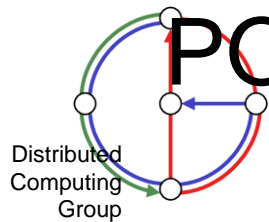


Chapter 12

POSITIONING



Mobile Computing
Winter 2005 / 2006

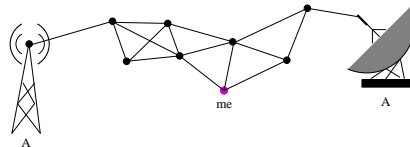
Overview

- Motivation
- Measurements
- Anchors
- Virtual Coordinates
- Heuristics
- Practice



Motivation

- Why positioning?
 - Sensor nodes without position information is often meaningless
 - Heavy and/or costly positioning hardware
 - Geo-routing



- Why **not GPS (or Galileo)**?
 - Heavy, large, and expensive (as of yet)
 - Battery drain
 - Not indoors
 - Accuracy?

- Solution: equip small fraction with GPS (**anchors**)

Measurements

Distance estimation

- Received Signal Strength Indicator (RSSI)
 - The further away, the weaker the received signal.
 - Mainly used for RF signals.
- Time of Arrival (ToA) or Time Difference of Arrival (TDoA)
 - Signal propagation time translates to distance.
 - RF, acoustic, infrared and ultrasound.

Angle estimation

- Angle of Arrival (AoA)
 - Determining the direction of propagation of a radio-frequency wave incident on an antenna array.
- Directional Antenna
- Special hardware, e.g., laser transmitter and receivers.



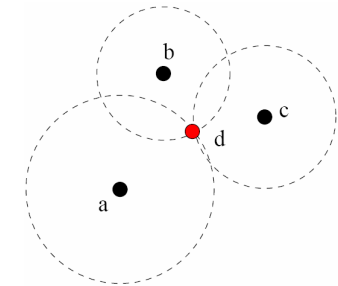
Positioning (a.k.a. Localization)

- Task: Given distance or angle measurements or mere connectivity information, find the locations of the sensors.
- **Anchor-based**
 - Some nodes know their locations, either by a GPS or as pre-specified.
- **Anchor-free**
 - Relative location only. Sometimes called virtual coordinates.
 - Theoretically cleaner model (less parameters, such as anchor density)
- **Range-based**
 - Use range information (distance estimation).
- **Range-free**
 - No distance estimation, use connectivity information such as hop count.
 - It was shown that bad measurements don't help a lot anyway.



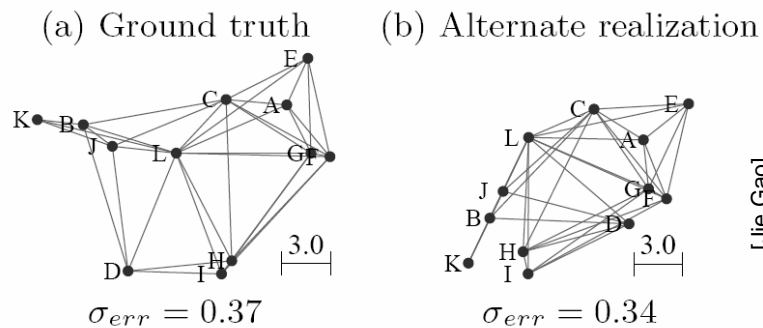
Trilateration and Triangulation

- Use geometry, measure the distances/angles to three anchors.
- **Trilateration**: use distances
 - Global Positioning System (GPS)
- **Triangulation**: use angles
 - Some cell phone systems
- How to deal with inaccurate measurements?
 - Least squares type of approach
 - What about strictly more than 3 (inaccurate) measurements?



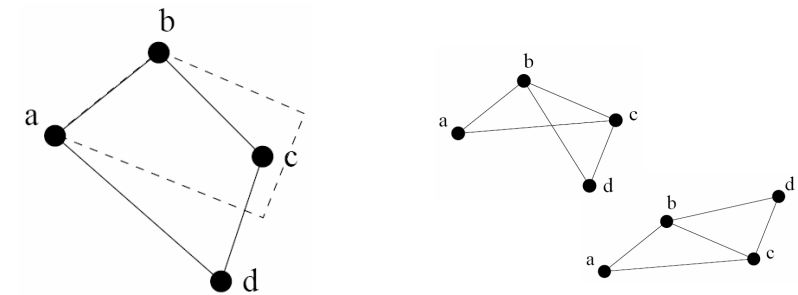
Ambiguity Problems

- Same distances, different realization.



[Jie Gao]

Continuous deformation, flips, etc.



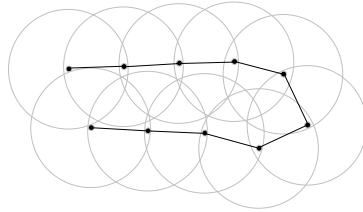
[Jie Gao]

- Rigidity theory: Given a set of rigid bars connected by hinges, rigidity theory studies whether you can move them continuously.



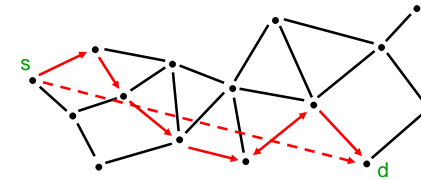
Simple hop-based algorithms

- Algorithm
 - Get graph distance h to anchor(s)
 - Intersect circles around anchors
 - radius = distance to anchor
 - Choose point such that **maximum error is minimal**
 - Find **enclosing circle** (ball) of minimal radius
 - Center is calculated location
- In higher dimensions: $1 < d \leq h$
 - Rule of thumb: **Sparse graph**
 - bad performance



How about no anchors at all...?

- In absence of anchors...
 - ...nodes are clueless about **real coordinates**.
- For many applications, real coordinates are not necessary
 - **Virtual coordinates** are sufficient
 - Geometric Routing requires only virtual coordinates
 - Require no routing tables
 - Resource-frugal and scalable



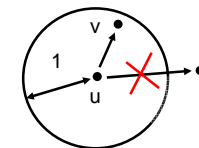
Virtual Coordinates

- Idea:
 - Close-by nodes have similar coordinates
 - Distant nodes have very different coordinates
- Similar coordinates imply physical proximity!
- Applications
 - Geometric Routing
 - Locality-sensitive queries
 - Obtaining meta information on the network
 - Anycast services („Which of the service nodes is closest to me?“)
 - Outside the sensor network domain: e.g., Internet mapping

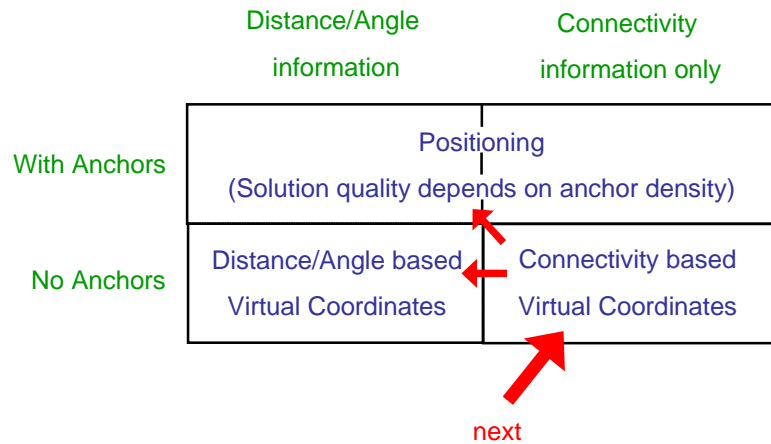


Model

- **Unit Disk Graph (UDG)** to model wireless multi-hop network
 - Two nodes can communicate iff Euclidean distance is at most 1
- Sensor nodes may not be capable of
 - Sensing directions to neighbors
 - Measuring distances to neighbors
- Goal: Derive topologically correct coordinate information from **connectivity information** only.
 - Even the simplest nodes can derive connectivity information



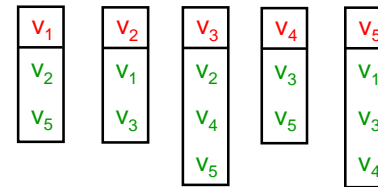
Context



Virtual Coordinates ↔ UDG Embedding



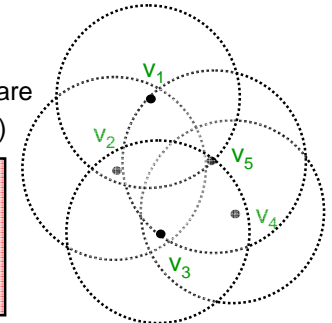
- Given the connectivity information for each node...



...and knowing the underlying graph is a UDG...

- ...find a UDG embedding in the plane such that all connectivity requirements are fulfilled! (→ Find a realization of a UDG)

This problem is NP-hard!
(Simple reduction to UDG-recognition problem, which is NP-hard)
[Breu, Kirkpatrick, Comp.Geom.Theory 1998]



UDG Approximation – Quality of Embedding



- Finding an exact realization of a UDG is NP-hard.
→ Find an embedding $r(G)$ which approximates a realization.
- Particularly,
→ Map adjacent vertices (edges) to points which are close together.
→ Map non-adjacent vertices („non-edges“) to far apart points.
- Define quality of embedding $q(r(G))$ as:

Ratio between longest edge to shortest non-edge in the embedding.

Let $\rho(u,v)$ be the distance between points u and v in the embedding.

$$q(r(G)) := \frac{\max_{\{u,v\} \in E} \rho(u,v)}{\min_{\{u',v'\} \notin E} \rho(u',v')}$$



UDG Approximation

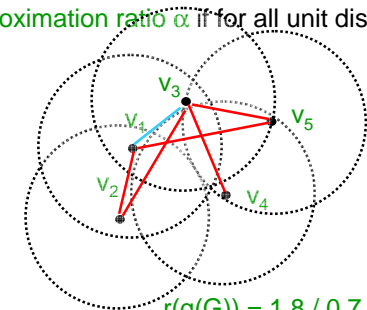
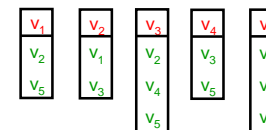


- For each UDG G , there exists an embedding $r(G)$, such that, $q(r(G)) \leq 1$.
(a realization of G)

$$q(r(G)) := \frac{\max_{\{u,v\} \in E} \rho(u,v)}{\min_{\{u',v'\} \notin E} \rho(u',v')}$$

- Finding such an embedding is NP-hard
- An algorithm ALG achieves approximation ratio α if for all unit disk graphs G , $q(r_{\text{ALG}}(G)) \leq \alpha$.

- Example:



$$q(r(G)) = 1.8 / 0.7 = 2.6$$



Some Results

- There are a few virtual coordinates algorithms
All of them evaluated only by **simulation on random graphs**
- In fact there is only one **provable approximation algorithm**

There is an algorithm which achieves an approximation ratio of $O(\log^{2.5} n \sqrt{\log \log n})$, n being the number of nodes in G .

- Plus there are **lower bounds on the approximability**.

There is no algorithm with approximation ratio better than $\sqrt{3/2} - \epsilon$, unless $P=NP$.



Approximation Algorithm: Overview

- Four major steps

1. Compute **metric** on MIS of input graph \rightarrow **Spreading constraints**
(Key conceptual difference to previous approaches!)
2. **Volume-respecting**, high dimensional **embedding**
3. **Random projection** to 2D
4. Final embedding

UDG Graph G with MIS M .

Approximate pairwise distances between nodes such that, MIS nodes are neatly spread out.

Volume respecting embedding of nodes in \mathbb{R}^n with small distortion.

Nodes spread out fairly well in \mathbb{R}^2 .

Final embedding of G in \mathbb{R}^2 .



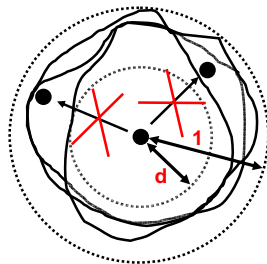
Lower Bound: Quasi Unit Disk Graph

- Definition **Quasi Unit Disk Graph**:

Let $V \in \mathbb{R}^2$, and $d \in [0,1]$. The symmetric Euclidean graph $G=(V,E)$, such that for any pair $u,v \in V$

- $\text{dist}(u,v) \leq d \Rightarrow \{u,v\} \in E$
- $\text{dist}(u,v) > 1 \Rightarrow \{u,v\} \notin E$

is called **d -quasi unit disk graph**.



- Note that between d and 1 , the existence of an edge is **unspecified**.



Reduction

- We want to show that finding an embedding with $q(r(G)) \leq \sqrt{3/2} - \epsilon$, where ϵ goes to 0 for $n \rightarrow \infty$ is NP-hard.
- We prove an equivalent statement:

Given a unit disk graph $G=(V,E)$, it is NP-hard to find a realization of G as a d -quasi unit disk graph with $d \geq \sqrt{2/3} + \epsilon$, where ϵ tends to 0 for $n \rightarrow \infty$.

- \rightarrow Even when allowing non-edges to be smaller than 1 , embedding a unit disk graph remains NP-hard!
- \rightarrow It follows that finding an approximation ratio better than $\sqrt{3/2} - \epsilon$ is also NP-hard.



Reduction

- Reduction from 3-SAT (each variable appears in at most 3 clauses)
- Given an instance C of this 3-SAT, we give a polynomial time construction of $G_C=(V_C, E_C)$ such that the following holds:

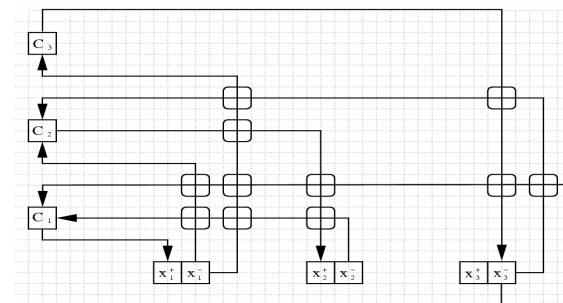
- C is satisfiable $\Rightarrow G_C$ is realizable as a unit disk graph
- C is not satisfiable $\Rightarrow G_C$ is not realizable as a d -quasi unit disk graph with $d \geq \sqrt{2/3} + \epsilon$

- Unless $P=NP$, there is no approximation algorithm with approximation ratio better than $\sqrt{3/2} - \epsilon$.



Proof idea

- Construct a grid drawing of the SAT instance.
- Grid drawing is *orientable* iff SAT instance is satisfiable.
- Grid components (clauses, literals, wires, crossings,...) are composed of nodes \rightarrow Graph G_C .
- G_C is *realizable as a d -quasi unit disk graph* with $d \geq \sqrt{2/3} + \epsilon$ iff grid drawing is orientable.



Summary

- Virtual coordinates problem is important!
- Natural formulation as unit disk graph embedding.
 \rightarrow Clear-cut optimization problem.

Upper Bound : $\alpha \in O(\log^{2.5} n \sqrt{\log \log n})$
 Lower Bound : $\alpha \geq \sqrt{3/2} - \epsilon$

\rightarrow **Gap** between upper and lower bound is huge!

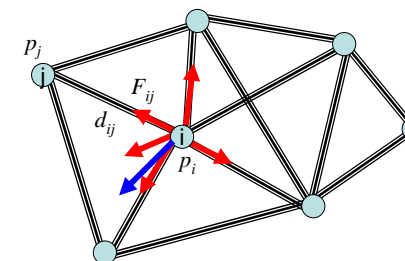
Open Problems:

- Diminish gap between upper and lower bound
- Distributed Algorithm



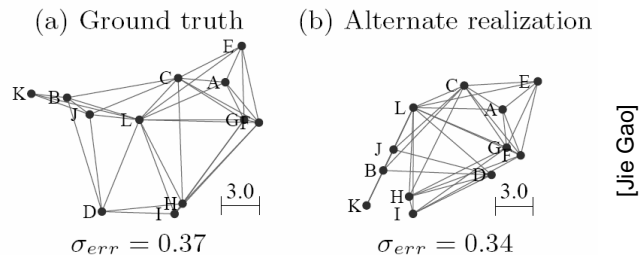
Heuristics: Spring embedder

- Nodes are "masses", edges are "springs".
- Length of the spring equals the distance measurement.
- Springs put forces to the nodes, nodes move, until stabilization.
- Force: $F_{ij} = d_{ij} - r_{ij}$, along the direction pp_j .
- Total force on n_i : $F_i = \sum F_{ij}$.
- Move the node n_i by a small distance (proportional to F_i).



Spring Embedder Discussion

- Problems:
 - may deadlock in local minimum
 - may never converge/stabilize (e.g. just two nodes)
- Solution: Need to start from a reasonably good initial estimation.



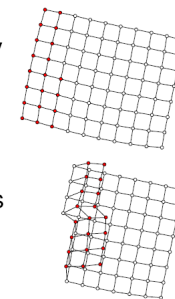
Heuristics: Priyantha et al.

N.B. Priyantha, H. Balakrishnan, E. Demaine, S. Teller:
Anchor-Free Distributed Localization
 in *Sensor Networks*, *SenSys*, 2003.

iterative process minimizes the layout energy

$$E(p) = \sum_{\{i,j\} \in E} \left(\|p_i - p_j\| - \ell_{ij} \right)^2$$

- fact: layouts can have *foldovers* without violating the distance constraints
- problem: optimization can converge to such a local optimum
- solution: find a good initial layout *fold-free* → already close to the global optimum (=“real layout”)



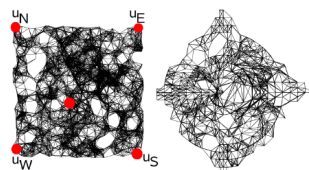
[Fleischer & Pich]



Continued

Phase 1: compute initial layout

- determine periphery nodes u_N, u_S, u_W, u_E
- determine central node u_C
- use polar coordinates



$$\rho_v = d(v, u_C) \quad \theta_v = \arctan \left(\frac{d(v, u_N) - d(v, u_S)}{d(v, u_W) - d(v, u_E)} \right)$$

as positions of node v

Phase 2: Spring Embedder

[Fleischer & Pich]

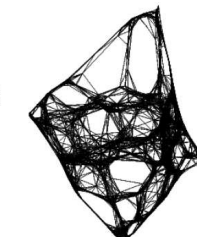
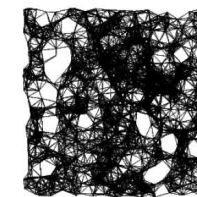


Heuristics: Gotsman et al.

C. Gotsman, Y. Koren [5]. **Distributed Graph Layout for Sensor Networks**, *GD*, 2004.

- initial placement: spread sensors $\frac{\sum_{\{i,j\} \in E} \exp(-\ell_{ij}) \|p_i - p_j\|^2}{\sum_{i < j} \|p_i - p_j\|^2} \rightarrow \min$
- linear algebra: minimized by second highest eigenvector v_2 of A where
$$a_{ij} = - \frac{\exp(-\ell_{ij})}{\sum_{j: \{i,j\} \in E} \exp(-\ell_{ij})}$$

$$a_{ii} = 1$$
- x, Ax, A^2x, A^3x, \dots converges to v_2
- $x_i \leftarrow \frac{1}{2} \left(x_i + \frac{\sum_{j: \{i,j\} \in E} \exp(-\ell_{ij} x_j)}{\sum_{j: \{i,j\} \in E} \exp(-\ell_{ij})} \right)$
- compute third eigenvector v_3 , use v_2, v_3 as coordinates

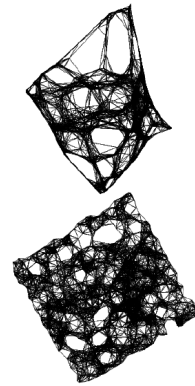


[Fleischer & Pich]



Continued

- ▶ distributed optimization (spring model)
- ▶ alternative: *majorization*
- ▶ compute sequence of layouts $p^{(0)}, p^{(1)}, p^{(2)}, \dots$ with $E(p^{(0)}) \geq E(p^{(1)}) \geq E(p^{(2)}) \geq \dots$
 - ▶ solve linear equation $L^{(t+1)} p^{(t+1)} = L^{(t)} p^{(t)}$ in distributed manner



[Fleischer & Pich]

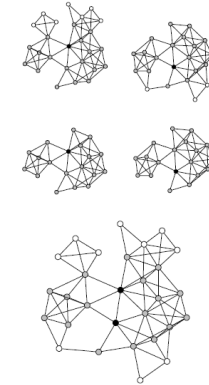


Heuristics: Shang et al.

Y. Shang, W. Ruml [7].

Improved MDS-based Localization, *IEEE Infocom*, 2004.

- ▶ compute a local map for each node (local MDS of the 2-hop neighborhood)
- ▶ merge local map patches into a global map (use incremental or binary-tree strategy)
- ▶ apply distributed optimization to the result



[Fleischer & Pich]



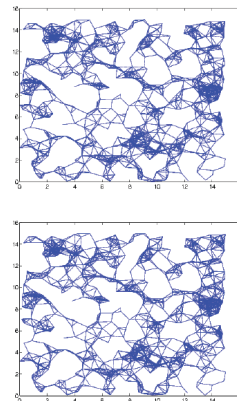
Heuristics: Bruck et al.

J. Bruck, J. Gao, A. Jiang [8]. **Localization and Routing in Sensor Networks by Local Angle Information**, *Mobile Ad Hoc Networking & Computing*, 2005.

- ▶ Choose an edge e as x -axis to obtain absolute angles.
- ▶ Form an LP whose variables are the edge lengths $\ell(e)$.
- ▶ For all edges $0 \leq \ell(e) \leq 1$.
- ▶ For any cycle e_1, \dots, e_p :

$$\sum_{i=1}^p \ell(e_i) \cos \theta_i = 0$$
 and

$$\sum_{i=1}^p \ell(e_i) \sin \theta_i = 0$$
- ▶ Non-adjacent node pair constraints.
- ▶ Crossing-edge constraints.



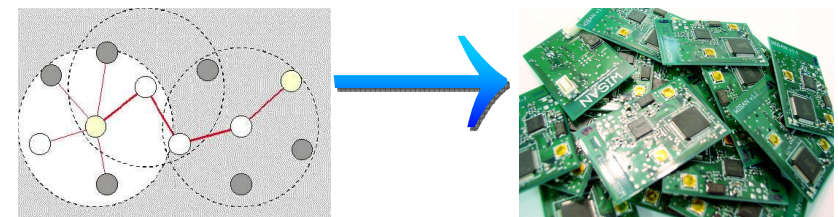
[Fleischer & Pich]



Practical lessons

Theory

Practice



- RSSI in sensor networks: good, but not for “reasonable” localization
- For exact indoor localization
 - Buy special hardware (e.g., UWB)
 - Place huge amount of short range anchors for single-hop localization

