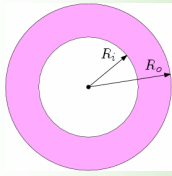


# Target Tracking with Binary Proximity Sensors

Raghu Mudumbai, Jaspreet Singh, Prof. Upamanyu Madhow, UC Santa Barbara

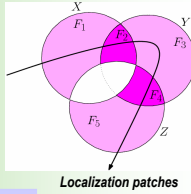
## Fundamental localization limits : One target scenario

(joint work with N. Shrivastava, Prof. S. Suri, UC Santa Barbara)



### Simple binary detection model

- detect objects within some radius
- only yes/no detection, no range info
- Practical imperfections**
- detection errors
- misses and false positives
- variations among sensors



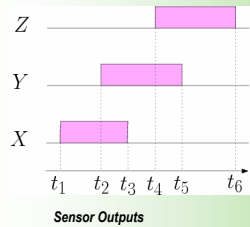
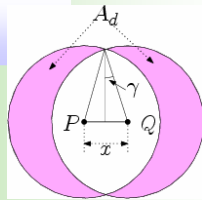
Localization patches

### How well can we localize with a network of sensors

- Each sensor can localize to within detection range
- Better localization by aggregating sensor data
- Intuition: more sensors → better localization
- Does larger detection radii improve localization?

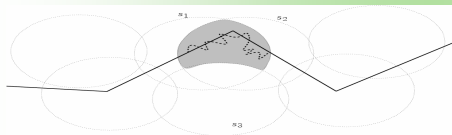
### Fundamental limits with geometric methods

- Probability that P & Q be distinguished
- For random deployment density  $\rho$
- Localization accuracy  $\propto \frac{1}{\rho R}$



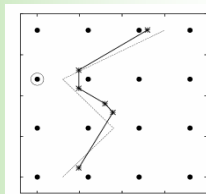
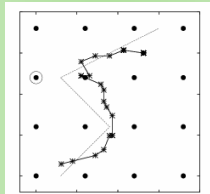
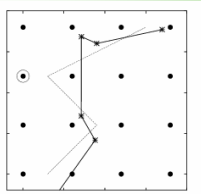
### Binary network for tracking

- Tracking → localization over time
- Binary sensors give sampled trajectory
- “Low-pass” version of actual path
- Simple geometric approach
  - motivated by Occam’s Razor
  - works well for smooth paths
- Does NOT work well with detection errors
- Combination of geometric and particle-filtering



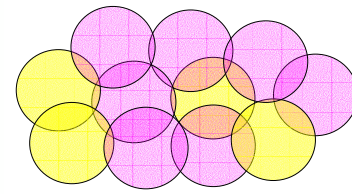
### Experiment with acoustic notes

- uses a tone-detection circuit for detection
- Extremely noisy and unreliable
- **One sensor failed completely!**
- Geometric approach not optimal
- Needs to be combined with particle filtering
  - works even with extreme imperfections



## Tracking multiple targets with binary sensors

(joint work with R. Kumar, Prof. S. Suri, UC Santa Barbara, R. Cagley, Toyon Research)

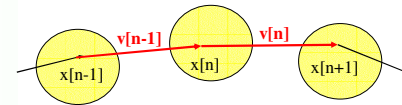
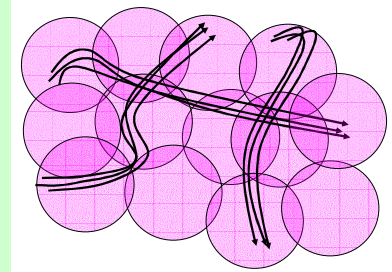


### Snapshot based inference

- significant ambiguity about not only the target locations, but also the number of targets
- greedy algorithm for minimal description in one dimension

### Tracking across snapshots : Particle-filter algorithm

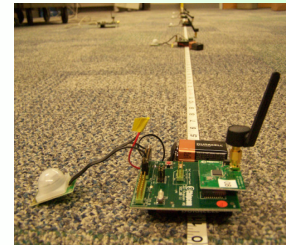
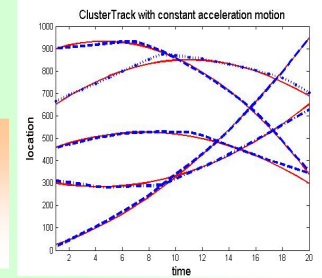
- Exploit the temporal correlations across snapshots : pick a cost function that penalizes variation in velocity
- To identify and track multiple targets, look for clusters of particles



$$\text{Additive cost function: } \sum \|v[n] - v[n-1]\|$$

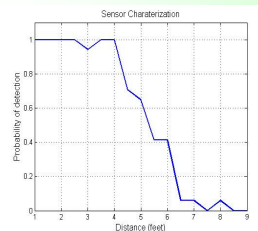
### Simulation results

- works really well if the targets separate out occasionally enough
- may generate some spurious trajectories, depending on the extent of overlap of the true paths



### Experiment with Passive Infrared Sensors

- each sensor transmits to base station when it changes state
- data gets time-stamped at the PC
- for ground truth: targets are provided separate sensor nodes equipped with localization engines



### Tracking performance

- Respectable performance even though one sensor completely missed a target

