Fast classification of Static and Dynamic Environment for Bayesian Occupancy Filter

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Outlines

 Environment monitoring in the Bayesian Occupancy Filter (BOF) framework

Fast Motion Detection

First results

- Conclusion



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/ADAS
- . Three main approaches for DATMO [*Petrovskaya11*]:



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Grid based DATMO

Grid-based DATMO:

- Occupancy grid framework [Elfes 89]
- Each cell as a probability of being occupied
- No "objects"
 - ➔ No data association





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



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Moving objects vs Static environnement

- Separating dynamic (moving) objects from static ones (map)
 - Simplify the tracking (less hypotheses)
 - Allow to reason on behaviors for further risk estimation
 - Toward semantic description of the scene
- Classic approaches :
 - background substraction, mostly in vision [Jain 1979, Li 2000, Taleghani 2009]
 - . SLAM-based approaches for occupancy grids [Wang 2007, Burlet 2007, Vu 2009]
 - SLAM + DATMO
 - . SLAMMOT
 - → the complete SLAM needs be solved !



Fast classification of the environment

- Fast classification of static/dynamic environment
 - Separate static from dynamic environment within the BOF framework
 - Not solving the complete SLAM problem
 - Fast and computationally efficient

→ First idea : filtering the grid using the cells' velocities estimated by the BOF



BOF: Velocities

- Cell antecedents: knowing antecedent of a cell tells its velocity
 - ➔ Relative velocities
- Problem:
 - No information about the robot's motion
 - . Too many tracking hypotheses
 - . Static objects are also tracked
 - . Convergence is slow in large regions



→ Solution: Finding static parts before using the BOF



BOF: Velocities

- Solution: Finding static parts before using the BOF
- Advantages:
 - Reduces the number of hypotheses for BOF filtering
 - Can provide pior information about velocities to the BOF



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





Grid Transformation

- The objective is to map a cell *j* in grid OG_{t-1} to cell *i* in grid O_t , with the hypothesys of static environnement
- Method :
 - Using motion data from IMU
 - Velocity
 - Angular velocity
 - and circular motion model find pose of O_t w.r.t O_{t-1}
 - Using a global coordinate system to avoid border effects





Initialization and update

- Initialization:
 - FreeCounter_t[i] = 1, if $OG_t[i] < 0.5$
 - OccupiedCounter_t[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_t[i] += FreeCounter_{t-1}[j]
 - OccupiedCounter_t[i] += OccupiedCounter_{t-1}[j]

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Motion grid

Motion

- MotionGrid[i] = $F(OG_t[i], FreeCounter_t[i], OccupiedCounter_t[i])$
- F can be a decision function, like:

$$MotionGrid_t[i] = \begin{cases} 1, & OG_t[i] > 0.5 \text{ and} \\ & FreeCount_t[i] > 2 * OccupiedCount_t[i] \\ 0, & \text{otherwise} \end{cases}$$

• Or F can be a probabilistic function, for further decision



First results



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









First results



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Results



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Next steps

- Find best way to use this in the BOF
 - Use velocities' prior information for the static environnment
 - Use only dynamic environnement in the BOF
 - Use a Bayesian combination of both approaches
 - Incorporate the classification process into the BOF
- Evaluate the influence on the clustering of the grid
- Manage uncertainty of the motion estimation





Thank You!

Questions?

