

Stable reconstruction of simple Riemannian manifolds from unknown interior sources

Inverse problems and nonlinearity

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In collaboration with:

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- 1 The problem
 - The setup
 - The data
 - The goal
- 2 Related problems
- 3 Labeled Gromov–Hausdorff distance
- 4 Theorems

The setup

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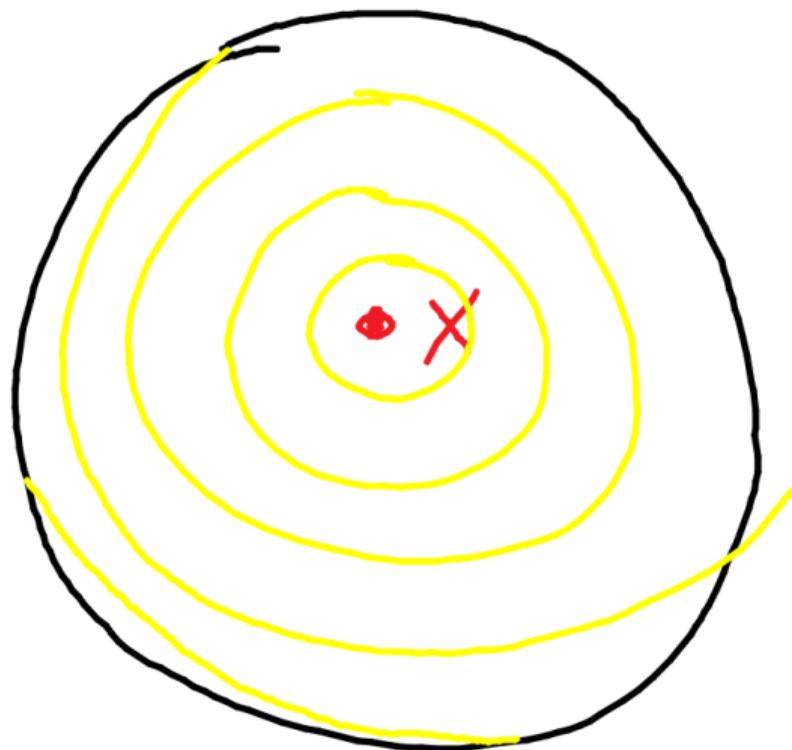
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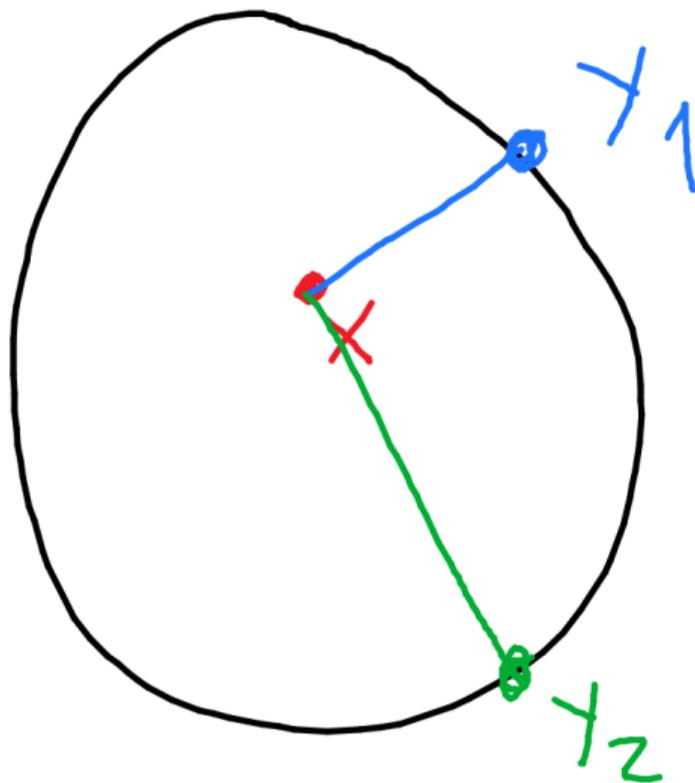
At least approximately?

The setup



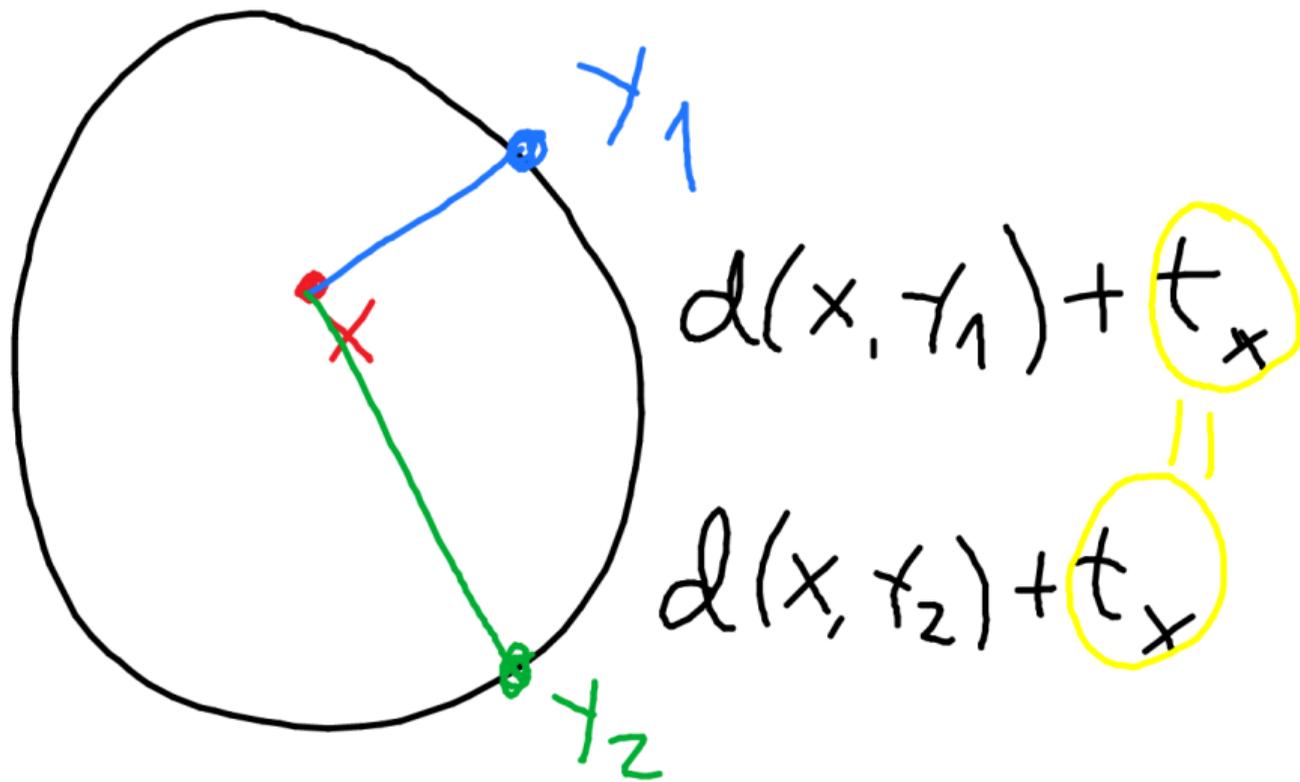
A point $x \in M$ sends waves.

The setup



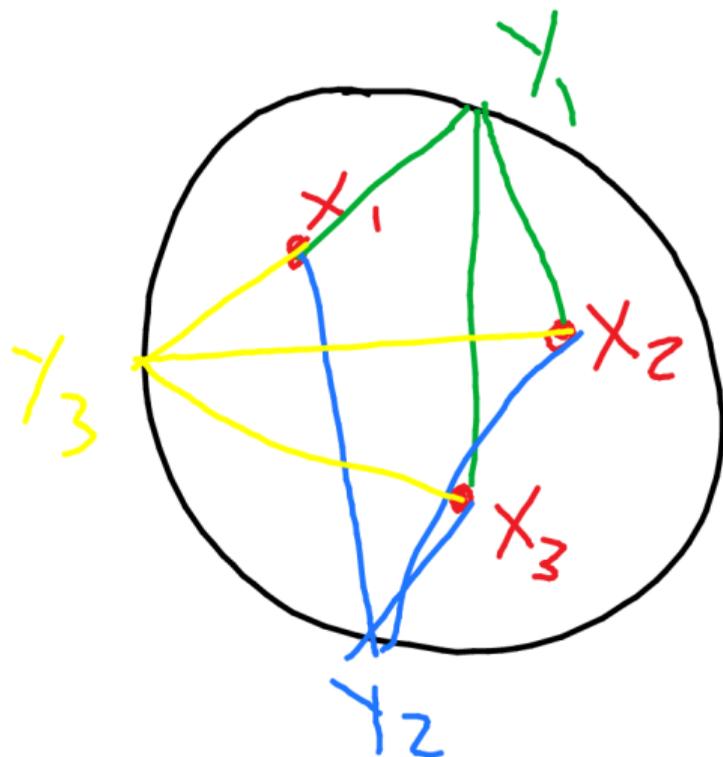
Arrival times from x are measured at y_1 and y_2 .

The setup



The data contains the distances offset by the time the source at x went off.

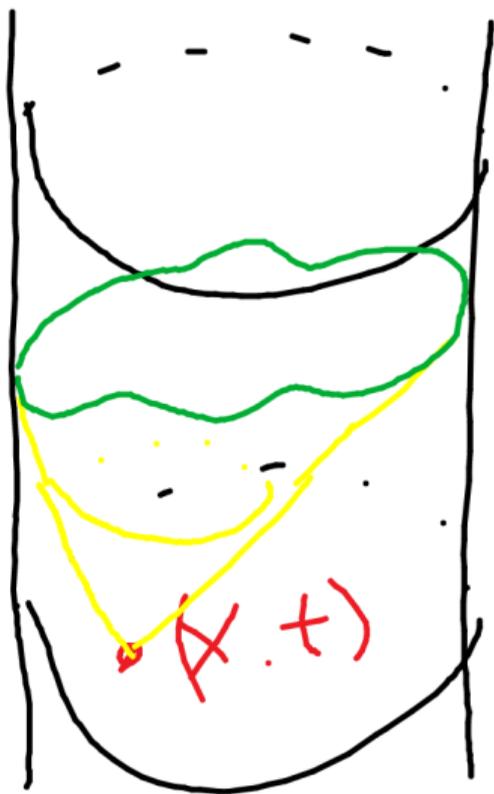
The setup



x_i goes off
at time t_i

Multiple sources measured from multiple stations. Offsets in time depend on x_i but not y_j .

The setup



observation for (x, t)
is light cone inter-
sected with boundary

$$Q(S) \subset \mathcal{M} \times \mathbb{R}$$

Sources and measurements in the spacetime.

The setup

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The data is the set of all arrival times without any labels:

$$Q(S) = \{(y, \tau(s) + d(y, \pi(s))); y \in \partial M, s \in S\} \subset \partial M \times \mathbb{R}.$$

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Question

To what extent does $Q(S)$ determine M ?

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- Approximate in what sense?
- If we approximate M as a metric space (Gromov–Hausdorff), then how is this reconstruction attached to the known boundary?

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 - Sources on the boundary
 - Interior sources with additional data
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- Sources everywhere on the boundary: *boundary rigidity*. (Michel '81, Gromov '83, Croke '91, Pestov–Uhlmann '05, Stefanov–Uhlmann '05, Burago–Ivanov '10)

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Interior sources with additional data

- Typically one assumes that geometric data is collected separately for each (unlabeled) interior point, but to us all the information comes as a single set.
- Known origin times, neatly separated sources everywhere: *boundary distance data*. (Kurylev '97, de Hoop-I-Lassas-Saksala '19)
- Unknown origin times, neatly separated sources everywhere: *boundary distance difference data*. (Lassas-Saksala '19, de Hoop-Saksala '19, Ivanov '20)

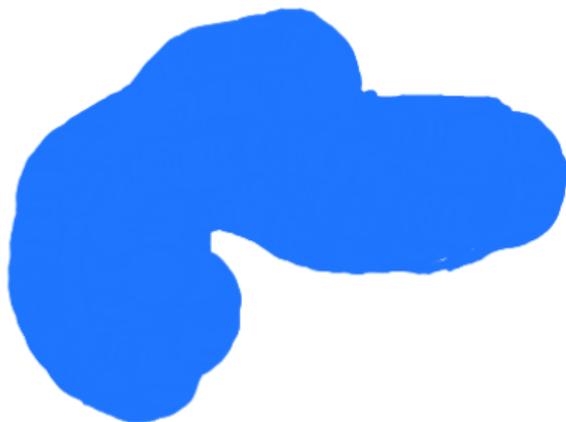
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 - Hausdorff distance
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Hausdorff distance

The Hausdorff distance is a distance between two compact subsets of a metric space.

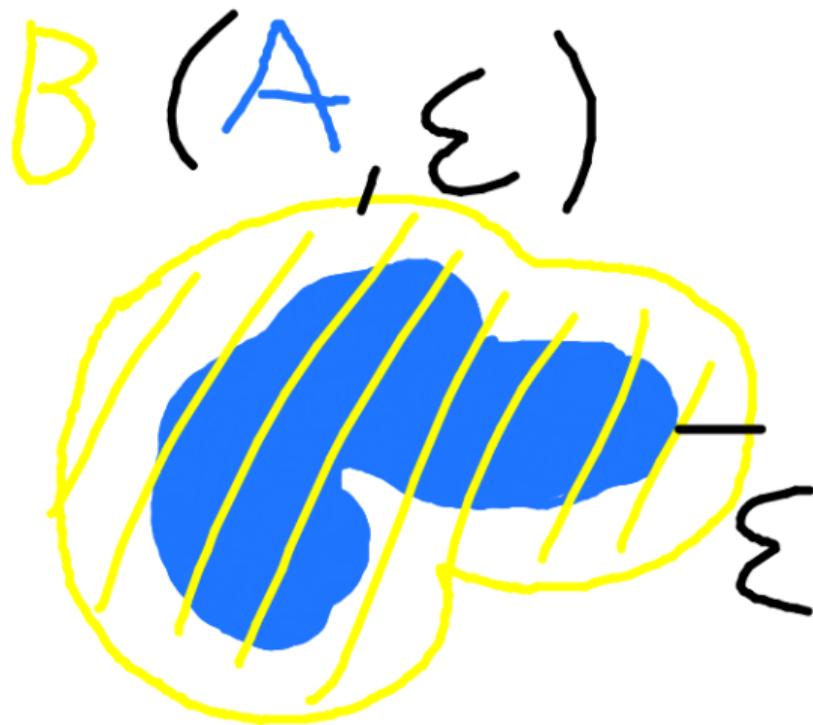
Hausdorff distance

$A \subset \mathbb{R}^2$



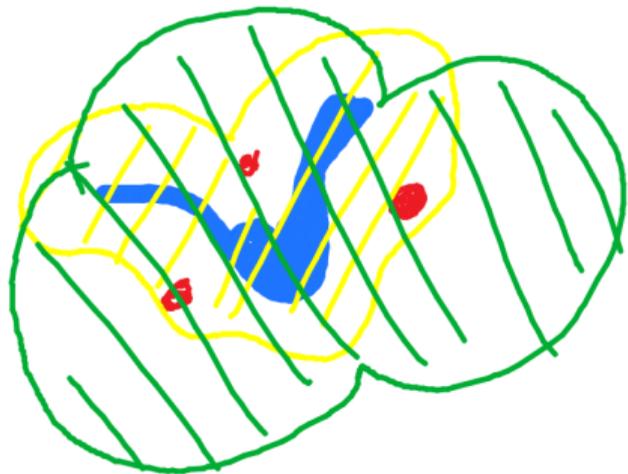
A compact set in the plane.

Hausdorff distance



A ball of radius ϵ around the set.

Hausdorff distance



$$\left. \begin{array}{l} A \subset B(A', \varepsilon) \\ A' \subset B(A, \varepsilon) \end{array} \right\} \Rightarrow d_H(A, A') \leq \varepsilon$$

Two sets within each other's balls.

Hausdorff distance

The Hausdorff distance between $A, A' \subset \mathbb{R}^2$ is the infimum of those ε s so that they are contained in each other's ε -balls.

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The Gromov–Hausdorff distance is a distance between two compact metric spaces.
(No fixed ambient geometry.)

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- The infimum (over all choices of Z) of these Hausdorff distances is the Gromov–Hausdorff distance $d_{GH}(X, Y)$.

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- The labeled Gromov–Hausdorff distance between (X, α) and (Y, β) is

$$d_{GH}^L(X, \alpha; Y, \beta) = \inf \{ d_{H,Z}(f(X), g(Y)) + \sup_{\ell \in L} d_Z(f(\alpha(\ell)), g(\beta(\ell))) \};$$

Z is a compact metric space,

$f: X \rightarrow Z$ and $g: Y \rightarrow Z$ are isometric embeddings.

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- Our construction of an approximate manifold P must come with an approximate boundary assignment function $\alpha: \partial M \rightarrow P$ so that

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- Then the approximate manifold is constructed in some relation to the known boundary, not floating unattached.

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- The labeled Gromov–Hausdorff distance is a metric:
 - 1 Symmetry.
 - 2 Triangle inequality.
 - 3 $d_{GH}^L(X, \alpha; Y, \beta) = 0$ if and only if there is an isometry $h: X \rightarrow Y$ so that $h \circ \alpha = \beta$.

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 - Quantitative simplicity
 - Perfect reconstruction of source points
 - Approximate reconstruction
 - Perfect reconstruction with infinite time

Quantitative simplicity

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A Riemannian manifold satisfies our estimates if and only if it is simple.

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$$\text{diameter} \times \sqrt{\max(\text{sectional curvature})} < \pi.$$

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This holds on many but not all simple manifolds.

Perfect reconstruction of source points

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We can also get some information on the distance to boundary points, but nothing perfect.

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Let M be a pretty simple Riemannian manifold and $S \subset M \times \mathbb{R}$ a discrete source set. The data set $Q(S)$ determines a metric space P and a map $\alpha: \partial M \rightarrow P$ so that

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where the error ε is explicit.

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where the error ε is explicit.

From the data and a priori geometric bounds we get an approximate finite model and an estimate on the reconstruction error!

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If we measure for an increasing amount of time, we get an increasingly good approximate reconstruction.

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Myers–Steenrod: Metric isometry implies smooth isometry.

Thank you!

Summary:

- Unknown source points, unknown origin times, unlabeled arrival time data.
- Approximate reconstruction in labeled Gromov–Hausdorff distance.
- Perfect reconstruction in infinite or increasing time.
- arXiv:2102.11799

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