

Scattered Data Interpolation in Complex Domains

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Outline

- Sensor networks as scattered data problem
- Overview of existing scattered data interpolation methods
- Representation of complex domains
- A distance metric for complex domains
- Visualizing sensor network data

Sensor Networks as Scattered Data Problem

Scattered Data Problem

- Given: A vector $S = ((p_1, v_1), \dots, (p_N, v_N))$ of N samples
 - p_i : Sample position (*site*) in some domain D
 - v_i : Sample value in some range R
- Samples are assumed to be measurements of some (unknown) function $F: D \rightarrow R$, i. e., $\forall i = 1, \dots, N : F(p_i) = v_i$
- Scattered data problem: Reconstruction of F based only on sample vector S

Interpreting Sensor Network Data

- p_i are sensor positions, v_i are sensor measurements
- For time-varying sensor data: S is time-varying, but the p_i stay fixed
- For mobile sensors: The p_i vary as well
- Analysis of sensor data requires reconstructing the measured field first

Interpreting Sensor Network Data

- Leads to three levels of scattered data problems
 - Static sites and static values
 - Static sites and dynamic values
 - Dynamic sites and dynamic values
- Additional fourth level: Optimizing sensor placement based on reconstruction result

Existing Scattered Data Methods

Scattered Data Interpolation

- Defined by distance metric $d: D \times D \rightarrow \mathbf{R}^+$
- Several classifications possible
 - Direct (gridless) methods: Rely only on d
 - Indirect (grid-based) methods: Construct grid on D (based on d), and use grid to reconstruct F
 - Global methods: Use all samples to evaluate $F(p)$
 - Local methods: Use only samples “close” to p to evaluate $F(p)$

Example Data Set

- Interpret color image as trivariate scattered data by sampling at 1600 “random” positions
- Source image: Digital photo (400×300 pixel)



Shepard's Global Method

- Shepard's formula:

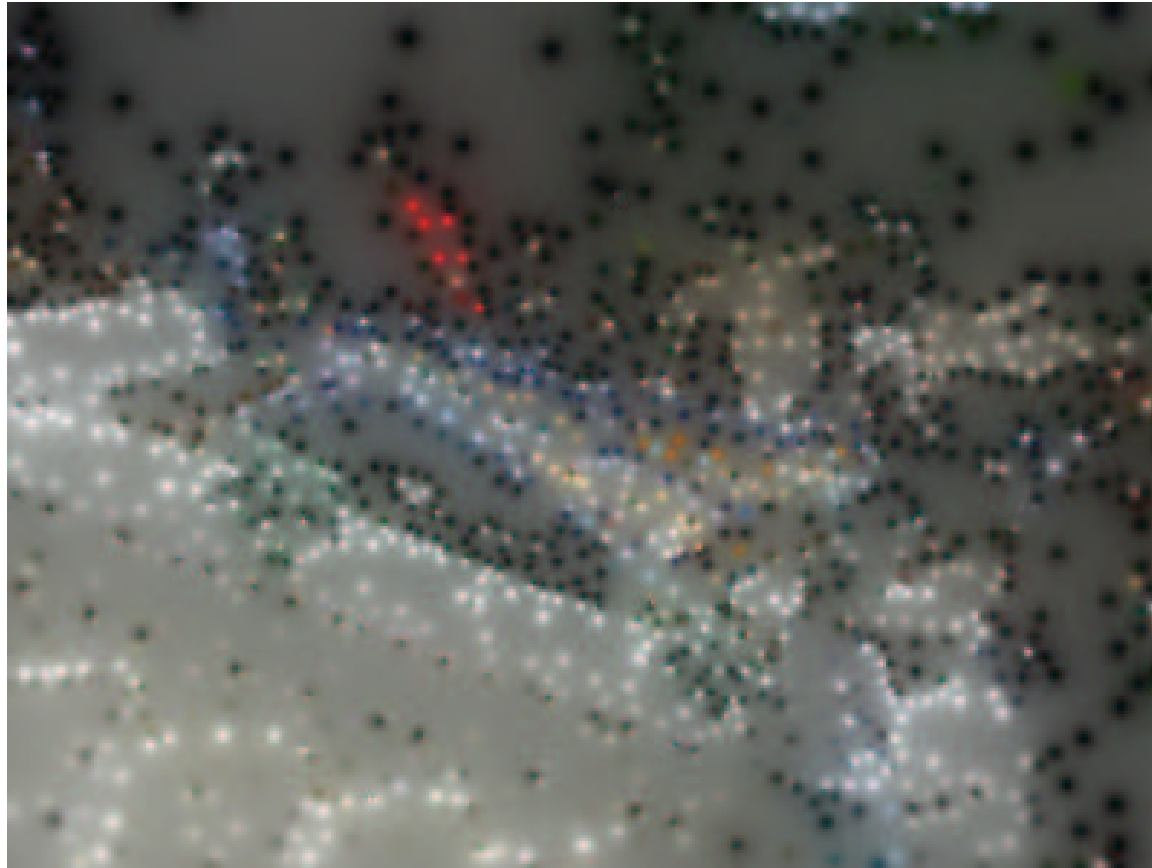
$$F(p) := \frac{\sum_{i=1}^N \frac{1}{d^2(p, p_i)} v_i}{\sum_{i=1}^N \frac{1}{d^2(p, p_i)}}$$

Shepard's Global Method

- Shepard's formula:

$$F(p) := \begin{cases} v_i & \text{if } \exists i : d(p_i, p) = 0 \\ \frac{\sum_{i=1}^N \frac{1}{d^2(p, p_i)} v_i}{\sum_{i=1}^N \frac{1}{d^2(p, p_i)}} & \text{otherwise} \end{cases}$$

Shepard's Global Method



Shepard's Local Method

- Find k nearest neighbours of point p in S
- Evaluate Shepard's formula for those k points

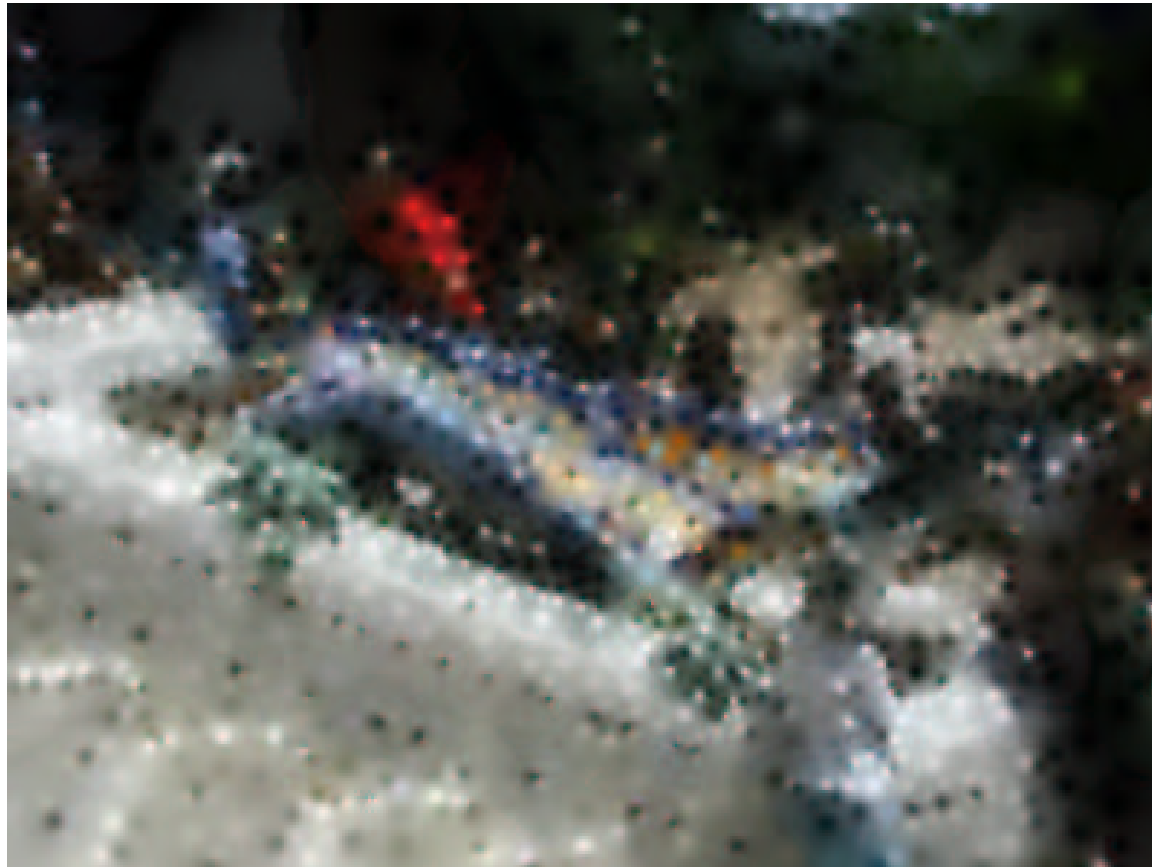
Shepard's Local Method

$k = 1$:



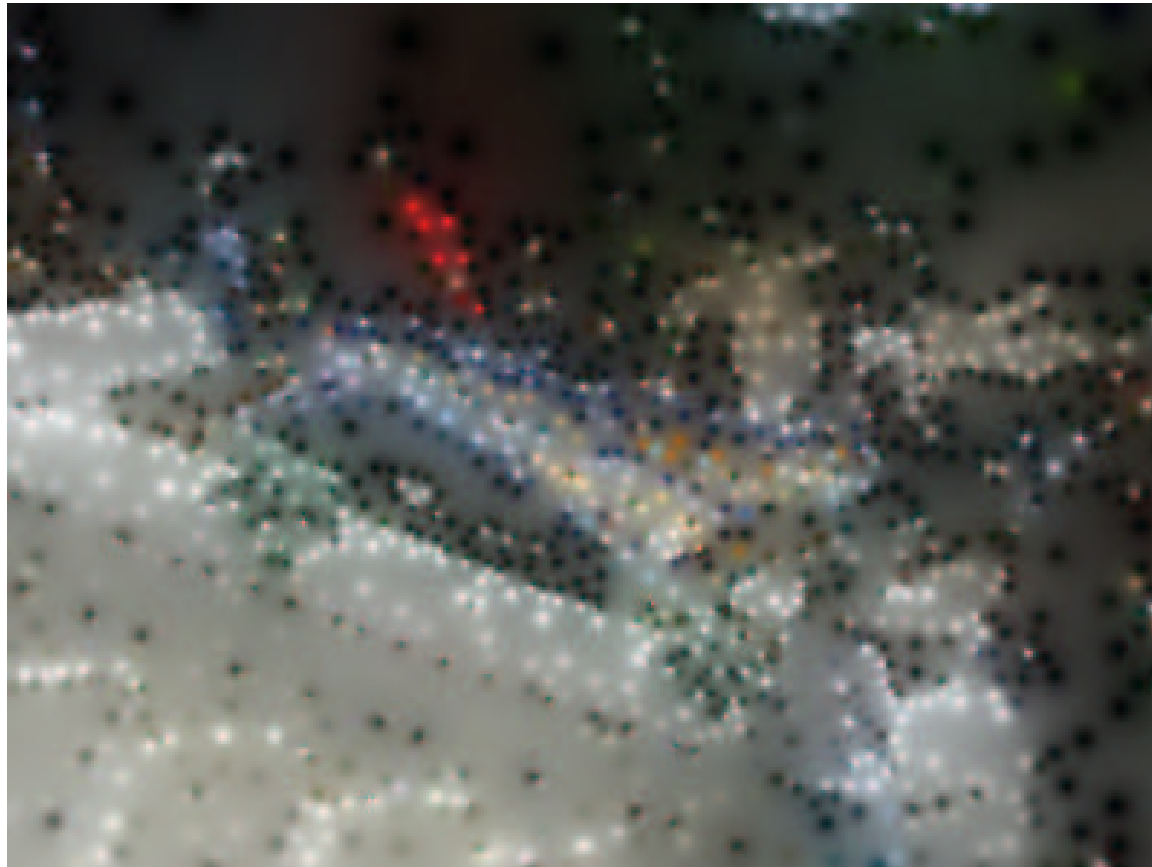
Shepard's Local Method

$k = 10$:



Shepard's Local Method

$k = 100$:



Hardy's Global Method

- Construct $N \times N$ -matrix $H := ((h_{ij}))$,
 $h_{ij} := (d^2(p_i, p_j) + R^2)^\alpha$ for parameters R^2, α
- Solve linear system
 $(v_1, \dots, v_N) = H \cdot (c_1, \dots, c_N)$
- Hardy's formula:

$$F(p) := \sum_{i=1}^N (d^2(p, p_i) + R^2)^\alpha c_i$$

Hardy's Global Method

$$R^2 = 1.0, \alpha = 0.5:$$



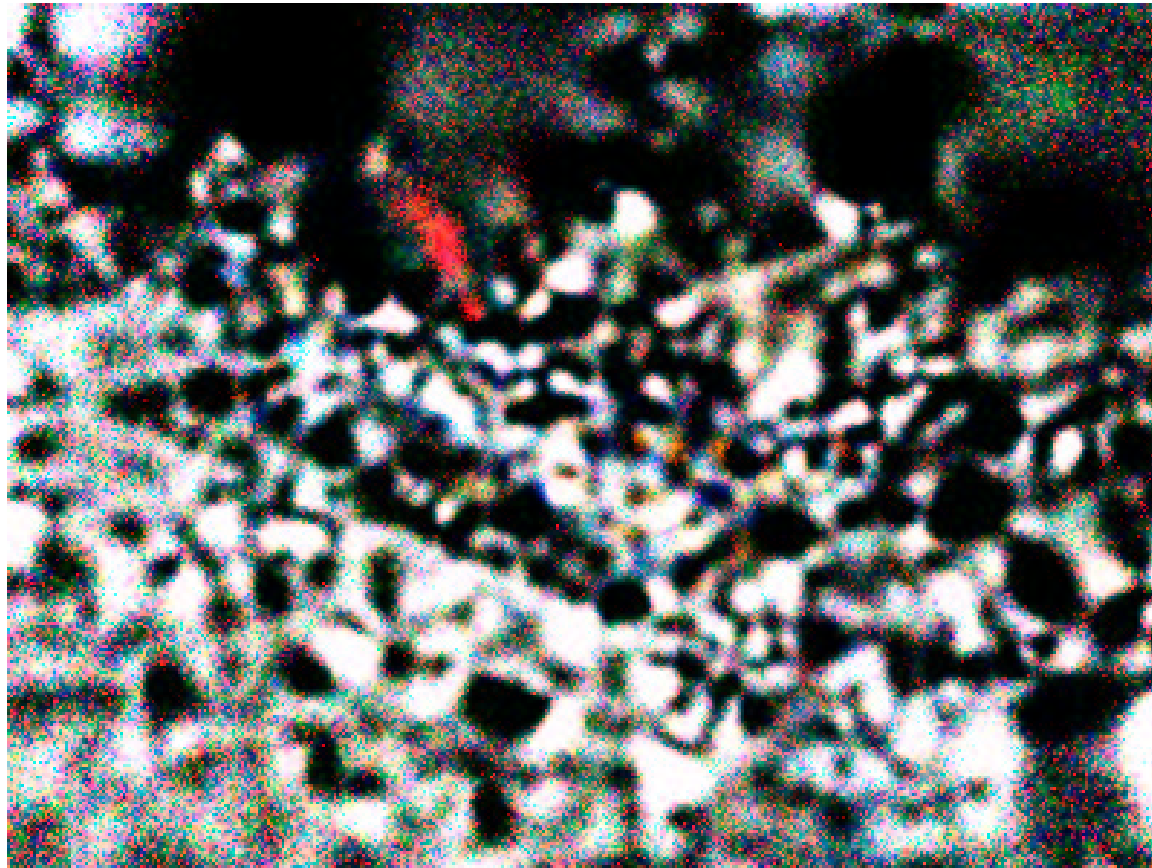
Hardy's Global Method

$$R^2 = 10.0, \alpha = 0.5:$$



Hardy's Global Method

$$R^2 = 100.0, \alpha = 0.5:$$



Hardy's Global Method

$$R^2 = 1.0, \alpha = 0.5:$$



Hardy's Global Method

$$R^2 = 1.0, \alpha = 0.1:$$



Hardy's Local Method

- Find k nearest neighbours of point p in S
- Construct matrix H for those k points
- Solve linear system for (c_1, \dots, c_k)
- Evaluate Hardy's formula for those k coefficients

Hardy's Local Method

$$R^2 = 1.0, \alpha = 0.5, k = 10:$$



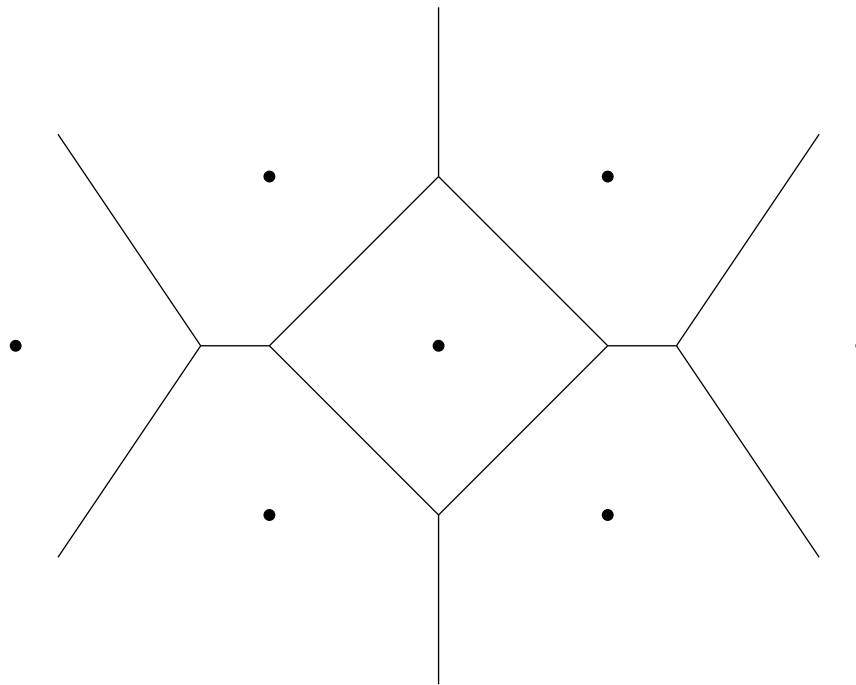
Hardy's Local Method

$$R^2 = 1.0, \alpha = 0.5, k = 100:$$



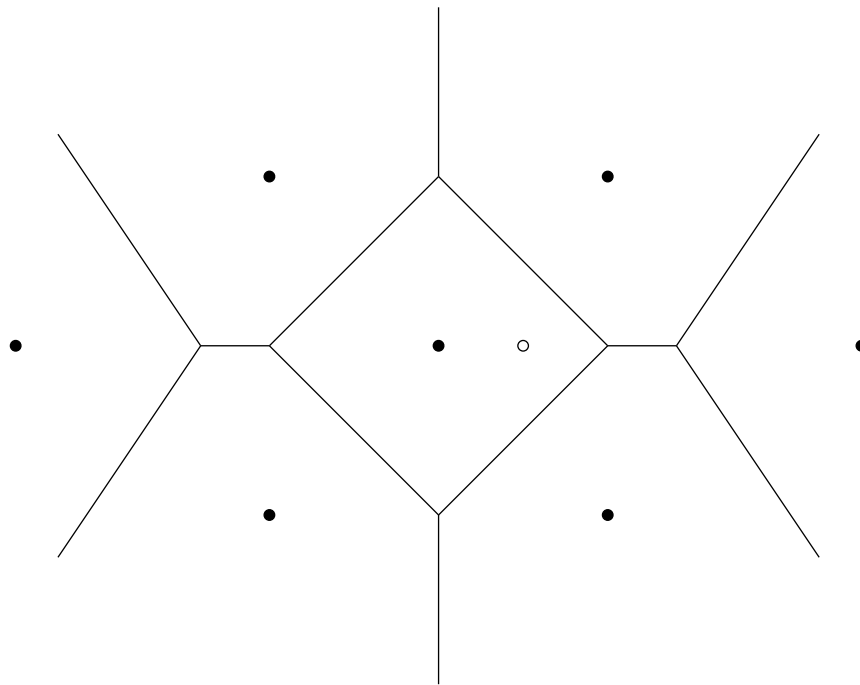
Sibson's Method

- Create Voronoï diagram V of points p_i



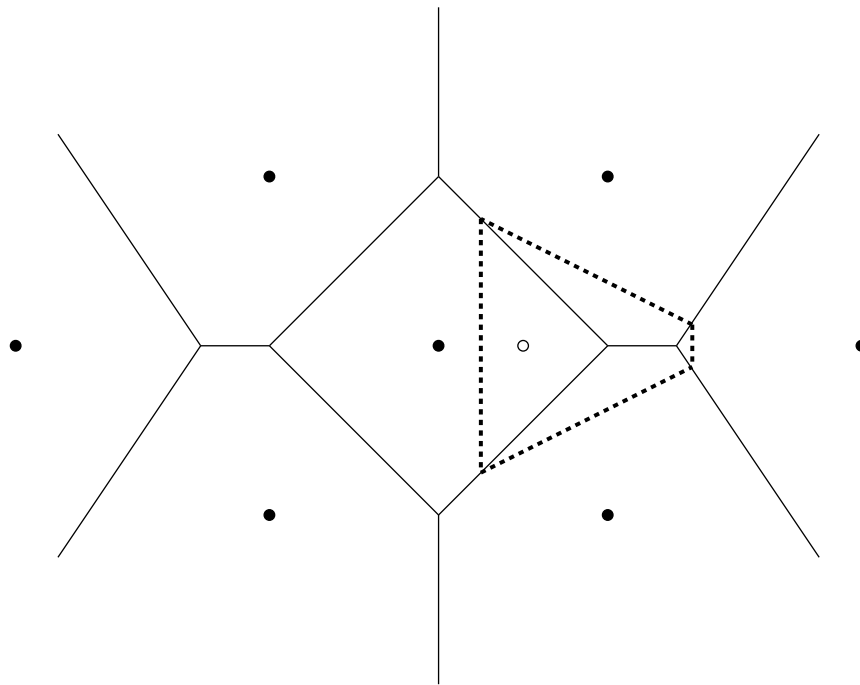
Sibson's Method

- “Tentatively” insert point p into V



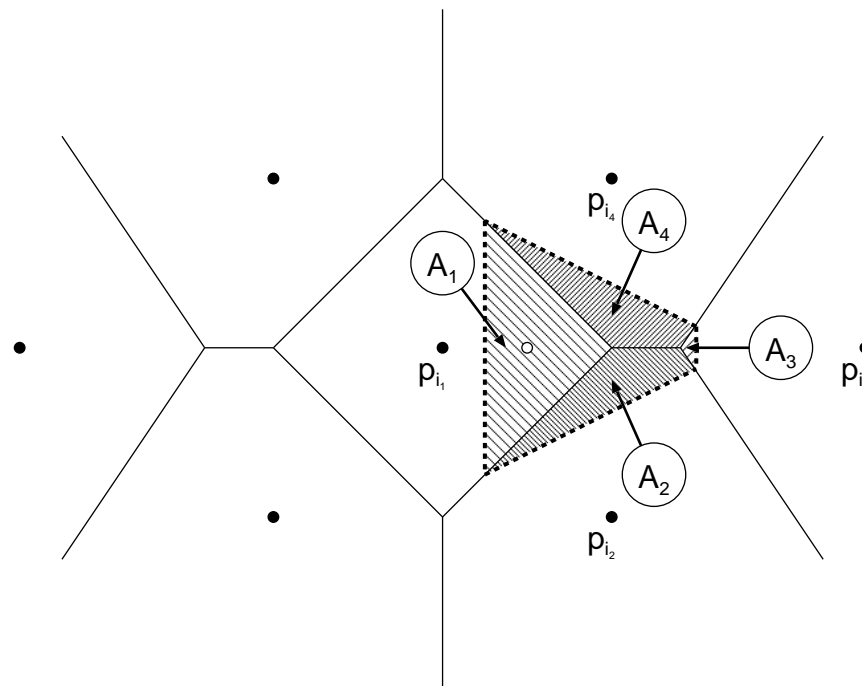
Sibson's Method

- “Tentatively” insert point p into V



Sibson's Method

- Calculate areas A_j of intersecting p 's Voronoï tile with tiles of k surrounding points p_{i_j}



Sibson's Method

- Sibson's formula:

$$F(p) = \frac{\sum_{j=1}^k A_j v_{i_j}}{\sum_{j=1}^k A_j}$$

Sibson's Method



Triangulation Method

- Create (Delaunay) triangulation T of points p_i
- Find triangle t_i containing point p
- Linearly interpolate values at vertices of t_i at position p

Triangulation Method



Optimal Triangulation Method

- Reconstruct F using triangulation of sites
- Optimize site positions based on estimated reconstruction error
- Can yield superior approximation quality using same number of samples

Optimal Triangulation Method



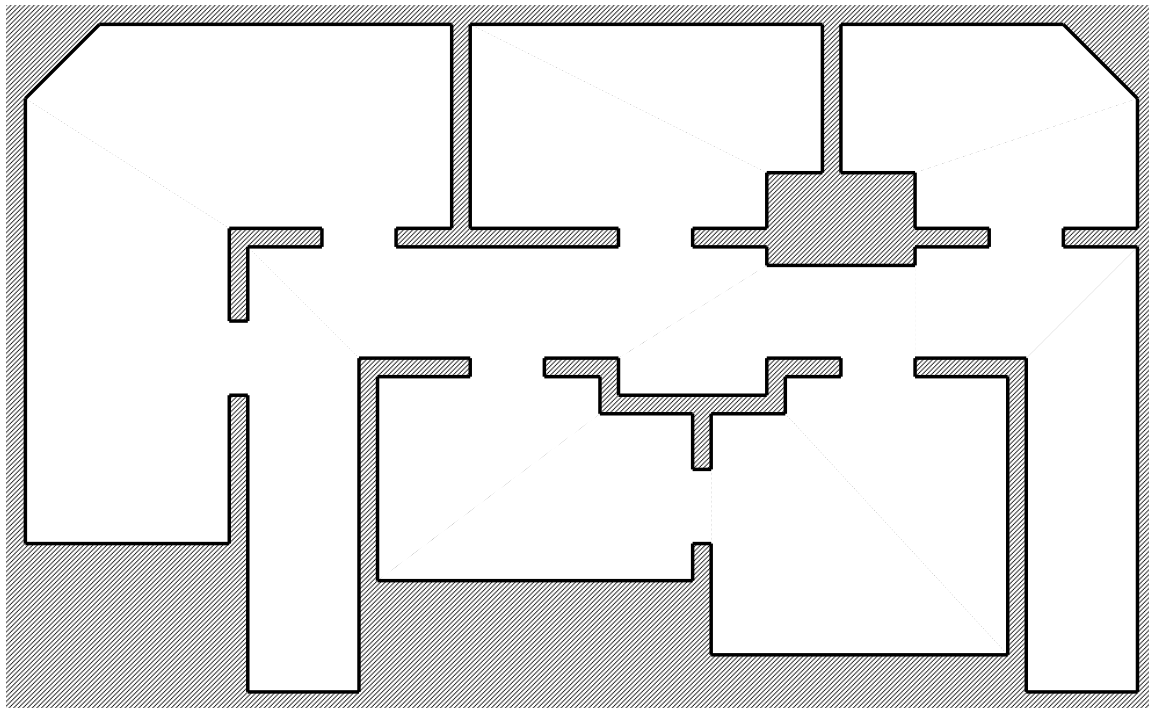
Limitations of Previous Methods

- Described methods are based on Euclidean metric $d(p, q) := (p - q)^2$
- Methods assume compact convex domain D , i. e., no obstacles
- Wrong assumption for building interiors!
- Solution: Define distance metric for buildings
- First: Need to represent complex domains

Representation of Complex Domains

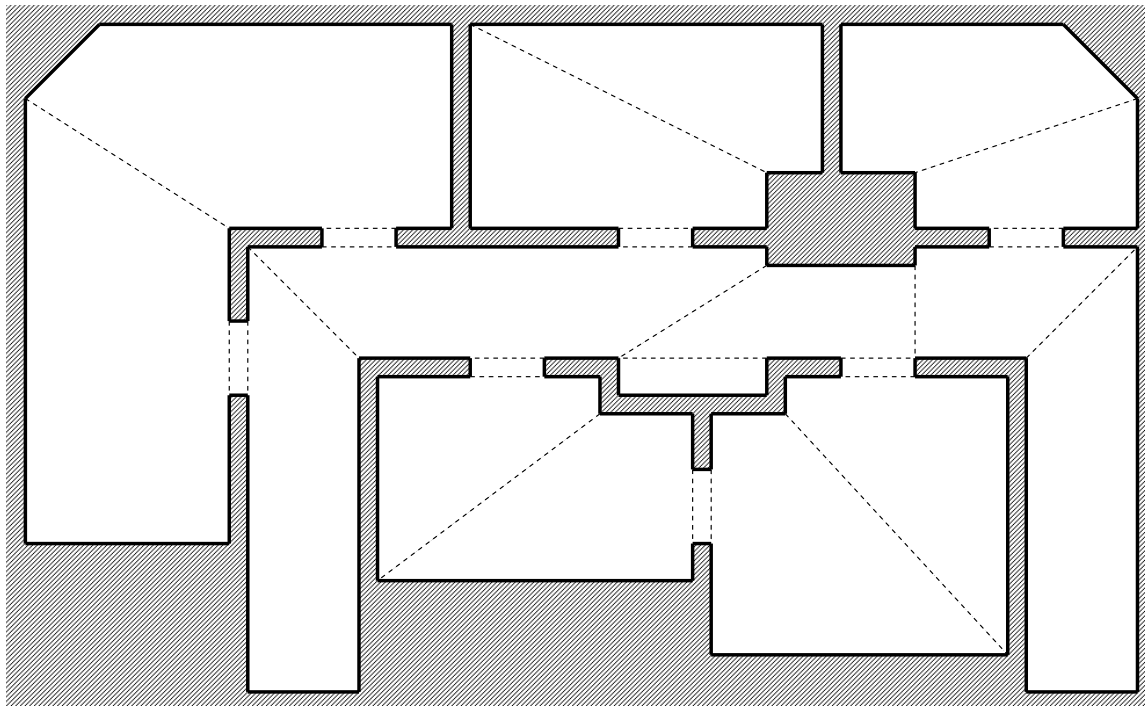
Representing Complex Domains

- Example domain: Hallway with offices



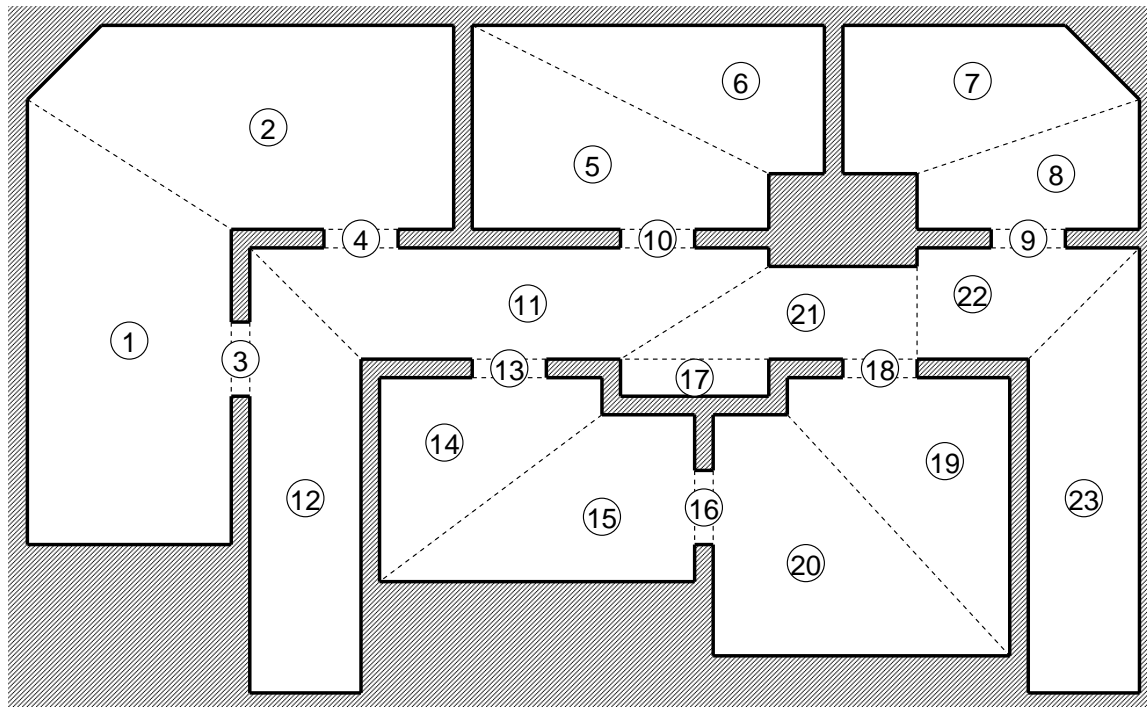
Representing Complex Domains

- Fundamental idea: Represent domain as union of convex polyhedral *sectors* explicitly connected by *portal* polygons



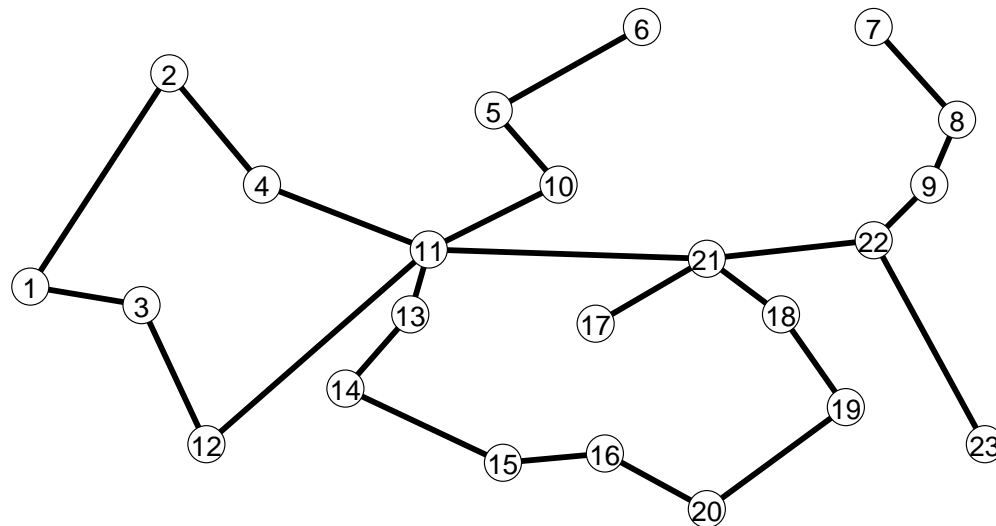
Representing Complex Domains

- Fundamental idea: Represent domain as union of convex polyhedral *sectors* explicitly connected by *portal* polygons



Representing Complex Domains

- Leads to representation of domain as undirected graph

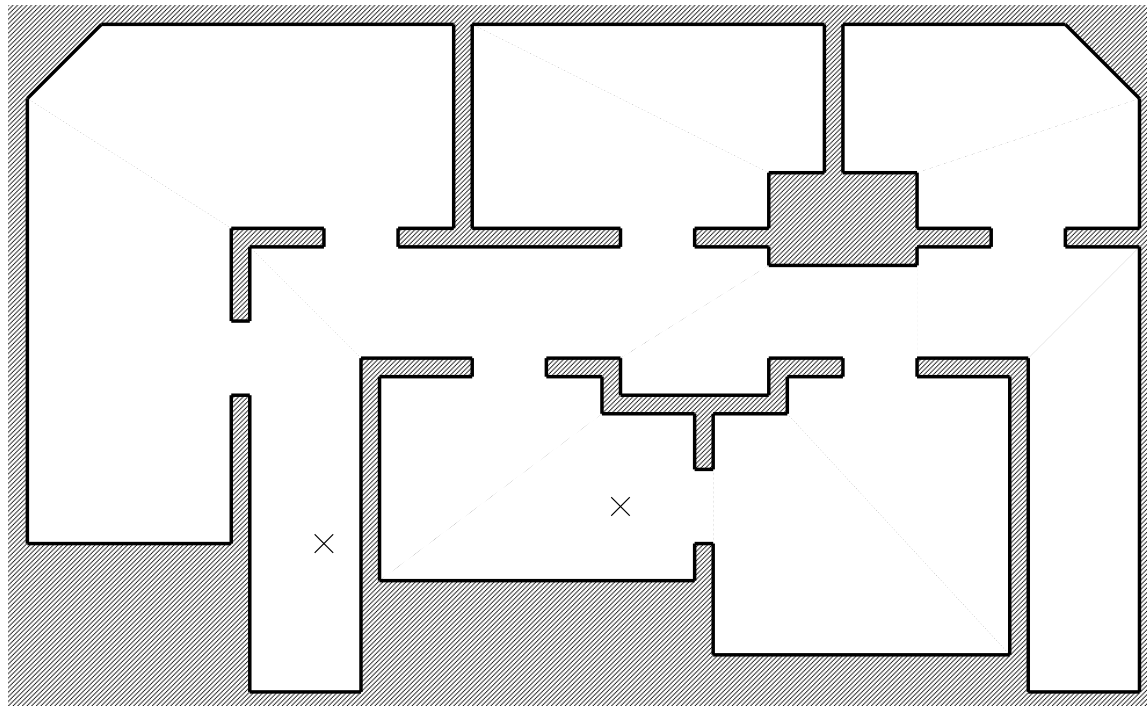


Current State and Outlook

- Domain representation data structure in place
- Need tool to generate domain representation from
 - Architectural blueprints
 - CAD models
 - . . .

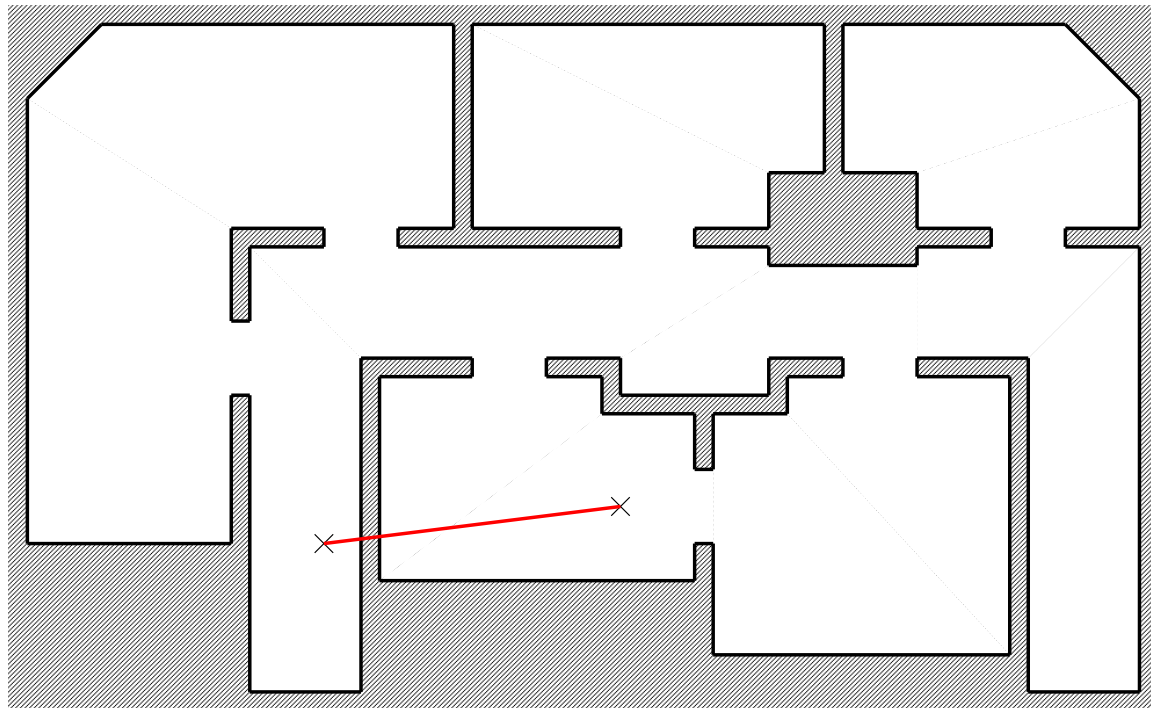
Distance Metric for Complex Domains

- Calculating distance between two query points



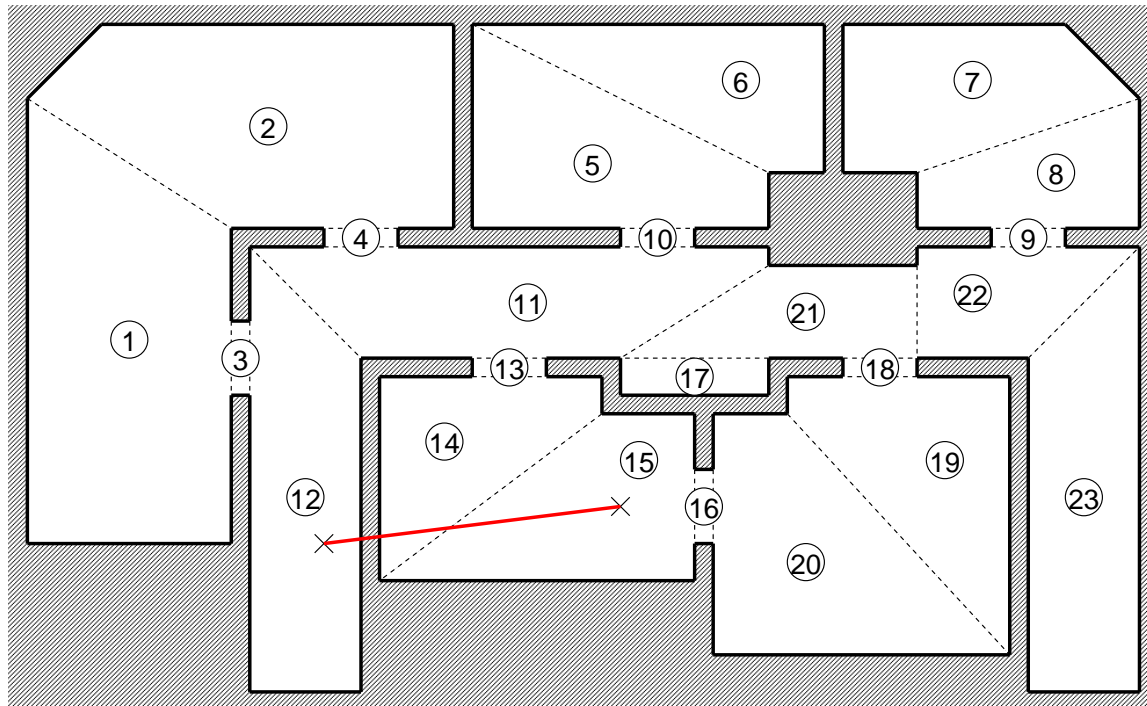
Distance Metric for Complex Domains

- Euclidean metric would ignore boundaries



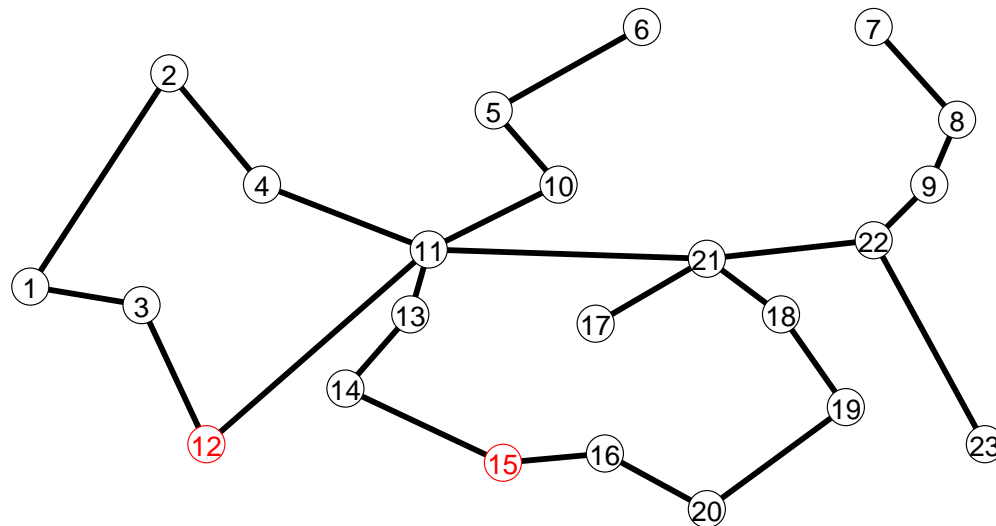
Distance Metric for Complex Domains

- Determine sector(s) containing query points



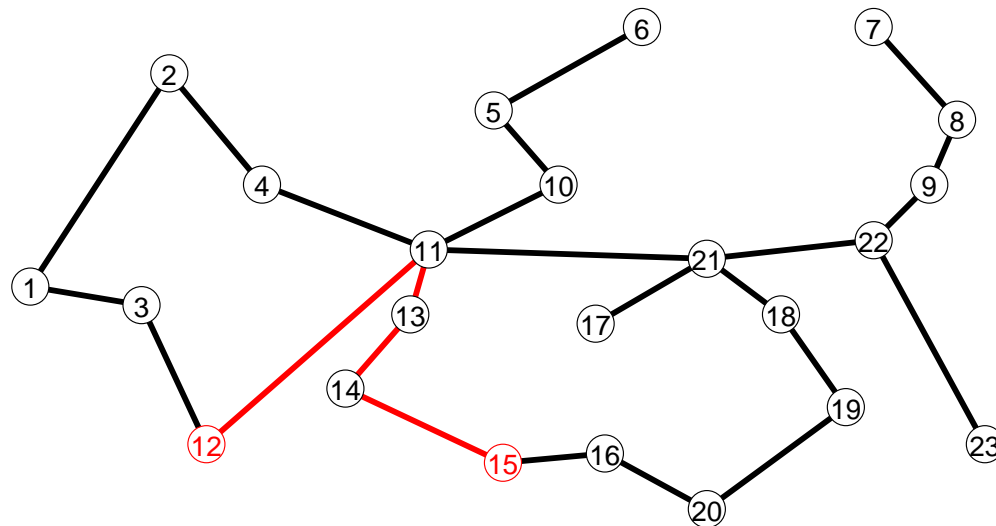
Distance Metric for Complex Domains

- Consider graph to find path between sectors



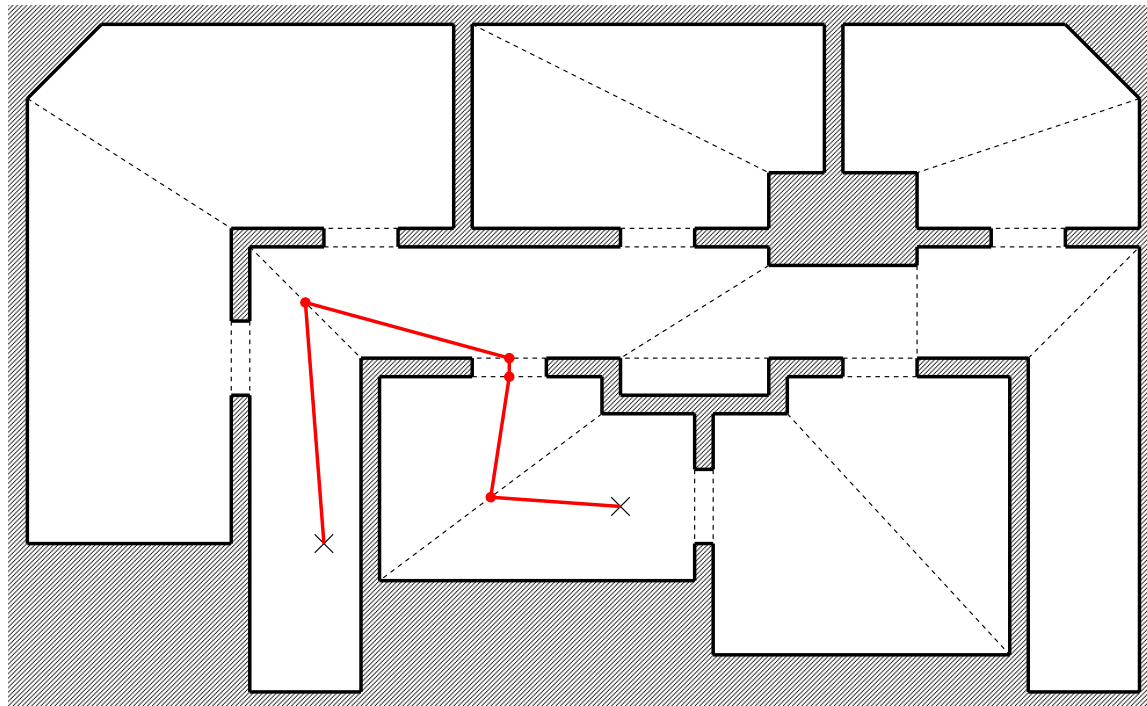
Distance Metric for Complex Domains

- Graph traversal yields sequence of portals to cross



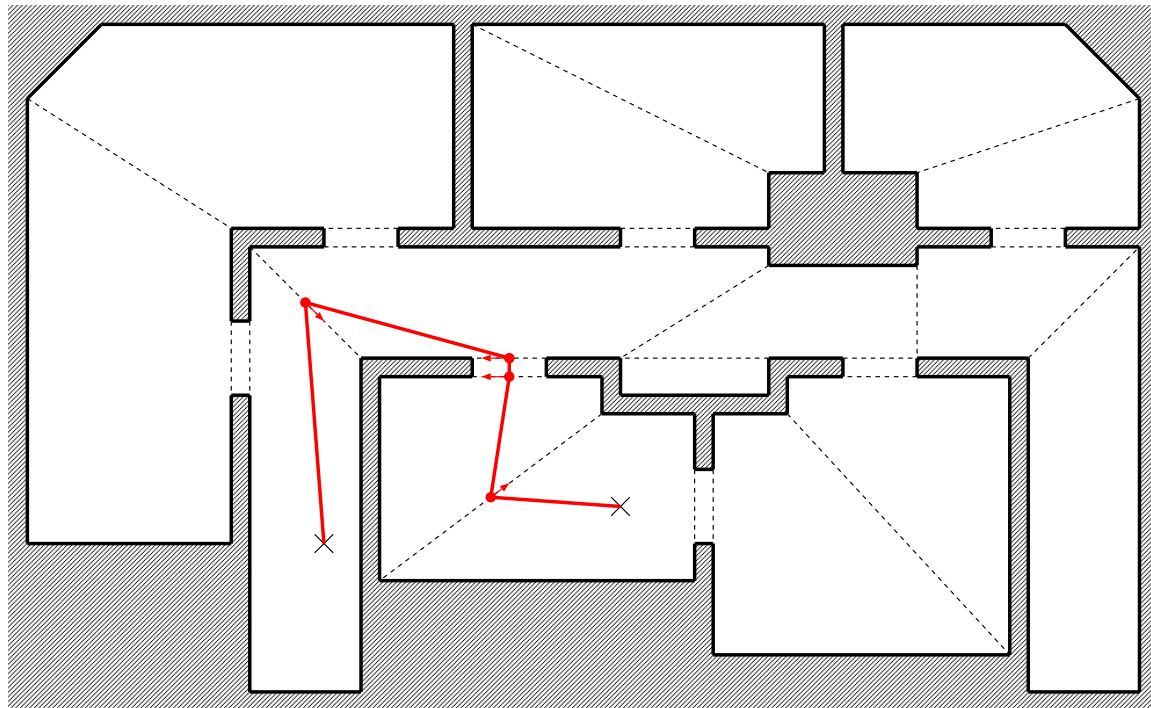
Distance Metric for Complex Domains

- Cross portals through their centroids initially



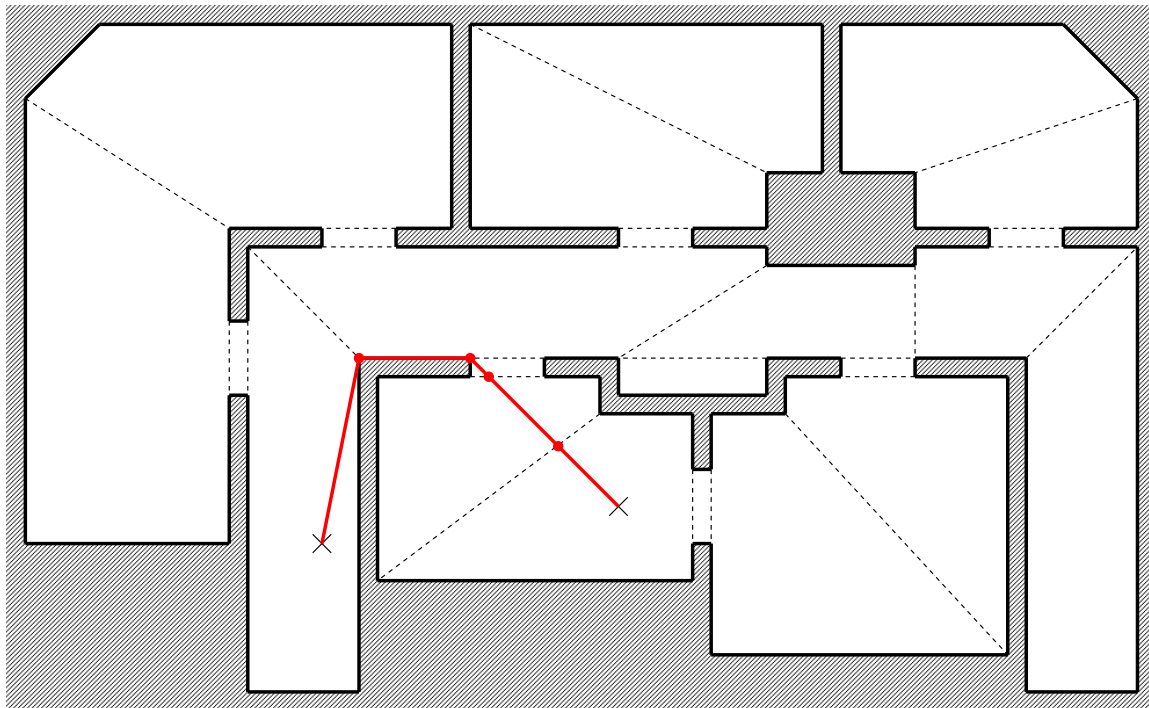
Distance Metric for Complex Domains

- Optimize placement of intermediate points



Distance Metric for Complex Domains

- Optimize placement of intermediate points



Metric Properties

- If there exists a direct line-of-sight between two query points, the described metric agrees with the Euclidean metric
 - ⇒ Artificial (and arbitrary) boundaries introduced by domain decomposition do not influence distances
- Metric can be computed efficiently

Metric Properties

- Metric can be used as drop-in replacement for Euclidean metric in Shepard's and Hardy's method
- Metric defines Voronoï diagram (and Delaunay triangulation) for Sibson's and triangulation method
 - Shape of Voronoï tiles?
 - Efficient computation?

Current State and Outlook

- One graduate student working on optimization algorithm
- One undergraduate student will soon start working on scattered data methods compatible with metric



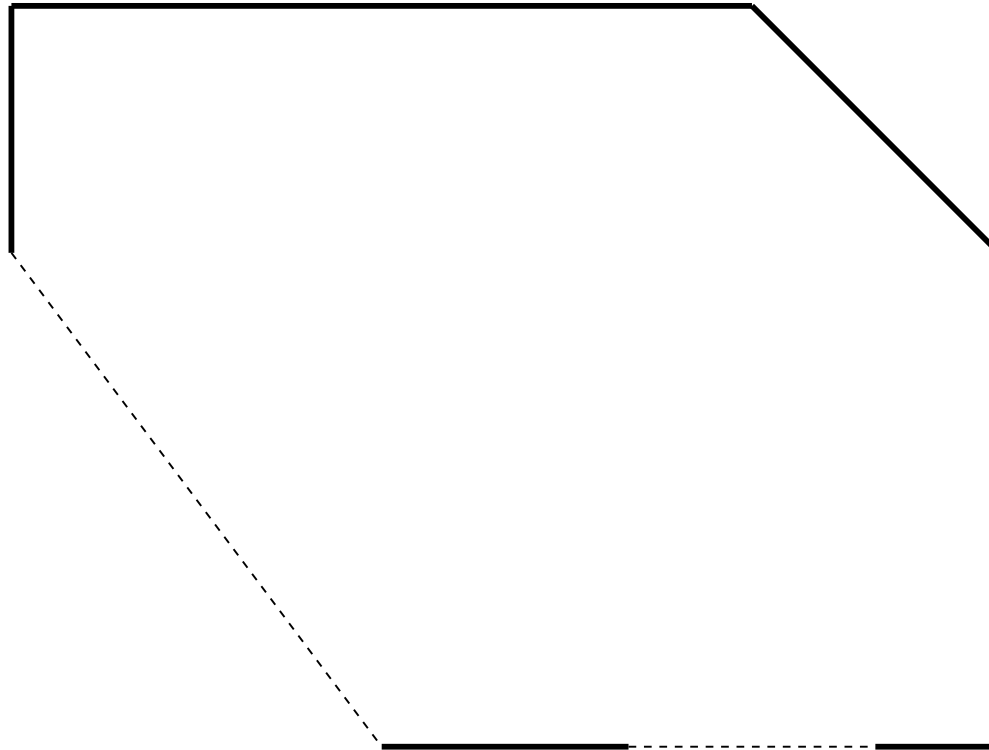
Visualizing Sensor Network Data

Visualization Strategy

- Basic idea: To visualize sensor data in context, combine rendering of domain boundaries with data visualization
- Domain representation is also efficient for interactive walk-throughs
- Sensor data can be reconstructed and visualized separately in each sector
- Since each sector is a convex polyhedron, standard visualization methods can be used

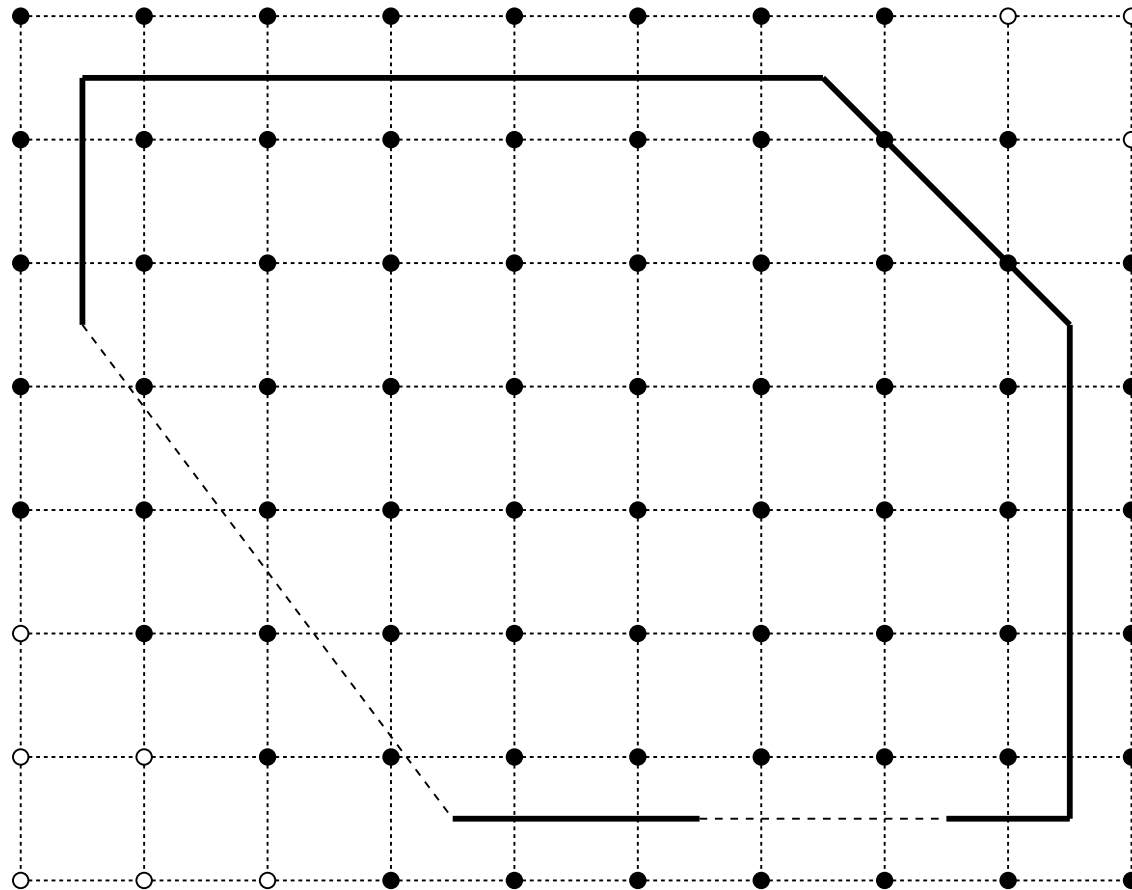
Per-Sector Visualization

- Single sector is convex polyhedron



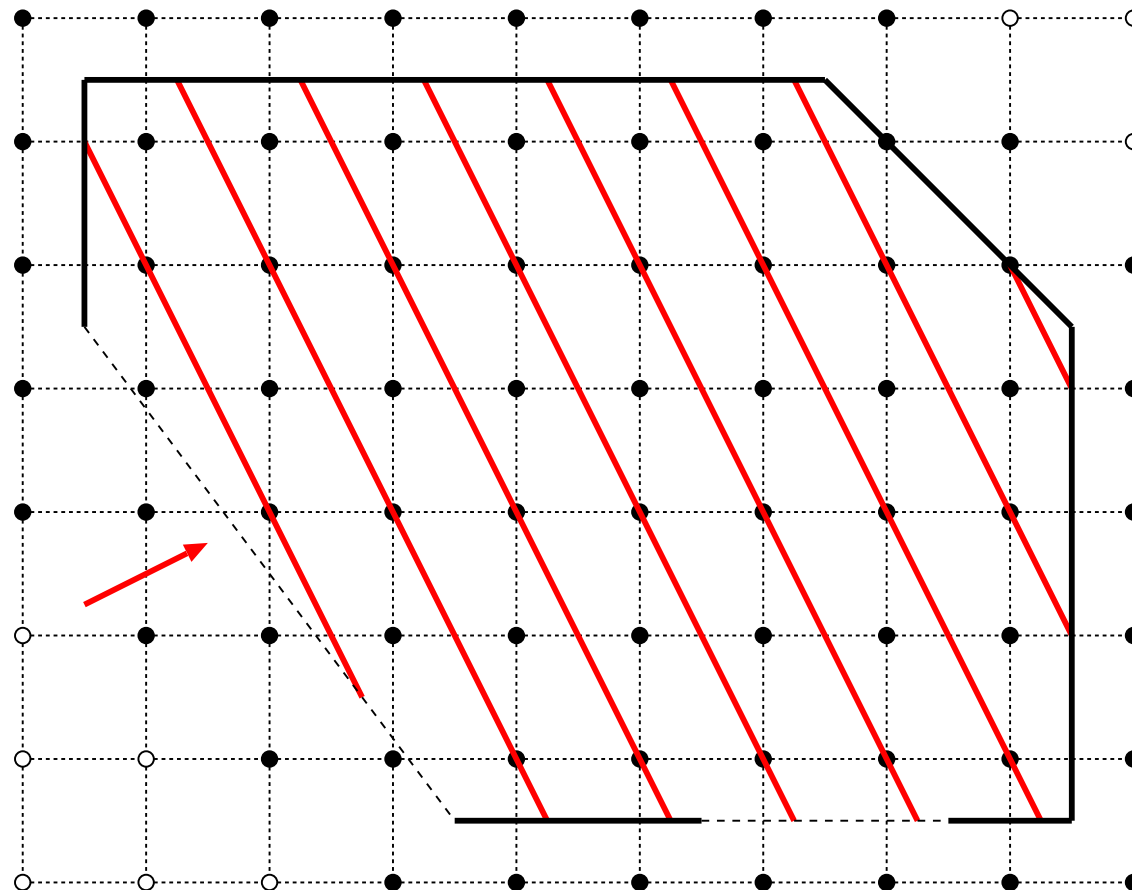
Per-Sector Visualization

- Overlay sector with regular grid

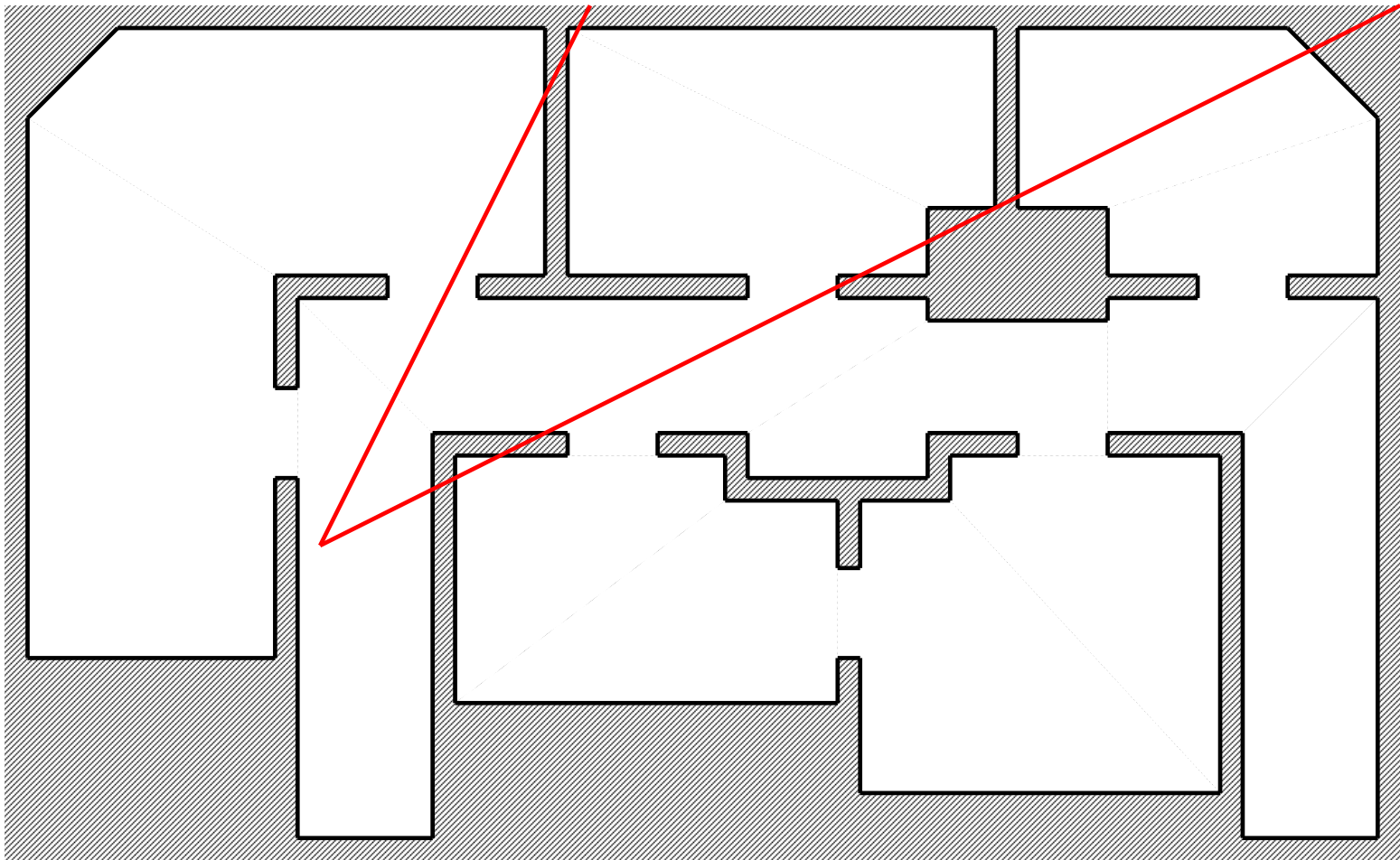


Per-Sector Visualization

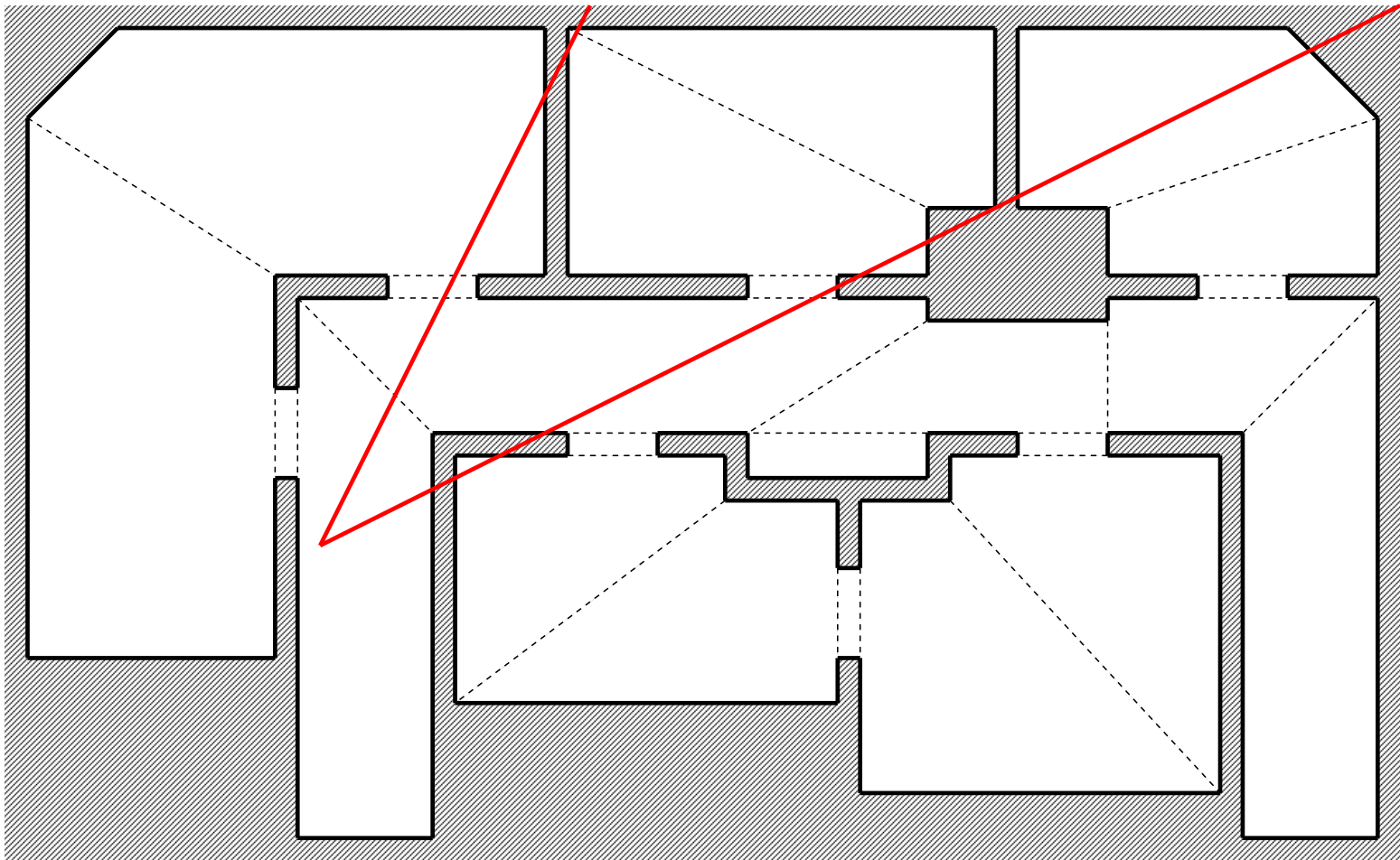
- Generate slices for volume rendering



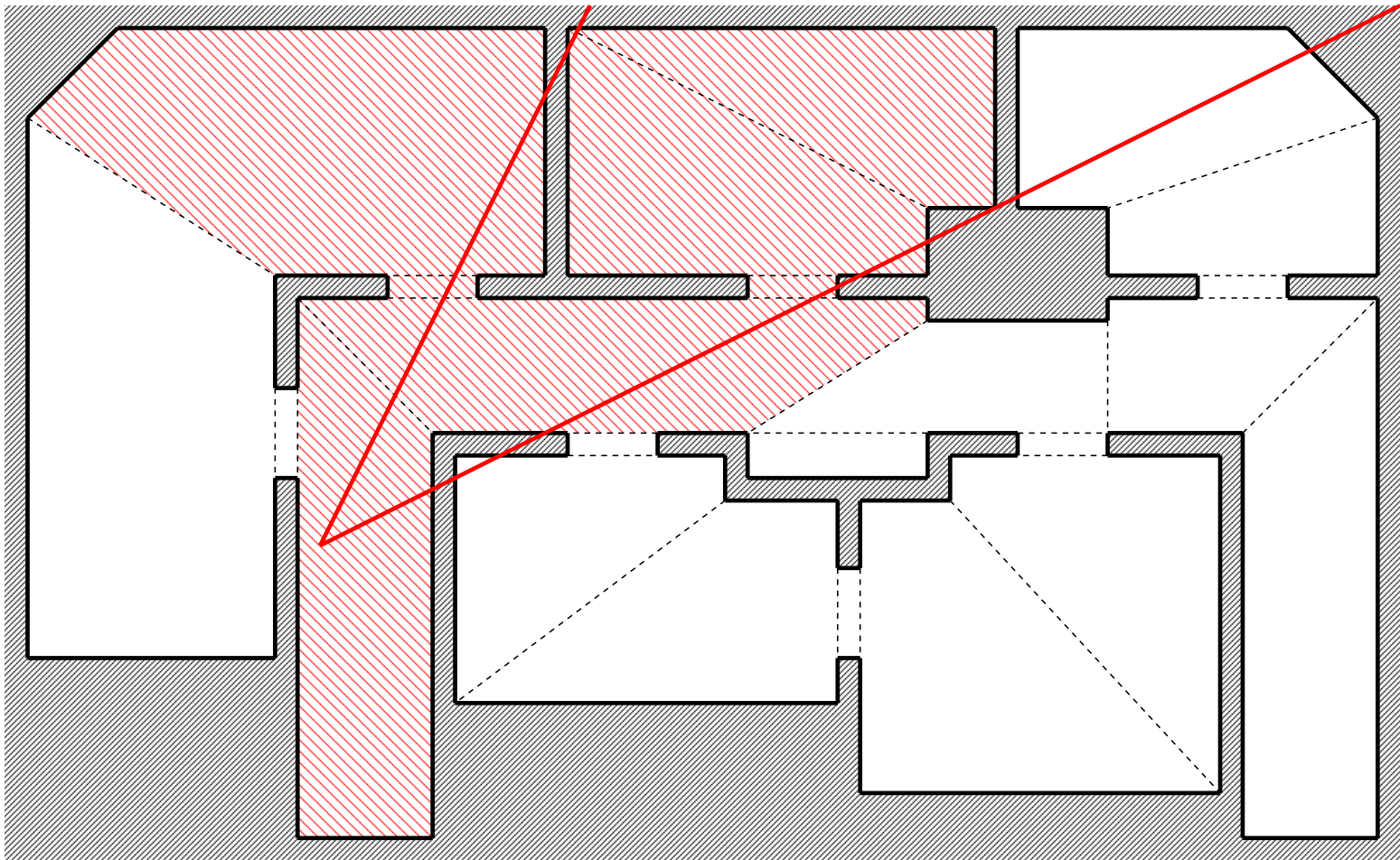
Visualizing Entire Domain



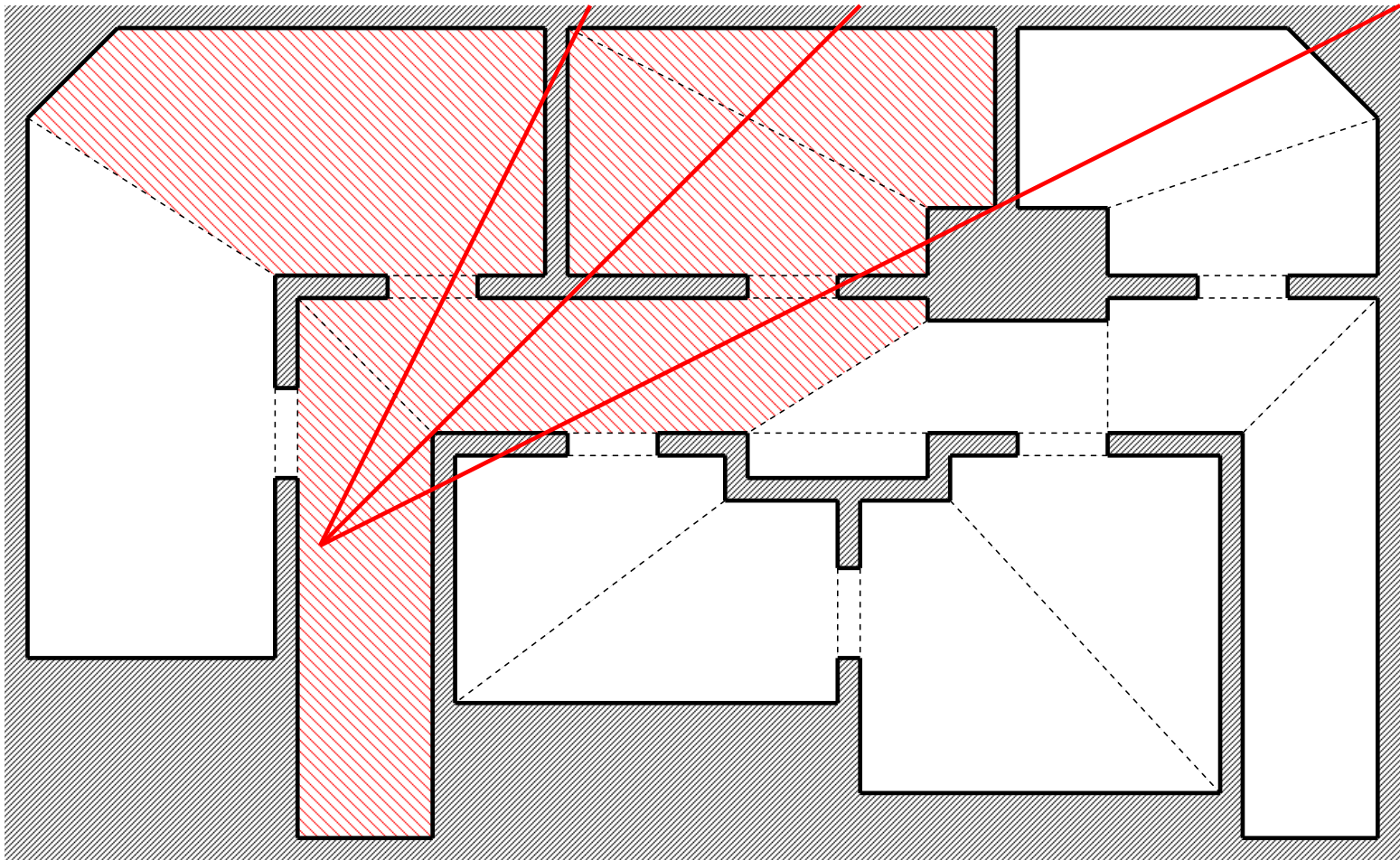
Visualizing Entire Domain



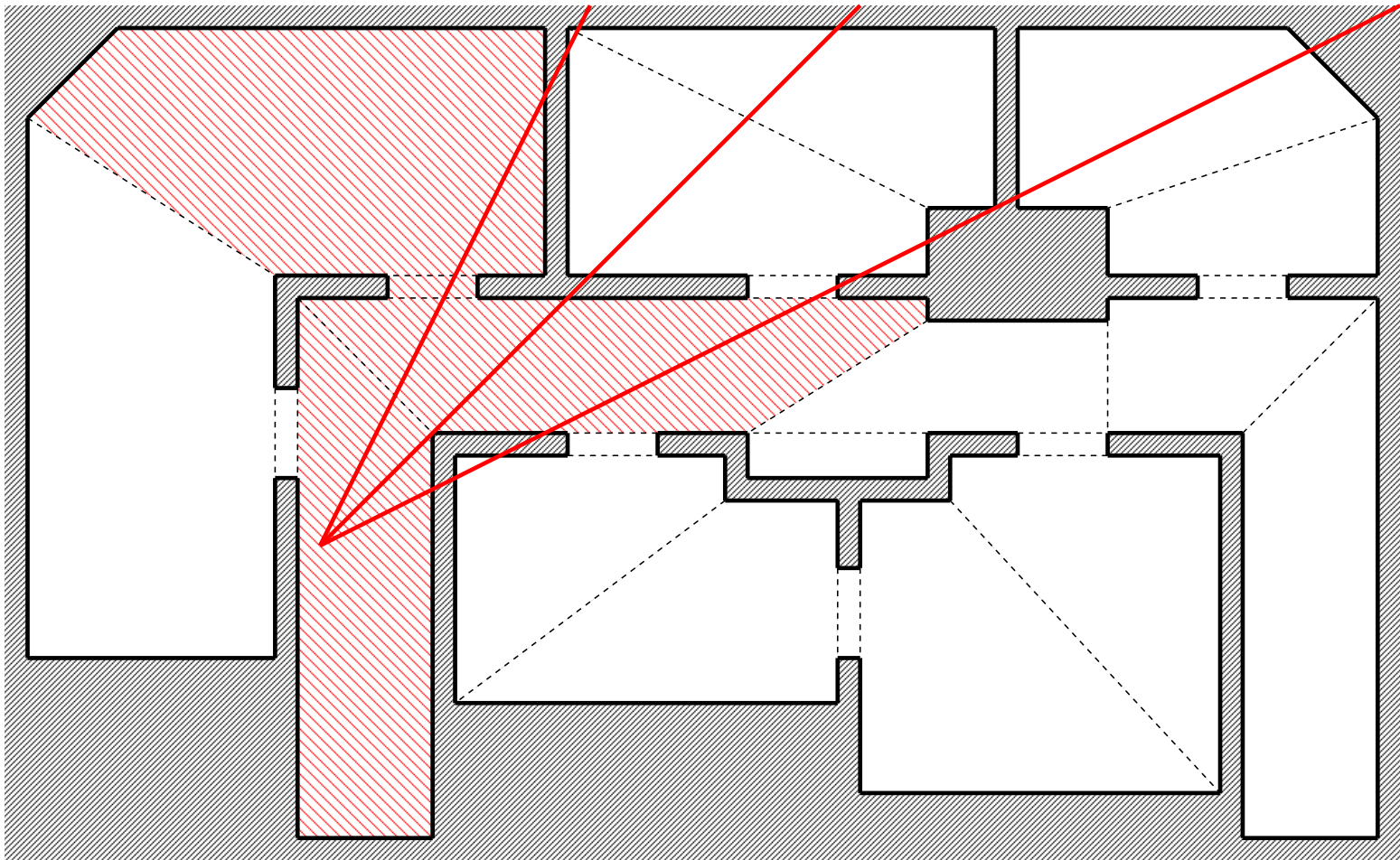
Visualizing Entire Domain



Visualizing Entire Domain



Visualizing Entire Domain



Current State and Outlook

- Walk-throughs of complex domains will be done Real Soon NowTM
- Walk-through will double as tool to “polish” sensor data input
- Sensor data visualization will be integrated when metric and scattered data method are in place
- Visualization will run in desktop and Virtual Reality environment

Conclusions

Short-Term Goal

- Combine
 - domain representation
 - distance metric calculation
 - (direct) scattered data interpolation method
- to visualize static sensor network data inside buildings

Long-Term Goals

- Analyze sensor network data based on reconstruction
- Move to indirect (grid-based) scattered data methods
- Move to time-varying and dynamic sensor data
- Optimize sensor placement based on reconstruction

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The End