



Spatial Considerations

A system that interacts with its environment should be "aware" of its surroundings



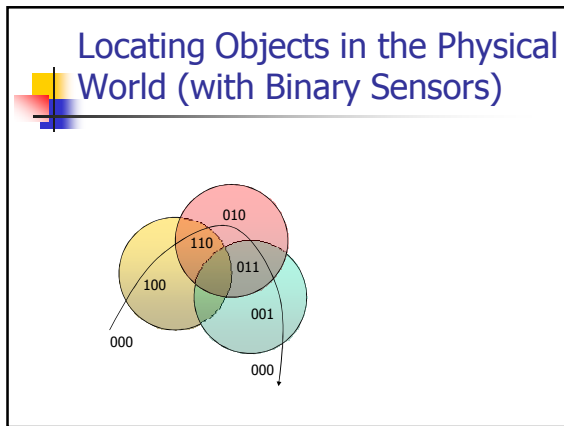
Target Tracking with Binary Proximity Sensors

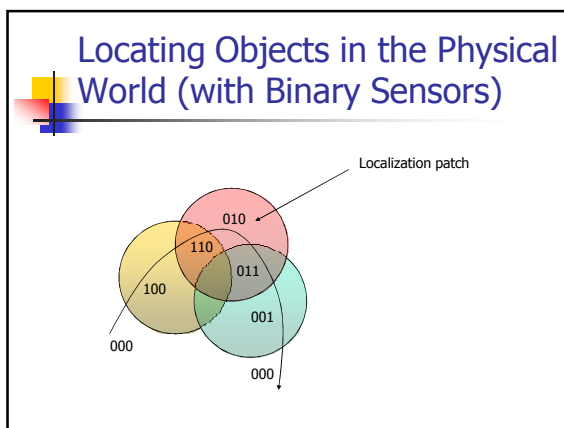
- Simplest sensor model:
 - Binary

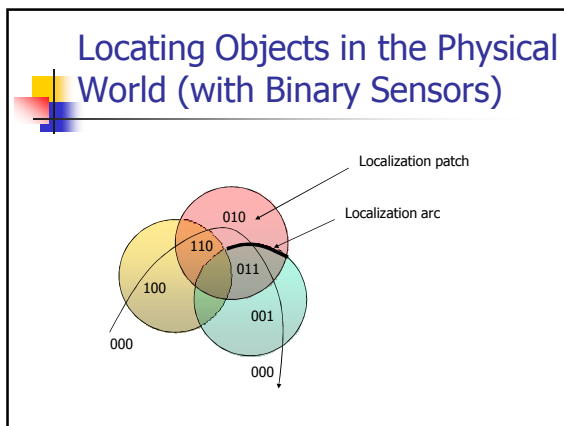


"Awareness" of the Physical World: Algorithmic Foundations

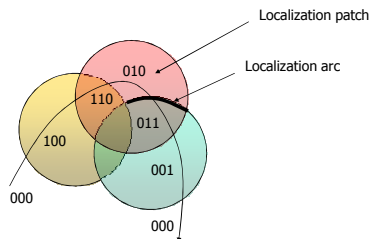
- Tracking (with perfect binary sensors)
- General tracking with imperfect sensors
- Multi-target tracking
- Hypothesis testing and data association
- Bayesian estimation
- Classification







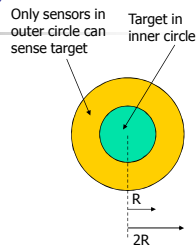
Locating Objects in the Physical World (with Binary Sensors)



Note: Can localize only to within a patch

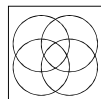
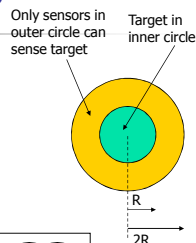
Patch Diameter?

- $N = \rho(4\pi R^2)$ sensors in outer circle



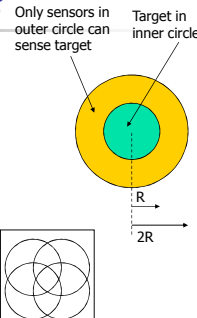
Patch Diameter?

- $N = \rho(4\pi R^2)$ sensors in outer circle
- There are at most $N(N-1) + 2$ patches in the inner circle



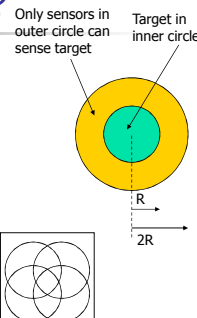
Patch Diameter?

- $N = \rho(4\pi R^2)$ sensors in outer circle
- There are at most: $N(N-1) + 2$ patches in the inner circle ($< N^2$)
- Patch area is at least $\pi R^2 / 16\rho^2\pi^2 R^4$



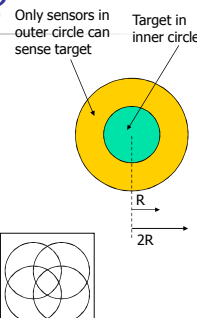
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- Area = $\Omega(1/\rho^2 R^2)$
- Diameter = $\Omega(1/\rho R)$



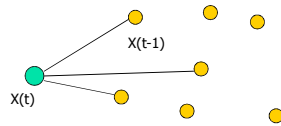
Result: Tracking Accuracy

Location error is at best
 $1/\rho R$

General Target Tracking with Imperfect Sensors

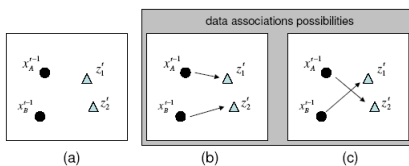
- The general single trajectory estimation problem

$$P(\text{target at } X(t)) = \int P(\text{target at } X(t) | \text{target was at } X(t-1)) * P(\text{target was at } X(t-1) | \text{observations}(t-1)) dX(t-1)$$



Multiple Target Tracking

- The data association problem:



Multiple Target Tracking

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(a) (b) (c)

N targets $\rightarrow N!$ possible associations

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N targets $\rightarrow N!$ possible associations
 T samples $\rightarrow (N!)^T$ possible associations

Multiple Hypothesis Testing

(a) (b) (c)

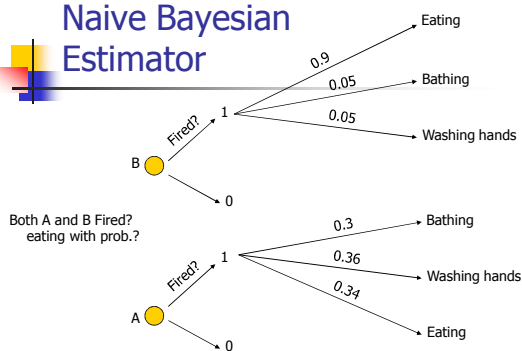
Consider $(N!)^K$ possible associations (where K is some window size \rightarrow keep the top M hypotheses)

Loop: Consider $M(N!)$ possible track extensions \rightarrow Keep top M

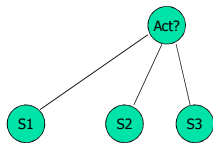
Spatial Activity Recognition

- Cameras and microphones?
 - Too intrusive
- Simple "switch" sensors
 - Perform activities
 - Collect vectors of sensor measurements for each activity
 - Find the probability that an activity is performed when a sensor is on.
 - Naive Bayesian Estimator: Multiply out the sensor probabilities.

Naive Bayesian Estimator



Naïve Bayesian Estimation



$$Act = \arg \max P(Act = a | S_1 = r_1, S_2 = r_2, \dots, S_n = r_n)$$

$$Act = \arg \max \frac{P(Act = a) \prod P(S_i = r_i | Act = a)}{\prod P(S_i = r_i)}$$



Classification

- Recursively find sensor with highest information gain

- Person vs. no person?

Motion sensor

ON: 70% person, 30% no person

OFF: 5% person, 95% no person

Magnetic sensor:

ON: 50% person, 50% no person

OFF: 55% person, 45% no person

Sounds sensor:

ON: 60% person, 40% no person

OFF: 50% person, 50% no person



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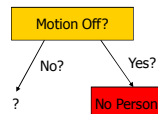
OFF: 55% person, 45% no person

Sounds sensor:

ON: 60% person, 40% no person

OFF: 50% person, 50% no person

- Pick motion sensor





Classification

- Find sensor with highest information gain

- Person vs. no person? Motion sensor ON and

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ON: 65% person, 35% no person

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graph TD; A[Motion Off?] -- Yes? --> B[No Person]; A -- No? --> C[Magnetic Off?]; C -- Yes? --> D[Person]; C -- No? --> E[?];
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