# **Introduction to Mobile Robotics**

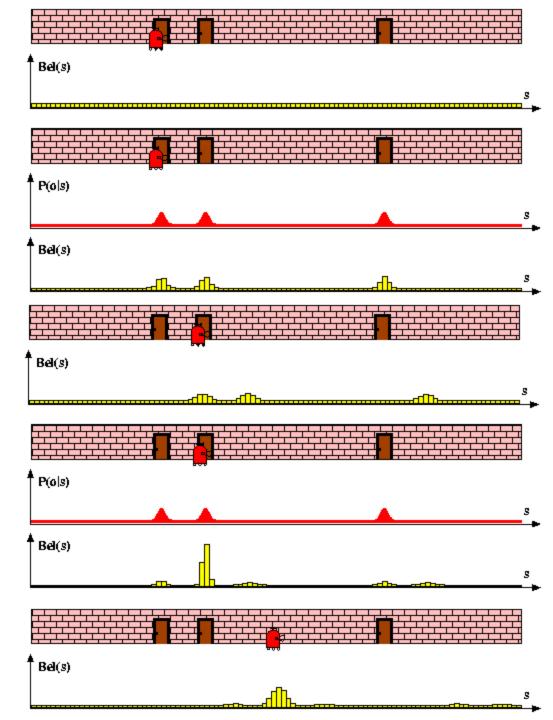
### **Bayes Filter – Discrete Filters**

Wolfram Burgard, Michael Ruhnke, Bastian Steder



$$Bel(x \mid z, u) = \alpha p(z \mid x) \int_{x'} p(x \mid u, x') Bel(x') dx'$$

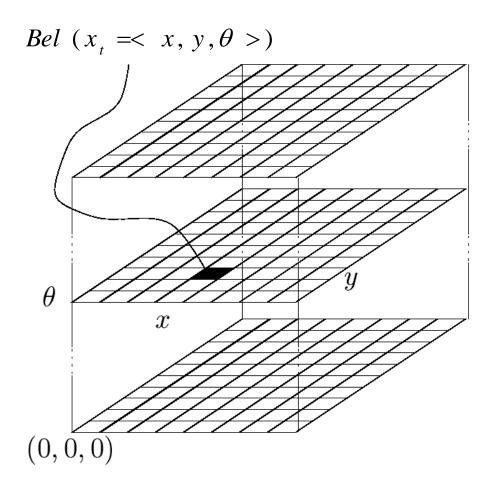
# Piecewise Constant



#### **Discrete Bayes Filter Algorithm**

```
Algorithm Discrete_Bayes_filter( Bel(x),d ):
2.
     \eta = 0
3.
     If d is a perceptual data item z then
4.
         For all x do
             Bel'(x) = P(z \mid x)Bel(x)
5.
6.
             \eta = \eta + Bel'(x)
7.
        For all x do
             Bel'(x) = \eta^{-1}Bel'(x)
8.
9.
      Else if d is an action data item u then
10.
         For all x do
              Bel'(x) = \sum P(x \mid u, x') Bel(x')
11.
    Return Bel'(x)
12.
```

# Piecewise Constant Representation



## Implementation (1)

- To update the belief upon sensory input and to carry out the normalization one has to iterate over all cells of the grid.
- Especially when the belief is peaked (which is generally the case during position tracking), one wants to avoid updating irrelevant aspects of the state space.
- One approach is not to update entire sub-spaces of the state space.
- This, however, requires to monitor whether the robot is de-localized or not.
- To achieve this, one can consider the likelihood of the observations given the active components of the state space.

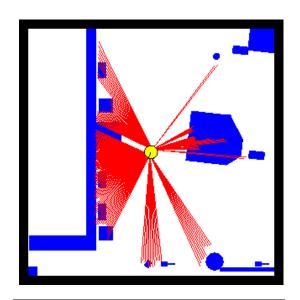
# Implementation (2)

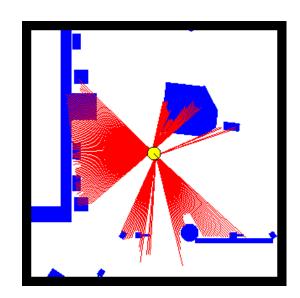
- To efficiently update the belief upon robot motions, one typically assumes a bounded Gaussian model for the motion uncertainty.
- This reduces the update cost from  $O(n^2)$  to O(n), where n is the number of states.
- The update can also be realized by shifting the data in the grid according to the measured motion.
- In a second step, the grid is then convolved using a separable Gaussian Kernel.
- Two-dimensional example:

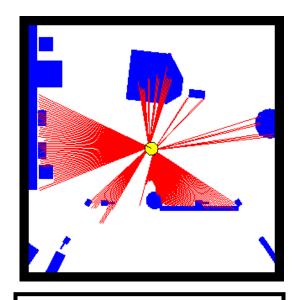
1/16	1/8	1/16		1/4				
1/8	1/4	1/8	<b>≅</b>	1/2	+	1/4	1/2	1/4
1/16	1/8	1/16		1/4				

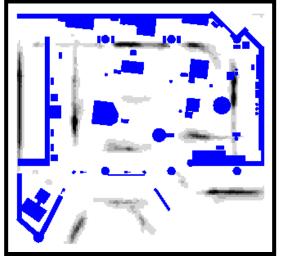
- Fewer arithmetic operations
- Easier to implement

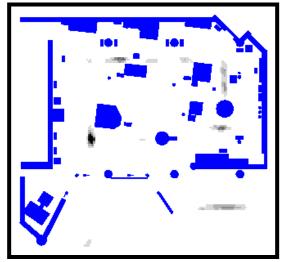
#### **Grid-based Localization**

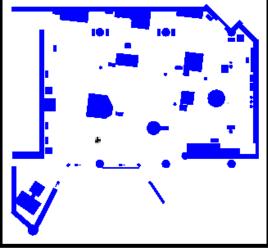




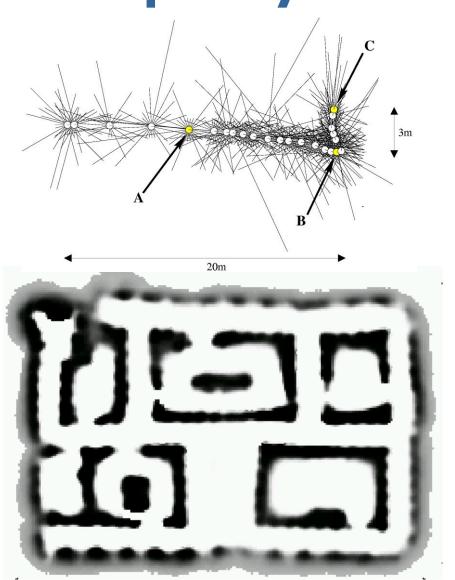


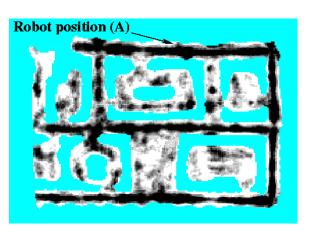


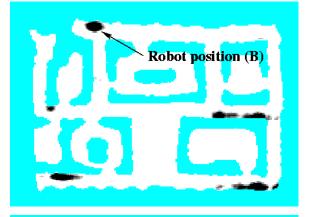


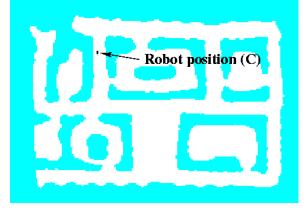


# Sonars and Occupancy Grid Map



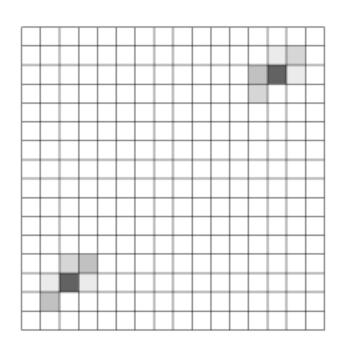


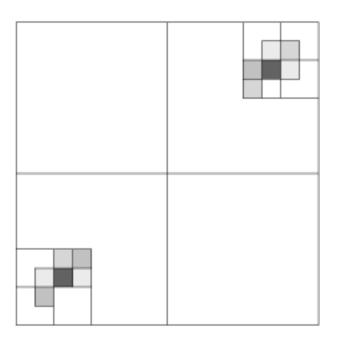




#### **Tree-based Representation**

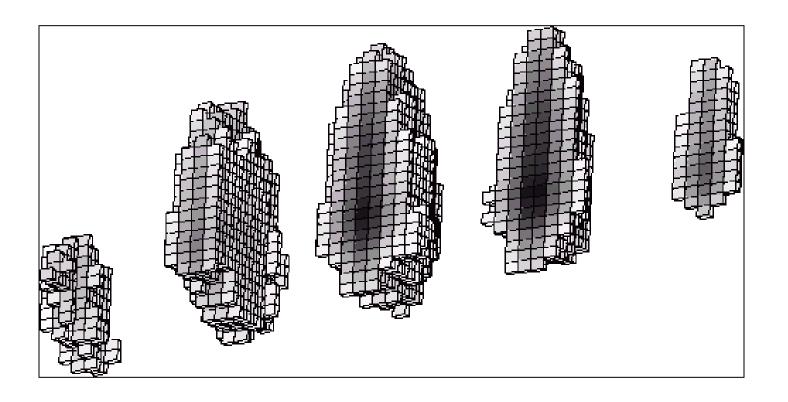
Idea: Represent density using a variant of octrees





# **Tree-based Representations**

- Efficient in space and time
- Multi-resolution



#### **Summary**

- Discrete filters are an alternative way for implementing Bayes Filters
- They are based on histograms for representing the density.
- They have huge memory and processing requirements
- Can easily recover from localization errors
- Their accuracy depends on the resolution of the grid.
- Special approximations need to be made to make this approach having dynamic memory and computational requirements.