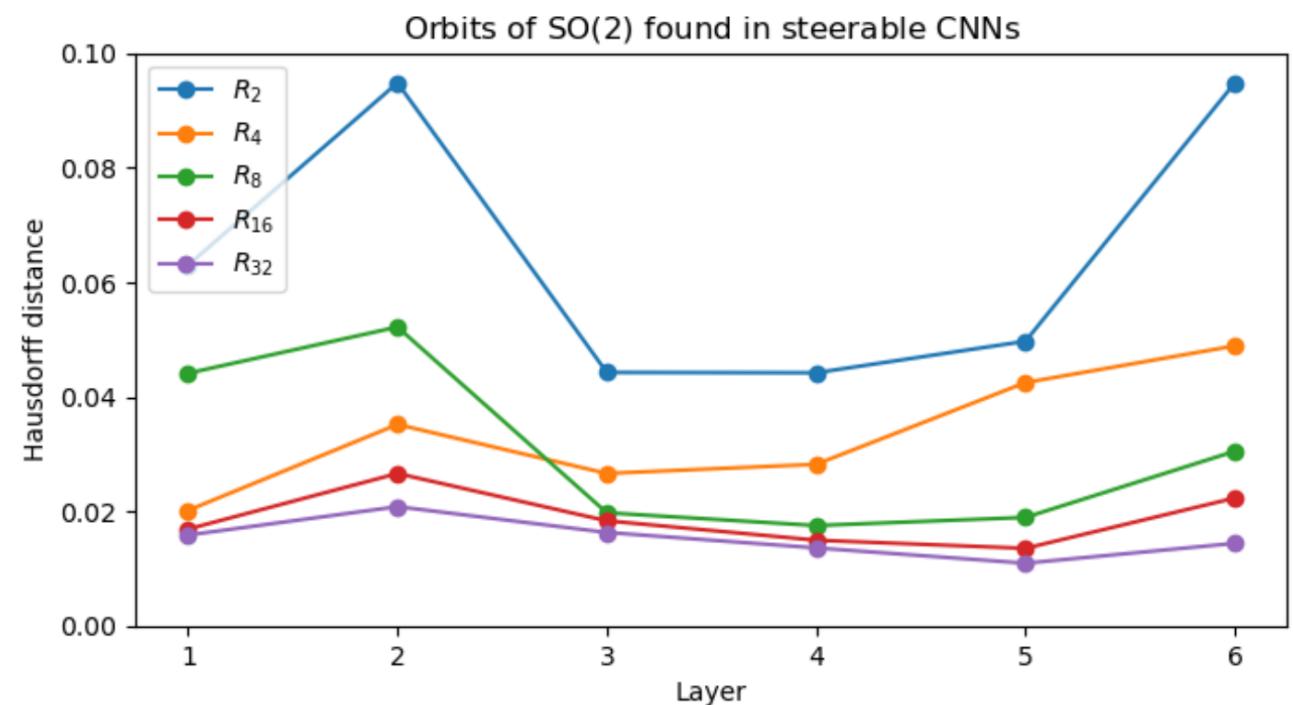
Denote $\mathcal{F}_i = f_1 \cdots f_i$ and $\mathcal{F} = \mathcal{F}_{p-1}$.

Say G acts linearly on V , via $\phi: G \rightarrow \text{GL}(V)$.

The network is **equivariant** if there exists representations $\phi_i: G \rightarrow \text{GL}(V_i)$ such that $\forall x \in V, \forall g \in G$,

$$\mathcal{F}_i(\phi(g)x) = \phi_i(g)\mathcal{F}_i(x).$$

Experiment: Consider **steerable CNNs** for several rotation groups R_n . We pick an image, and generate 500 rotations. In each of the layers, we apply our algorithm to find a linear-orbit of $\text{SO}(2)$.



<https://arxiv.org/abs/2309.03086>

<https://github.com/HLovisiEnnes/LieDetect>

Thanks!

Detection of **actions** via the induced representation on space of vector fields

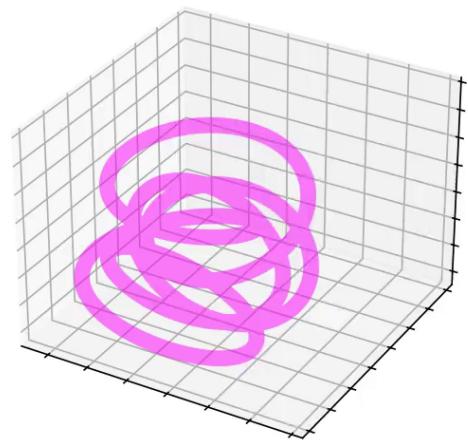
$$\begin{array}{ccc} G & \xrightarrow{\phi} & \text{Diff}(\mathcal{M}) \\ \exp \uparrow & & \uparrow \exp \\ \mathfrak{g} & \xrightarrow{d\phi} & \mathcal{X}(\mathcal{M}) \end{array}$$

Statistical guarantees to test the linear-orbit hypothesis.

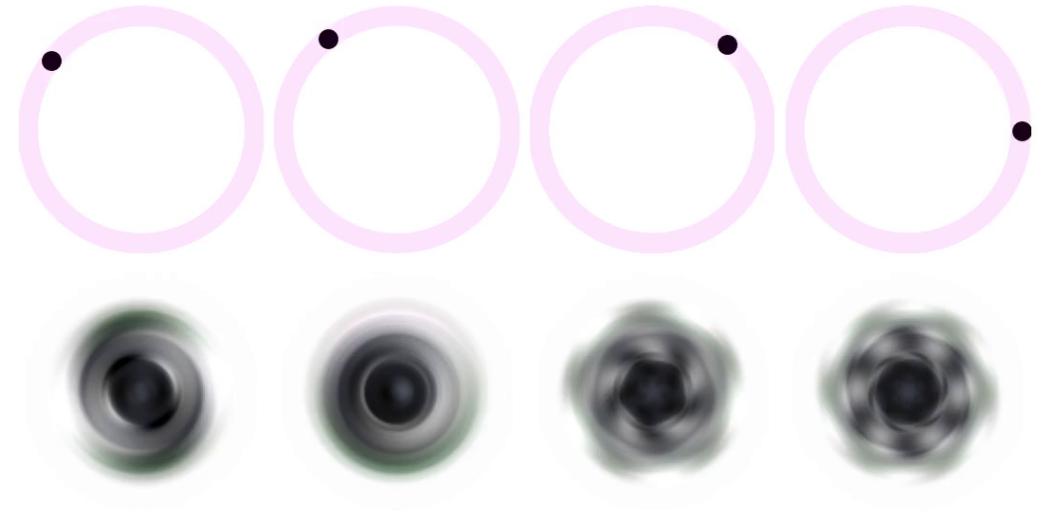
Rotations of $m \times m$ RGB image



Embedding in $\mathbb{R}^{m \times m \times 3}$



Projection in eigenplanes



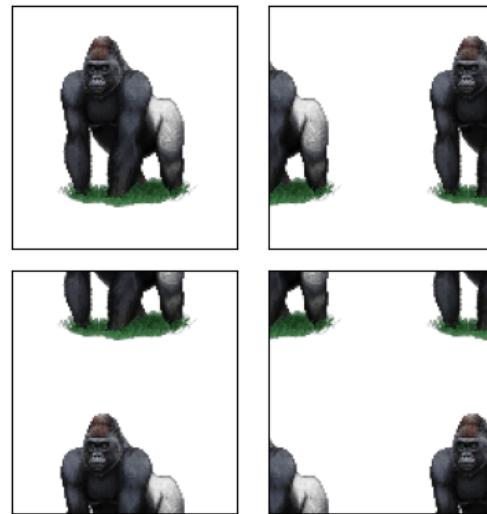
Eigenvalues of the point cloud's covariance matrix:

0.155, 0.155, 0.11, 0.11, 0.041, 0.041, 0.04, 0.04, 0.038, 0.038, 0.026, 0.026, ...

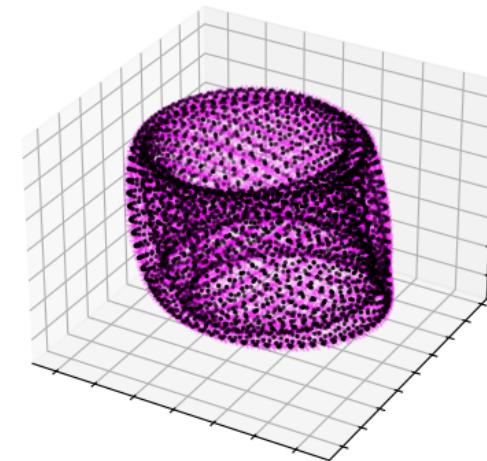
In these eigenplanes, the orbit is close to

$$\theta \mapsto \begin{pmatrix} \mu_1 \cos \omega_1 \theta \\ \mu_1 \sin \omega_1 \theta \\ \mu_2 \cos \omega_2 \theta \\ \mu_2 \sin \omega_2 \theta \\ \vdots \\ \mu_k \cos \omega_k \theta \\ \mu_k \sin \omega_k \theta \end{pmatrix} = \begin{pmatrix} \cos \omega_1 \theta & -\sin \omega_1 \theta \\ \sin \omega_1 \theta & \cos \omega_1 \theta \\ & & \cos \omega_2 \theta & -\sin \omega_2 \theta \\ & & \sin \omega_2 \theta & \cos \omega_2 \theta \\ & & & & \ddots \\ & & & & \cos \omega_k \theta & -\sin \omega_k \theta \\ & & & & \sin \omega_k \theta & \cos \omega_k \theta \end{pmatrix} \begin{pmatrix} \mu_1 \\ 0 \\ \mu_2 \\ 0 \\ \vdots \\ \mu_k \\ 0 \end{pmatrix}$$

Translations of $m \times m$ RGB image



Embedding in $\mathbb{R}^{m \times m \times 3}$



Covariance matrix eigenvalues: 0.228, 0.228, 0.142, 0.142, 0.108, 0.108, 0.022, 0.022, ...

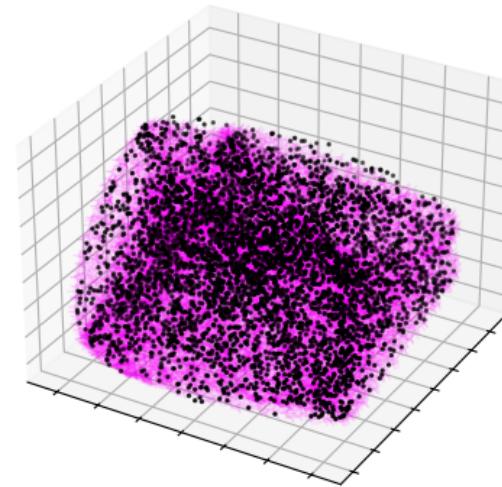
In these eigenplanes, the orbit is close to

$$\theta^{(1)}, \theta^{(2)} \longmapsto \begin{pmatrix} \mu_1 \cos(\omega_1^{(1)}\theta^{(1)} + \omega_1^{(2)}\theta^{(2)}) \\ \mu_1 \sin(\omega_1^{(1)}\theta^{(1)} + \omega_1^{(2)}\theta^{(2)}) \\ \mu_2 \cos(\omega_2^{(1)}\theta^{(1)} + \omega_2^{(2)}\theta^{(2)}) \\ \mu_2 \cos(\omega_2^{(1)}\theta^{(1)} + \omega_2^{(2)}\theta^{(2)}) \\ \vdots \\ \mu_k \cos(\omega_k^{(1)}\theta^{(1)} + \omega_k^{(2)}\theta^{(2)}) \\ \mu_k \cos(\omega_k^{(1)}\theta^{(1)} + \omega_k^{(2)}\theta^{(2)}) \end{pmatrix} = \text{linear action of } T^2 \text{ on } \begin{pmatrix} \mu_1 \\ 0 \\ \mu_2 \\ 0 \\ \vdots \\ \mu_k \\ 0 \end{pmatrix}$$

Rotations of $m \times m \times m$ greyscale object



Embedding in $\mathbb{R}^{m \times m \times m}$



Covariance matrix eigenvalues: 0.246, 0.239, 0.234, 0.058, 0.057, 0.056, 0.055, 0.054 ...

In these eigenplanes, the orbit is close to

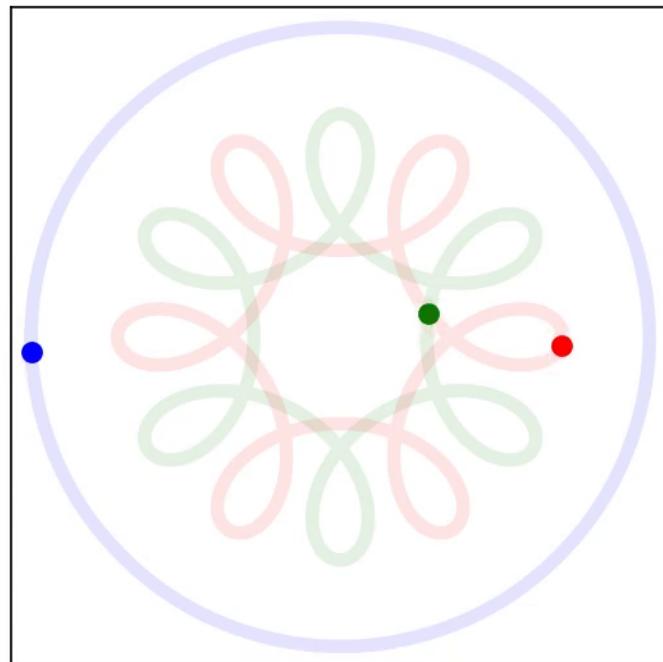
$$\theta^{(1)}, \theta^{(2)}, \theta^{(3)} \longmapsto \text{linear action of } \text{SO}(3) \text{ on } \begin{pmatrix} \mu_1 \\ 0 \\ \vdots \\ \mu_2 \\ 0 \\ \vdots \end{pmatrix}$$

In 1975, Roger Broucke found several periodic orbits.

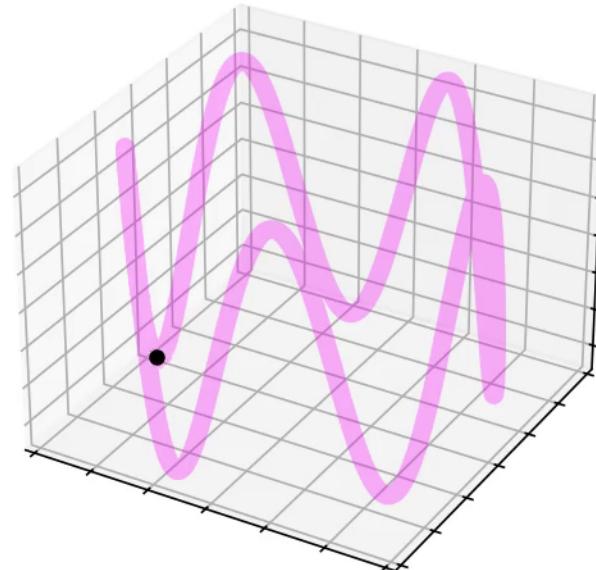
Let $x_1(t), x_2(t), x_3(t)$ be the three bodies, and define $z(t) = (x_1(t), x_2(t), x_3(t)) \in \mathbb{R}^6$.

Orbit A3

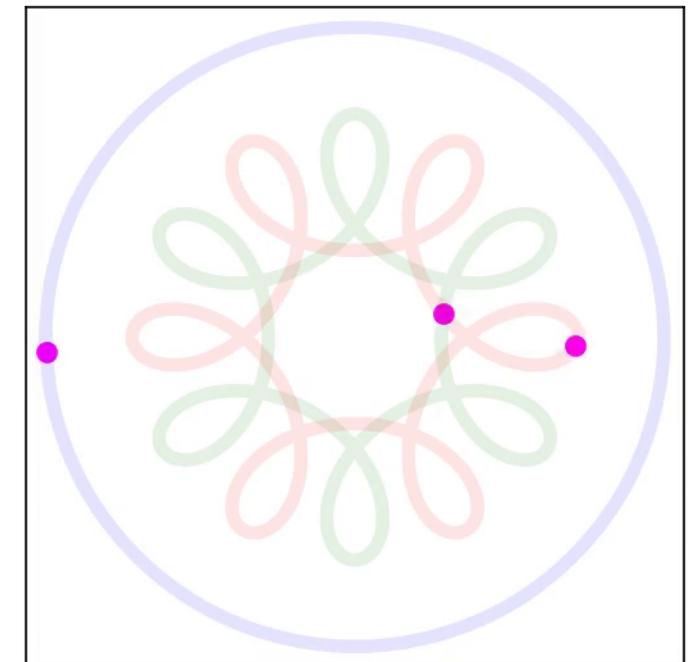
Trajectory of x_1, x_2, x_3
(found by integration)



Trajectory of z



Reconstructed orbit of $\text{SO}(2)$

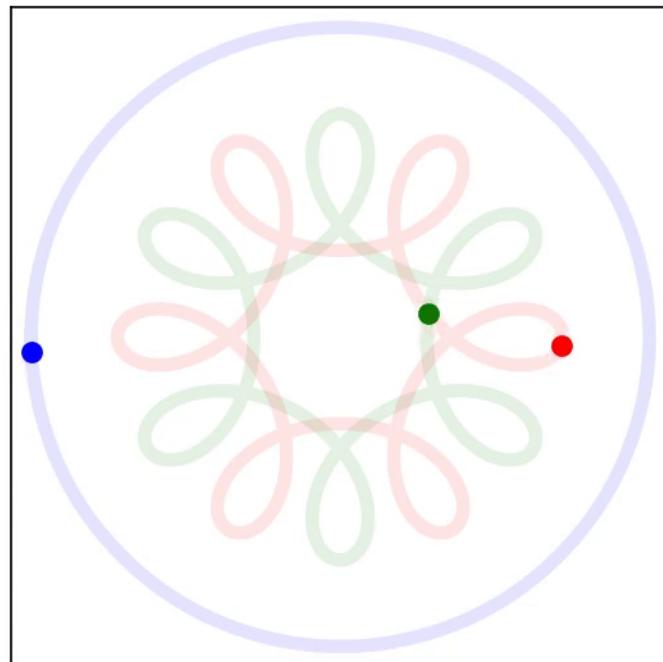


In 1975, Roger Broucke found several periodic orbits.

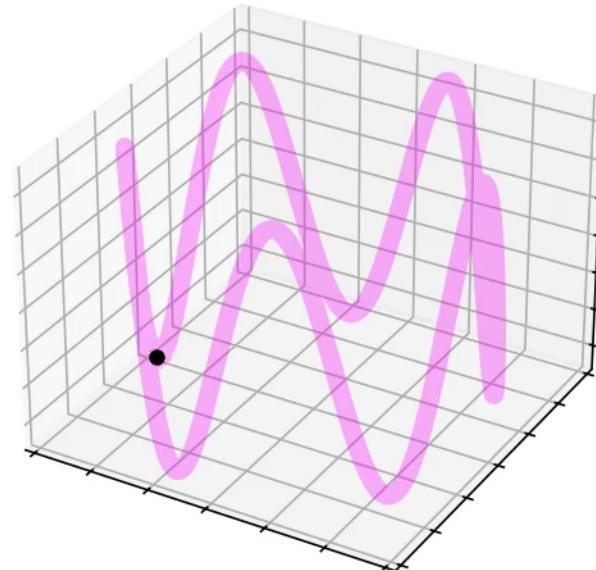
Let $x_1(t), x_2(t), x_3(t)$ be the three bodies, and define $z(t) = (x_1(t), x_2(t), x_3(t)) \in \mathbb{R}^6$.

Orbit R2

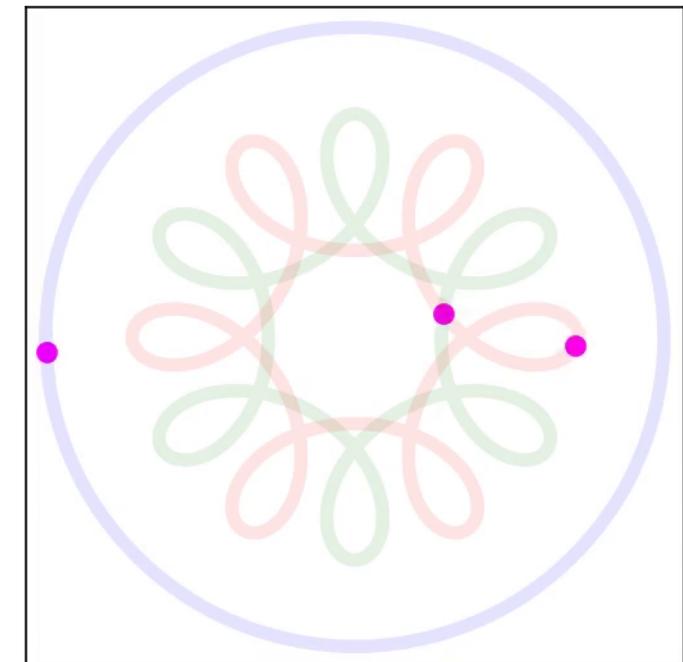
Trajectory of x_1, x_2, x_3
(found by integration)



Trajectory of z



Reconstructed orbit of $\text{SO}(2)$



Let $G = \mathrm{SO}(2)$, whose dimension is $d = 1$. The output \hat{h} of LiePCA is a skew symmetric $n \times n$ matrix A .

Suppose that n is even. The representations of $\mathrm{SO}(2)$ in \mathbb{R}^n take the form

$$\phi_{(\omega_1, \dots, \omega_{n/2})}(\theta) = \begin{pmatrix} R(\omega_1\theta) & & \\ & \ddots & \\ & & R(\omega_{n/2}\theta) \end{pmatrix} \quad \text{where} \quad R(\theta) = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

and where $(\omega_1, \dots, \omega_{n/2}) \in \mathbb{Z}^{n/2}$. In practice, we fix a maximal frequency $\omega_{\max} \in \mathbb{N}$.

The corresponding pushforward Lie algebra is spanned by the matrix

$$B_{(\omega_1, \dots, \omega_{n/2})} = \begin{pmatrix} L(\omega_1) & & \\ & \ddots & \\ & & L(\omega_{n/2}) \end{pmatrix} \quad \text{where} \quad L(\omega) = \begin{pmatrix} 0 & -\omega \\ \omega & 0 \end{pmatrix}$$

In this context, the minimization problem reads

$$\min \| \mathrm{proj}[A] - \mathrm{proj}[OB_{(\omega_1, \dots, \omega_{n/2})}O^\top] \| \quad \text{s.t.} \quad \begin{cases} (\omega_1, \dots, \omega_{n/2}) \in \mathbb{Z}^{n/2}, \\ O \in \mathrm{O}(n). \end{cases}$$

This is equivalent to

$$\min \| A \pm OB_{(\omega_1, \dots, \omega_{n/2})}O^\top \| \quad \text{s.t.} \quad \begin{cases} (\omega_1, \dots, \omega_{n/2}) \in \mathbb{Z}^{n/2}, \\ O \in \mathrm{O}(n). \end{cases}$$

We recognize a **two-sided orthogonal Procrustes problem with one transformation**.

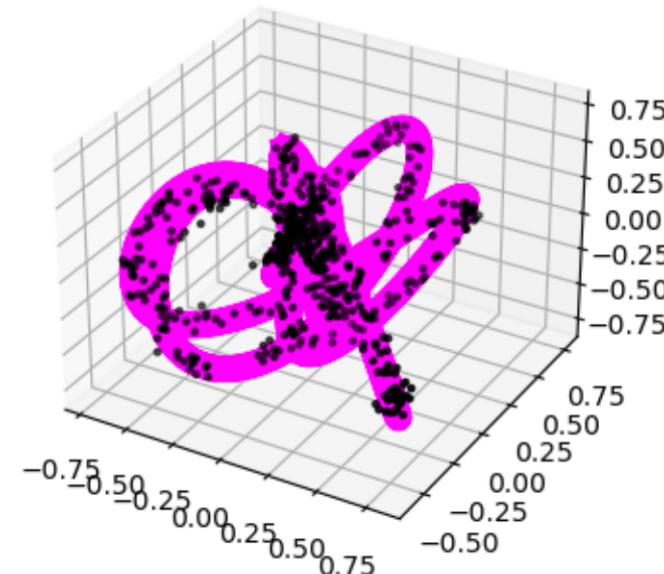
Example: We consider a representation of $\text{SO}(2)$ in \mathbb{R}^{10} with frequencies $(2, 4, 5, 7, 8)$ and sample 600 points on one of its orbits, that we corrupt with a Gaussian additive noise of deviation $\sigma = 0.03$.

We perform the minimization over all representations of $\text{SO}(2)$ in \mathbb{R}^{10} , with parameter $\omega_{\max} = 10$.

Representation	$(2, 4, 5, 7, 8)$	$(2, 5, 6, 8, 9)$	$(3, 5, 7, 9, 10)$	$(3, 6, 7, 9, 10)$	$(3, 5, 6, 8, 9)$	$(2, 4, 5, 6, 7)$
Cost	0.028	0.032	0.037	0.037	0.038	0.044
Representation	$(3, 5, 6, 9, 10)$	$(2, 5, 7, 9, 10)$	$(2, 3, 4, 5, 6)$	$(2, 5, 6, 9, 10)$	$(2, 6, 7, 9, 10)$	$(3, 5, 6, 8, 10)$
Cost	0.046	0.055	0.057	0.058	0.058	0.058

The correct representation is found.

The Hausdorff distance between the point cloud and the estimated orbit is $d_H(X \mid \hat{\mathcal{O}}) \approx 0.231$.



Let $G = T^d$ the torus of dim d . The output of LiePCA is a d -tuple (A_1, \dots, A_d) of skew symmetric matrices.

The representations of T^d in \mathbb{R}^n take the form

$$\phi_{(\omega_i^j)}(\theta_1, \dots, \theta_d) = \sum_{j=1}^d \phi_{(\omega_1^j, \dots, \omega_{n/2}^j)}(\theta_j)$$

where $(\omega_i^j)_{1 \leq i \leq n/2}^{1 \leq j \leq d}$ is a $n/2 \times d$ matrix with integer coefficients.

The push-forward Lie algebra is spanned by

$$B_{(\omega_1^1, \dots, \omega_{n/2}^1)}, \quad B_{(\omega_1^2, \dots, \omega_{n/2}^2)}, \quad \dots, \quad B_{(\omega_1^d, \dots, \omega_{n/2}^d)}.$$

In this context, the minimization problem reads

$$\min \left\| \text{proj}[\langle A_i \rangle_{j=1}^d] - \text{proj}[\langle OB_{(\omega_1^j, \dots, \omega_{n/2}^j)} O^\top \rangle_{j=1}^d] \right\| \quad \text{s.t.} \quad \begin{cases} (\omega_i^j)_{1 \leq i \leq n/2}^{1 \leq j \leq d} \in \mathbb{Z}^{n/2 \times d}, \\ O \in O(n). \end{cases}$$

This is linked to the **simultaneous reduction of a tuple of skew-symmetric matrices**.

Lemma: Denote by $(\rho_i)_{i=1}^d$ the coefficients of an optimal simultaneous reduction of the matrices $(A_i)_{i=1}^d$ in normal form. Then the problem is equivalent to

$$\min_{(\omega_i^j)} \sum_{k=1}^d f\left((\rho_i^k)_{i=1}^{n/2}, (\omega_i^k)_{i=1}^{n/2}\right) \quad \text{where} \quad f(x, y) = \|x/\|x\| - y/\|y\|\|^2.$$

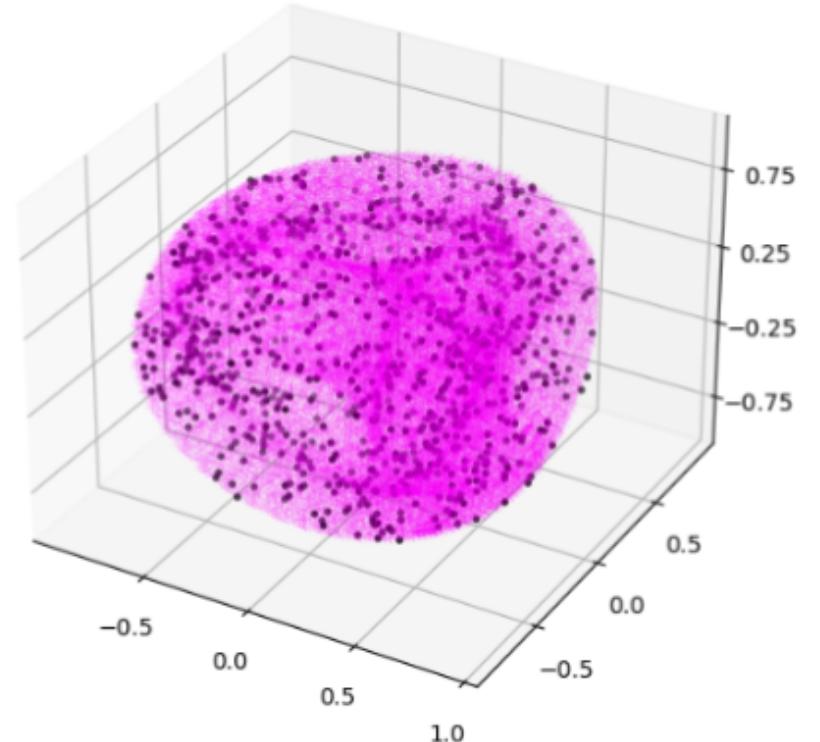
Example: Let X be a uniform 750-sample of an orbit of the representation $\phi_{\left(\begin{smallmatrix} 1 & 1 & 2 \\ 1 & 2 & 1 \end{smallmatrix}\right)}$ of the torus T^2 in \mathbb{R}^6 .

We apply the algorithm with $G = T^2$ restrict to representations with frequencies at most $\omega_{\max} = 2$.

Representation	$\left(\begin{smallmatrix} 0 & 1 & 1 \\ 2 & -2 & 1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 1 & 1 & 2 \\ -2 & 2 & -1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 2 \\ 2 & -2 & -1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 1 \\ 1 & -2 & 0 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 1 \\ 1 & -2 & -1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 2 \\ 2 & -2 & 1 \end{smallmatrix}\right)$
Cost	0.036	0.136	0.198	0.233	0.244	0.312
Representation	$\left(\begin{smallmatrix} 0 & 1 & 2 \\ 1 & -2 & -2 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 2 \\ 1 & -2 & -1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 1 & 2 & 2 \\ -2 & -2 & 1 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 1 & 1 & 1 \\ -2 & -1 & 2 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 2 \\ 1 & -2 & 0 \end{smallmatrix}\right)$	$\left(\begin{smallmatrix} 0 & 1 & 1 \\ 1 & -2 & 1 \end{smallmatrix}\right)$
Cost	0.331	0.348	0.388	0.447	0.457	0.472

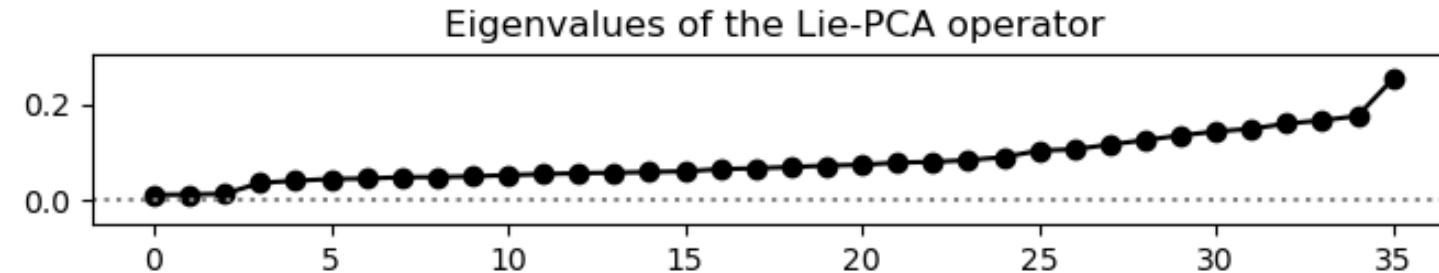
The algorithm's output is $\left(\begin{smallmatrix} 0 & 1 & 1 \\ 2 & -2 & 1 \end{smallmatrix}\right)$. It is equivalent to $\phi_{\left(\begin{smallmatrix} 1 & 1 & 2 \\ 1 & 2 & 1 \end{smallmatrix}\right)}$.

Moreover, the Hausdorff distance is $d_H(X|\widehat{\mathcal{O}}) \approx 0.071$.



When the underlying group is unknown, we can guess it from LiePCA or test several candidates.

Example: Let X be a 1500-sample of an orbit of the representation $(1, 5)$ of $SU(2)$ in \mathbb{R}^6 .



We see a Lie algebra of dimension 3. One expects the torus T^3 , $SO(3)$ or $SU(2)$.

Representation of $SU(2)$	$(1, 5)$	$(1, 1, 1, 3)$	$(1, 1, 4)$	$(3, 3)$
Cost	8.6×10^{-5}	0.007	0.008	0.015

Representation of T^3	$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
Cost	0.014

Representation $(1, 5)$: we get the (non-symmetric) Hausdorff distance $d_H(X|\hat{\mathcal{O}}) \approx 0.062$.

Representation $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$: we get the (non-symmetric) Hausdorff distance $d_H(X|\hat{\mathcal{O}}) \approx 0.751$.

Ideal covariance matrix: Suppose that \mathcal{O} is an orbit of the representation $\phi: G \rightarrow \mathrm{M}_n(\mathbb{R})$, and $\mu_{\mathcal{O}}$ the uniform measure on it. With $x_0 \in \mathcal{O}$ an arbitrary point, the covariance matrix can be written

$$\Sigma[\mu_{\mathcal{O}}] = \int (\phi(g)x_0) \cdot (\phi(g)x_0)^{\top} d\mu_G(g).$$

Now, let $\mathbb{R}^n = \bigoplus_{i=1}^m V_i$ be the decomposition of ϕ into irreps, and denote as $(\Pi[V_i])_{i=1}^m$ the projection matrices on these subspaces. We can decompose

$$\Sigma[\mu_{\mathcal{O}}] = \sum_{i=1}^m C_i \quad \text{where} \quad C_i = \int \phi_i(g) \left(\Pi[V_i](x_0) \cdot \Pi[V_i](x_0)^{\top} \right) \phi_i(g)^{\top} d\mu_G(g).$$

If ϕ is orthogonal, then by Schur's lemma, the C_i are homotheties:

$$\Sigma[\mu_{\mathcal{O}}] = \sum_{i=1}^m \sigma_i^2 \Pi[V_i] \quad \text{where} \quad \sigma_i^2 = \frac{\|\Pi[V_i](x_0)\|^2}{\dim(V_i)}.$$

This shows that, in general, important quantities are:

- The variance $\mathbb{V}[\|\mu_{\mathcal{O}}\|]$, a measure of *deviation from orthogonality* of \mathcal{O}
- The ratio $\sigma_{\max}^2/\sigma_{\min}^2$, a measure of *homogeneity* of \mathcal{O} .

Proposition: Let $\mathcal{O} \subset \mathbb{R}^n$ be the orbit of a representation, potentially non-orthogonal, $\mu_{\mathcal{O}}$ its uniform measure, $\Pi[\langle \mathcal{O} \rangle]$ the projection on its span, and $\sigma_{\max}^2, \sigma_{\min}^2$ the top and bottom nonzero eigenvalues of $\Sigma[\mu_{\mathcal{O}}]$.

Besides, let ν be a measure, $\Sigma[\nu]$ its covariance matrix, $\epsilon > 0$ and $\Pi_{\Sigma[\nu]}^{>\epsilon}$ the projection on the subspace spanned by eigenvectors with eigenvalue at least ϵ .

If $W_2(\mu_{\mathcal{O}}, \nu)$ is small enough, we have the following bound between the pushforward measures after Step 1:

$$\begin{aligned} W_2\left(\sqrt{\Sigma[\mu_{\mathcal{O}}]^{+}}\Pi[\langle \mathcal{O} \rangle]\mu_{\mathcal{O}}, \sqrt{\Sigma[\nu]^{+}}\Pi_{\Sigma[\nu]}^{>\epsilon}\nu\right) \\ \leq 8(n+1)^{3/2}\left(\frac{\sigma_{\max}^3}{\sigma_{\min}^3}\right)\left(\frac{W_2(\mu_{\mathcal{O}}, \nu)}{\sigma_{\min}}\right)^{1/2}\left(\left(\frac{\mathbb{V}[\|\mu_{\mathcal{O}}\|]}{\sigma_{\min}^2}\right)^{1/2} + \frac{W_2(\mu_{\mathcal{O}}, \nu)}{\sigma_{\min}}\right)^{1/2}. \end{aligned}$$

Proof: Consequence of Davis-Kahan theorem, together with

$$\|\Sigma[\mu_{\mathcal{O}}]^{-1/2} - \Sigma[\nu]^{-1/2}\|_{\text{op}} \leq \frac{\sqrt{2}}{\sigma_{\min}^2} \cdot \left(2\mathbb{V}[\|\mu_{\mathcal{O}}\|]^{1/2} + W_2(\mu_{\mathcal{O}}, \nu)\right)^{1/2} \cdot W_2(\mu_{\mathcal{O}}, \nu)^{1/2}.$$

LiePCA operator: Say we observe $X = \{x_1, \dots, x_N\} \subset \mathbb{R}^n$, assumed close to \mathcal{O} .

$$\text{Define } \Lambda: M_n(\mathbb{R}) \rightarrow M_n(\mathbb{R}) \text{ as} \quad \Lambda(A) = \frac{1}{N} \sum_{1 \leq i \leq N} \widehat{\Pi}[N_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle]$$

where

- $\widehat{\Pi}[N_{x_i} X]$ are estimations of projection matrices onto the normal spaces $N_{x_i} \mathcal{O}$,
- $\Pi[\langle x_i \rangle]$ are projection matrices on the lines $\langle x_i \rangle$.

Explanation: On the one hand, $\mathfrak{sym}(\mathcal{O}) = \{A \in M_n(\mathbb{R}) \mid \forall x \in \mathcal{O}, Ax \in T_x \mathcal{O}\}$. Thus,

$$\mathfrak{sym}(\mathcal{O}) = \bigcap_{x \in \mathcal{O}} S_x \mathcal{O} \quad \text{where} \quad S_x \mathcal{O} = \{A \in M_n(\mathbb{R}) \mid Ax \in T_x \mathcal{O}\}.$$

On the other hand, considering only X , one has

$$\bigcap_{i=1}^N S_{x_i} \mathcal{O} \approx \ker \left(\sum_{i=1}^N \Pi[(S_{x_i} \mathcal{O})^\perp] \right),$$

Last, the authors showed that $\Pi[(S_{x_i} \mathcal{O})^\perp](A) = \Pi[N_{x_i} \mathcal{O}] \cdot A \cdot \Pi[\langle x_i \rangle]$.

Ideal Lie-PCA: Suppose that \mathcal{O} is an orbit of the representation $\phi: G \rightarrow M_n(\mathbb{R})$, and $\mu_{\mathcal{O}}$ its uniform measure. We define

$$\Lambda_{\mathcal{O}}(A) = \int \Pi[N_x \mathcal{O}] \cdot A \cdot \Pi[\langle x \rangle] d\mu_{\mathcal{O}}(x).$$

Proposition: Its kernel is equal to $\mathfrak{sym}(\mathcal{O})$. Moreover, when $\mathcal{O} = S^{n-1}$, its nonzero eigenvalues are exactly δ_n and δ'_n where

$$\delta_n = \frac{2(n-1)}{n(n(n+1)-2)} \quad \text{and} \quad \delta'_n = \frac{1}{n}.$$

Proof: Show that $\Lambda_{\mathcal{O}}$ is equivariant with respect to the action of G by conjugation:

$$\phi(g)\Lambda(A)\phi(g)^{-1} = \Lambda\left(\phi(g)A\phi(g)^{-1}\right)$$

Then use Schur's lemma.

Empirical observation: More generally, the nonzero eigenvalues of $\Lambda_{\mathcal{O}}$ belong to $[1/n^2, 1/n]$ when \mathcal{O} is *homogenous*, i.e., $\sigma_{\max}^2/\sigma_{\min}^2 = 1$.

Stability: Comparing

$$\Lambda(A) = \sum_{1 \leq i \leq N} \widehat{\Pi}[\mathbf{N}_{x_i} X] \cdot A \cdot \Pi[\langle x_i \rangle] \quad \text{and} \quad \Lambda_{\mathcal{O}}(A) = \int \Pi[\mathbf{N}_x \mathcal{O}] \cdot A \cdot \Pi[\langle x \rangle] d\mu_{\mathcal{O}}(x).$$

amounts to quantifying the quality of normal space estimation. We use local PCA:

$$\widehat{\Pi}[\mathbf{N}_{x_i} X] = I - \Pi_{x_i}^{l,r}[X],$$

where $\Pi_{x_i}^{l,r}[X]$ is the projection matrix on any l top eigenvectors of the *local covariance matrix* $\Sigma_{x_i}^r[X]$ centered at x_i and at scale r , itself defined as

$$\Sigma_{x_i}^r[X] = \frac{1}{|Y|} \sum_{y \in Y} (y - x_i)(y - x_i)^\top,$$

where $Y = \{y \in X \mid \|y - x_i\| \leq r\}$, the set input points at distance at most r from x_i .

Measure-theoretic formulation: If μ is a measure on \mathbb{R}^n , we define its *local covariance matrix* centered at x at scale r as

$$\Sigma_x^r[\mu] = \int_{\mathcal{B}(x, r)} (y - x)(y - x)^\top \frac{d\mu(x)}{\mu(\mathcal{B}(x, r))}.$$

Bias-variance tradeoff: Let $\mu_{\mathcal{M}}$ be measure on a submanifold $\mathcal{M} \subset \mathbb{R}^n$ of dimension l , $x \in \mathcal{M}$, ν a measure on \mathbb{R}^n and $y \in \text{supp}(\nu)$. We decompose

$$\left\| \frac{1}{l+2} \Pi[T_x \mathcal{M}] - \frac{1}{r^2} \Sigma_y^r[\nu] \right\|_F \leq$$

$$\underbrace{\left\| \frac{1}{l+2} \Pi[T_x \mathcal{M}] - \frac{1}{r^2} \Sigma_x^r[\mu_{\mathcal{M}}] \right\|_F}_{\text{consistency}} + \underbrace{\left\| \frac{1}{r^2} \Sigma_x^r[\mu_{\mathcal{M}}] - \frac{1}{r^2} \Sigma_y^r[\mu_{\mathcal{M}}] \right\|_F}_{\text{spatial stability}} + \underbrace{\left\| \frac{1}{r^2} \Sigma_y^r[\mu_{\mathcal{M}}] - \frac{1}{r^2} \Sigma_y^r[\nu] \right\|_F}_{\text{measure stability}}$$

Lemma: If the parameters are chosen correctly, this is

$$\lesssim r + \|x - y\| + \left(\frac{W_2(\mu, \nu)}{r^{l+1}} \right)^{\frac{1}{2}}.$$

Corollary: We deduce a bound between Lie-PCA operators:

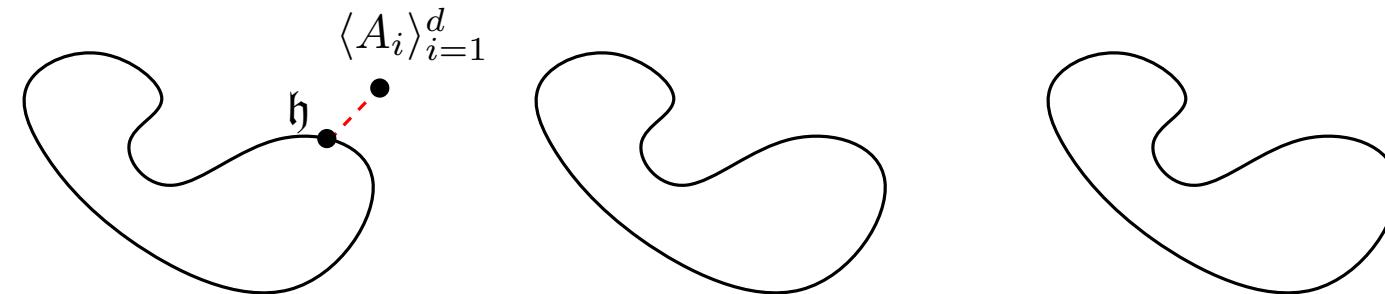
$$\|\Lambda_{\mathcal{O}} - \Lambda\|_{\text{op}} \leq \sqrt{2}\rho \left(r + 4 \left(\frac{W_2(\mu_{\mathcal{O}}, \mu_X)}{r^{l+1}} \right)^{1/2} \right).$$

In Step 3, we consider the bottom eigenvectors A_1, \dots, A_d of Lie-PCA, and solve

$$\min \sum_{i=1}^d \|\Lambda(A_i)\|^2 \quad \text{s.t.} \quad \langle A_1, \dots, A_d \rangle \in \mathcal{G}^{\text{Lie}}(G, \mathfrak{gl}(n)).$$

with $\mathcal{G}(G, \mathfrak{so}(n))$ the subspace of $\mathfrak{so}(n)$ consisting of the Lie subalgebras pushforward of G by a representation.

The set $\mathcal{G}(G, \mathfrak{so}(n))$ has many connected components, one for each *orbit-equivalence* class of representations. We want to make sure that the minimizer belongs to the correct connected component.

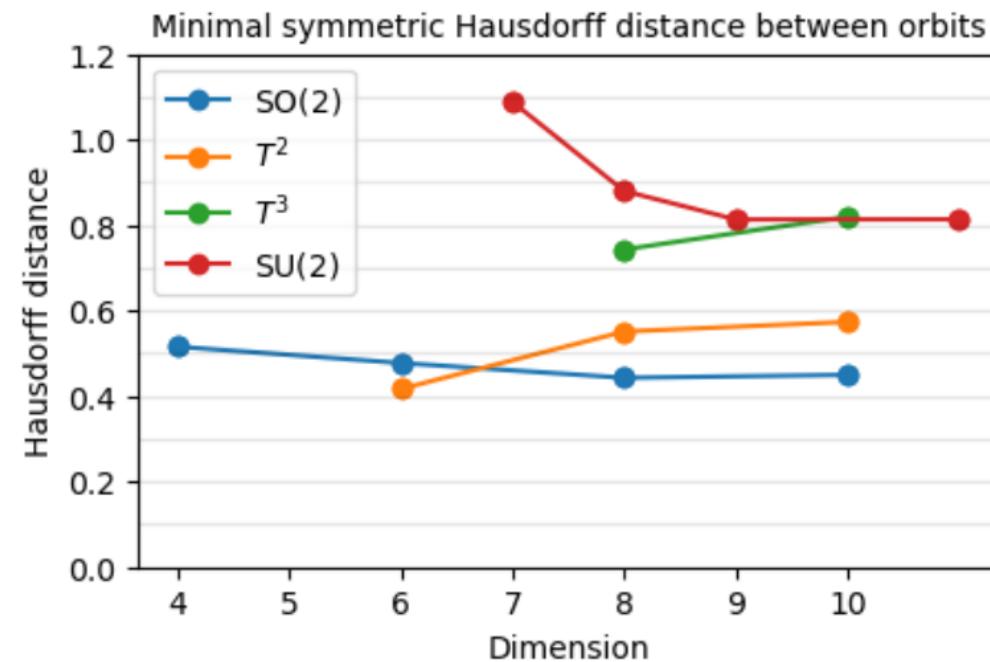
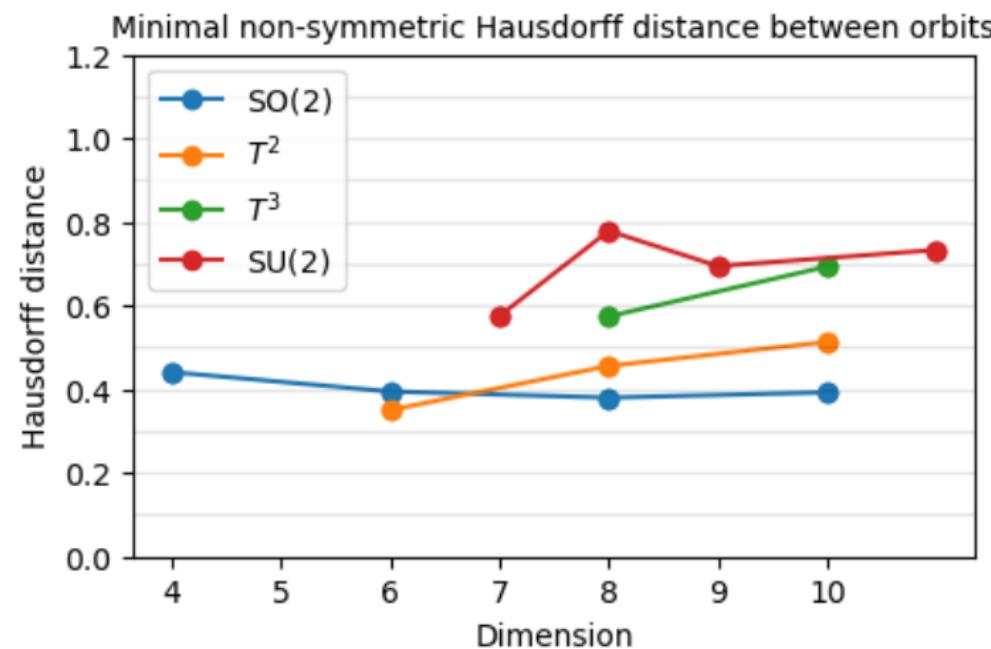


The distance from $\langle A_i \rangle_{i=1}^d$ to \mathfrak{h} must be lower than the *reach* of $\mathcal{G}(G, \mathfrak{so}(n))$. In this context, it is called *rigidity*:

$$\Gamma(G, n) = \inf \|\Pi[\mathfrak{h}] \Pi[\mathfrak{s}^\perp]\|^2 \quad \text{s.t.} \quad \mathfrak{h} \in \mathcal{G}^{\text{Lie}}(G, \mathfrak{gl}(n)), \mathfrak{s} \in \mathcal{G}^{\text{Lie}}(H, \mathfrak{gl}(n)), s \not\simeq \mathfrak{h}.$$

Lemma: Consider the subset of $\mathcal{G}(G, \mathfrak{so}(n))$ with weights at most ω_{\max} . Then

$$\Gamma(G, n, \omega_{\max}) \geq 4/(n\omega_{\max}^2).$$



Left: empirical estimation of the minimal non-symmetric Hausdorff distance $d_H(\widehat{\mathcal{O}}_x^1 | \widehat{\mathcal{O}}_x^2)$ between two orbits of a same initial point x for two non-orbit equivalent representations ϕ_1, ϕ_2 of a compact Lie group G in \mathbb{R}^n . The minimal value is approximately 0.35.

Right: same for the symmetric Hausdorff distance $d_H(\widehat{\mathcal{O}}_x^1, \widehat{\mathcal{O}}_x^2)$. The minimal value is 0.42.

Running time (in seconds or minutes) and success rate (percentage) of full execution of `LieDetect`, as a function of the input group, and the dimension of the ambient Euclidean space. The input of the algorithm is a point cloud sampled from the uniform measure on an orbit chosen randomly.

For the Abelian groups $\text{SO}(2)$, T^2 , and T^3 , the representations are considered up to a maximal frequency, 100 runs of the algorithm are performed, and the results are averaged. For $\text{SU}(2)$, 10 runs have been performed.

Dimension	4	6	8	10
Running time	0.04s	0.05s	0.08s	0.14s
Success	100.0%	100.0%	100.0%	100.0%

(a) $\text{SO}(2)$

Dimension	6	8	10
Running time	0.24s	0.63s	4.03s
Success	82.0%	100.0%	98.0%

(b) T^2

Dimension	8	10
Running time	1.44s	5.98s
Success	100.0%	100.0%

(c) T^3

Dimension	4	5	7	8	9	10
Running time	0.6s	5.04s	4min 21s	13min 7s	16min 9s	10min 53s
Success	100.0%	100.0%	90.0%	100.0%	100.0%	100.0%

(d) $\text{SU}(2)$