

Distributed Consensus on Manifolds using the Riemannian Center of Mass

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August 16, 2023

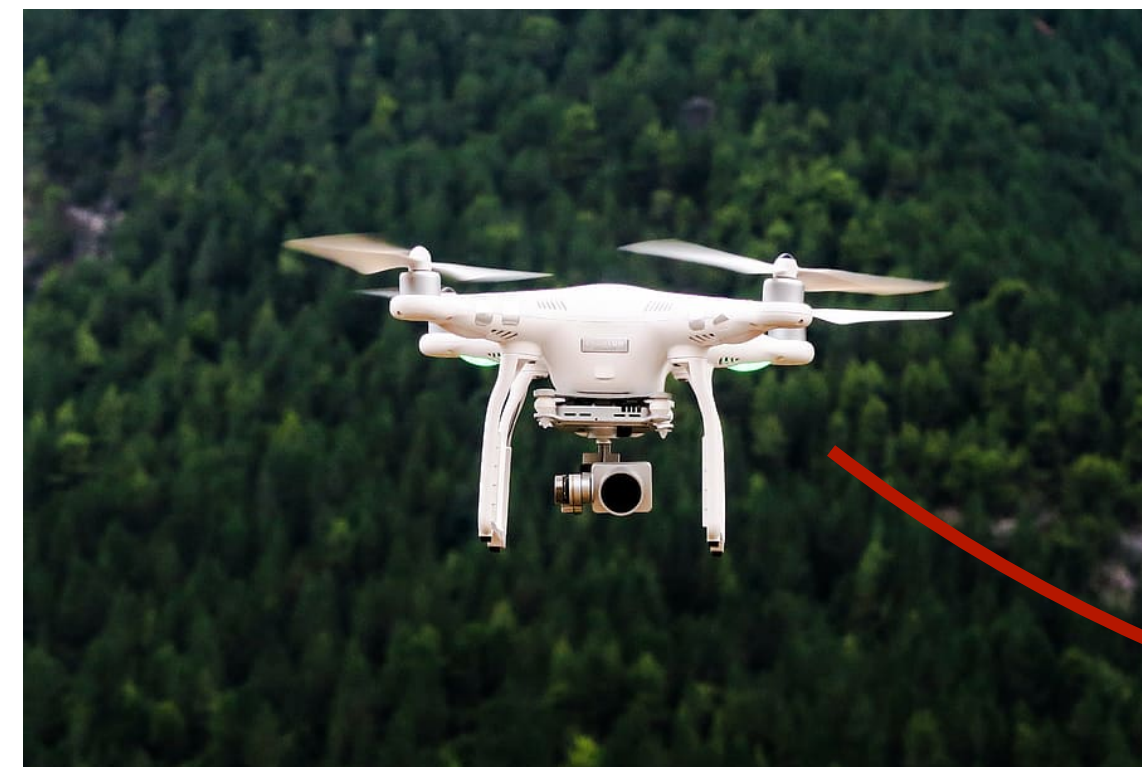


motivation for consensus on manifolds

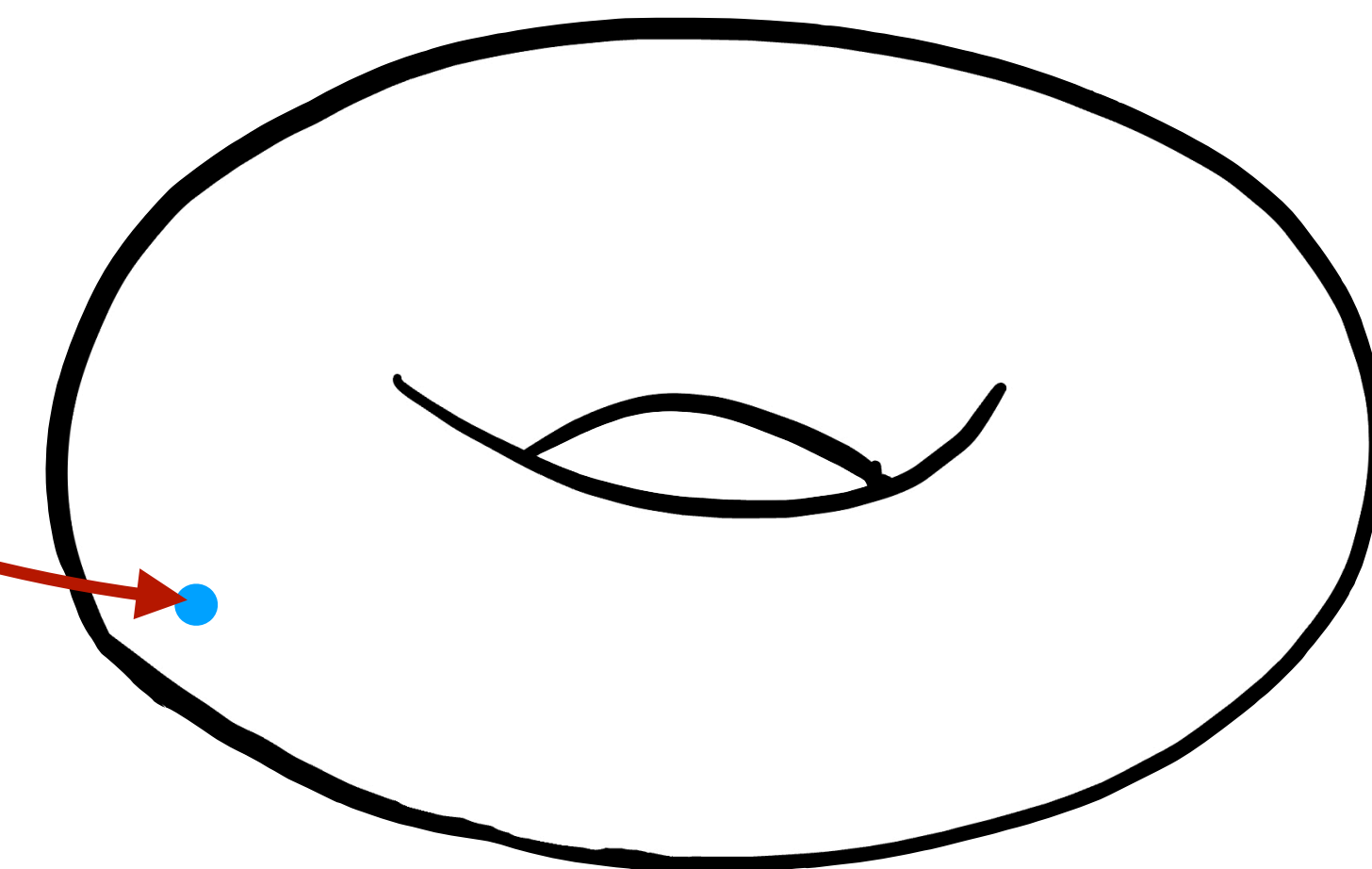
- consensus is central to **distributed computation**
 - Steer a set of agents to a **single point**
 - Studied mostly on Euclidean spaces
- Robots with **non-Euclidean** state spaces
- **Formation control** on manifolds



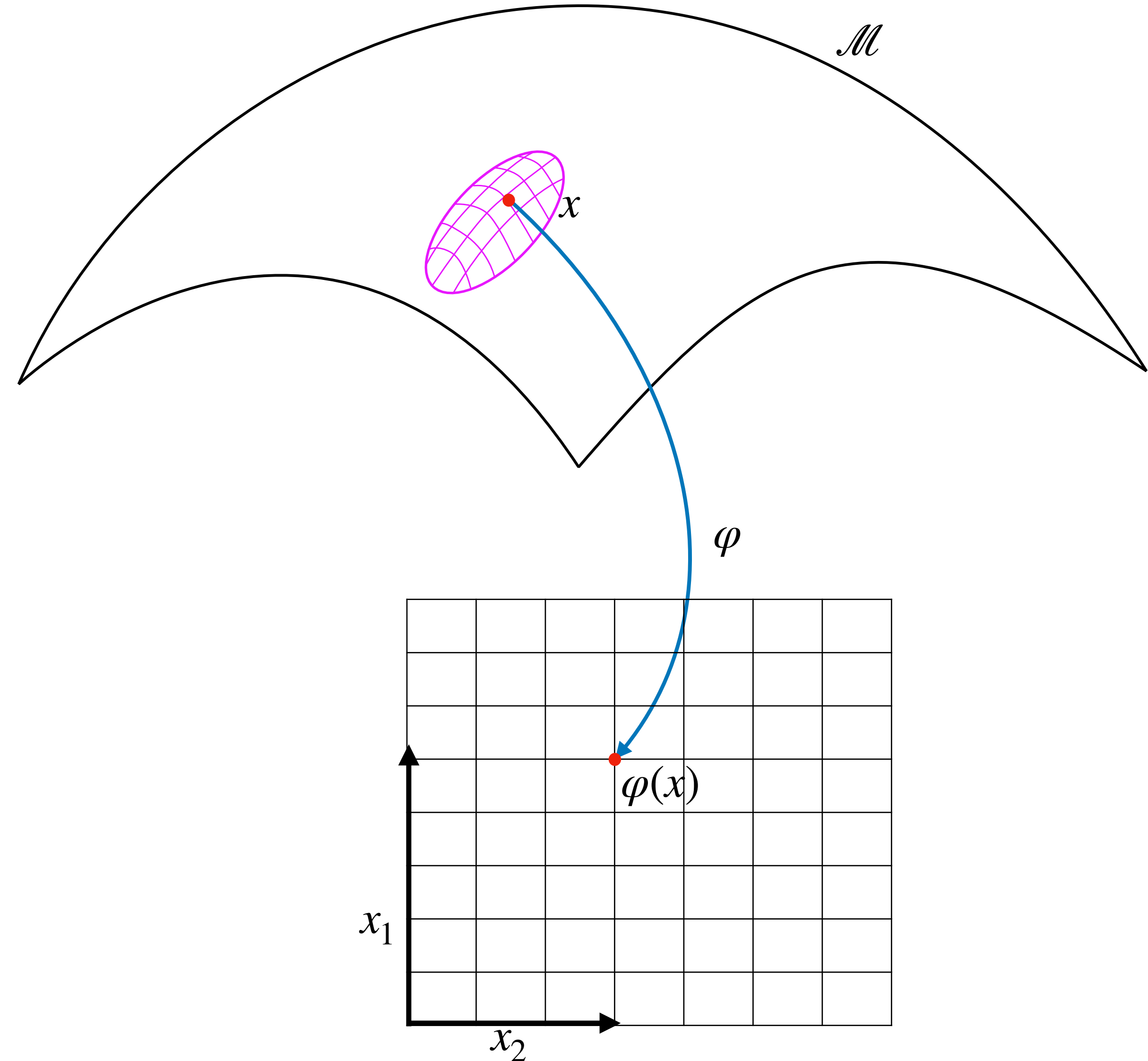
The Canadarm2 has a **non-Euclidean** state space.



The state space of a dome camera is a **torus** $S^1 \times S^1$.



- **smooth manifold:**
 - topological space $\mathcal{M} \subset \mathbb{R}^n$
 - locally Euclidean
- **Riemannian metric:**
 - induces metric space structure $d_g(\cdot, \cdot)$
- intrinsic vs extrinsic quantities



what is an average on a Riemannian manifold?

- average solves $\min_{x \in \mathbb{R}^n} \sum_{i=1}^N \|x - z_i\|^2$

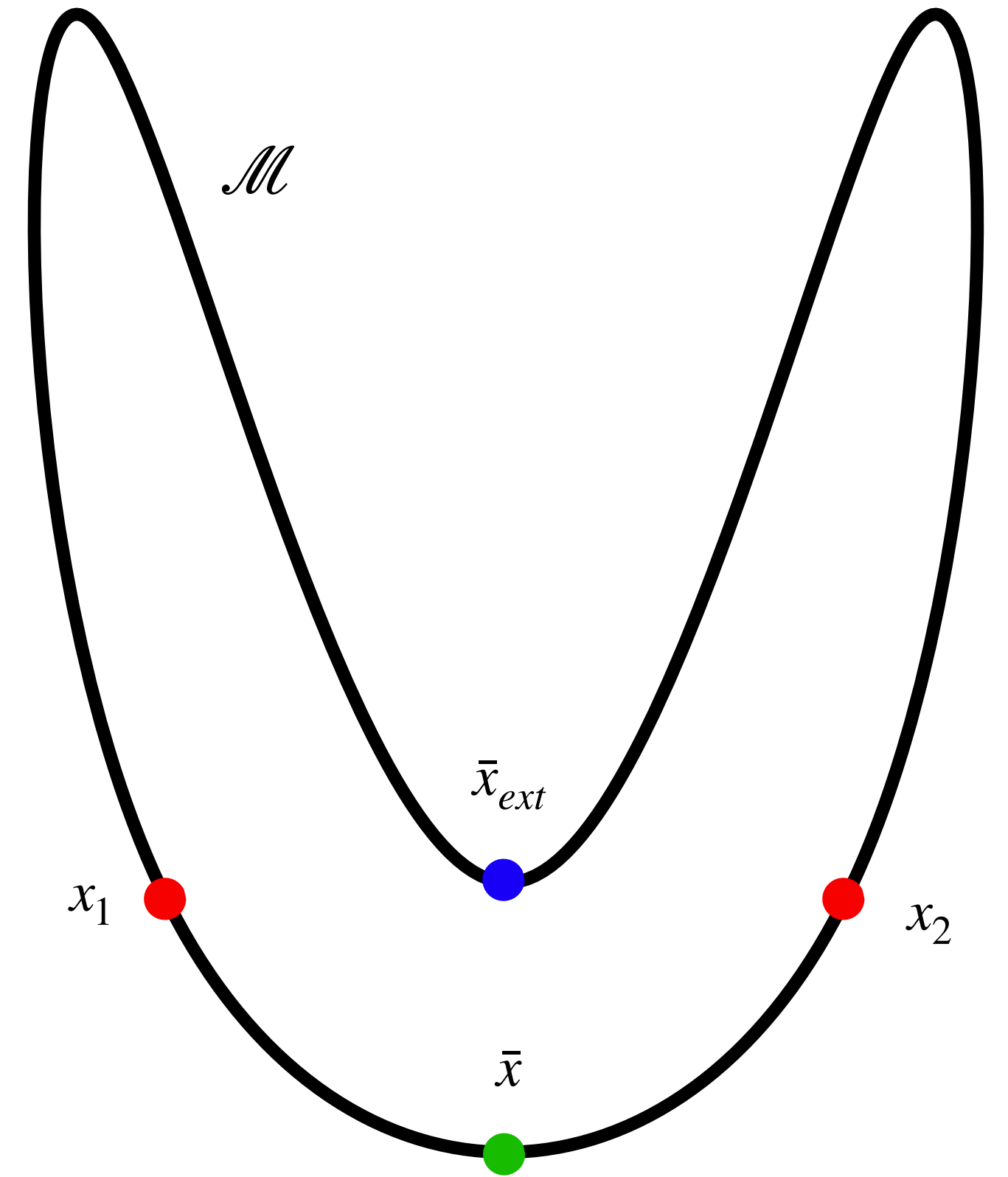
- Riemannian center of mass (RCM) \bar{z} solves

$$\min_{x \in \mathcal{M}} f(x) := \sum_{i=1}^n d_g(x, z_i)^2$$

- intrinsic better than extrinsic: $\bar{z}_{ext} = \text{proj}_{\mathcal{M}} \left(\frac{1}{N} \sum_{i=1}^N z_i \right)$

- applications

- medical imaging
- weather



The RCM (green) vs. extrinsic mean (blue) of a pair of points (red) [Afsari, 2011]

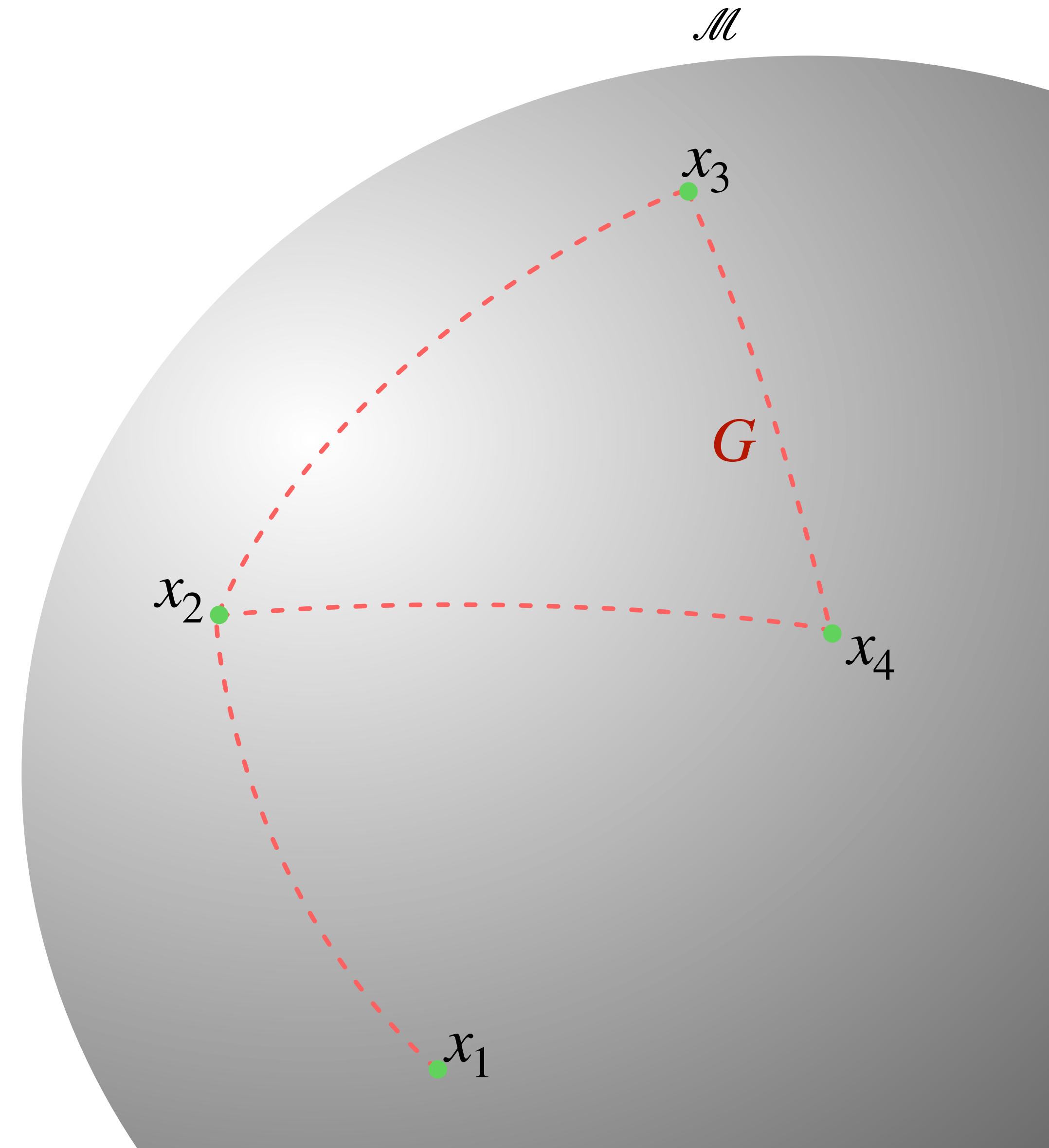
problem formulation

consider:

- Riemannian manifold \mathcal{M}
- agents $x_1, \dots, x_N \in \mathcal{M}$
 - communication graph G
 - $x_i(k+1) = F_i(k, x_m(k) : m \sim i)$

problem:

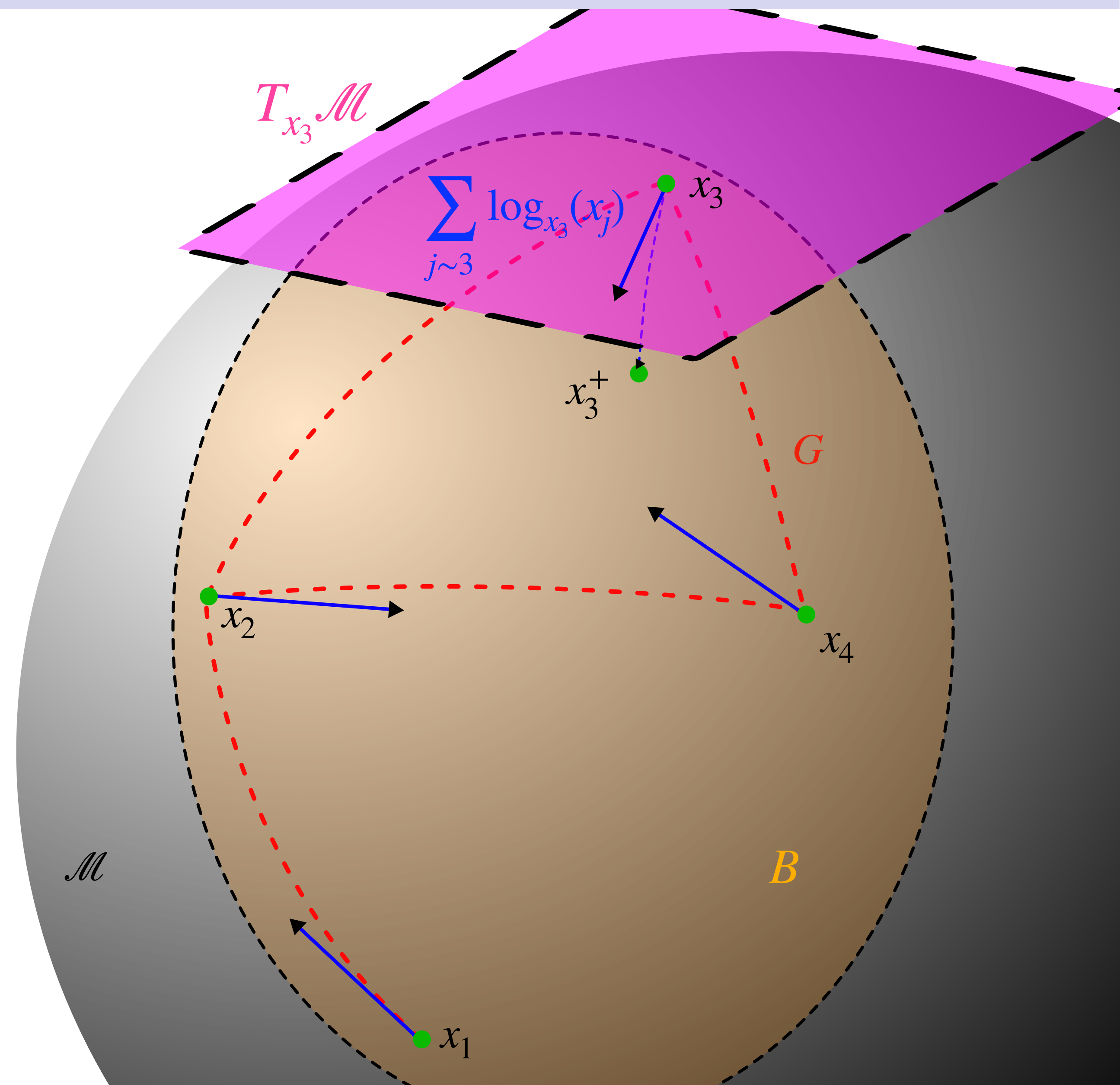
- design **intrinsic** + **distributed** dynamics to **synchronize** agents
 - no projection, no embedding



consensus on manifolds

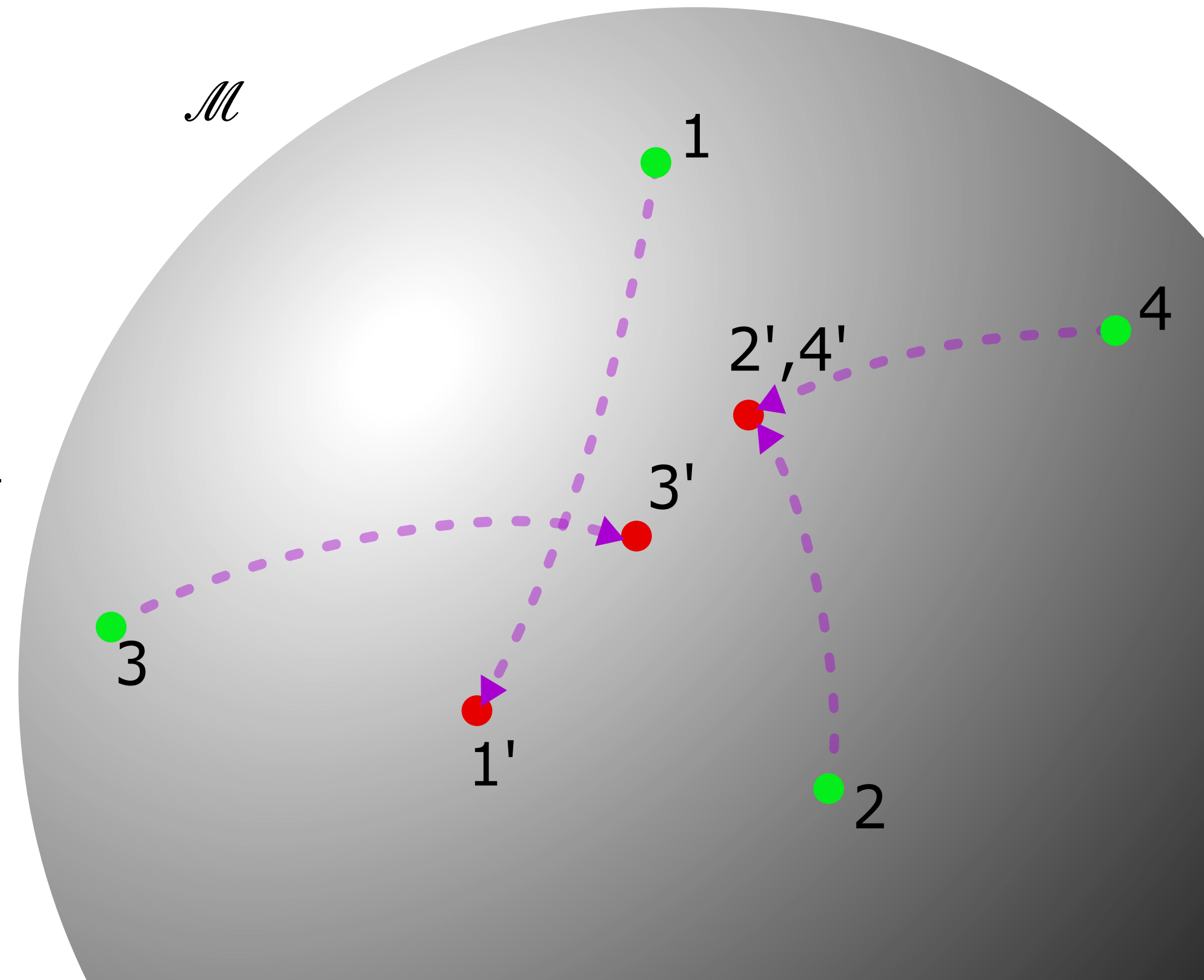
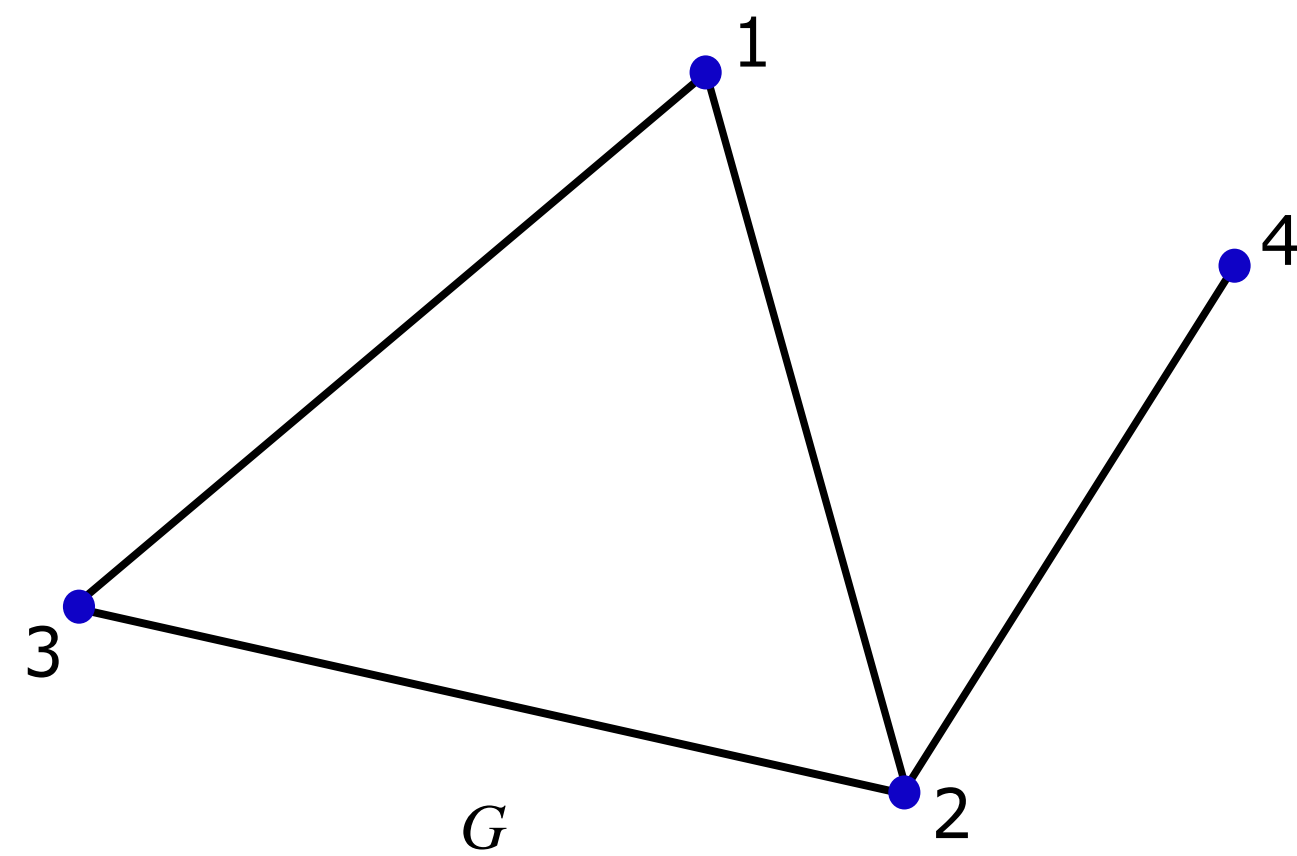
- Consensus dynamics for Euclidean space:
 - $x_i(0) = z_i$
 - $x_i^+ = x_i + \epsilon \sum_{j \sim i} (x_j - x_i)$
- Simplest consensus dynamics for a manifold [Tron et al, 2013]:
 - $x_i(0) = z_i$
 - $x_i^+ = \exp_{x_i} \left(\epsilon \sum_{j \sim i} \log_{x_i}(x_j) \right)$

theorem: if $\{x_i(0)\}$ are initialized within a geodesically convex ball B and $\{x_i(k)\} \subset B$ for all k , then consensus is guaranteed.



- **intuition:** each agent moves towards neighbors' average
- **Round robin approach:** one agent moves per time step

$$x_i(k+1) = \begin{cases} \text{RCM}(x_j(k) : j \sim i) & i - 1 \equiv k \pmod{N} \\ x_i(k) & \text{else} \end{cases}$$



guarantees

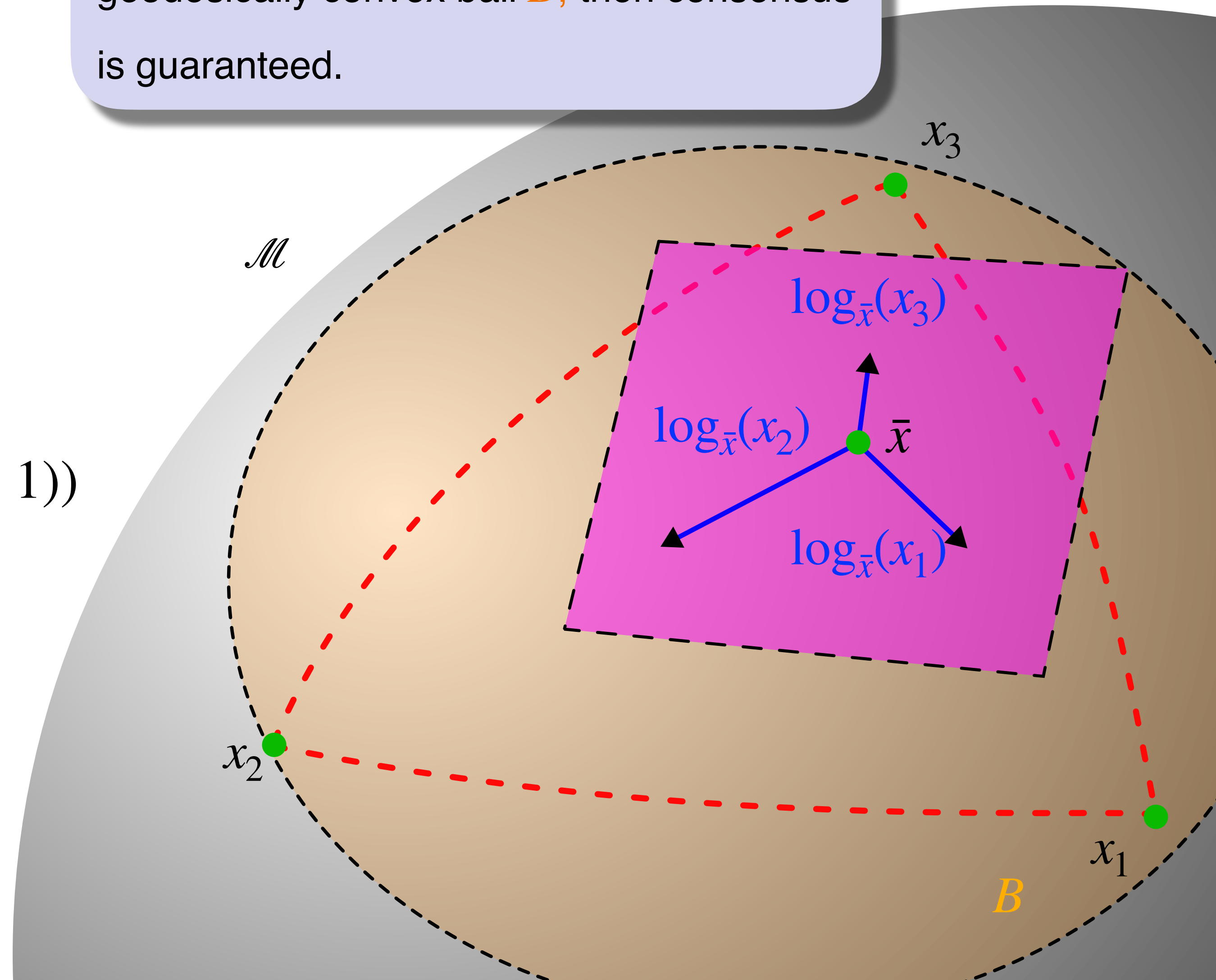
- $x_i(k + 1)$ is in strict interior of geodesic convex hull of $\{x_j(k) : j \sim i\}$ (B) (**not true for extrinsic mean**)

- geodesic variance

$$\varphi(\mathbf{x}(k)) = \frac{1}{2} \sum_{\{i,j\} \in E} d_g(x_i(k), x_j(k))^2 > \varphi(\mathbf{x}(k + 1))$$

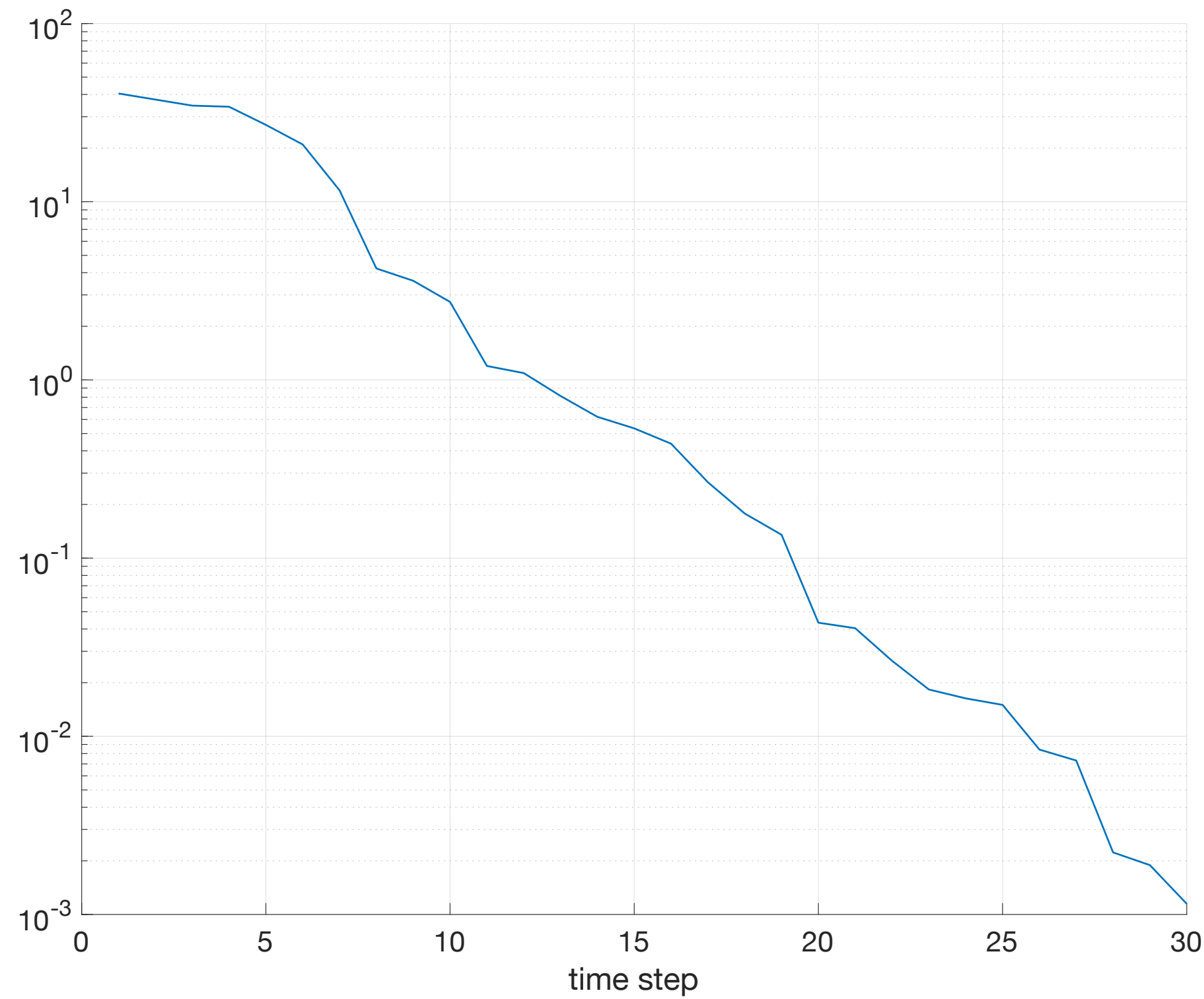
- $\varphi(\mathbf{x}(k + N)) = \varphi(\mathbf{x}(k))$ iff $x_1(k) = \dots = x_N(k)$

theorem: if $\{x_i(0)\}$ are initialized within a geodesically convex ball B , then consensus is guaranteed.



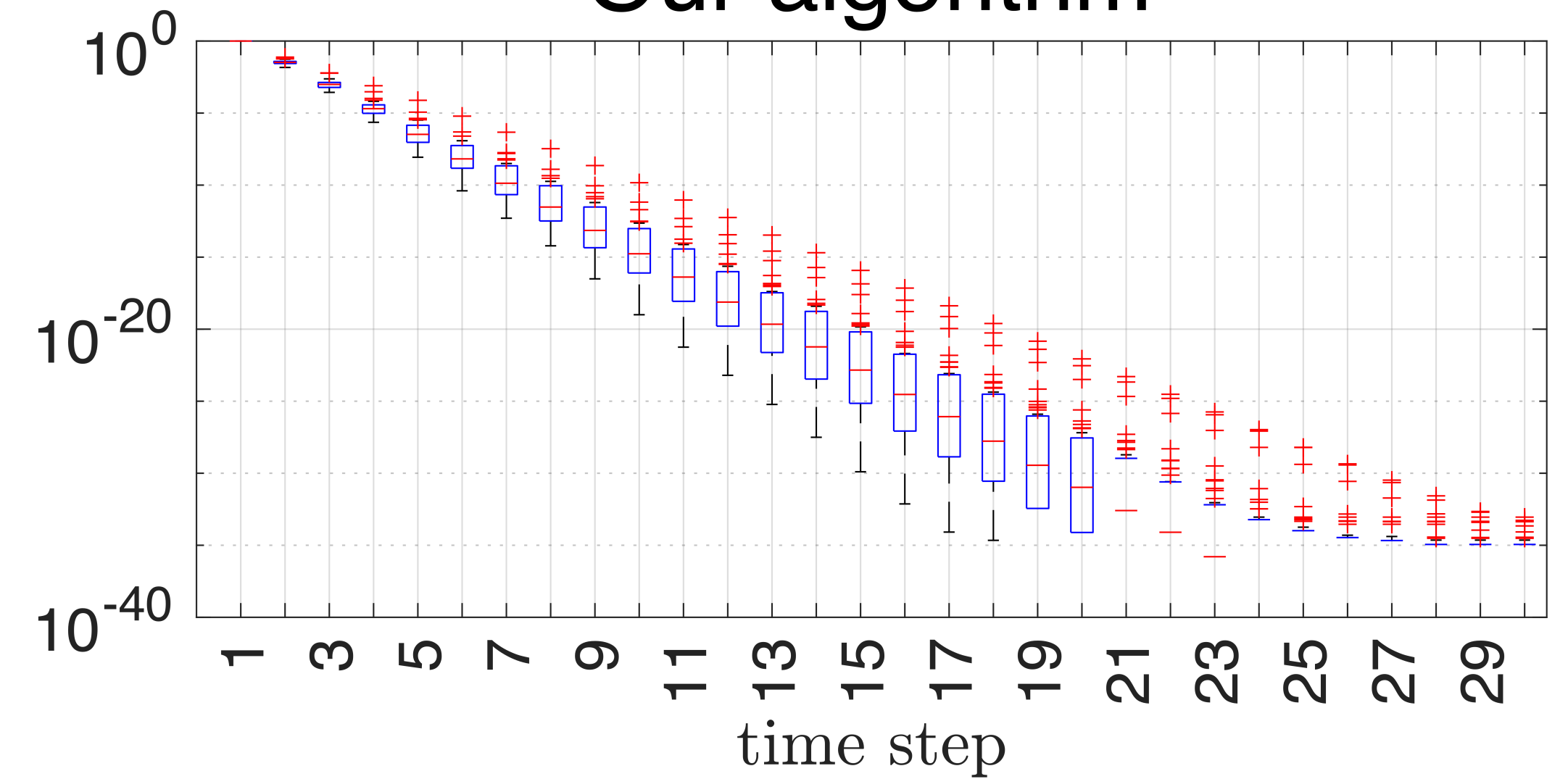
numerical simulation

- average cost over 40 runs
- our algorithm (top) vs [2] (bottom)

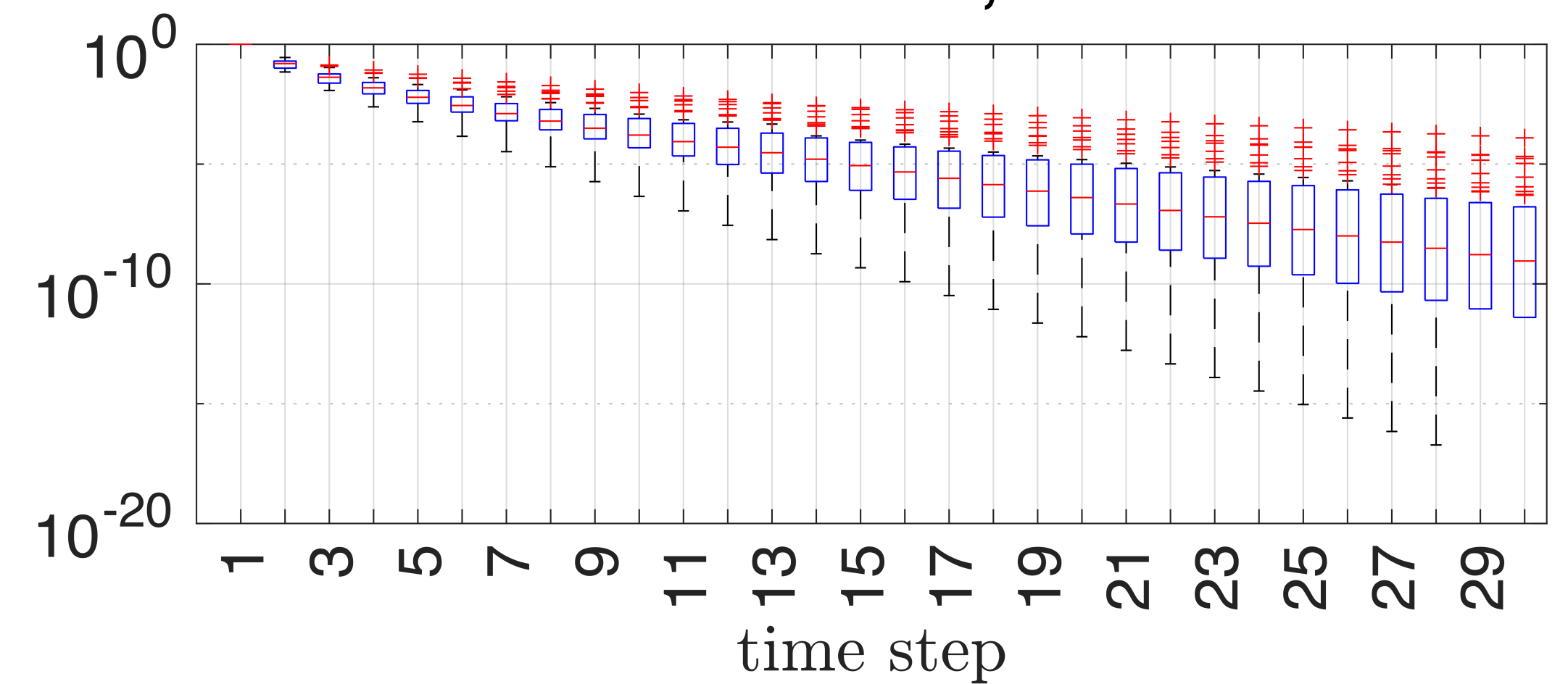


linear decay of geodesic variance

Our algorithm



Tron et al., 2013



concluding remarks

Euclidean consensus

dynamics:

$$x_i^+ = x_i + \epsilon \sum_{j \sim i} (x_j - x_i)$$

Results

$$\lim_{k \rightarrow \infty} x_i(k) = \frac{1}{N} \sum_{i=1}^N z_i$$

Riemannian consensus

dynamics:

$$x_i^+ = \exp_{x_i} \left(\epsilon \sum_{j \sim i} \log_{x_i}(x_j) \right)$$

Results

$$\lim_{k \rightarrow \infty} x_i(k) = ? \neq \text{RCM}(z_1, \dots, z_N)$$

Goal: Given $z_1, \dots, z_N \in M$ and a communication graph $\mathcal{G} = ([N], E)$ find discrete dynamics

$$x_i(k+1) = F_i \left(x_j(k) : j \in N_i \cup \{i\} \right)$$

such that

$$\lim_{k \rightarrow \infty} x_i(k) = \text{RCM}(z_1, \dots, z_N).$$

- **in this talk:**
 - consensus on Riemannian manifolds
 - asynchronous approach with convergence guarantees
- **to be continued:**
 - agents move in random order
 - consensus to the Riemannian center of mass (CDC)