

Using Fast Classification of Static and Dynamic Environment for Improving Bayesian Occupancy Filter (BOF) and Tracking

Qadeer Baig, Mathias Perrollaz, Jander Botelho Do Nascimento,
Christian Laugier

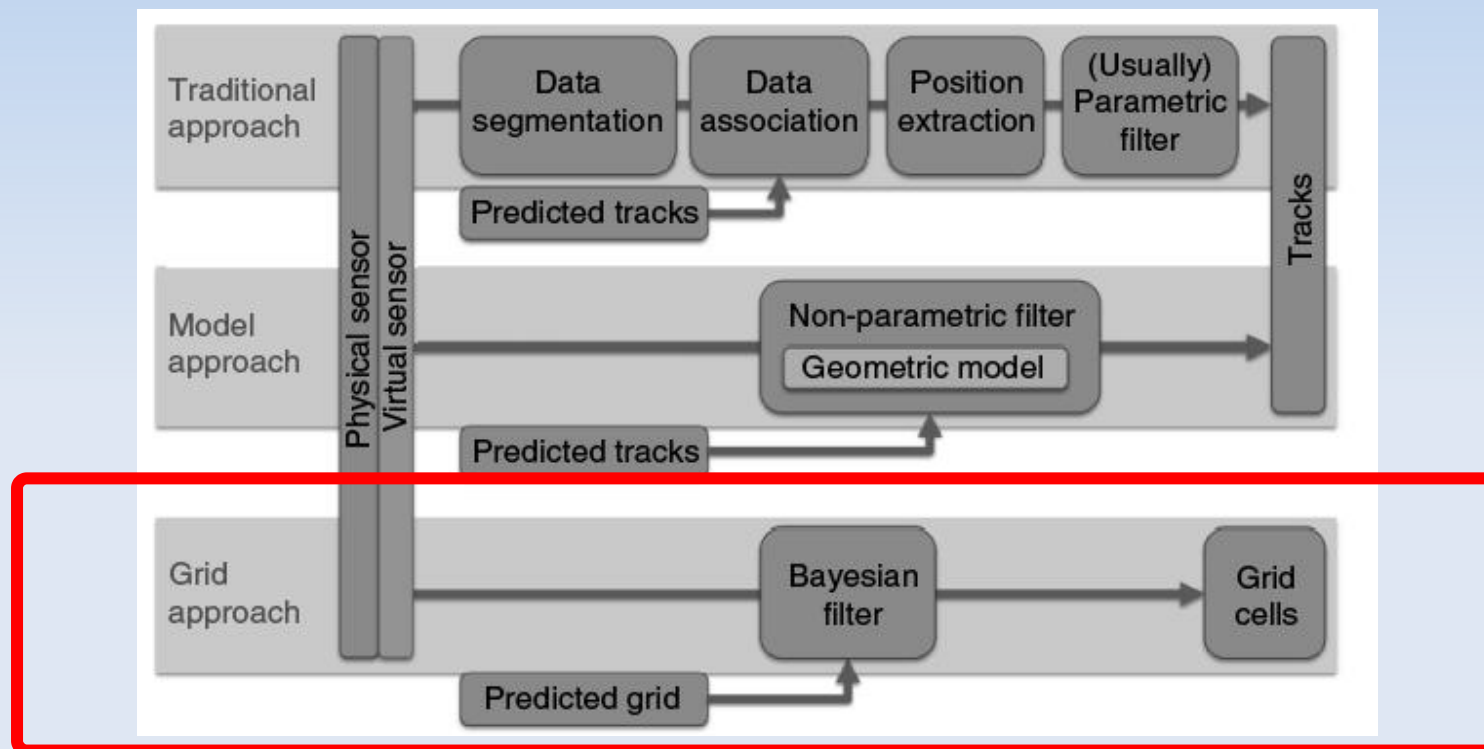
e-Motion team,
Inria Rhône-Alpes (Grenoble), France

Outlines

- ◆ Environment monitoring in the Bayesian Occupancy Filter (BOF) framework and FCTA
- ◆ Fast Motion Detection
- ◆ Integration with BOF and FCTA
- ◆ Results

Grid based DATMO

- ◆ For mobile robots: Detection and Tracking of Moving Objects (DATMO) is essential for navigation
- ◆ For intelligent vehicles : DATMO is essential for risk estimation
- ◆ Three main approaches for DATMO [*Petrovskaya11*]:



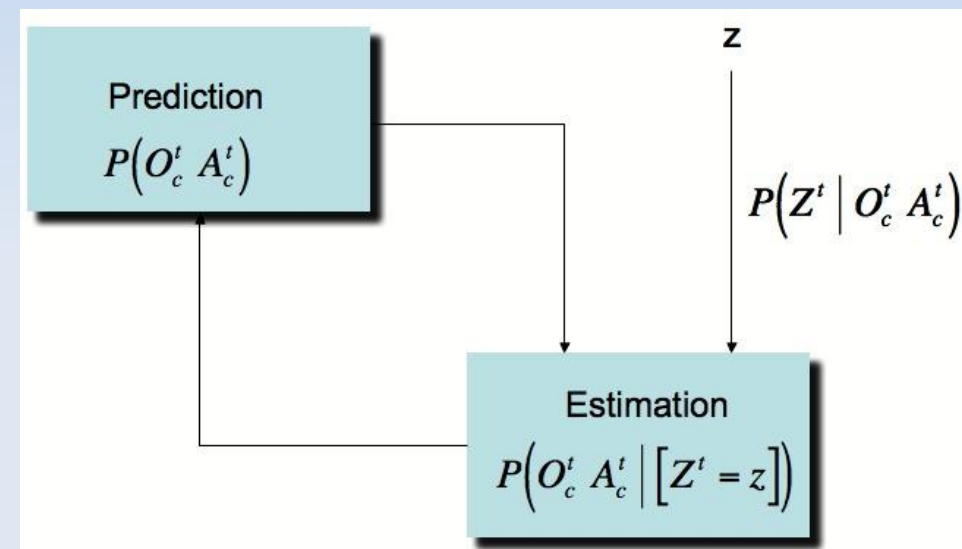
Bayesian Occupancy Filter (BOF)

[Coué IJRR 2005]

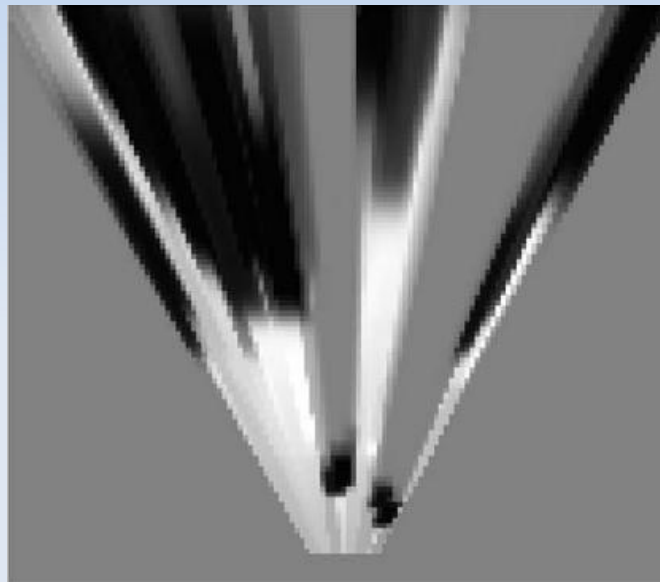
- Grid-based approach for Bayesian Filtering
 - **Prediction/estimation** loop
 - Each cell has an estimated **occupancy** and a probability distribution over possible **velocities**

Allows to estimate velocities from grid measurements

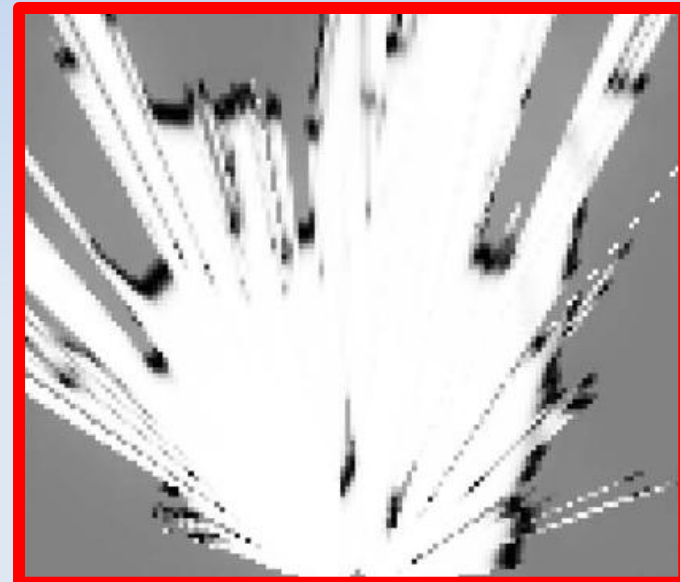
- **Prediction:** propagates occupancy and velocity to neighboring cells, using dynamic models
- **Estimation:** corrects predicted grids using observation grids computed using sensor model



BOF : input grids



Using stereo-vision
Perrollaz, T-ITS 2012



Using multi-layers laser scanners
Adarve, ICRA 2012

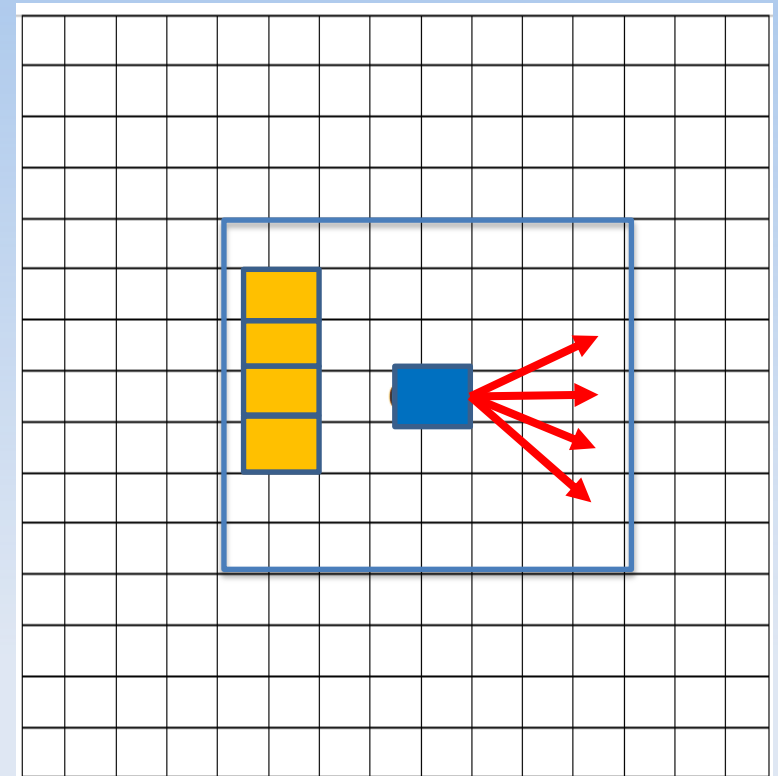
BOF Velocities and FCTA

◆ Cell antecedents: knowing antecedent of a cell tells its velocity

- . Antecedent is the cell at $t-1$
- . Antecedent only in a neighborhood
- . Distribution over all antecedents
- . Relative velocities

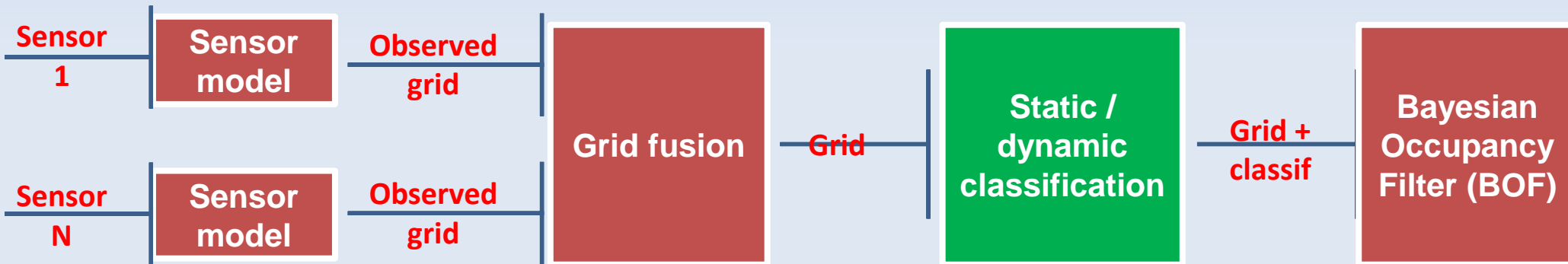
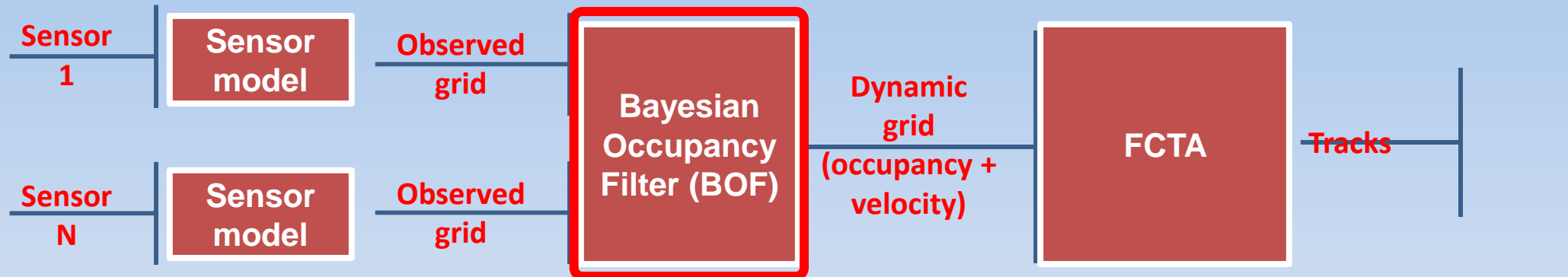
◆ FCTA: *[Mekhnacha 2008]*

- . Fast Clustering and Tracking Algo
- . Clustering based on cell occ and vel
- . Too many tracking hypotheses
- . Static objects are also tracked
- . Many parameters to tune
- . Results depend on parameters values
- . Convergence is slow in large regions



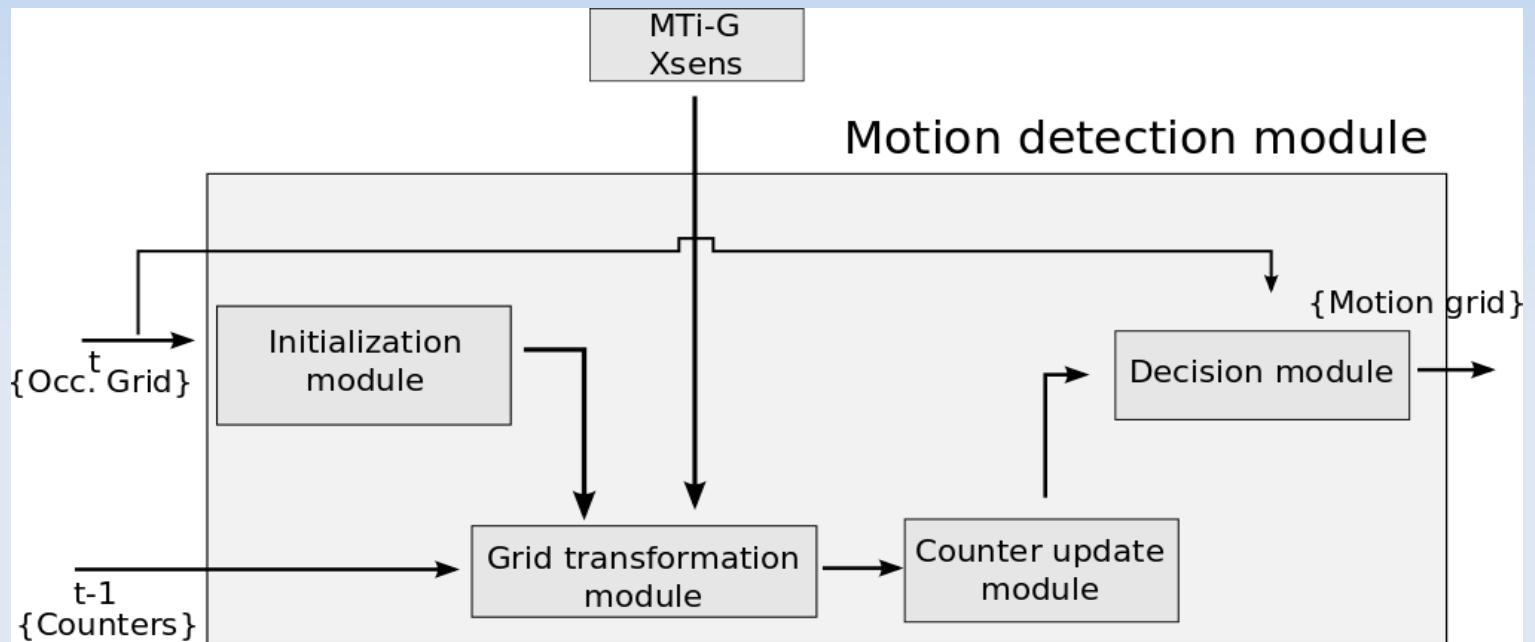
Solution: Finding static parts before using the BOF

System Architecture



Fast Motion Detection

- ◆ Main idea: How many times a cell is observed as free and how many times occupied, in a global coordinate system
 - . Use free/occupied counters for each cell
 - . Map cells from $t-1$ to t , using robot's motion
 - . Update counters at each timestep
 - . Framework:

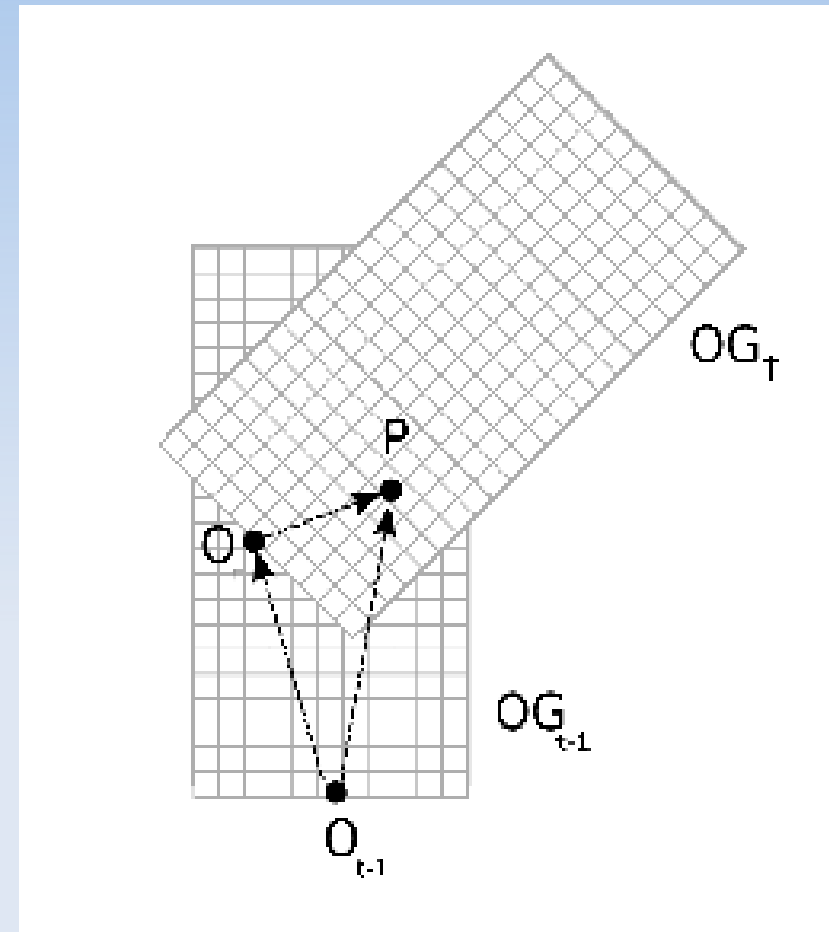


Grid Transformation

◆ The objective is to map a cell j in grid OG_{t-1} to cell i in grid O_t , with the hypothesis of static environment

◆ Method :

Using motion data from IMU
Velocity, Angular velocity
and circular motion model find
pose of O_t w.r.t O_{t-1}



Initialization and update

◆ Initialization:

- .FreeCounter $[i] = 1$, if $OG_t[i] < 0.5$
- .OccupiedCounter $[i] = 1$, if $OG_t[i] > 0.5$

◆ Updating counters from previous time step :

- .Mapping of cells of grid at time $t-1$ to grid at t
- .Update counters:
 - .FreeCounter $[i] +=$ FreeCounter $_{t-1}[j]$
 - .OccupiedCounter $[i] +=$ OccupiedCounter $_{t-1}[j]$

◆ Decision

- .MotionGrid $[i] = F(OG_t[i], \text{FreeCounter}[i], \text{OccupiedCounter}[i])$

Very fast: Do not solve complete SLAM

Results: Platform



- .2 IBEO Lux laser scanners
- .(4 layers each)
- .1 TYXZ stereo camera
- .(baseline 22cm)
- .Xsens MTI-G inertial sensor



Motion Detection Results



Integration with BOF and FCTA

With BOF: Modified velocity update step, update only if cell is moving. This means remove velocity information for static cells

$$P(A_i^t | A_i^{t-1}) = \begin{cases} (1 - \epsilon)P(A_{A_i^{t-1}}^{t-1}) + \epsilon / \|A_i\| & \text{if } MotionGrid_t[i] > 0 \\ 1 / \|A_i\| & \text{otherwise} \end{cases}$$

With FCTA: Modified clustering step, take a cell into account for clustering if it has velocity information

Result: More than 78% false positives removed with relaxed FCTA parameters

Results

Videos

Thank You!

Questions?