Crowds in Hitman: Absolution

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Highlevel goals

- Quality over quantity
  - Around 1200 agents per crowd, 500 on-screen
  - Player should not distinguish between crowd & npc actors

- Ambient crowd behaviors
  - Mill around
  - Be aware of points of interest & react to player actions
  - Level designer has partial control of placement & movement flows

- Panic crowd behaviors
  - Evacuate the crowd area
  - Help enhance the action experience of the game
  - Never get in the way of the player during action
Crowds in general

- Global knowledge/solution vs. Agent based
  - Ex. Global: Continuum based crowds with dynamically updated potential fields
  - Ex. Agent: No goals, very local behavior, can appear "erratic"

- Crowds in a game
  - The "fun factor" is the most important thing
  - Perfect simulation (no intersections or stopping up) becomes secondary
  - Must be very dynamic and react to player actions
  - Level designers must have quite a lot of manual control
  - Each agent must visually be of an acceptable quality, even when viewed up close

- My opinion
  - The best approach is that of a traditional, but lightweight, AI system
Agenda for today

● What to expect from this hour...
● Cover the main components of our crowd system
  ● Framework: The cell map, agent model, tools
  ● AI: Steering & behavior selection
  ● Animation
  ● Believability: integration with core gameplay features
● Performance
The cell map

- We could just add X agents to the world, but:
  - We need very fast navigation mesh queries (can we walk at a certain position??)
  - We need very fast checks for walls & other static obstacles
  - We need very fast neighborhood queries
- We overlay a regular grid on top of the nav mesh
  - This means that the crowd area is only 2.5D (no overlaps in height)
  - Memory usage scales with area of a rectangle, even if walkable region is sparse
  - Each cell stores
    - Walkable/unwalkable flag
    - Current agents in cell (stored as an intrusive singly linked list)
  - Can also annotate the map with additional info, as needed for gameplay
The cell map

- Cell annotations
  - Exclusion zones
  - Panic only cells
  - Ambient flow vectors
  - Teleporters
  - Exit zones
Agent model

- Craig W. Reynolds
- Agent "particles"
  - Position
  - Radius
  - Forward vector
  - Speed
  - Steering input
Tools: Initial agent placement

- Agents are distributed onto the cell map as:
  - Manually placed individuals
  - Groups of agents, manually placed and configured
  - Randomly placed agents
- Manually placed individuals
  - Originally a debugging tool, but ended up being used quite a lot by level designers
- Groups
  - This is what designers really wanted!
  - Position and shape: spherical or square
  - Agent count
  - Optionally: A list of POIs.
  - Optionally: A list of idle animation overrides
The crowd framework
Crowd AI

- Based on a state machine
  - Steering states: Idle, “pending walk”, walk
  - Behavior/gameplay states: Alert, dead, possessed, prone, scared etc.

- State specific memory
  - Each state can define its own “memory class”, which stores arbitrary state-dependent AI memory data
  - Placement new’ed into fixed-size memory block on each agent
  - Wiped & initialized when entering state

- Every frame the agent “Thinks”
  - Steps the AI, using current state and current state memory
  - Ask current state if a state change is wanted
  - Check external factors to see if a state change is needed (more later...)
  - In some cases, change state randomly
Steering: Pending walk

- **Used when**
  - Agent is standing still, but wants to be moving.

- **Purpose**
  - Find the best valid direction and time to start moving
  - Since agent is usually in a crowded place this requires some AI logic

- **Sub-phases**
  - "Search for direction"
    - Send out probes to check for wall collisions and other obstructing agents
    - Probe direction is changed every frame, favoring directions in front of agent
  - "Wait for clear"
    - Wait until agent can start moving
    - Communicate a wanted state change to the agent (into walk state)
Steering: Walk

- Used when
  - Milling about
- Purpose
  - Move agent around, avoiding collisions with walls and other agents
- Algorithm
  - Find preferred direction
    - Check for walls, and steer to avoid collision
    - Check for avoid zones and ambient flows
    - Apply wander behavior (Reynolds)
  - Sample neighborhood for dynamic obstacles, select worst threat (Reynolds)
  - Do "unaligned collision avoidance" to get actual steering direction (Reynolds+)
  - Either accept the steering, or communicate a wish to stop moving
Steering: Key learnings

- This turned out to be hard in dense environments!
  - Lots of "magic numbers" to tweak
  - Especially hard when having multiple movement speeds

- Using speed for steering
  - Turned out to be critical!
  - First decide on an initial preferred and max speed (for example: walk relaxed and walk fast)
  - Each steering component (wall or dynamic avoidance) then reports:
    - New preferred speed
    - New maximum possible speed
  - Decision is based on, for example, distance to wall or whether or not a speed change can resolve a dynamic collision
  - A real human often prefers slightly changing speed over changing direction

- Favor stopping to radically changing direction
Steering: Panic

- Same as “Pending walk” / “Walk”, but tweaked differently
  - Higher speed means different settings for probing for walls, collecting neighbors etc
- Panic steering relies heavily on “panic flows”
  - Each exit in the crowd becomes one separate “flow channel”
  - When cell map is generated, each flow channel is calculated
  - Each cell stores: direction to exit, along shortest path, and total cost to reach that exit
  - Each agent dynamically switches between flow channels to quickly flee the map
- Needs some manual guidance/annotation in narrow spaces
  - Panic flows are based on modified Dijkstra algorithm
  - Shortest path generates choke points around corners
Video: Steering behaviors
Behavior selection

- Navigation AI automatically handles state changes
- More specific AI states are handled differently
  - A data-driven system makes the crowd react to various players actions
  - For example: aiming a gun, shooting, acting suspicious
- A player action spawn up to 3 user-configured zones
  - Radius & angle (spherical or cone)
  - Agent reaction type: (POI, avoid, alert, scare, go prone)
  - Reaction types are listed in "order of importance", and a zone can override less important zones
Behavior zones
Behavior zone pulses

● Zones continuously send “pulses” into the crowd
  ● This way each zone “pushes” itself on the affected agents
● When an agent is hit by a behavior pulse
  ● Is this now the currently most important behavior zone?
    ● Check current agent mood (ambient, alerted, scared, paniced, dead)
    ● Check “inflicted mood” from zone (derived from reaction type)
  ● During “Think”
    ● If mood for current zone is strictly worse than the current agent mood, then we change AI state
● Benefits of system
  ● Level designer configures the behavior on a per-crowd basis
  ● Quick and easy way to handle multiple inputs to the agents
Video: Behavior selection
Animation: First attempt

- What and how?
  - Fit animations on top of simulation
  - Share a number of looping clips between all agents (idle, walk, run etc)
  - At any time: animation state for an agent is two animation IDs and a blend weight

- Why?
  - Concerned about animation performance
  - Simple to implement
Animation: First attempt

- **Pros**
  - Performance was great
  - Navigation logic was stable
  - Agents can move at any speed!

- **Cons**
  - Overall robotic look and feel
  - Foot sliding in transitions: idle -> walk -> idle
  - No turn/banking animations
  - Agent animation looks synchronized
    - So we added multiple loops per animation, started at random times...
  - Tedious and manual approach to controlling animation state from AI code
  - Code involved in adding new animations to the system
  - Hard to avoid animation glitches and blend errors

- **Overall**
  - The approach was valid, but we had higher ambitions than that....
Animation: Second try

- **What and how?**
  - Ambitious goal - 500 agents on screen with no foot sliding plus support for transition and turn/banking animations
  - Based on heavily modified version of "Near-optimal Character Animation with Continuous Control"
    - Annotated motion clips, high-level steering inputs, data driven
  - Agents are now moved by a trajectory channel in the animations, rather than from steering velocity!
  - Each visible agent now needs a uniquely blended animation pose, much like an ordinary NPC

- **Why?**
  - Player gets very close to the individual agents
  - We felt that having a high quality of animation on each individual was needed for achieving a believable crowd experience
  - Avoid the robotic feel
Animation: Second try

- **Pros**
  - Looks much better 😊
  - Completely removed tedious animation management code from the crowd AI
  - Greatly simplified the AI code itself

- **Cons**
  - Took a lot of work to implement and optimize
  - In rare cases a bit more control over the animations can be useful
  - And very importantly: Agents react *much* slower to steering input, which makes it harder to avoid collisions and intersections!

- **Overall conclusion**
  - It was a great success!
  - The approach we used for crowd agents might be how we control real NPCs in future games...
Animation

- Check GDC Vault for: “Animation Driven Locomotion for Smoother Navigation” for further inspiration! (Gabriel Leblanc, Shawn Harris, Bobby Anguelov)
Believability

● **Main challenges:**
  ● Core game mechanics: close combat, human shield etc
  ● Detail animation
  ● Visual variety
Core game mechanics

- No wish to have duplicate implementation
- Possession system
  - On-demand upgrade agent to full NPC AI
  - Allocates small pool of invisible NPCs
  - Simple API allows game programmers to switch between crowd agent and NPC
  - Made it trivial to support advanced gameplay mechanics
Detail animation

- Head IK
- Crowd acts
  - Talk on phone, smoke, sit on bench
  - Uses possession system and existing cut-scene tools
  - Spawns randomly near player
- Upper body acts
  - Lightweight overlay anims: cough, wave etc.
  - Can play while agent walks around
Visual variety

- Unique scaling factor for each agent
  - Small amount: ~5%
  - Softens up horizon
  - Does wonders for perceived diversity of crowd

- Diffuse texture overrides
  - Simply replace the diffuse texture of material
  - Cheap way of having red shirt, yellow shirt etc..
Performance: PS3

- Some numbers: 1200 agents simulated, 500 on-screen
- PPU: 5ms
  - Animation system: ~2ms
  - Crowd AI / steering: ~2ms
  - Framework: ~1ms
- SPU: ~20ms, distributed across multiple SPUs
  - Animation sampling
  - Animation blending
  - Animation selection logic
  - Frustum and occlusion culling
  - Crowd AI sensors (more later)
- GPU: 8ms
  - Listed here as an example, but obviously very dependent on render tech and meshes used
  - In G2: the vertex shader is limiting factor on PS3 due to skinning massive amount of vertices
Performance

- **Scaling?**
  - System has very low general overhead
  - Scales nearly linearly with number of agents in crowd
  - Culled/on-screen ratio also affects performance, due to animation cost

- **Memory layout: Agent data**
  - On the PS3 the memory layout is one of the most important things for performance
  - AI: code is pretty simple, but called many times and:
    - Performs a lot of neighborhood searches
    - Inspects properties on all neighbor agents
  - Size of a full agent ~256 bytes
  - Separate out “agent core”. Stores the most basic properties: position, speed etc. 36 bytes
  - Each agent object has a pointer to its corresponding core
  - Allocate all cores as a single, 128 byte aligned, block of memory (1200 agents: 42kb)
  - Reduces cache missing during simulation and fits on SPUs
Memory layout: Cell map

- Conceptually each cell stores many different pieces of data:
  - Walkable/non-walkable (and other "cell flags")
  - Flow vectors
  - Heights
  - Head pointer of linked list of current occupiers
- Bad implementation
  - Implement class ZCell, map is an array of ZCell objects
- Good implementation
  - Map is 4 arrays, each storing a different attribute
- Why?
  - Array of struct vs. struct of arrays
  - Usually an algorithm is only interested in one of the attributes
    - Which can then be 128 byte aligned
    - Which can (more easily) fit on SPU local store
    - Spans less memory, in turn causing less cache misses
Crowd AI & steering on SPUs

- Moving the entire AI code to SPU is hard
  - Has many dependencies between components in the system
  - Virtual methods

- Profiling showed a few hotspots
  - Neighborhood gathering
  - Raycasting through cell map
  - Selecting “worst threat” for steering
  - All hotspots are isolated algorithms, working on a limited input!

- Added sensor system
  - Sensor input: position and radius for neighborhood, raycast requests etc
  - Sensor output: Current neighborhood, current worst threat, ray results
Steering with sensor data

- Sensor input is usually fixed
  - Probe a certain distance ahead of agent for walls
  - Collect around agent
  - Select worst threat
- Sensor input is usually configured once when entering AI state
- Actually, sensor output is not 1 frame delayed
  - (except for first frame in state)
Sensor updates on SPUs

- Each job updates X number of agents
  - So it fans out on multiple SPUs
- Needed data on local storage
  - Agent cores: ~42kb
  - For ray casts: ~16kb
    - Our crowds have around 16k cells
    - Cell flags: Array of bytes
  - For neighborhood searches: ~32kb
    - Head pointers from cell map (stored as 16bit indices)
    - Linked list is intrusive, stored in agent cores
  - Sensor input/output for each of the X agents: ~3kb (30 agents per job)
- In total: ~93kb of data needed. Plenty of room for code.
Conclusions

- We managed to create a new crowd system that is a significant step up from our previous system
- We managed to achieve very good performance, which was necessary since the crowd has to integrate with a full game
- Having a proper layout of data is critical for performance when handling massive amount of characters
- It is a time consuming task to tweak all the magic numbers in steering code
- Having proper animation on characters in very dense crowds is very hard, since steering relies on quick reactions from the characters
Questions?

● (Also feel free to email me at: kasperbf@ioi.dk)

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