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CIS 700

2017

Simulating Heterogeneous Crowds

Based on a Eurographics 2014 Tutorial

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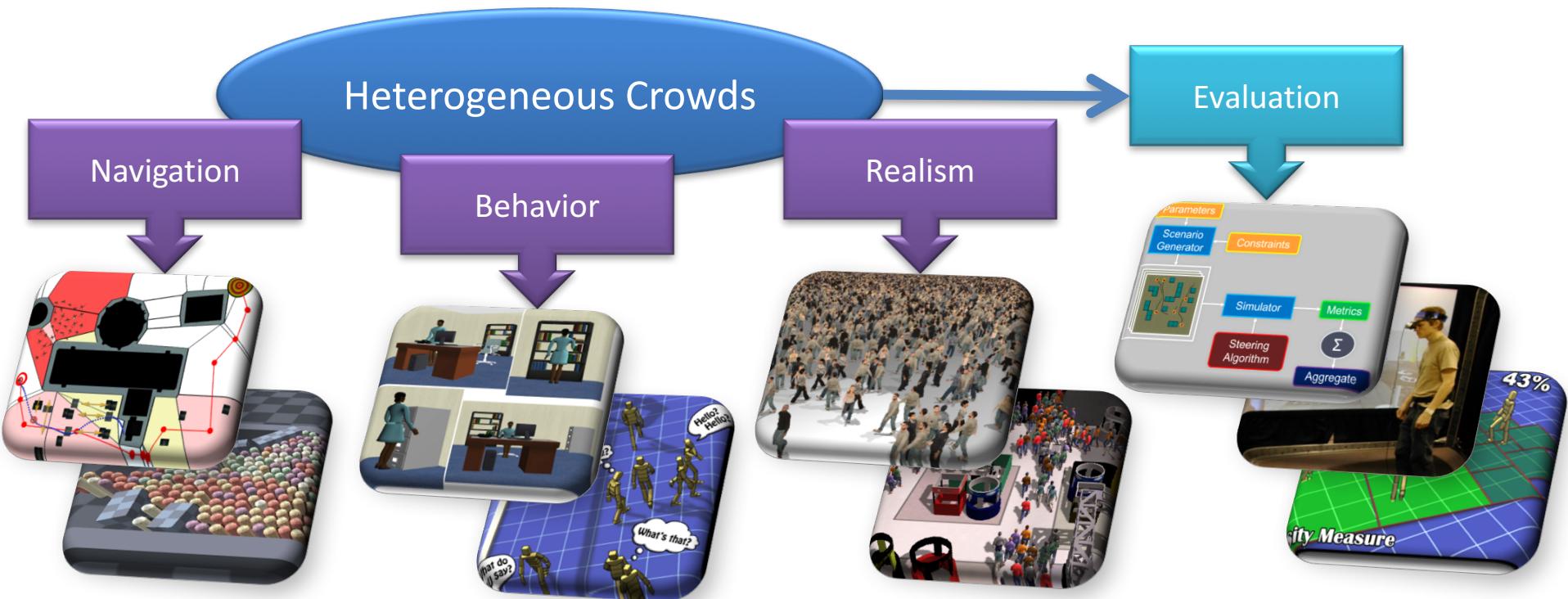
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Overview



Simulating Heterogeneous Crowds

NAVIGATION

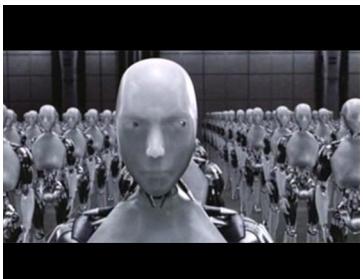
Stephen J. Guy
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Crowd Simulation Challenges

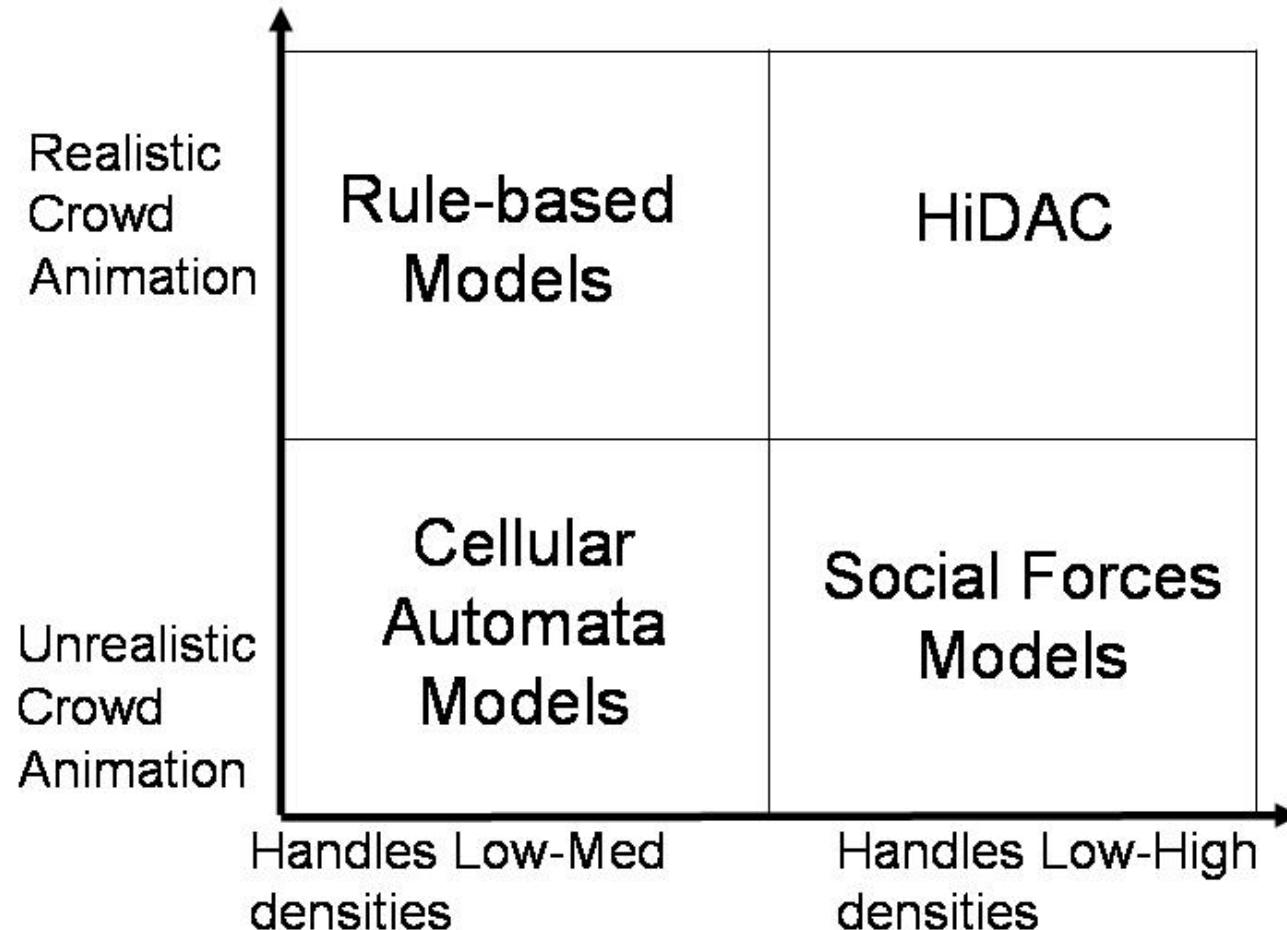
- ◆ People navigate in very difficult circumstances
 - Complex environments
 - Fast paced motion
 - Very dense crowds



How can we *reproduce* this ability in virtual environments?



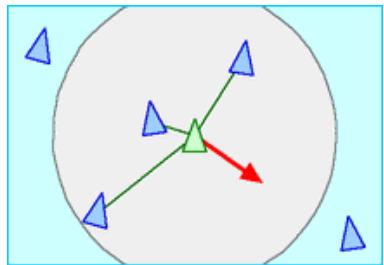
Approaches to Crowd Animation



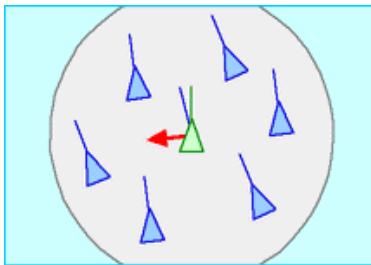
Rules: Simulating Flocks & Herds

◆ Boids Algorithm [Reynolds '87]

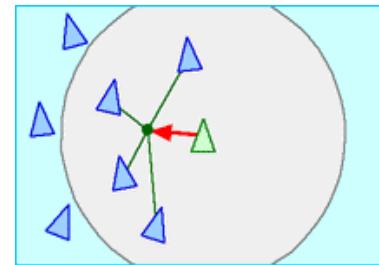
- Three simple rules, applied to each timestep, in a local neighborhood:



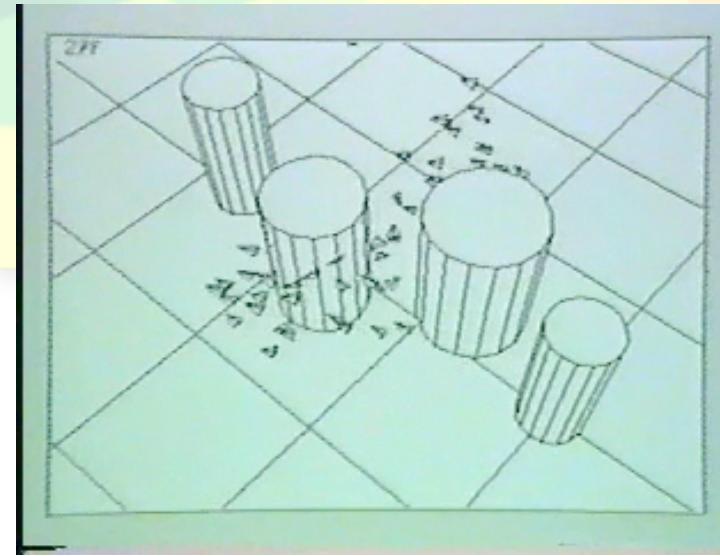
Separation



Alignment



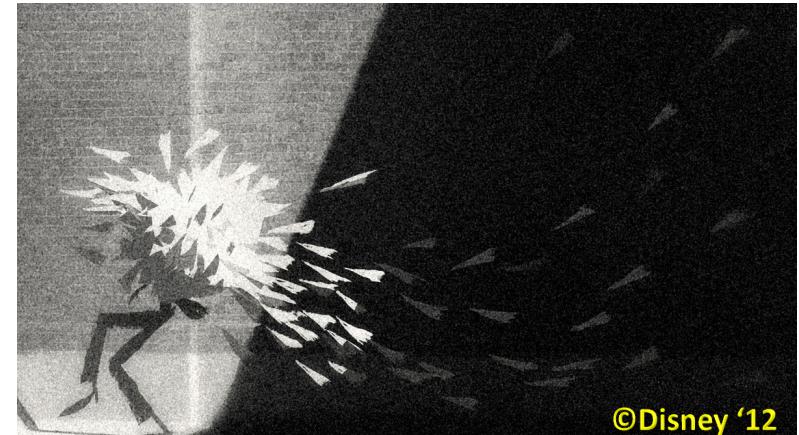
Cohesion



◆ Commonly used for CGI animals:



©Disney '95



©Disney '12

The HiDAC Model: Forces plus Geometry (Pelechano, Allbeck, Badler, SCA 2007)

- Forces influencing the direction of movement (for the i^{th} agent at the n^{th} time step):

$$\mathbf{F}_i^{To}[n] = \underbrace{\mathbf{F}_i^{To}[n-1]}_{\text{Current direction}} + \underbrace{\mathbf{F}_i^{At}[n]w_i^{At}}_{\text{Attractor}} + \underbrace{\sum_w \mathbf{F}_{wi}^{Wa}[n]w_i^{Wa}}_{{\text{Walls}}} + \underbrace{\sum_k \mathbf{F}_{ki}^{Ob}[n]w_i^{Ob}}_{{\text{Obstacles}}} + \underbrace{\sum_{j(\neq i)} \mathbf{F}_{ji}^{Ot}[n]w_i^{Ot}}_{{\text{Other Agents}}}$$

- Forces used to compute new agent position at time $n+1$.

HiDac Animations



SCA 07

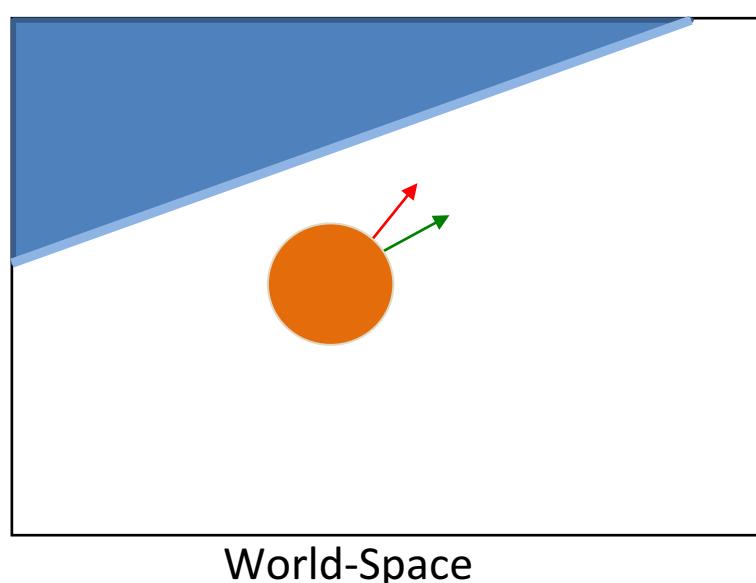
High Density Autonomous Crowds

- Each 2D agent has its own desired velocity (vector in the same color of the agent)
- yellow lines = other agents avoidance.
- cyan lines = obstacles and walls avoidance.
- pink rectangle = rectangle of influence for one of the agents (updates density)
- red points = agents that affect the perceived density
- circle in same color as agent = waiting rules

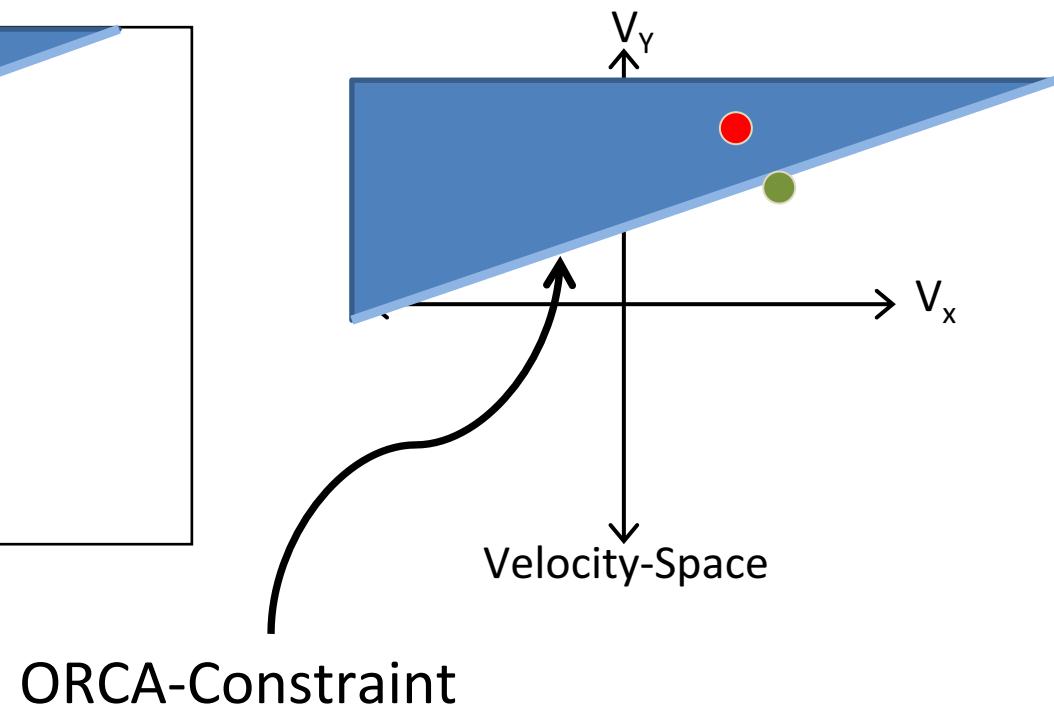
Optimal Reciprocal Collision Avoidance (ORCA)

(J. van den Berg, S. J. Guy, M. Lin, D. Manocha, STAR 2011)

- ◆ Identify the set of agent velocities which might collide soon, and then avoid them!
- ◆ Key Idea: These velocities can be found geometrically, followed by an optimization step.



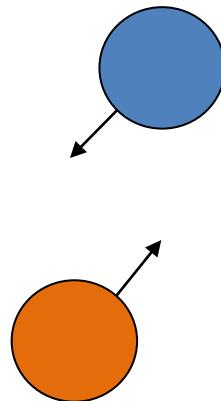
World-Space



ORCA-Constraint

Agent-Agent Interactions

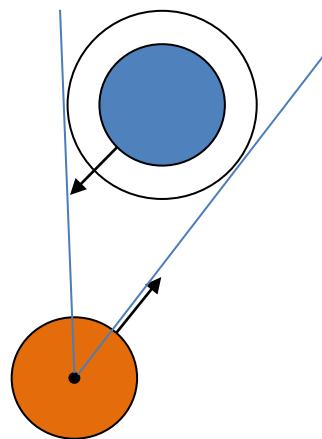
- ◆ Linear velocity constraints can be formulated for agent pairs
 - Key Idea: Reciprocity – Each agent must avoid at least $\frac{1}{2}$ the collision.



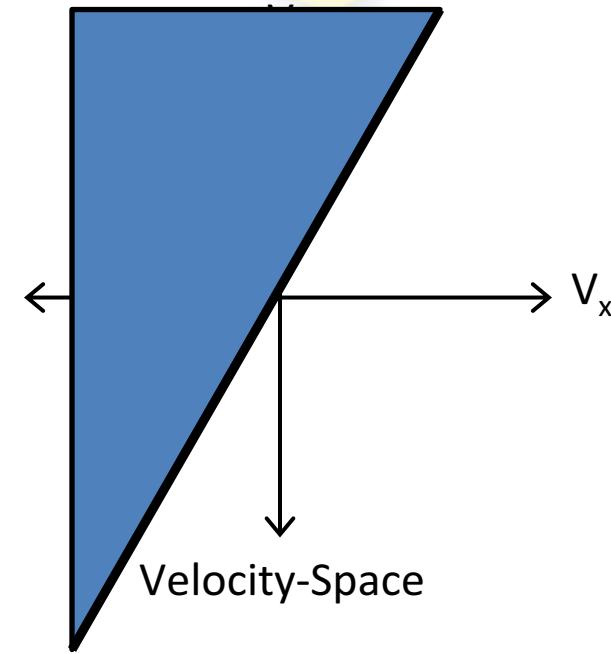
World-Space

Agent-Agent Interactions

- ◆ Linear velocity constraints can be formulated for agent pairs
 - So expand radius of other agent by “my” radius (Minkowski sum of agent outlines).



World-Space



- ◆ <https://www.youtube.com/watch?v=Hc6kng5A8IQ>

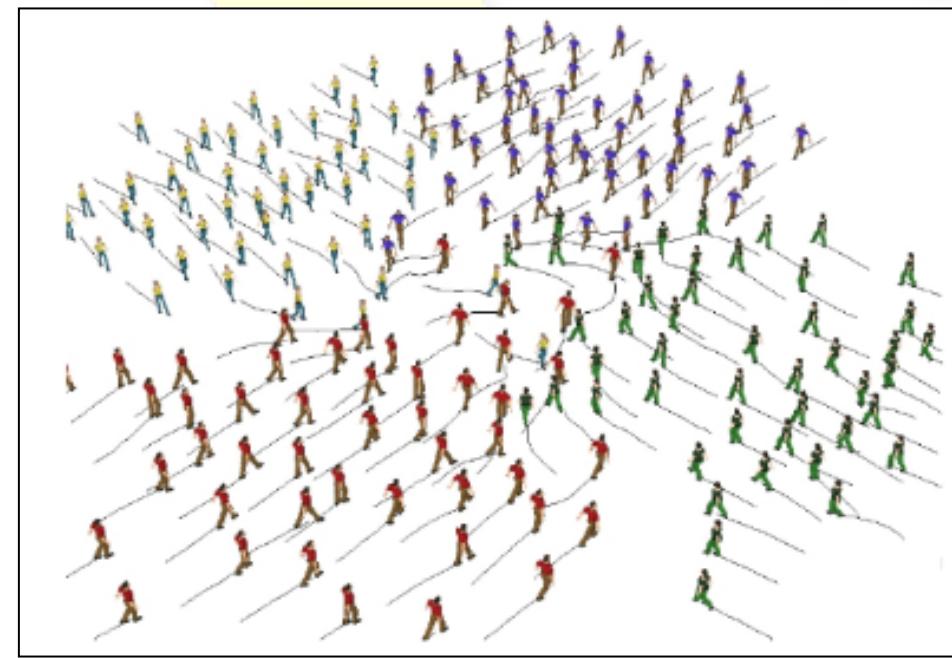
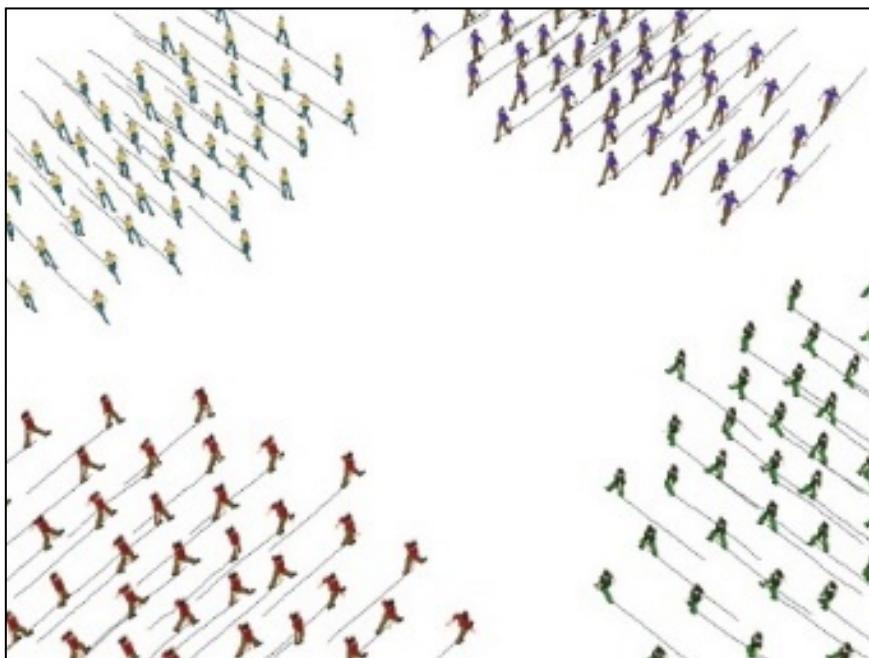
Continuum Crowds

(Treuille and Popović, SIGGRAPH 2006)



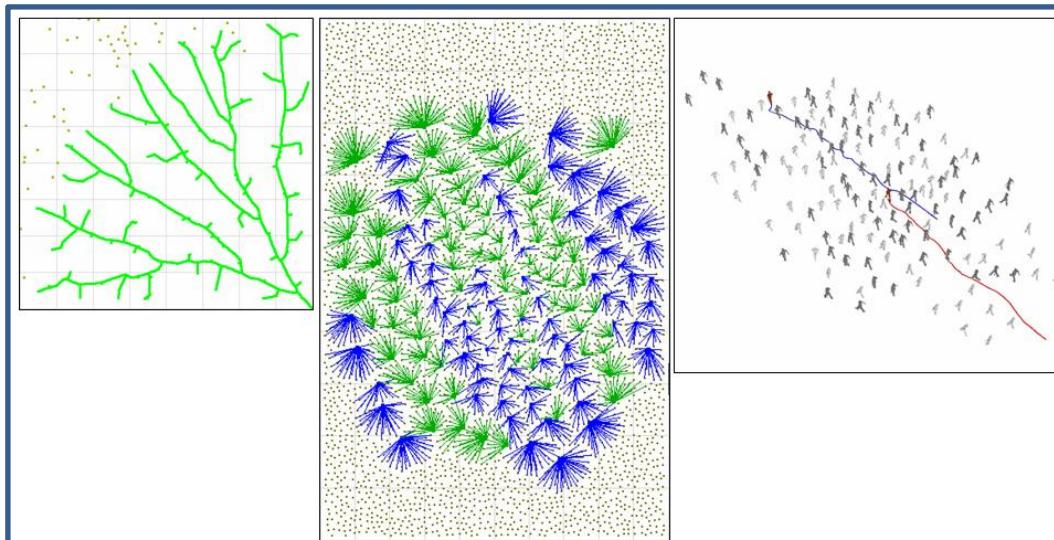
BioCrowds (Soraia Musse)

- ◆ Crowd simulation based on biological patterns.

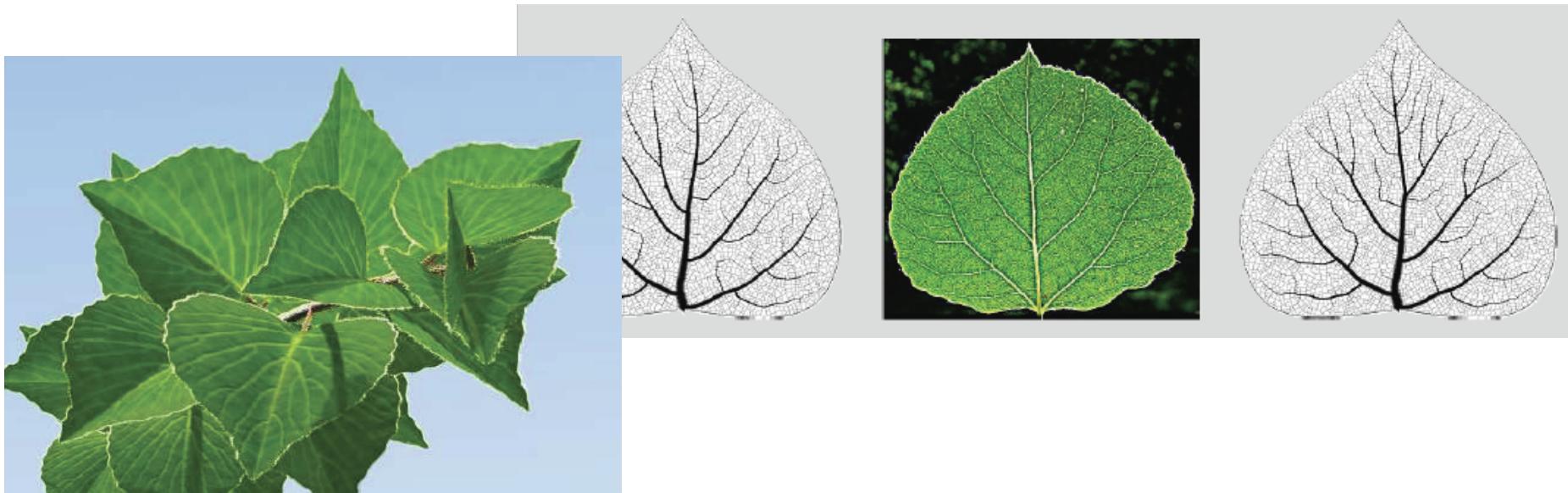
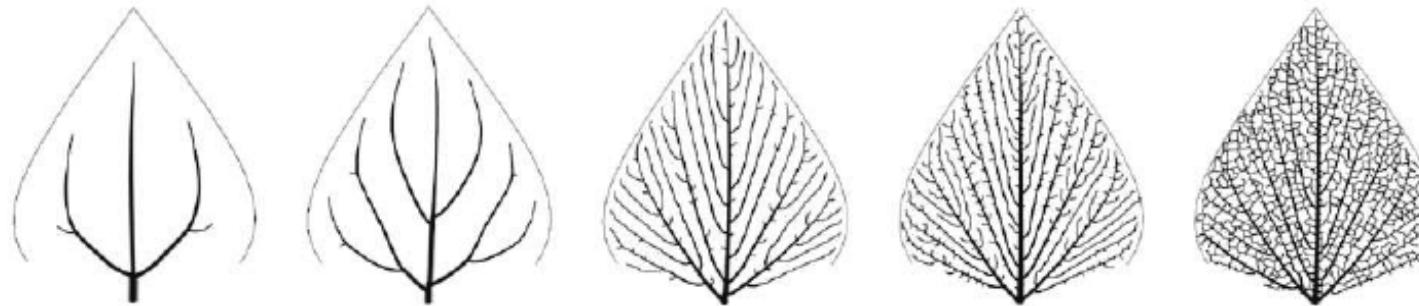


BioCrowds

- ◆ [doi:10.1016/j.cag.2011.12.004](https://doi.org/10.1016/j.cag.2011.12.004) (Bicho et al, CG&A, 2012)
- ◆ A method for crowd simulation based on a biologically motivated space colonization algorithm which was originally introduced to model leaf venation patterns and the branching architecture of trees
- ◆ BioCrowds operates by simulating the competition for space between agents



Results (Runions' method)



A. Runions, M. Fuhrer, B. Lane, P. Federl, A.-G. Rolland-Lagan, P. Prusinkiewicz

Modeling and visualization of leaf venation patterns. ACM Trans Graph, 24 (3) (2005), pp. 702–711

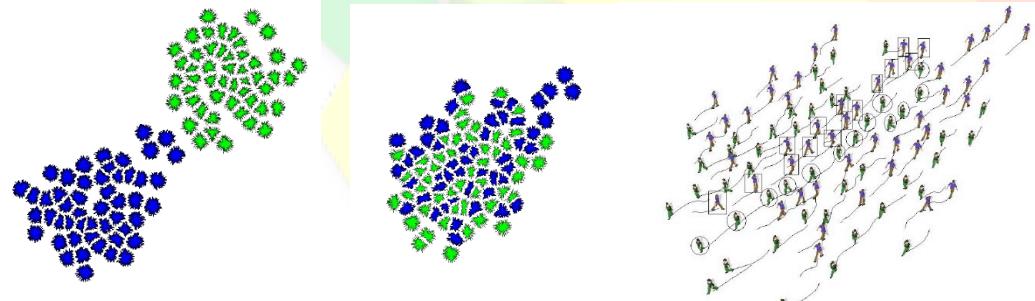
BioCrowds



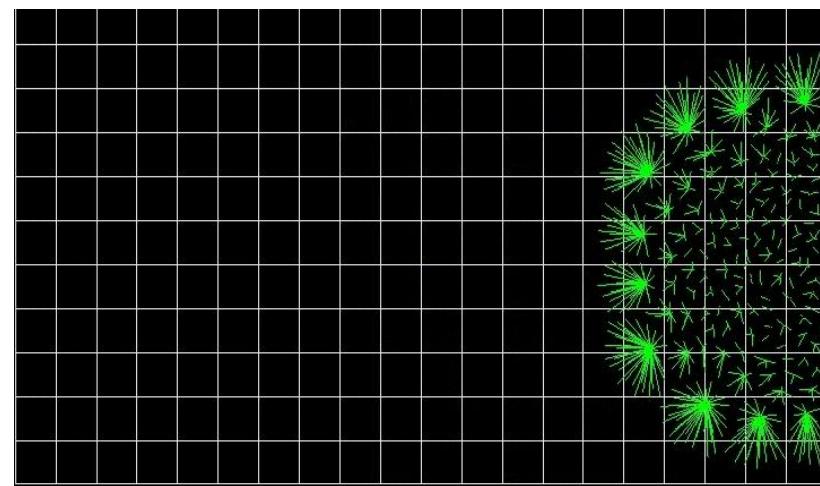
- ◆ Parameters:
 - N Markers (*Dart-throwing*);
 - *Agent i positions $x_i(t)$, goals $g_i(t)$ and desired maximum speed s_{max}^i*
 - Perception field (Circular region) R_i .
- ◆ At each timestep and for a given agent i :
 - $A = \{a_1, a_2, \dots, a_N\}$, markers that are closer to agent i than any other agent
 - $m = \sum_{k=1}^N w_k(a_k - x)$, motion vector
 - $w_k = \frac{f(g-x, a_k-x)}{\sum_{i=1}^N f(g-x, a_i-x)}$, f should prioritize markers that lead the agent directly to its goal
 - $v = s \frac{m}{\|m\|}$, where $s = \min\{\|m\|, s_{max}\}$, v is agent displacement
 - Next agent position is $x(t + 1) = x(t) + v$
- ◆ See <https://www.youtube.com/watch?v=jqfedsvqNg>

BioCrowds: Examples of Emergent Behaviors

- ◆ Lanes formation:



- ◆ Arc formation:



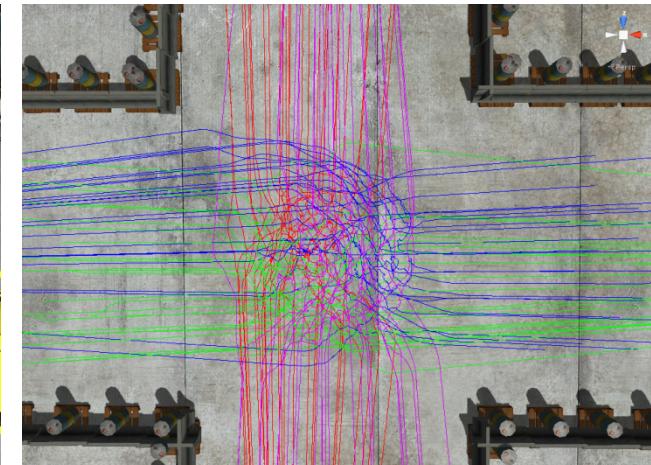
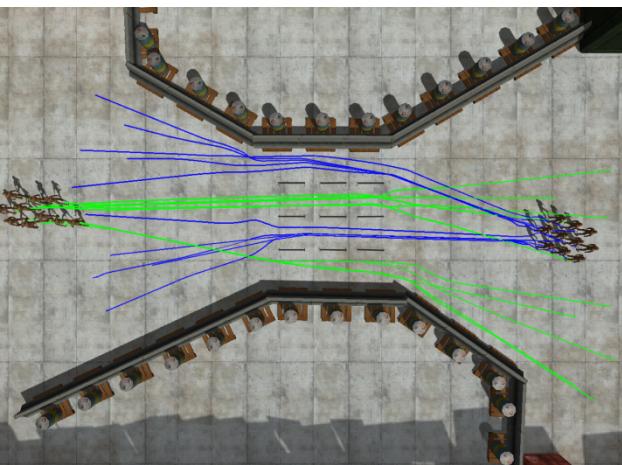
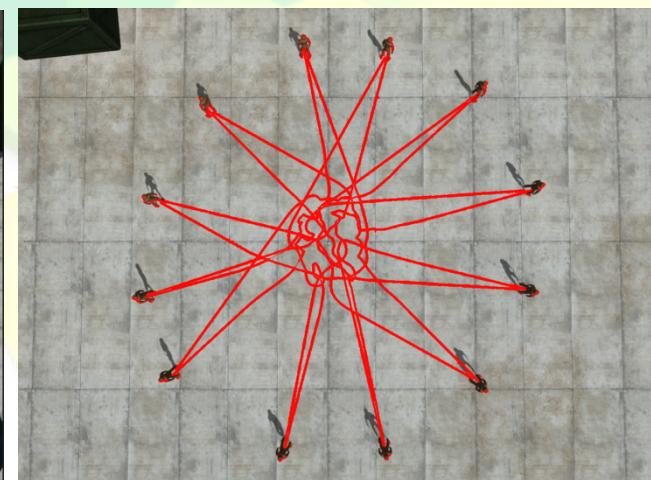
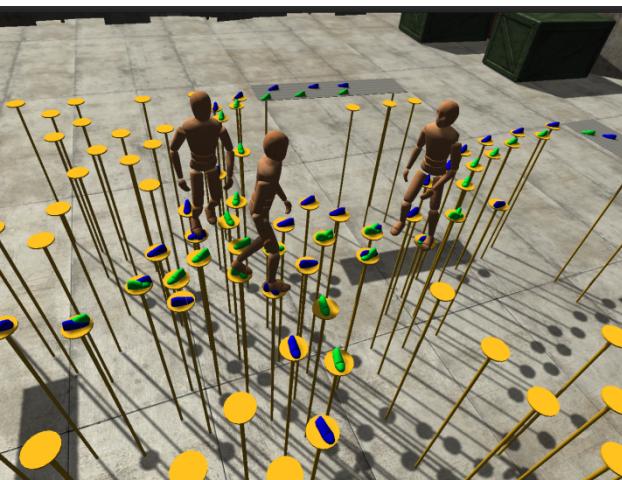
- ◆ Integrate with environment:

<https://www.youtube.com/watch?v=EP4RSGR4Mh0>

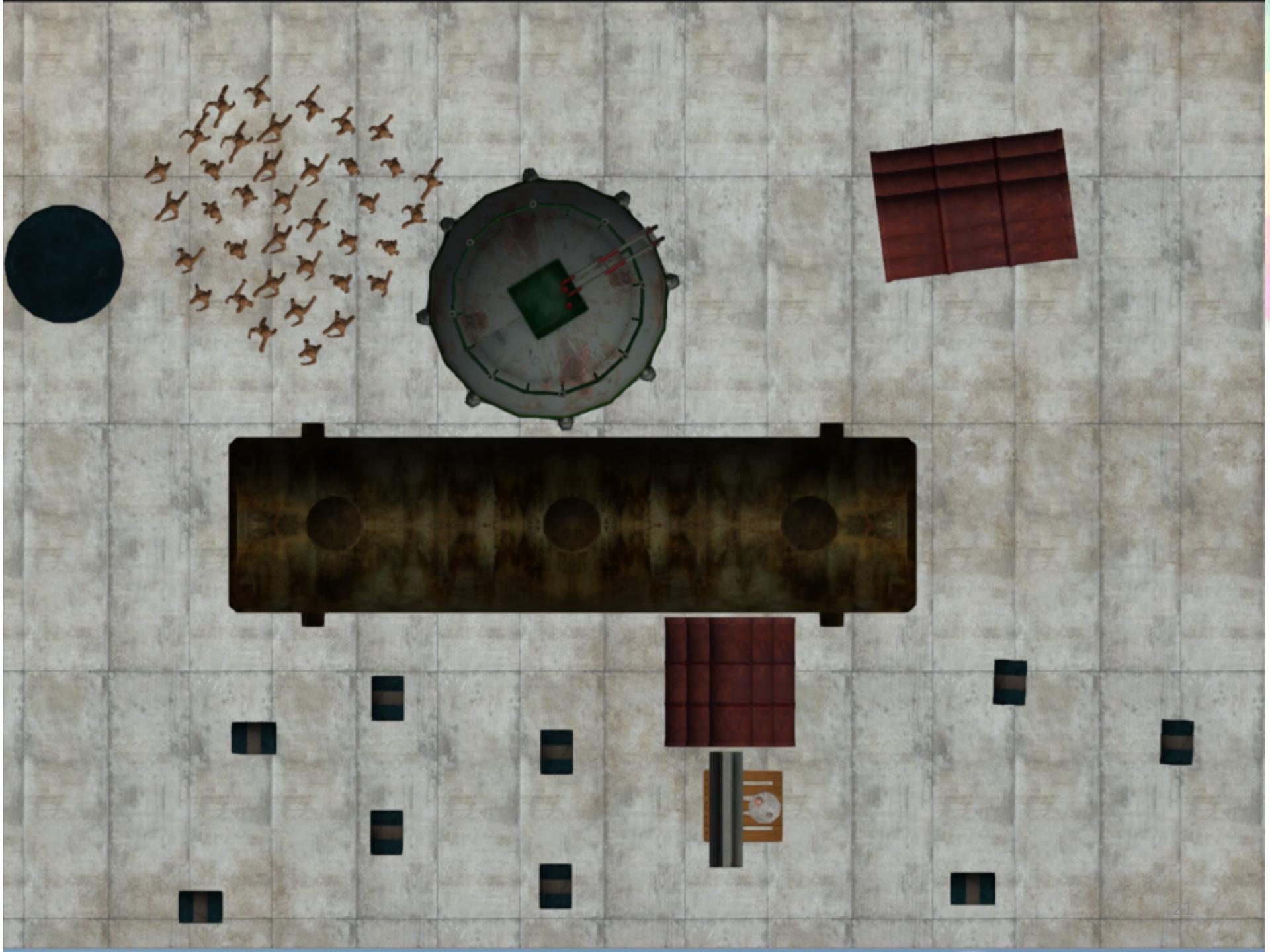
BioCrowds: Problems

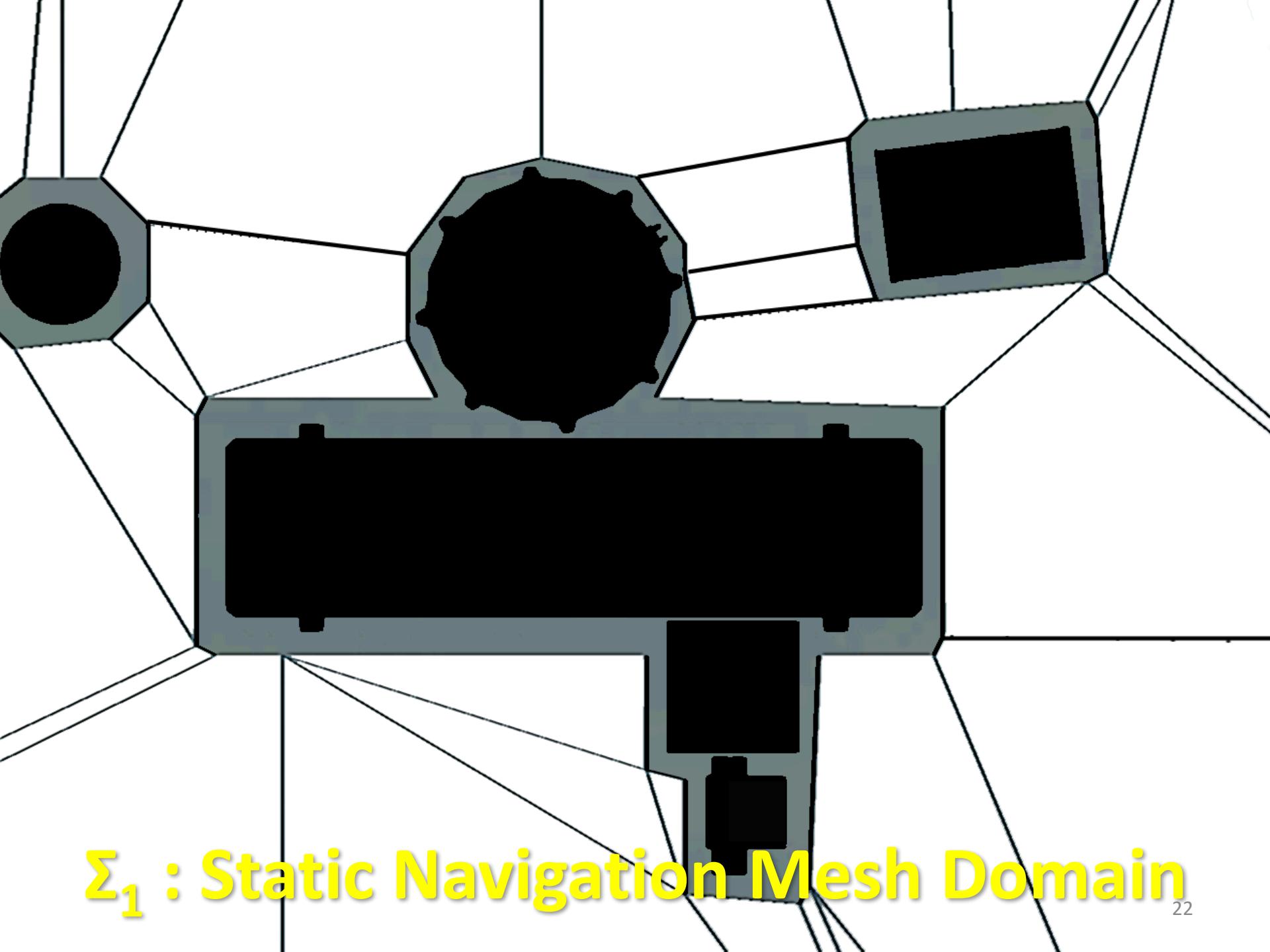
- ◆ An agent does not see any other
- ◆ Local minimum traps (there was no path planning)

Need for high fidelity navigation in complex, dynamic virtual environments

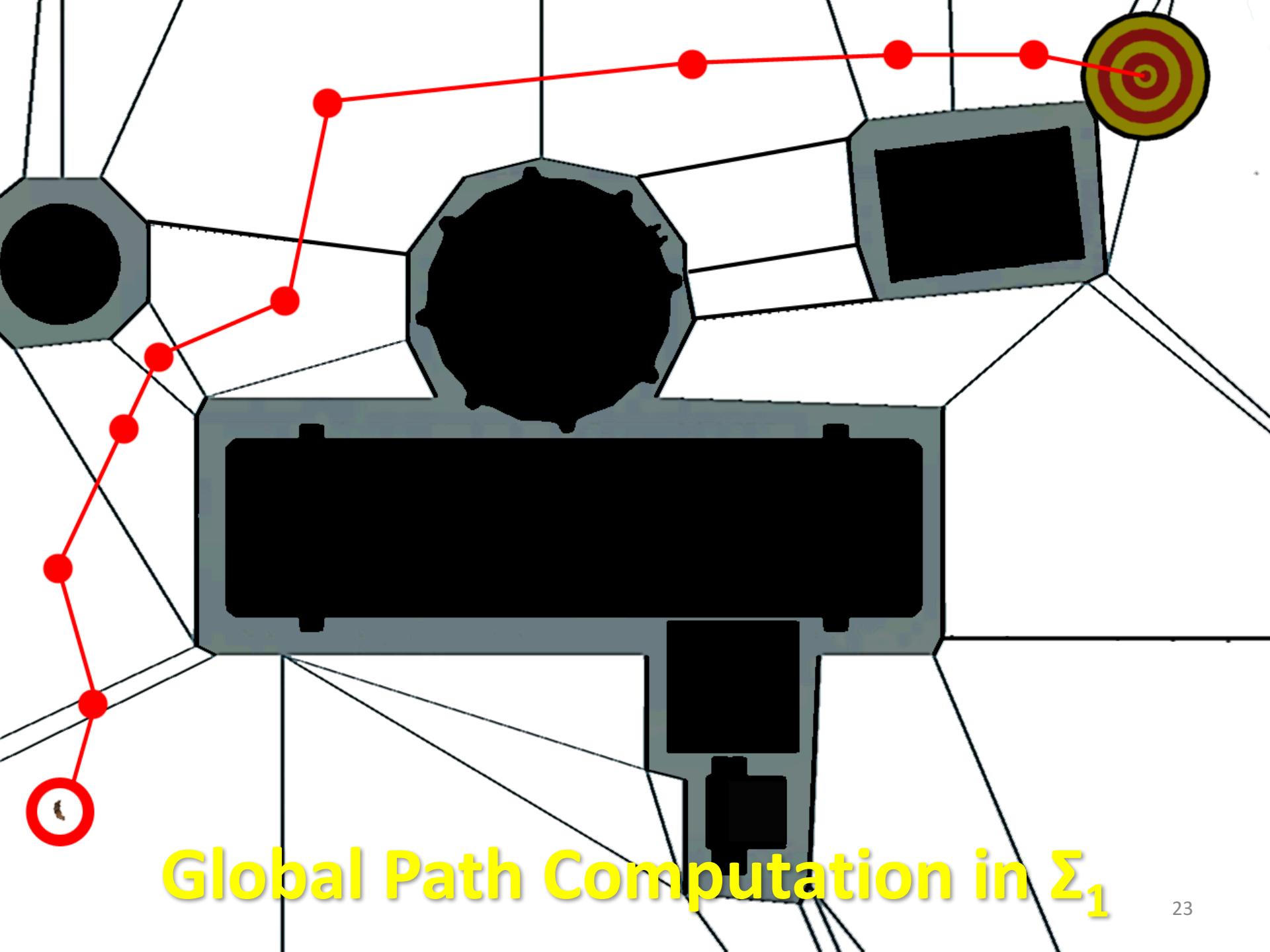


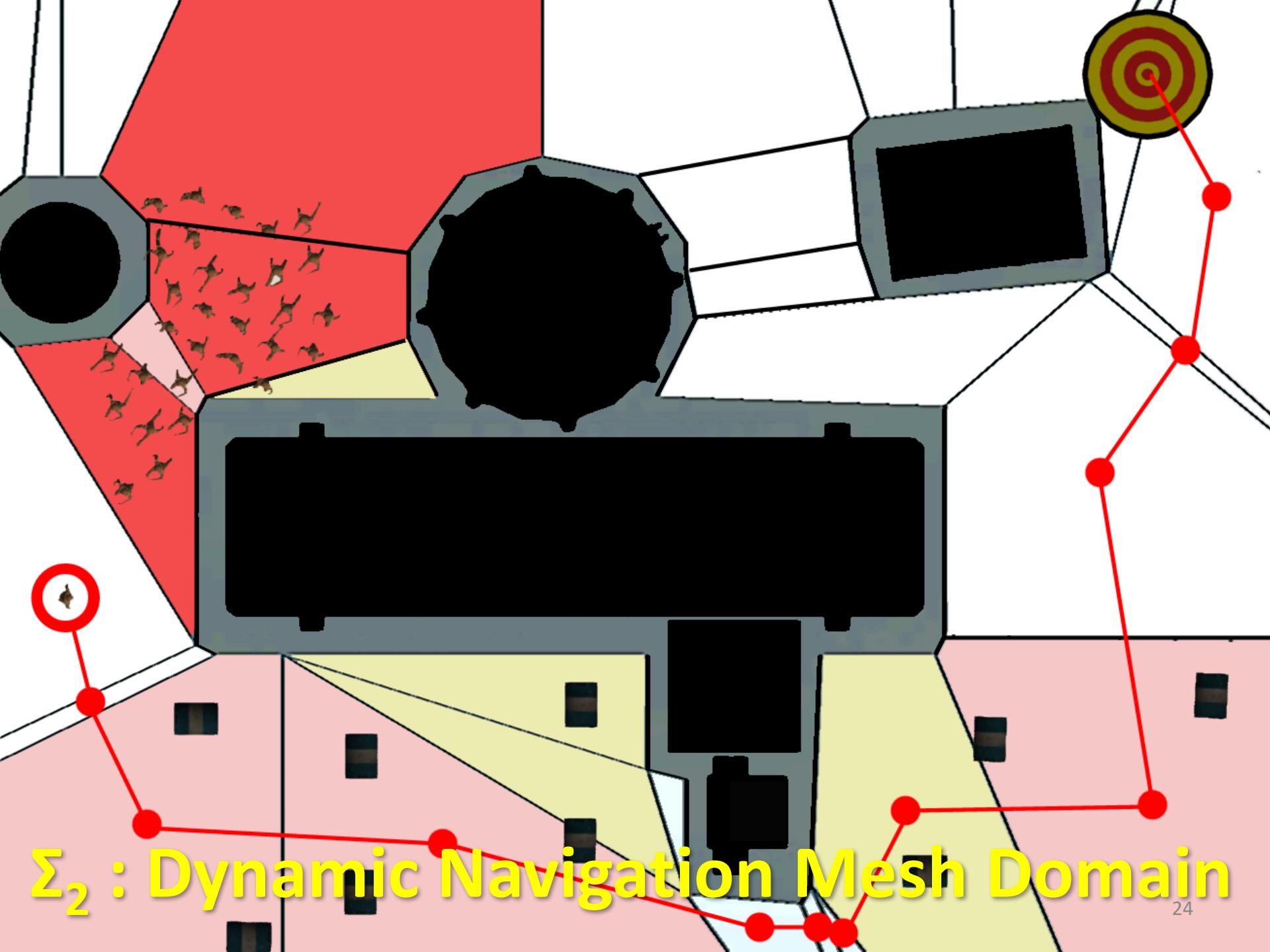
Real-time system that can handle many characters, without compromising control fidelity.



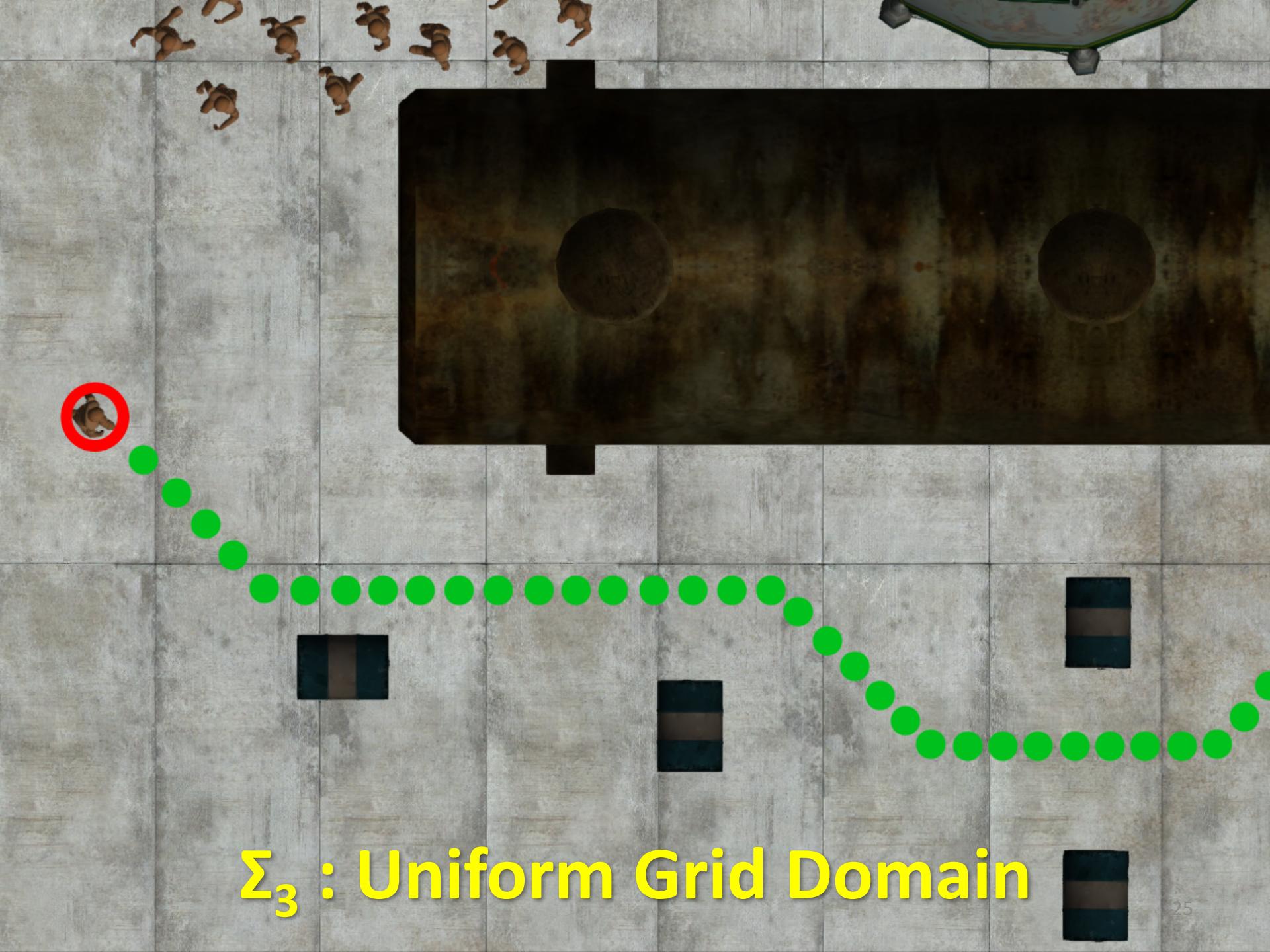


Σ_1 : Static Navigation Mesh Domain

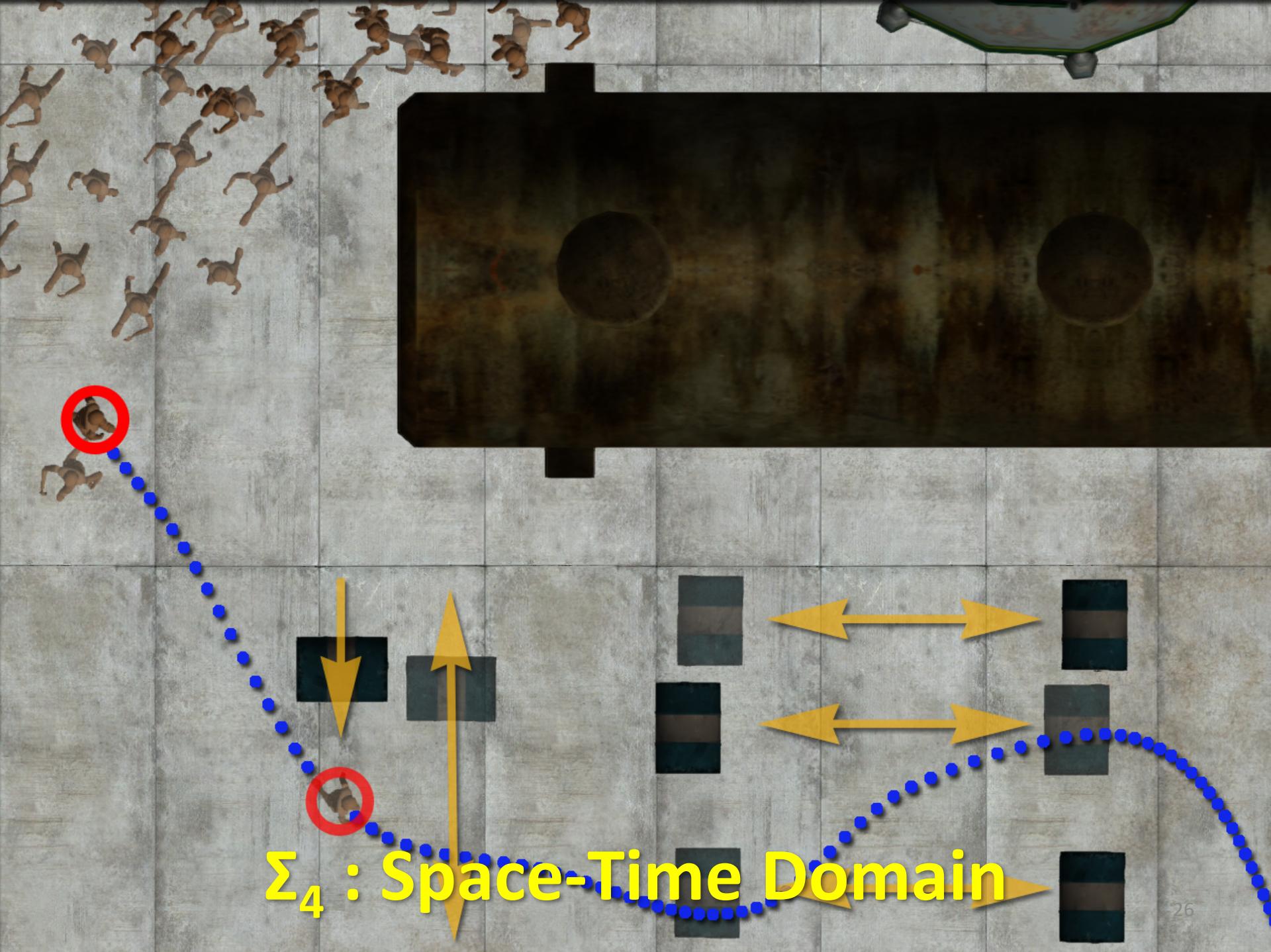




$\Sigma_2 : \text{Dynamic Navigation Mesh Domain}$



Σ_3 : Uniform Grid Domain



Footstep Level Navigation Planning



Multi-Domain Planning (example)



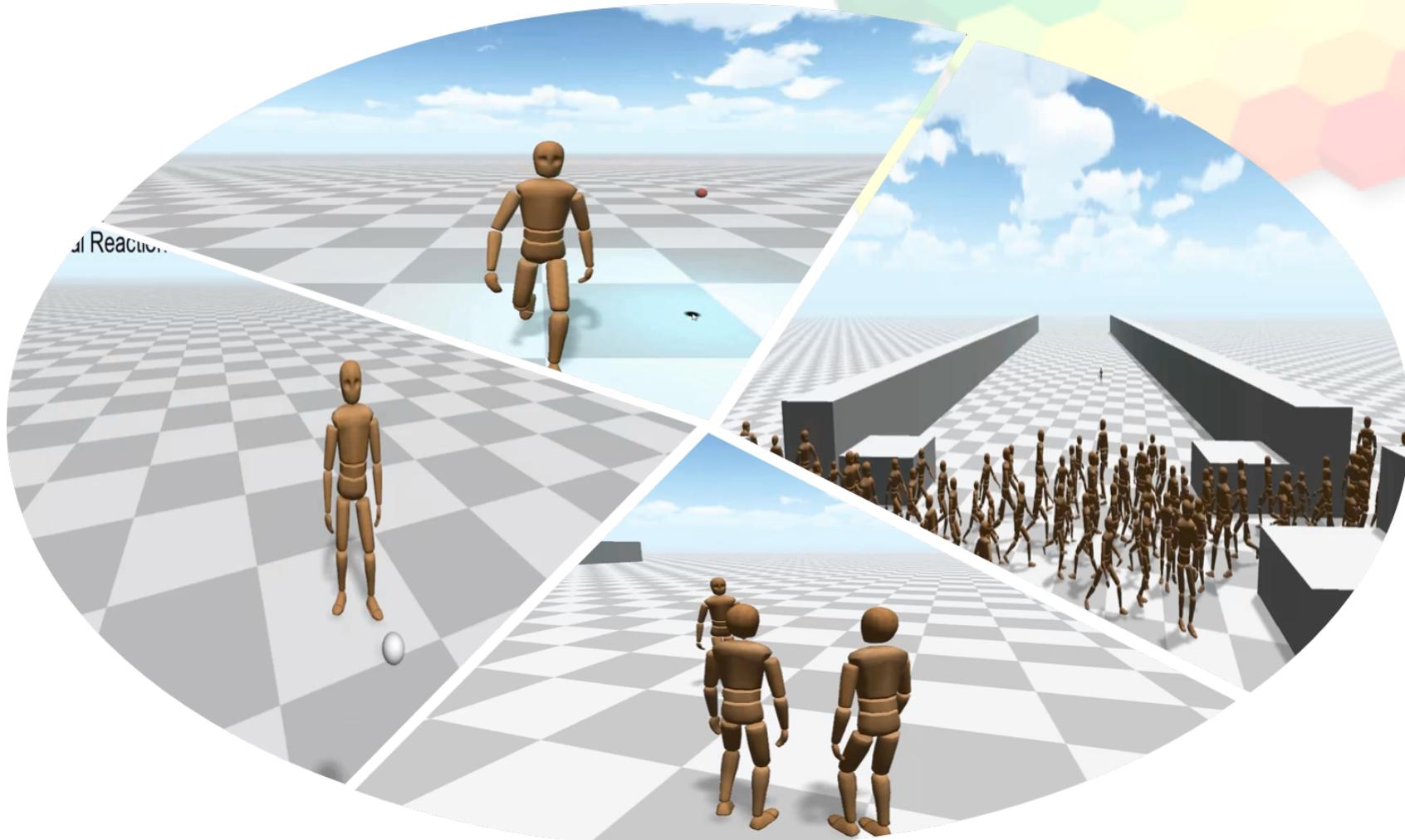
Path Constraints and Relations

Constraint-Aware Navigation
in Dynamic Environments

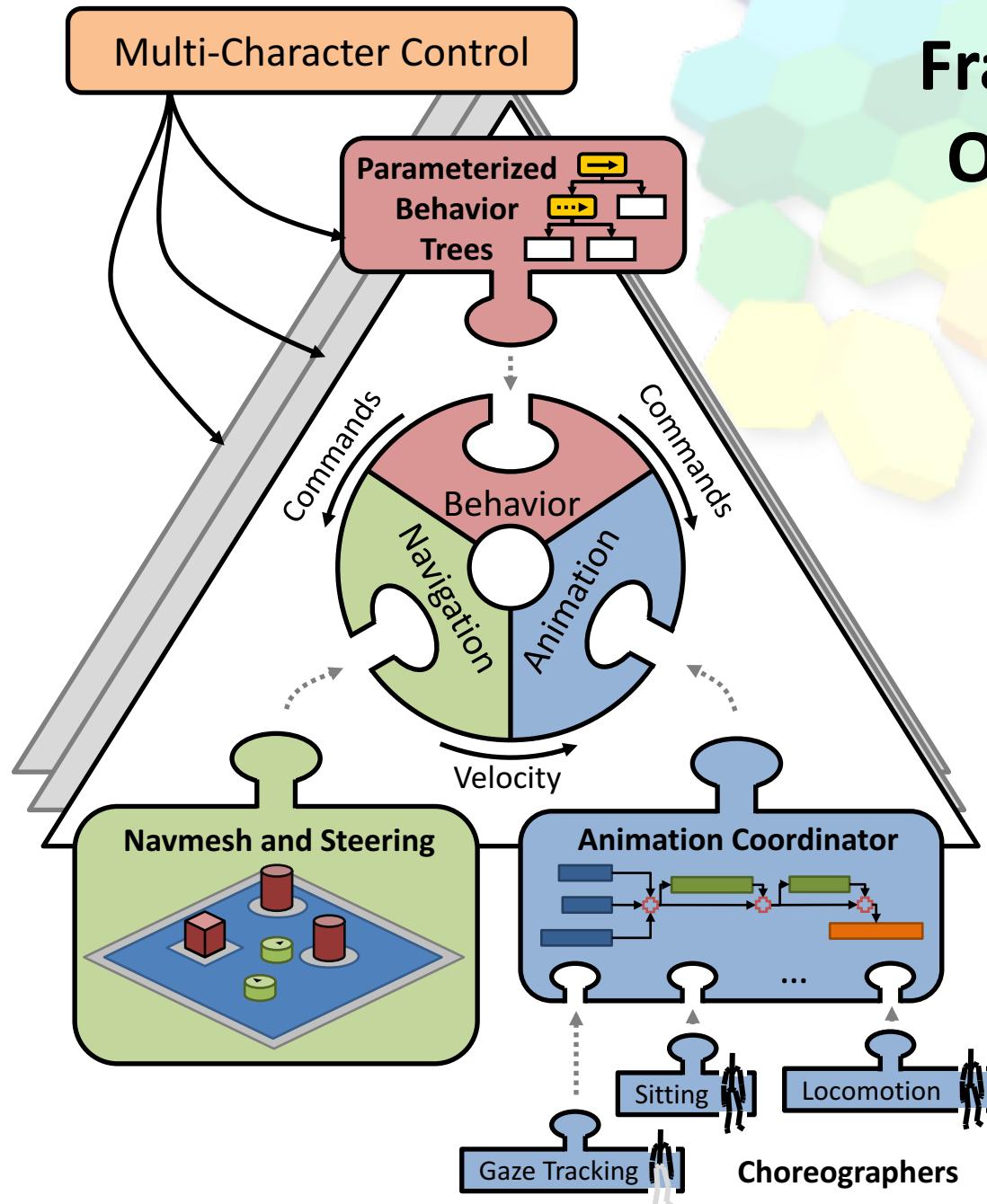
Behavior

ADAPT: The Agent Development and Prototyping Testbed

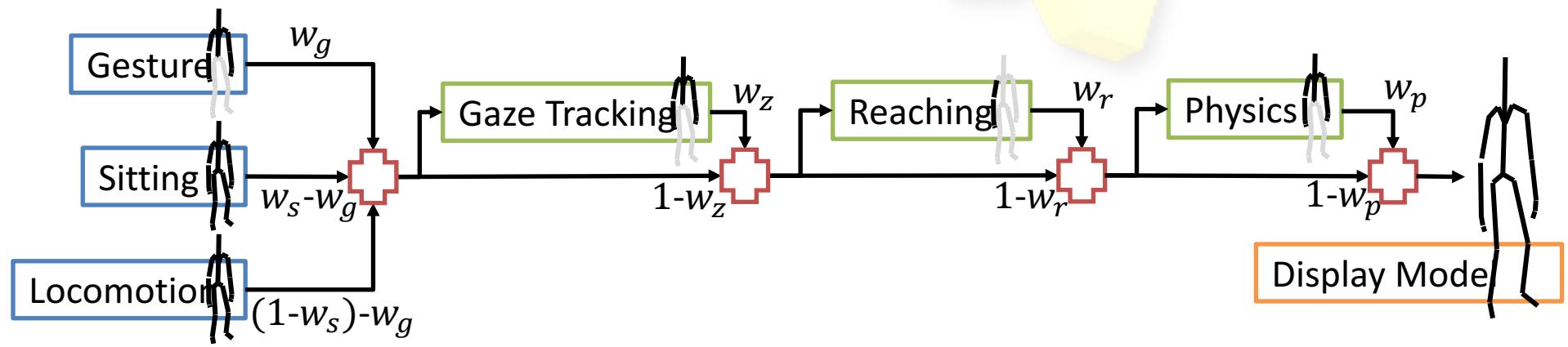
*Mubbasir Kapadia
Disney Research, Zurich*



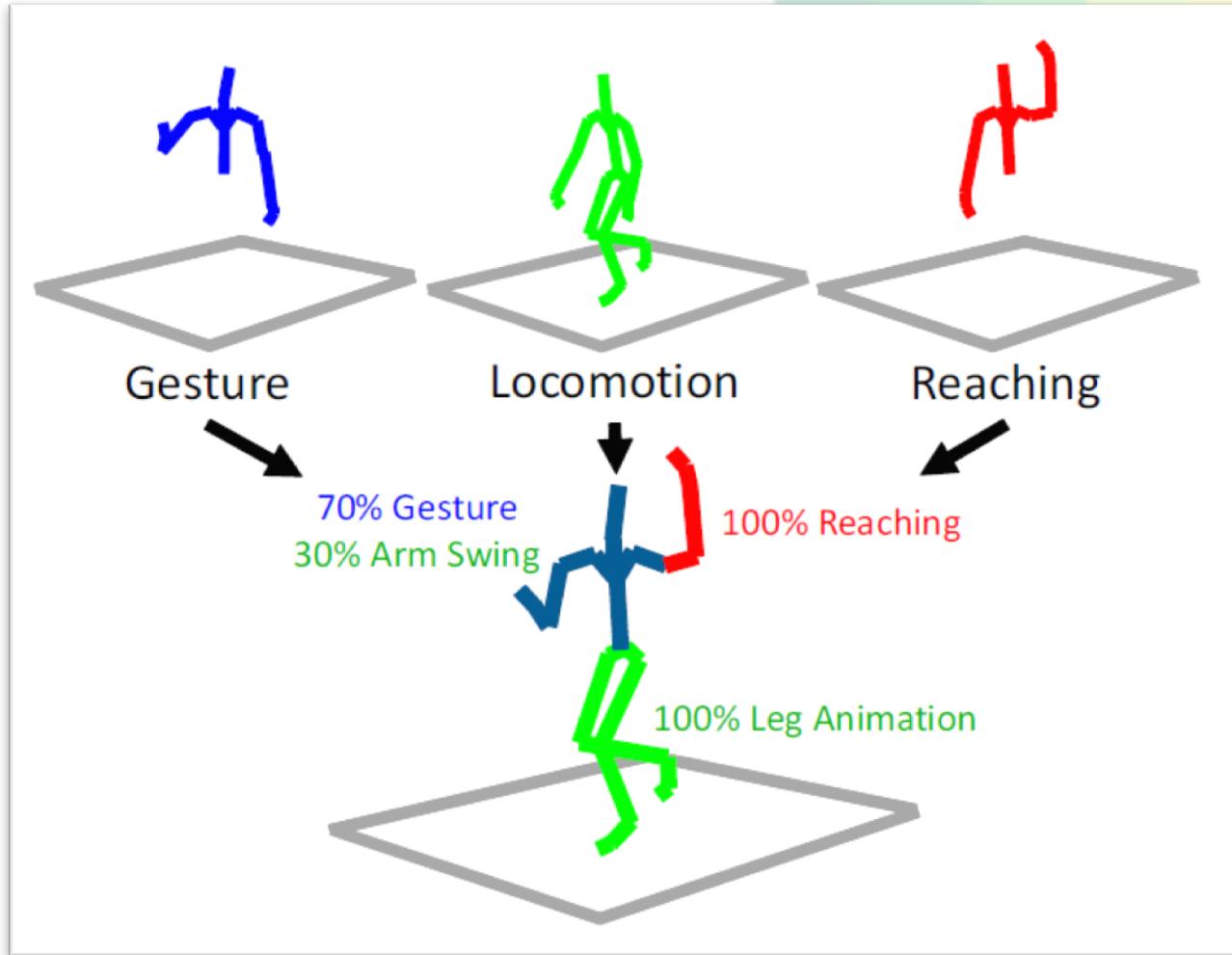
Framework Overview

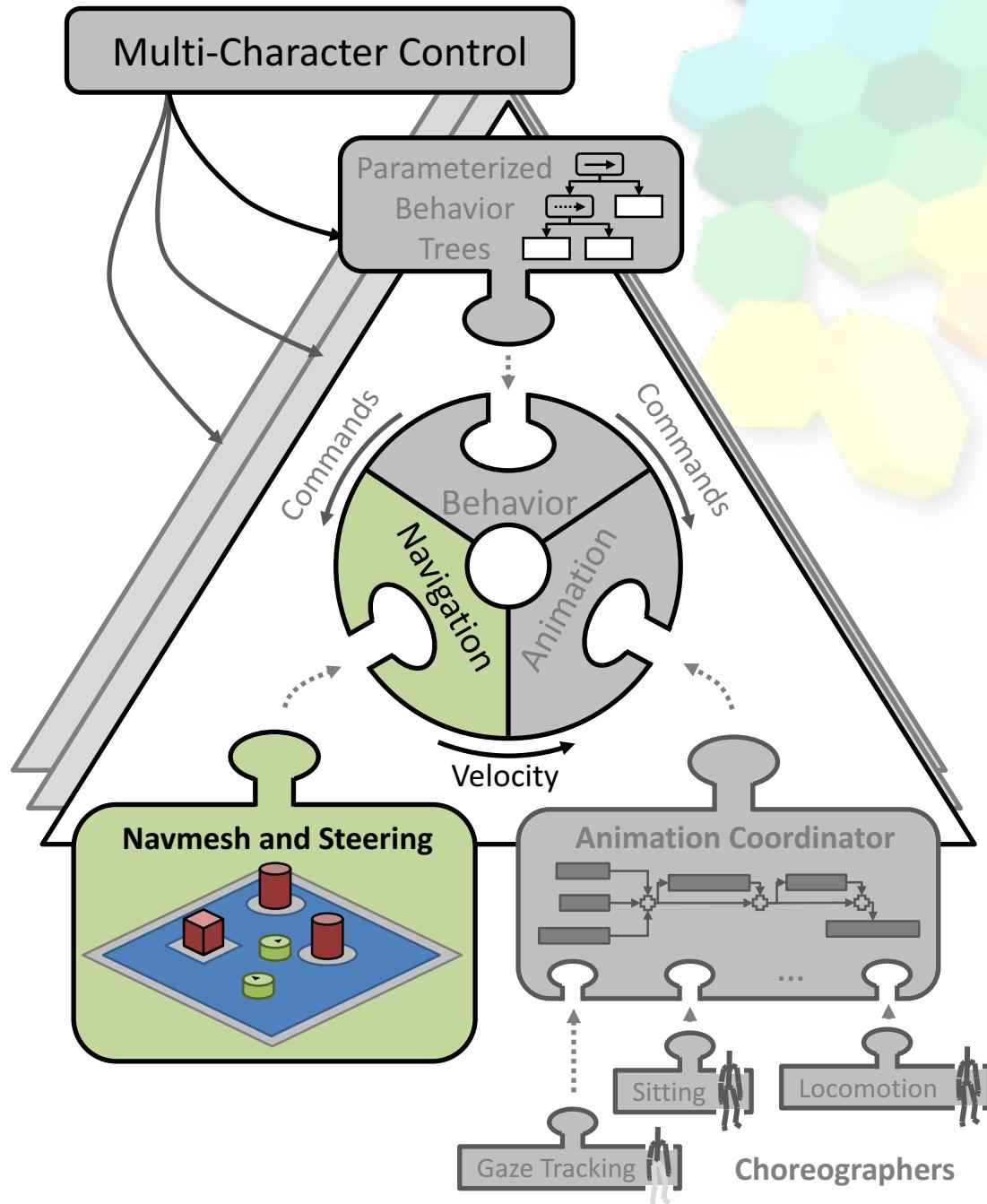


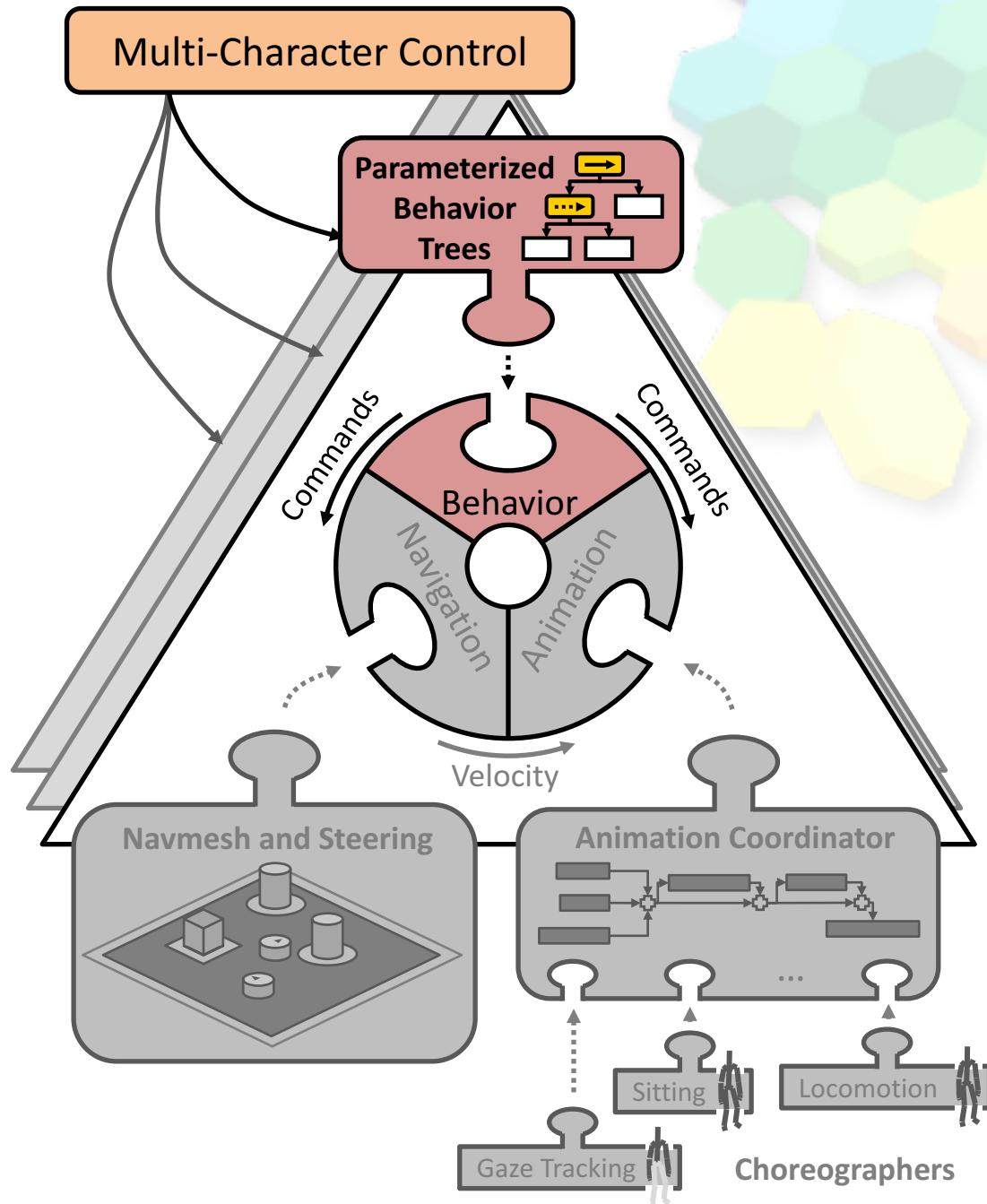
Choreographer Data-flow Diagram



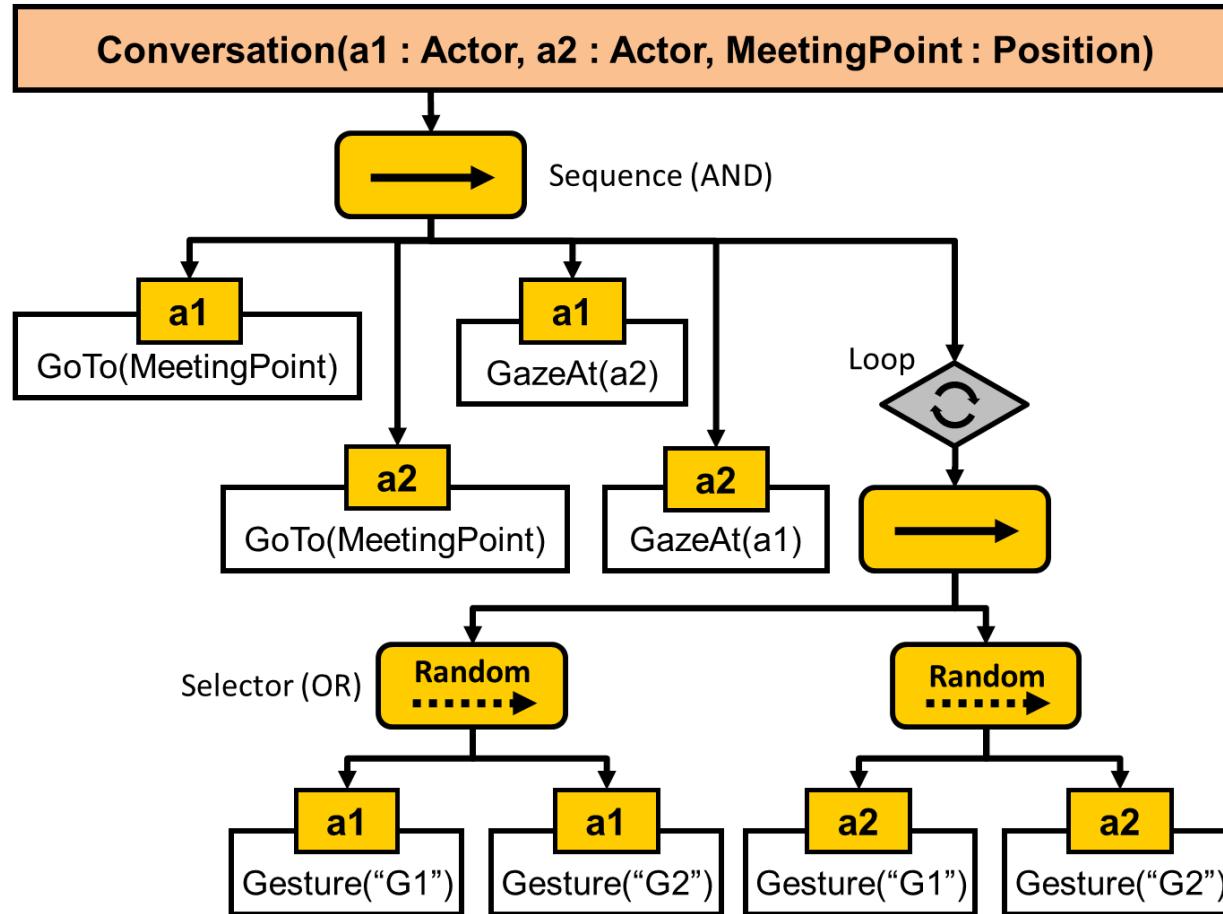
Shadow Blending





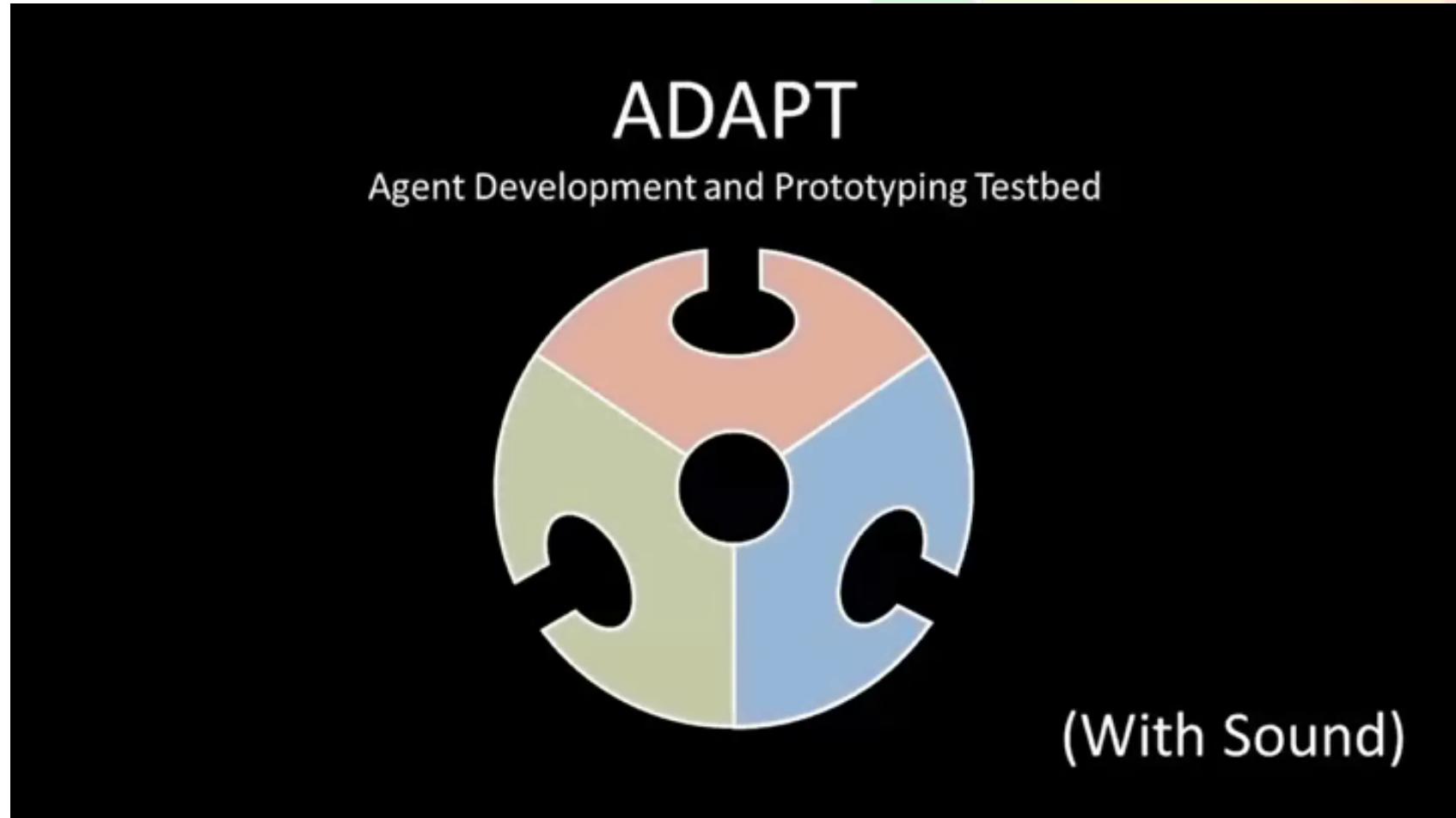


Parameterizing Behavior Trees



Alexander Shoulson, Francisco Garcia, M. Jones, Norman I. Badler
Fourth International Conference on Motion in Games (MIG), 2011

ADAPT Animations



Behavior

Functional Crowds and Psychological models

Jan Allbeck

George Mason University



POPULATIONS WITH PURPOSE

- Allbeck, J.M. CAROSA: A Tool for Authoring NPCs. In Proceedings of Motion in Games. Springer, 2010, pages 182-193.
- Allbeck, J.M. and Badler, N.I. Simulating Human Activities for Synthetic Inputs to Sensor Systems. In Distributed Video Sensor Networks. B. Bhanu, C.V. Ravishankar, A.K. Roy-Chowdhury, H. Aghajan, and D.Terzopoulos (Eds). Springer. 2011, pages 193-206.
- Li, W. and Allbeck, J.M. Populations with Purpose. In Proceedings of Motion in Games. Springer, pages 133-144, 2011.

Goal

- ◆ *Functional crowds* that are feasibly created and modified by non-programmers.
 - Heterogeneous: plausible, contextual variations in behaviors
 - Meaningful interactions with environment and other agents
 - *Assumes underlying crowd simulator*



Autonomous Pedestrians

(Shao and Terzopoulos, Graphical Models, 2007)



CAROSA (Allbeck)

- ◆ Use a Parameterized Action Representation (PAR) to add semantics of actions, objects, and agents.
- ◆ Construct a resource manager to allocate object participants.
- ◆ Define four types of actions (scheduled, reactive, need-based and aleatoric) as mechanisms for behavior selection.
- ◆ Create mechanisms for suspending and preempting actions to promote user interaction.
- ◆ Use roles to specify behaviors at a higher level.
- ◆ Create richer *natural* behaviors through the addition of other social psychological models (e.g. use need levels to establish action priorities)

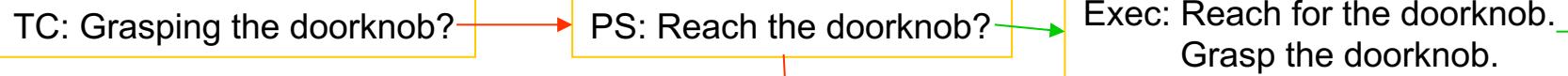
PAR Actions

- ◆ core semantics: motion, force, state-change, paths
- ◆ participants: agent, objects
- ◆ purpose: state to achieve, action to generate, etc.
- ◆ manner: how to perform action (e.g. “carefully”)
- ◆ type: aleatoric, reactive, need-based
- ◆ duration: timing, iteration, or extent; e.g., “for 6 seconds”, “between 5 and 6 times”
- ◆ sub-steps: actions to perform to accomplish action (includes parallel constructs)
- ◆ next-step: next action to be performed
- ◆ super-step: parent action
- ◆ conditions: *prior, post*

Action: Open the door



Action: Grasp the doorknob



Action: Walk to the doorknob



Action: Stand up



END

Yes
No

Resource Manager

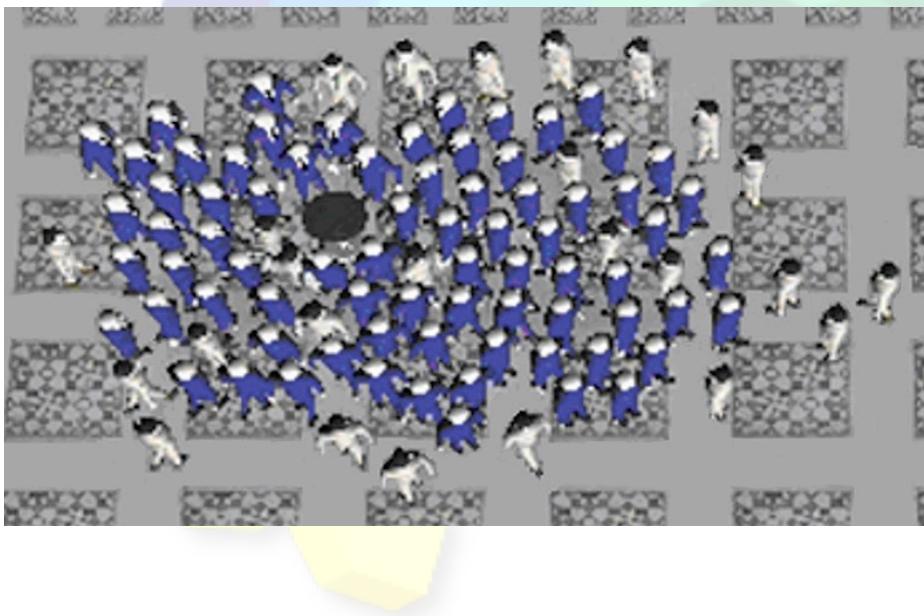
- ◆ Set up resources as a part of environment loading
- ◆ Associate resources with spaces (e.g. rooms)
- ◆ Allocate object participants for actions
- ◆ Take into account: object types, object states and properties, possessions, habits, preferences, ease of access, knowledge of environment
- ◆ Allocate automatically as a part of assigning actions to agents (or groups of agents) and process them for execution
- ◆ Automatically update the state of objects and de-allocate resources as post assertions of actions
- ◆ Failure is captured and conveyed to the agent process

Action Types

- ◆ Mechanisms for behavior selection
- ◆ Scheduled: arise from specified roles for individuals or groups
- ◆ Reactive: are triggered by contextual events or environmental constraints
- ◆ Need-based: arise from explicit goals and priorities
- ◆ Aleatoric: are random but structured by choices, distributions, or parametric variations

Summary

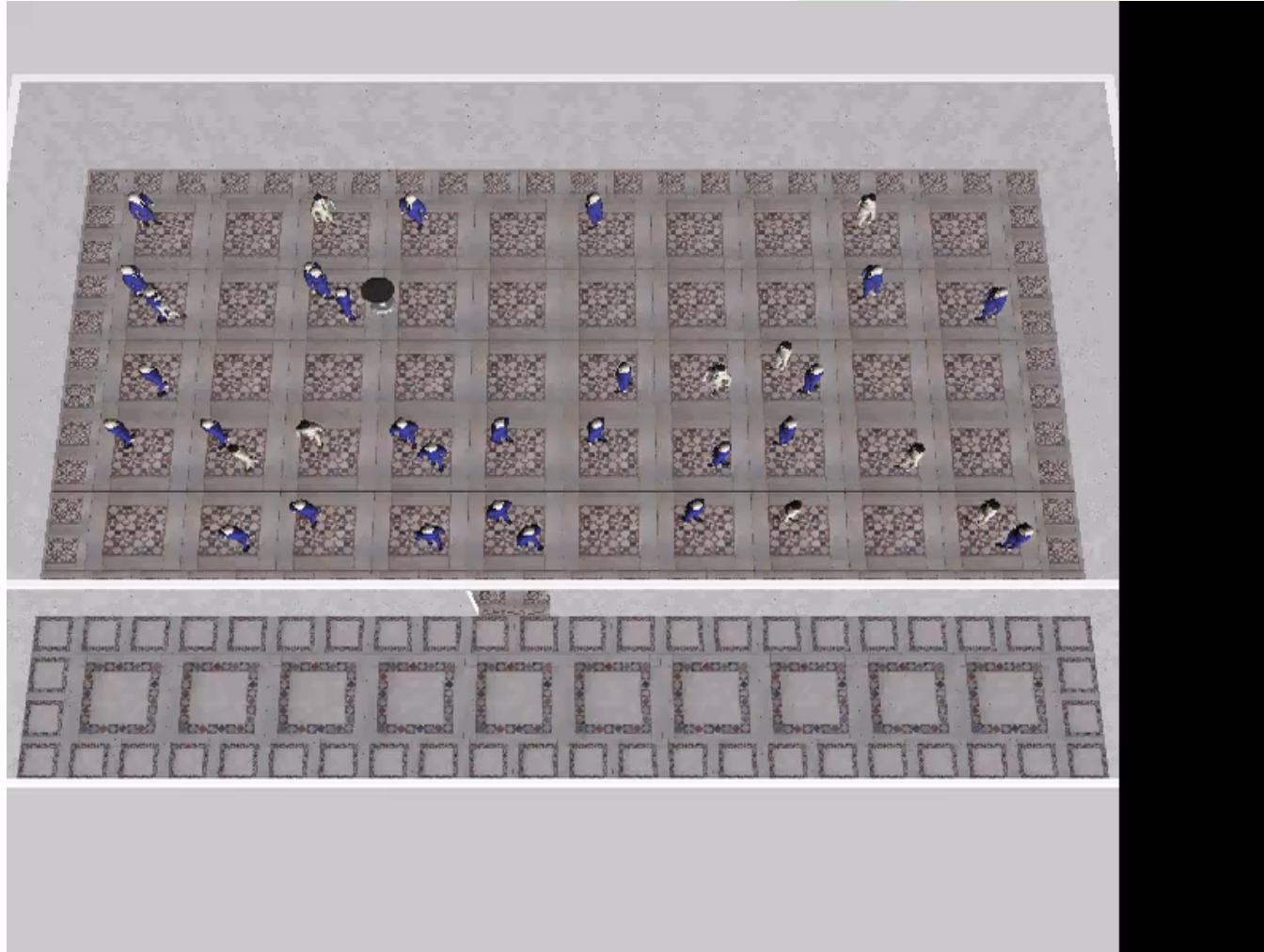
- ◆ *Functional crowds* are heterogeneous populations in which agents have meaningful interactions with their environment.
- ◆ Representing the semantics of actions and objects facilitates authoring (control vs. autonomy).
- ◆ Simple behavior selection mechanisms can lead to interesting, complex scenes.
- ◆ Defining roles, along with personality types and needs, can promote more consistent, compelling long term behaviors



HOW THE OCEAN PERSONALITY MODEL AFFECTS THE PERCEPTION OF CROWDS

- Durupinar, J. Allbeck, N. Pelechano, and N. Badler. Creating Crowd Variation with the OCEAN Personality Model. Proceedings of Autonomous Agents and Multi-Agents Systems 2008. pp 1217-1220.
- Durupinar, F., Pelechano, N., Allbeck, J., Gudukbay, U., and Badler, N. The Impact of the OCEAN Personality Model on the Perception of Crowds. IEEE Computer Graphics and Applications. vol. 31, no. 3, pp. 22-31, May/June, 2011.

Personalities



Five Factor (OCEAN) Model of Personality

- ◆ Personality is a pattern of a person's behavioral, temperamental, emotional, and mental traits.
- ◆ **Openness (closed)**: imaginative and creative
- ◆ **Conscientious (not conscientious)**: organized and careful
- ◆ **Extroverted (introverted)**: outgoing and sociable
- ◆ **Agreeable (disagreeable)**: friendliness and generosity
- ◆ **Neurotic (stable)**: emotional instability and negative emotions

Representation

- ◆ An agent personality: $\pi = \langle \Psi_O, \Psi_C, \Psi_E, \Psi_A, \Psi_N \rangle$
- ◆ Distribution: $\Psi_i = N(\mu_i, \sigma_i^2)$, for $i \in \{O, C, E, A, N\}$ where $\mu \in [0, 1]$, $\sigma \in [-0.1, 0.1]$
- ◆ Behavior:
 - $\beta = (\beta_1, \beta_2, \dots, \beta_n)$
 - $\beta_j = f(\pi)$, for $j = 1, \dots, n$

Parameter Mapping

Agent behavior	Personality factor	OCEAN
Leadership	Assertive, social, unsocial, calm, fearful	E, N
Trained or untrained	Informed, ignorant	O
Communication	Social, unsocial	E
Panic	Oversensitive, fearful, calm, orderly, predictable	N, C+
Impatience	Rude, assertive, patient, stubborn, tolerant, orderly	E+, C, A
Pushing	Rude, kind, harsh, assertive, shy	A, E
Right preference	Cooperative, predictable, negative, contrary, changeable	A, C
Avoidance or personal space	Social, distant	E
Waiting radius	Tolerant, patient, negative	A
Waiting timer	Kind, patient, negative	A
Exploring environment	Curious, narrow	O
Walking speed	Energetic, lethargic, vigorless	E
Gesturing	Social, unsocial, shy, energetic, lethargic	E

Panic

- ◆ Linked to stability and conscientiousness

$$\beta_i^{\text{Panic}} = \omega_{\text{NP}} \Psi_i^N + \omega_{\text{CP}} f(\Psi_i^C)$$

$$f(\Psi_i^C) = \begin{cases} -2\Psi_i^C + 2 & \text{if } \Psi_i^C \geq 0, \\ 0 & \text{otherwise} \end{cases}$$

where $\beta_i^{\text{Panic}} \propto N$, $\beta_i^{\text{Panic}} \propto^{-1} C+$, and $\beta_i^{\text{Panic}} \in [0, 1]$

Pushing

- ◆ Disagreeable and extroverted agents tend to push others

$$P_i(\text{Pushing}) = \omega_{EP} \Psi_i^E + \omega_{AP} (1 - \Psi_i^A)$$

$$\beta_i^{\text{Pushing}} = \begin{cases} 1 & \text{if } P_i(\text{Pushing}) \geq 0.5 \\ 0 & \text{otherwise} \end{cases}$$

where $P_i(\text{Pushing}) \propto E$, $P_i(\text{Pushing}) \propto^{-1} A$, and

$$\beta_i^{\text{Pushing}} \in \{0, 1\}$$

Waiting Radius

- ◆ Influenced by kindness and consideration
- ◆ Linked to agreeableness

$$\beta_i^{\text{WaitingRadius}} = \begin{cases} 0.25 & \text{if } \Psi_i^A \in \left[\frac{0}{3}, \frac{1}{3} \right) \\ 0.45 & \text{if } \Psi_i^A \in \left[\frac{1}{3}, \frac{2}{3} \right] \\ 0.65 & \text{if } \Psi_i^A \in \left(\frac{2}{3}, \frac{3}{3} \right] \end{cases}$$

where $\beta_i^{\text{WaitingRadius}} \propto A$ and $\beta_i^{\text{WaitingRadius}} \in \{0.25, 0.45, 0.65\}$

Evaluation

- ◆ 15 videos
 - Various OCEAN model settings
 - Evacuation drills, cocktail parties, museum galleries
- ◆ Map HiDAC parameters to OCEAN factors using trait-descriptive adjectives
- ◆ Determine correspondence between our mapping and users' perception of traits
- ◆ 70 participants (21 female and 49 males, ages 18-30)
- ◆ Shown videos and complete questionnaire

Analysis

