## Managing Swarms of Robots: From Centralized to Decentralized Scheduling

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## OS Support for Mobile Robots

 Set of heterogeneous devices (with different capabilities)



## **OS Support for Mobile Robots**

 Management, control and coordination by the operating system



## OS Support for Mobile Robots

 Multiple (distributed) applications executed (simultaneously on the "swarm"













#### Definition

- Virtual movement: "Logic" movement of application
- Physical movement: Real physical movement

# **Application Model**

#### Building blocks

- Flexible composition of actions at runtime
- Connect input / output
- Actions
  - Constrainable in space and time
  - E.g., take picture, measure temperature
- Execution
  - Concurrent / sequential execution of (in-)dependent actions
  - Transparent execution across node boundaries



# Scheduling of ActionSuites

• Input: ActionSuite as, dependency graph g<sup>d</sup>

- Action  $a \in as = (g, t_{min}, t_{max}, d)$
- $g^d$  is a directed, acyclic graph
- Output: Scheduled job  $(j^a, j^t)$  for each  $a_i \in as$

• 
$$j^a = (p, t, r)$$
  
•  $j^t = [(x_1, y_1, t_1), .$ 

- World
  - Static obstacles O<sup>s</sup>
  - Dynamic obstacles  $O^m$ ,  $f: T \to \mathbb{R}^2$

.]

## Determine Slot Candidate

- Determine slot candidate
  - Euclidean distance:  $s = ||\vec{x}_{a_i} \vec{x}_{a_i}||$
  - Overhead of detour:  $\Delta s := (s_1 + s_2) s_{12}$
  - Select slot candidate with minimal detour
- Temporal constraints
  - *t<sub>min</sub>*, *t<sub>max</sub>* are action constraints
  - t1, t2 define begin / end of free slot



## Path Planner

- Computes trajectory
  - .. from  $\vec{x}_{a_1}$  to  $\vec{x}_{a_3}(s_1)$
  - .. from  $\vec{x}_{a_3}$  to  $\vec{x}_{a_2}(s_2)$
- Resulting problem is a Trajectory Planning Problem (TPP)
- Solution: Path-Velocity-Decomposition
  - Static Path Planning Problem (PPP)
  - Dynamic Velocity Planning Problem (VPP)
- Path Planner obtains input from the job scheduler
  - loc: location candidate (P)
  - *slot*: (*r*, [*t*<sub>1</sub>, *t*<sub>2</sub>])



## Forbidden Regions

- Compute possible collisions with dynamic obstacles and mark them as forbidden regions
- $O^m$  crosses robot path  $\pi$
- The result is a forbidden region in  $s \times t$  space
- *n* robot segment and *m* segments of  $O^m$  creates  $n \times m$  tiles



## Velocity Planning Problem (VPP)

- Computes velocity profile along path  $\pi$
- s × t-space shows forbidden regions (dynamic obstacles)
- Find time based mapping  $s: T \rightarrow S, S = [s_I, s_F] = [0, 8]$ s(2) = 1 or s(2) = 4
- Extend to:  $ec{x}_{\pi}: \ec{T} 
  ightarrow ec{x}$   $\Leftrightarrow$   $ec{x}_{\pi}: (\ec{T} 
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#### • VPP induces high complexity

- Meshing to obtain velocity graph
- Check of visibility

 $\rightarrow$  Execution time depends on  $O^m$ ,  $\pi^m$ ,  $\pi$  (segments) planning

#### • However, scalability problem still remains

- Very large worlds
- Huge amount of robots

- GPGPU computing suitable for trajectory planning
- Approaches from autonomous driving, TP with time horizon  $\rightarrow$  Loose guarantees and predictability, deadlocks
- $\bullet$  Scalability  $\rightarrow$  Decentralized Approaches
  - Static Approach
  - Dynamic Approach

### Static Approach

- Splitting the world into static regions
- Each regions has its own scheduler
- Scheduling decentralized and in parallel



### Static Approach

- Cross-boundary movement possible
- .. requires locking
- Full locking  $\rightarrow$  degenerates to centralized version  $\rightarrow$  unlikely



### Static Approach

- ~250 trajectory segments
- 17 segments with cross-boundary movement
- $\bullet$   $\approx$  93.2%



## Dynamic Approach

- Regions emerge dynamically based on scheduling requests
- Formation of "scheduler groups"
- Run distributed and in parallel
- Getting destroyed after request



## Dynamic Approach

- Each group creates a region lock
- Bounding box of group
- If overlap  $\rightarrow$  dependency (waiting) graph
- Variants
  - One node becomes scheduler
  - All nodes become schedulers (first one wins)
  - All nodes become schedulers (consensus / majority voting)



## **Preliminary Results**

- 25 nodes: Core i3-6100U CPU @ 2.30GHz, dual core
- Time: Centralized vs. decentralized
- f(o), number of obstacles
- 10,000 jobs
- More obstacles create more path segments
   → Complexity of forbidden regions increase



## **Preliminary Results**

- 25 nodes: Core i3-6100U CPU @ 2.30GHz, dual core
- Time: Centralized vs. decentralized
- *f*(*r*), region size
- 10,000 jobs



## **Preliminary Results**

- 50 nodes: Core i3-6100U CPU @ 2.30GHz, dual core
- Acceptance rate: Centralized vs. decentralized
- f(n), number of nodes
- Non-linear increase of acceptance rate
- (De-)centralized approach has only little impact



- OS support for mobile robot swarms
- TPP has high complexity
- Address scalability by decentralization
  - Static approach
  - Dynamic approach
- Locking vs non-locking
- Distributed data structures: transfer only delta schedule
  - After scheduling
  - On request