

Game Technology

Lecture 14 – Artificial Intelligence in Games



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Skipped

Skipped slides will be
marked with this
symbol and are not
relevant for the exam

Final timetable



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Lecture No.	Date	Topic
1	17.10.2014	Basic Input & Output
2	24.10.2014	Timing & Basic Game Mechanics
3	31.10.2014	Software Rendering 1
4	07.11.2014	Software Rendering 2
5	14.11.2014	Basic Hardware Rendering
6	21.11.2014	Animations
7	28.11.2014	Physically-based Rendering
8	05.12.2014	Physics 1
9	12.12.2014	Physics 2
10	19.12.2014	Procedural Content Generation
11	16.01.2015	Compression & Streaming
12	23.01.2015	Multiplayer
13	30.01.2015	Audio
14	06.02.2015	AI
15	13.02.2015	Scripting - Optional

Current news

Lecture and Exercise 15, Scripting

- will be optional and not part of the exercise bonus points or the exam

Exercise points and overall bonus points will be completed soon

Example exam

- Will be provided next week

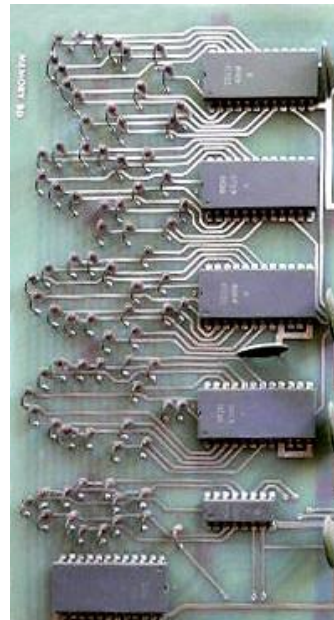
Feedback

- Thank you for filling out the feedback sheets!
- We will discuss some of the feedback and how we plan to adress it

Introduction

Computer Space (1971)

- Joel Bushnell, Ted Dabney
- Players control a spaceship
- AI UFOs
 - Fire into the quadrant in which the player spaceship is



Introduction



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Pac Man (1980)

- Tōru Iwatani
- Four different ghost „personalities“

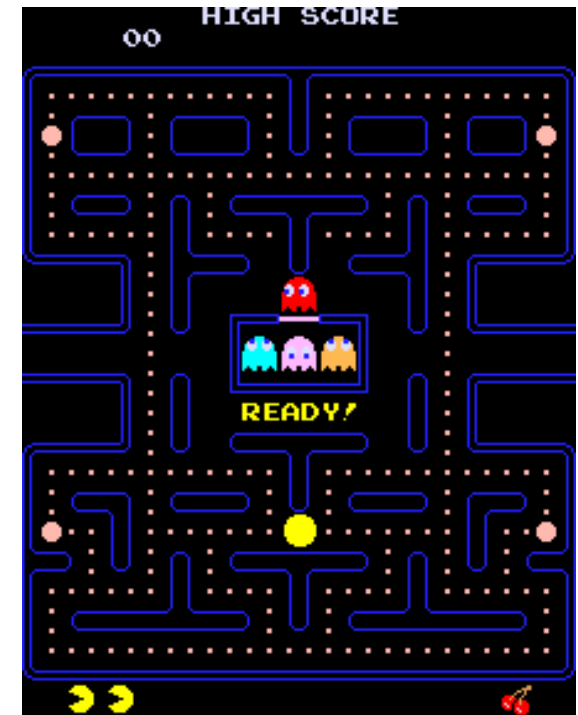
CHARACTER / NICKNAME		CHARACTER / NICKNAME	
	- SHADOW "BLINKY"		OIKAKE - - - "AKABEI"
	- SPEEDY "PINKY"		MACHIBUSE - - "PINKY"
	- BASHFUL "INKY"		KIMAGURE - - "AOSUKE"
	- POKEY "CLYDE"		OTOBOKE - - - "GUZUTA"

E.g. red ghost

- „Shadow“
 - Target is always Pac-Man (vs. where he is going to)

More info

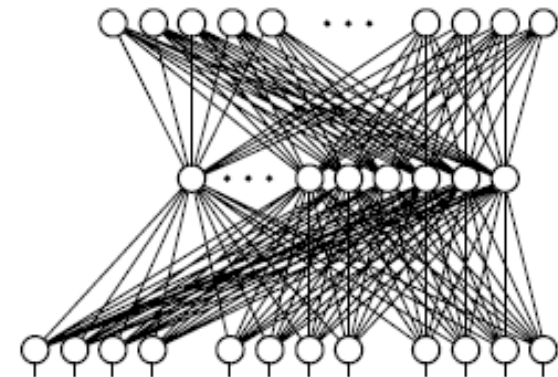
- <http://gameinternals.com/post/2072558330/understanding-pac-man-ghost-behavior>



Introduction

Creatures (1996)

- Steve Grand
- Train and breed creatures
- Used [neural networks](#) for AI



Introduction

Today

- AI central component in almost all games
- Gets more importance (= computation time), is integrated into game design decisions
- AI engines and AI middleware



Model

AI gets its
information

AI gets given processor time

Execution management

World interface

Group AI

Strategy

Character AI

Decision making

Movement

Content creation

Scripting

AI has implications
for related technologies

Animation

Physics

AI gets turned into on-screen action

Simple things can look good

- ELIZA (1966)
- > Hello. My name is Eliza. How may I help you?
- * I want to give a lecture on AI for games.
- > Why do you want to give a lecture on AI for games?
- * It is part of a lecture series on game technology.
- > Say, do you have any psychological problems?
- * Maybe...
- > You don't seem quite certain.

Pac Man

CHARACTER / NICKNAME		CHARACTER / NICKNAME	
 - SHADOW	"BLINKY"	 OIKAKE - - -	"AKABEI"
 - SPEEDY	"PINKY"	 MACHIBUSE - -	"PINKY"
 - BASHFUL	"INKY"	 KIMAGURE - -	"AOSUKE"
 - POKEY	"CLYDE"	 OTOBOKE - - -	"GUZUTA"

Complexity



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More complex things can easily look bad



Kinds of AI in games

„Hacks“

- Make random adjustments to movement
- Add an animation to show the emotional state instead of simulating the resulting actions

Heuristics

- Instead of pathfinding, turn and go towards the goal
- In strategy games, assign values to units instead of computing their capabilities

Algorithms

- Re-usable approaches to AI
- E.g. movement, pathfinding, decision making

Movement

AI gets its
information

AI gets given processor time

Execution management

World interface

Group AI

Strategy

Character AI

Decision making

Movement

Content creation

Scripting

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for related technologies

Animation

Physics

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Movement

Input

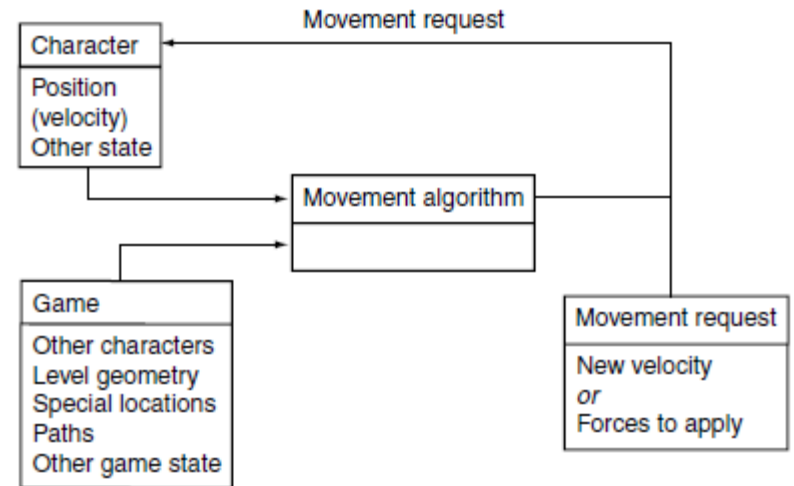
- Character's current state
- Game information, e.g. the current goal

Output

- Movement requests

Kinematic vs. Dynamic movement

- Kinematic – Update the velocity directly
- Dynamic – Take into account velocities, change via accelerations (see physics lectures)

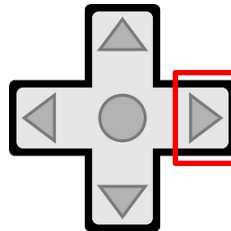


Movement comparison

Kinematic



Dynamic



Dimensionality

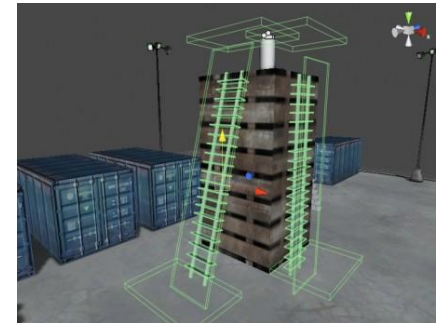
2D

- Easiest
- Often applicable to 3D rendered games
- 2D-Position + Orientation



2.5D

- 3D-Position + 2D orientation
- E.g. for characters walking and jumping/climbing ledges



3D

- Required if motion is truly 3D
- E.g. flight simulators



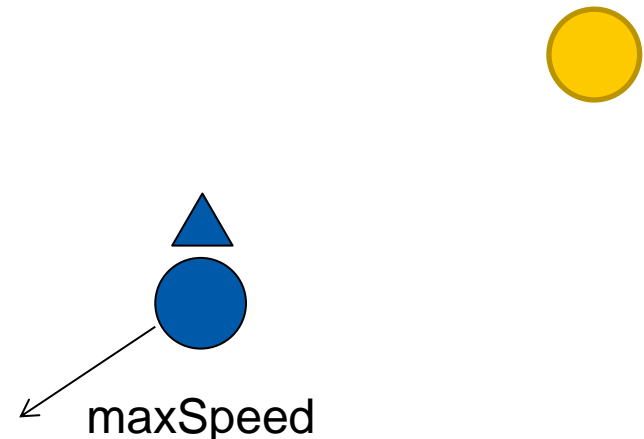
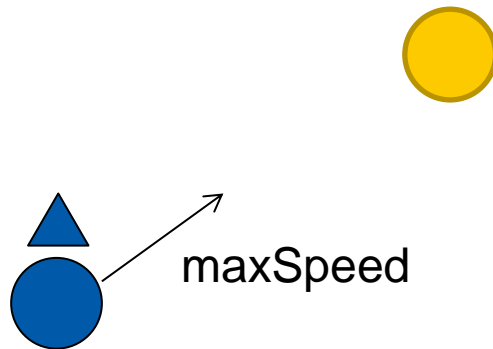
Simple movement behaviours (kinematic)

Seek

- Compare to target and current position
- Output a velocity going there directly at maximum speed

Flee

- Reverse direction from Seek



Simple movement behaviours

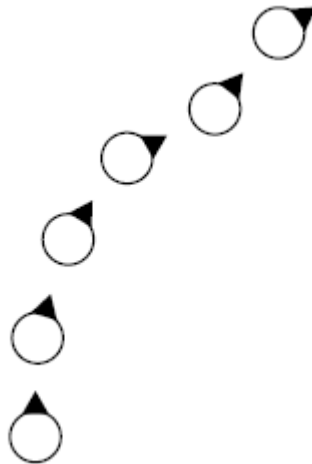
Arrive

- Define a radius when the goal is reached
- Slow down as the character gets closer
- Outside of radius
 - Calculate speed to reach target in fixed amount of time
 - Clip to maximum speed
 - → Will slow down as it gets closer
- Inside the radius
 - No movement (probably other behaviour appropriate, e.g. attacking)

Simple movement behaviours

Wander

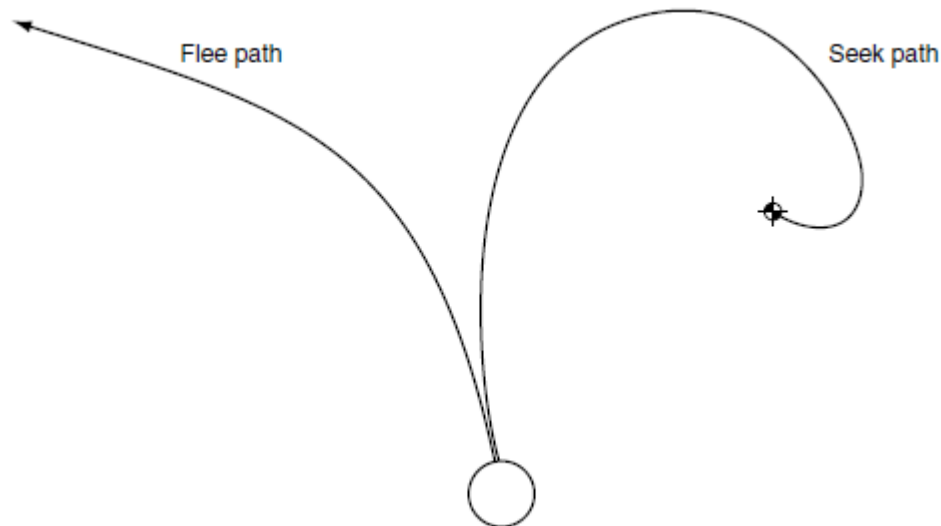
- Change the orientation randomly in a small range each time the behaviour is called



Take into account rotation and velocity

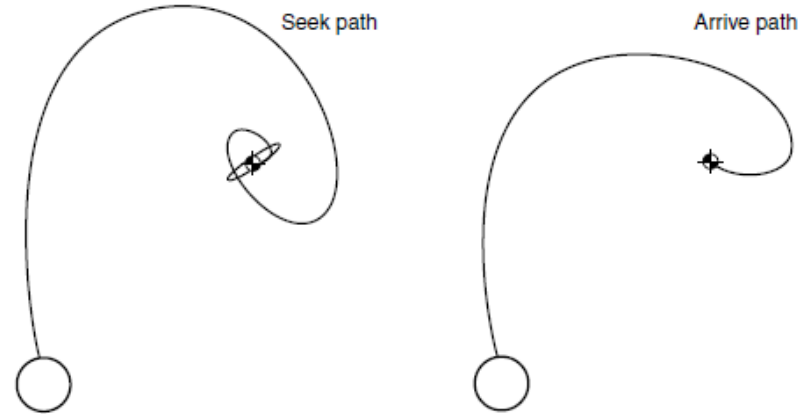
Seek and Flee

- As before: Choose direction to goal
- Output an acceleration in that direction
- Maximum speed: Either via clipping or via drag



Arrive

- Seek can overshoot and oscillate
- Define a radius in which the character should slow down
- Slowing down is done by calculating an acceleration that leads to the target velocity



Align

- Align to the same orientation as the target

Velocity matching

- Match the velocity of the target

Delegated behaviours

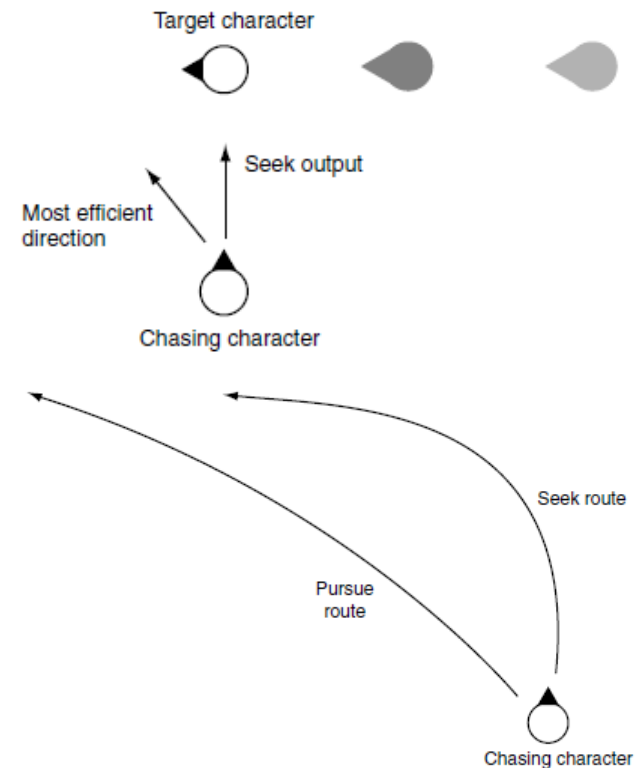
Based on the primitive behaviours we have looked at until now

Pursue

- Calculate where the target will be if it continues to move with the current velocity
- A Steer behaviour is used to move towards this new target

Evade

- Opposite of pursue



Delegated behaviours

Face

- Align to the target

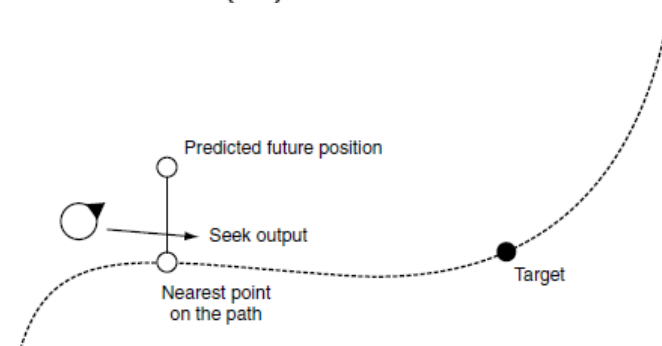
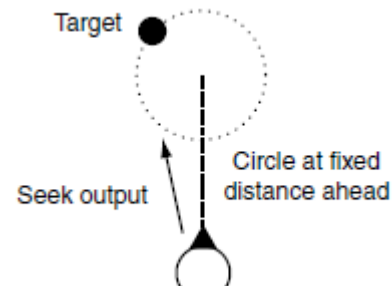
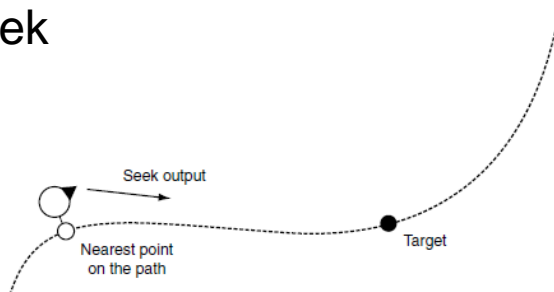
Looking where you are going

Wander

- Uses Seek

Path following

- Uses seek

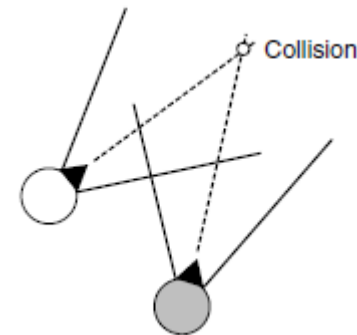
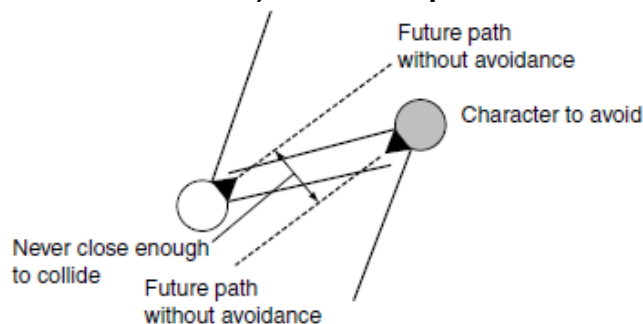


Separation

- Stay away from nearby other characters
- $\text{strength} = \text{maxAcceleration} * (\text{threshold} - \text{distance}) / \text{threshold}$

Collision Avoidance

- For characters that move on crossing trajectories
- a) Only do separation inside a cone in front of the character
- b) Only handle collisions that will take place (similar to collision detection in the physics lectures) → no „panic“ reactions



Obstacle and Wall Avoidance

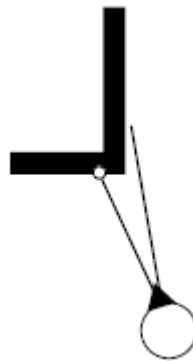
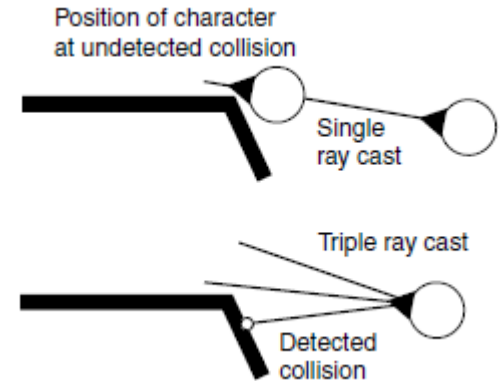
Similar, but distinct from collision detection in physics

Basic mechanism: Use ray casts

- One ray cast often not enough
- Use multiple casts or swept volumes

Handling corners

- Deadlock even for several rays
- Possible solutions
 - Adaptive ray fan sizes
 - Special-case code for corners



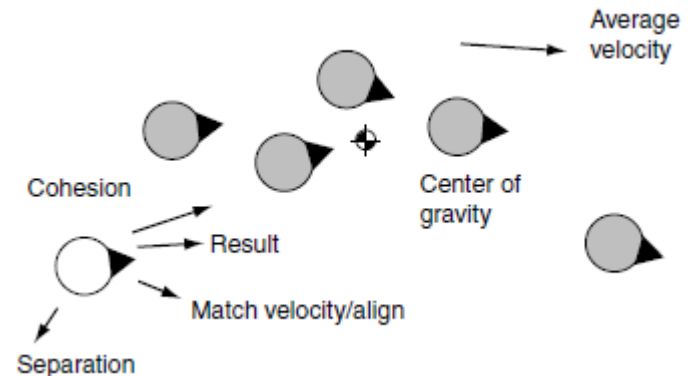
Combining steering behaviours

(Weighted) blending

- Take the (weighted) average of basic steerings
- Good weights? Always the same weights?

Flocking

- Simulation of fish schools, bird flocks, ...
- Simulated entities referred to as boids
- Created by Craig Reynolds, 1987
- Blend of 3 steering factors
 - Separation (move away from close boids)
 - Move in the same way as the flock (match avg. velocity and orientation)
 - Cohesion (steer towards center of the flock)



Combining steering behaviours

Equilibria

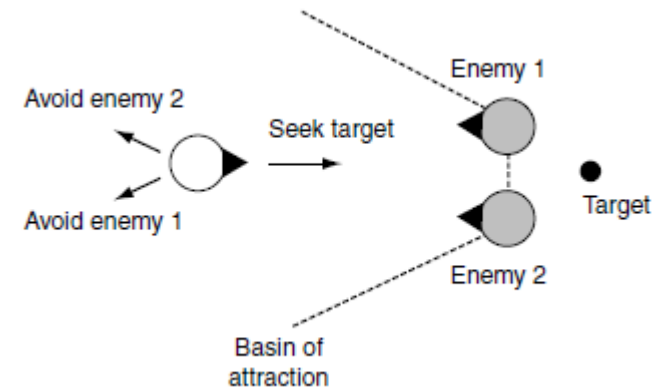
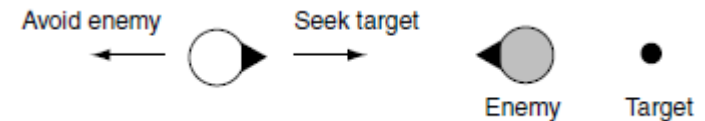
- Conflicting steering behaviours lead to deadlocks

Unstable equilibria

- Situation is inpropable
- Basin of attraction is small

Stable equilibria

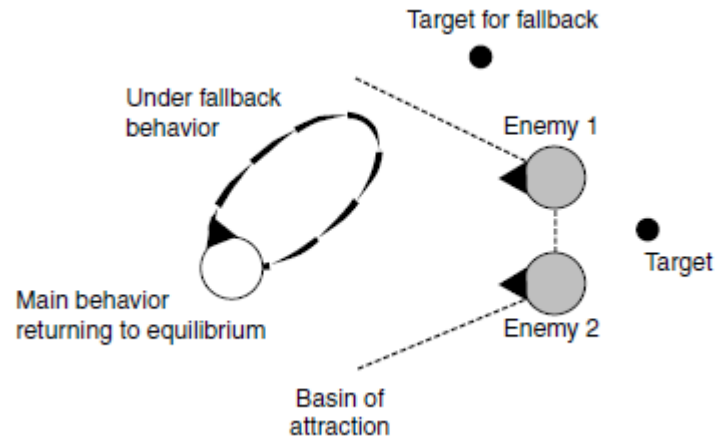
- Less likely to be broken by numerical inaccuracies



Define priorities for steering behaviours

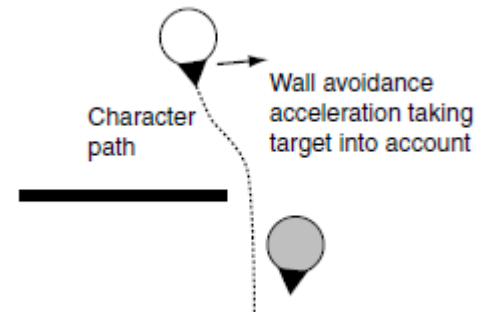
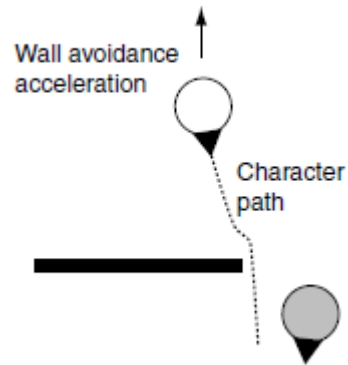
E.g. collision avoidance has priority, if it has an output, it is taken

Can still oscillate for stable equilibria



Cooperative arbitration

Communication between steering behaviours



Pathfinding

AI gets its
information

AI gets given processor time

Execution management

World interface

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Strategy

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Decision making

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AI gets turned into on-screen action

Higher-level movement planning

- Generate a path for an AI to follow

Based on a graph

- Basic connectivity: point B can be reached from A if there is a connection between them
- Weights/costs: How „hard“ is it to get from A to B?
- Directions: In which direction is the connection? (E.g. jumping down an unclimbable ledge)

Finding a path

A*-Algorithm is the prevalent solution

Not required to compute it in the exam, but you should know how it works

- Requires an adjacent-list and connections between nodes (e.g. tiles)

calculate cost of CurrentNode

add CurrentNode to the Open-Node-List

while (Open-Node-List not empty)

 CurrentNode = node from Open-Node-List with lowest costs

if (CurrentNode == TargetNode)

 Path completed

else

for each AdjacentNode next to CurrentNode

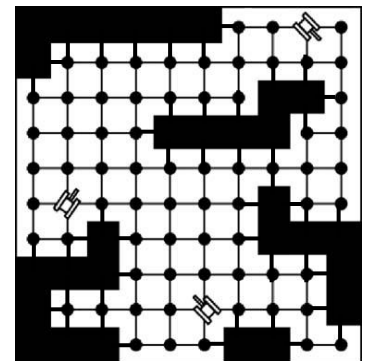
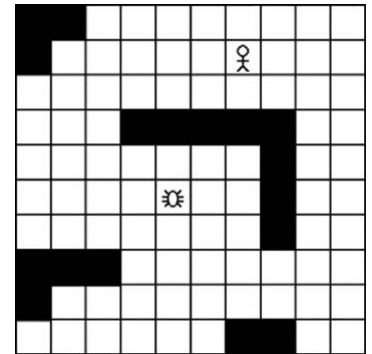
if (AdjacentNode **not** in Open-Node-List **and**
 AdjacentNode **not** in Visited-Node-List **and**
 AdjacentNode **is not** an obstacle)

 calculate cost of AdjacentNode

 link AdjacentNode to CurrentNode

 add AdjacentNode to Open-Node-List

 add CurrentNode to Visited-Node-List





World representation

Quantization

- Find the appropriate node for the current world position

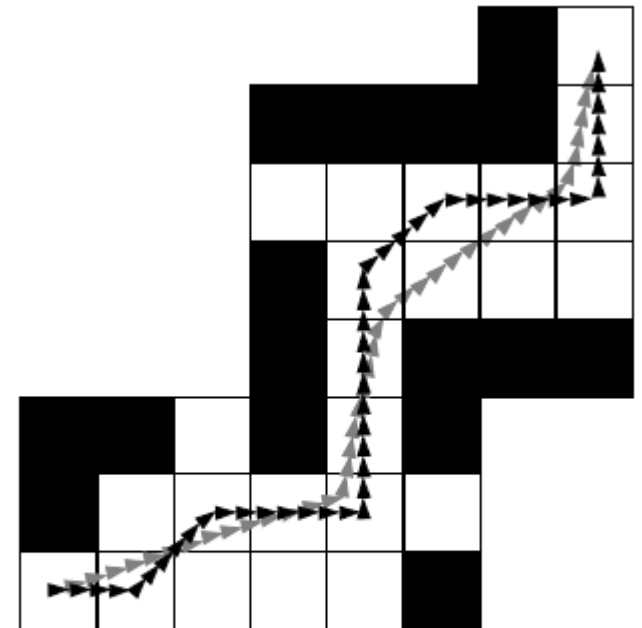
Localization

- Find the world location of a graph node



Tile-based

**Well-suited if game world is
already represented by tiles (e.g.
heightmap)**

**Plans can be blocky due to
rectangular nature**

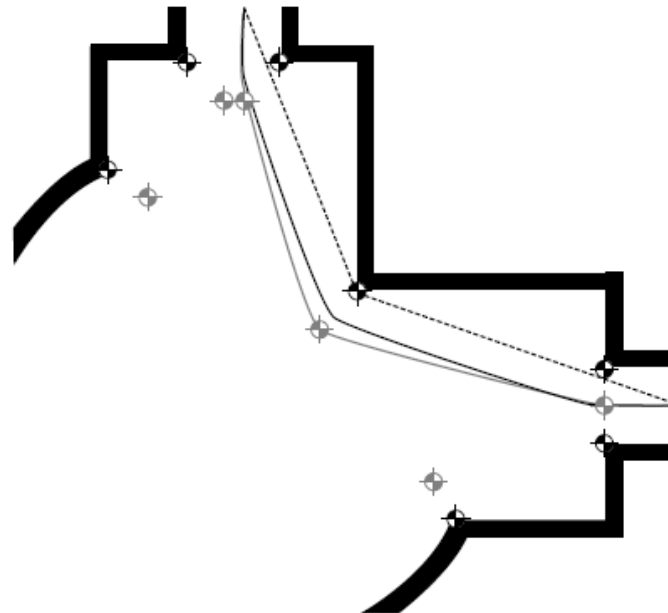


Key

-  Output blocky plan
-  Ideal direct plan

Points of visibility

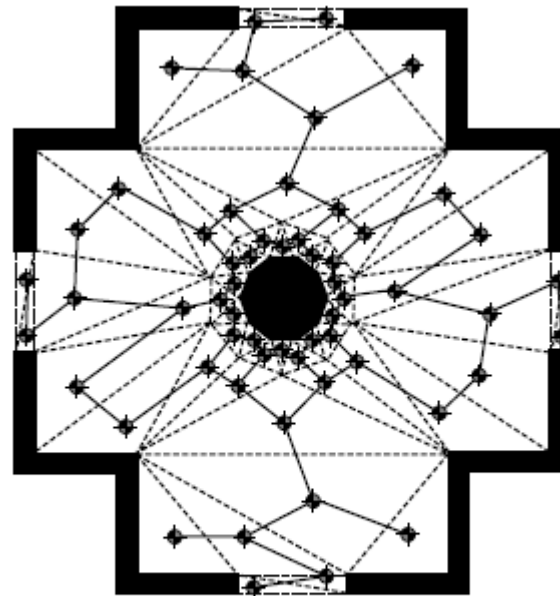
Based on the edges of the world



- Key**
- Optimal path for zero-width character
 - Path for character with width
 - Path using vertex offsets
 - ⬤ Original characteristic points at vertices
 - ⬤ Offset characteristic points

Navigation mesh

Use the mesh of the floor as the
basis



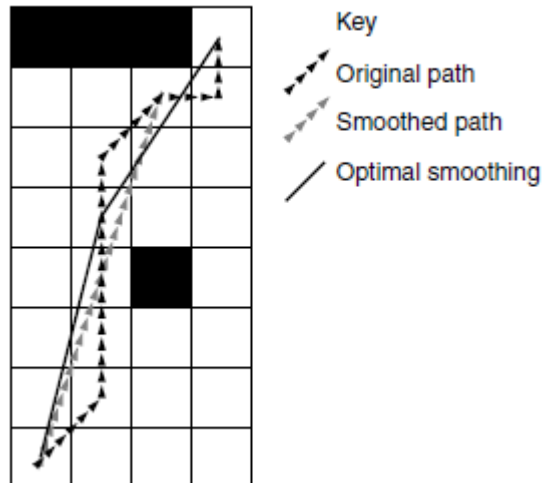
Key

- Edge of a floor polygon
- Connection between nodes

Path smoothing

Filter paths to remove sharp edges

- E.g. by checking connectivity between waypoints by ray casts

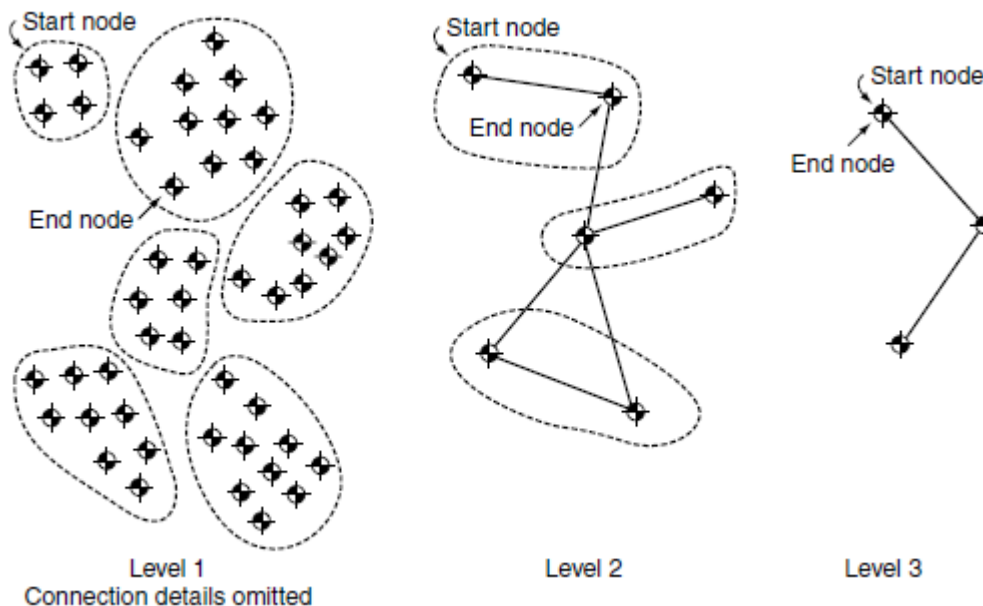


Hierarchical pathfinding

Cluster waypoints (e.g. by rooms, areas, ...)

Compute paths between the clusters

Only when the character needs to traverse a cluster, find a path through it

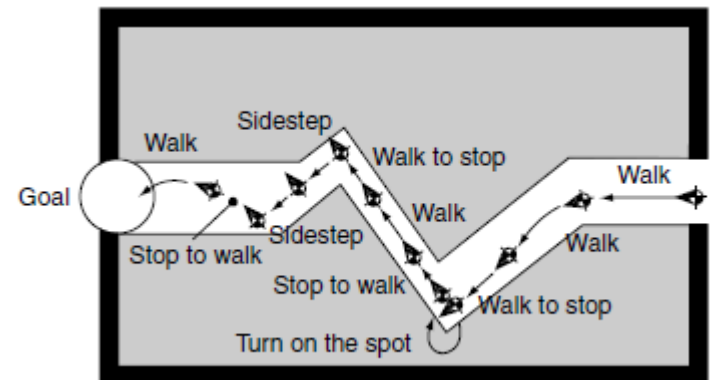
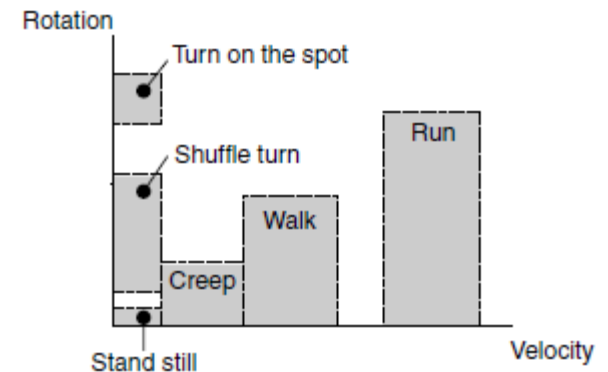


Movement planning

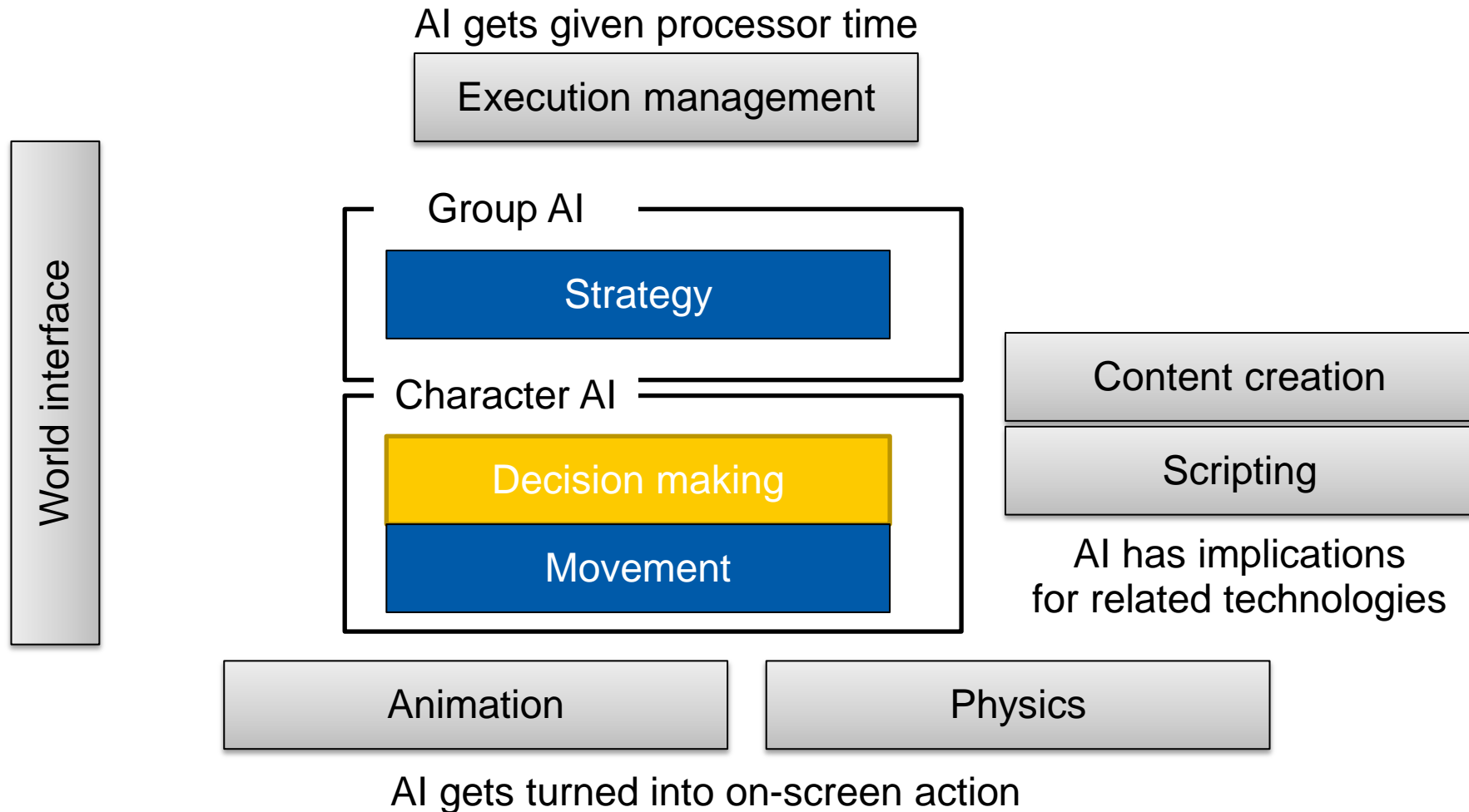
A character that is animated has only a finite set of animations

- Limited to certain directions
- Limited to certain speeds

Use A* variant to search good combinations



Decision Making



Decision making

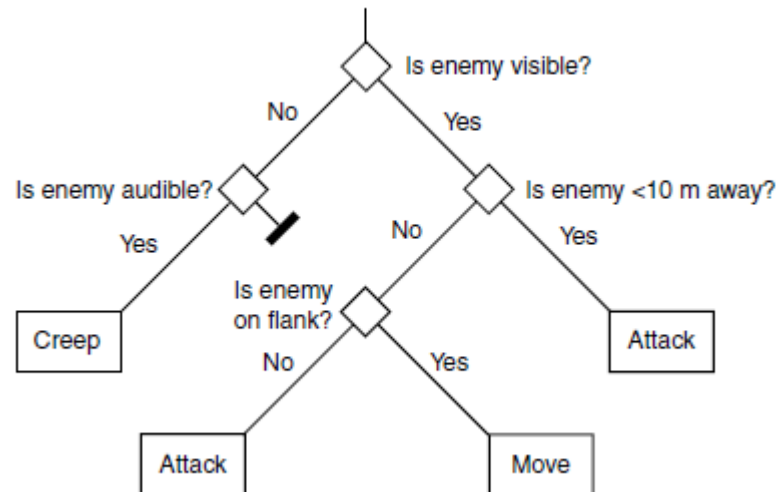


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Structure decisions as a series of conditions and actions as leaves

- Created by designer
- Condition nodes can have more than 2 branches
- Performance
 - Try to keep trees balanced
 - Use computation-heavy conditions further down in the tree



Random decision trees

We do not want to have the same (predictable) behaviour each time

Add randomness to the graph

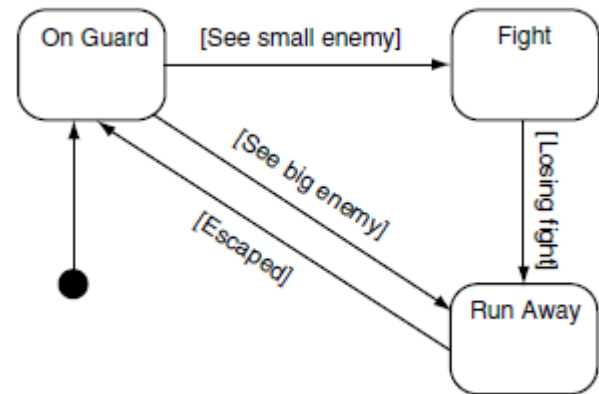
- Watch out for oscillations
- Save previous choices and only change after a timeout

Generalization of the decision trees

- States govern the behaviour of the agent
- Transitions are triggered if the condition matches

Created by designers

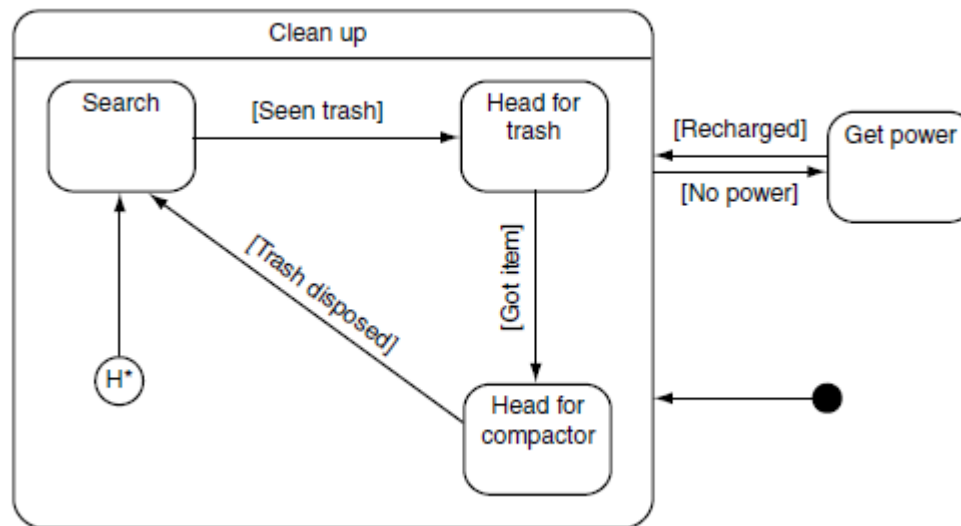
Hard-coded vs. flexible



Hierarchical state machines

Problem: „Alarm“ states

- Normal behaviour is interrupted, e.g. by seeing the player
- Can be from any of the normal states → many transitions to alarm state
- Later return to normal state



Behaviour Trees

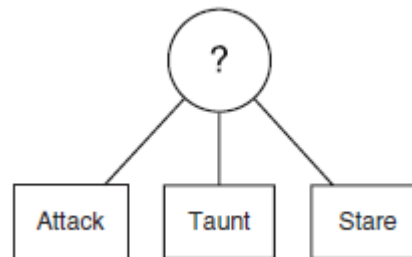
Combination of other techniques

Easy to understand and create for non-programmers

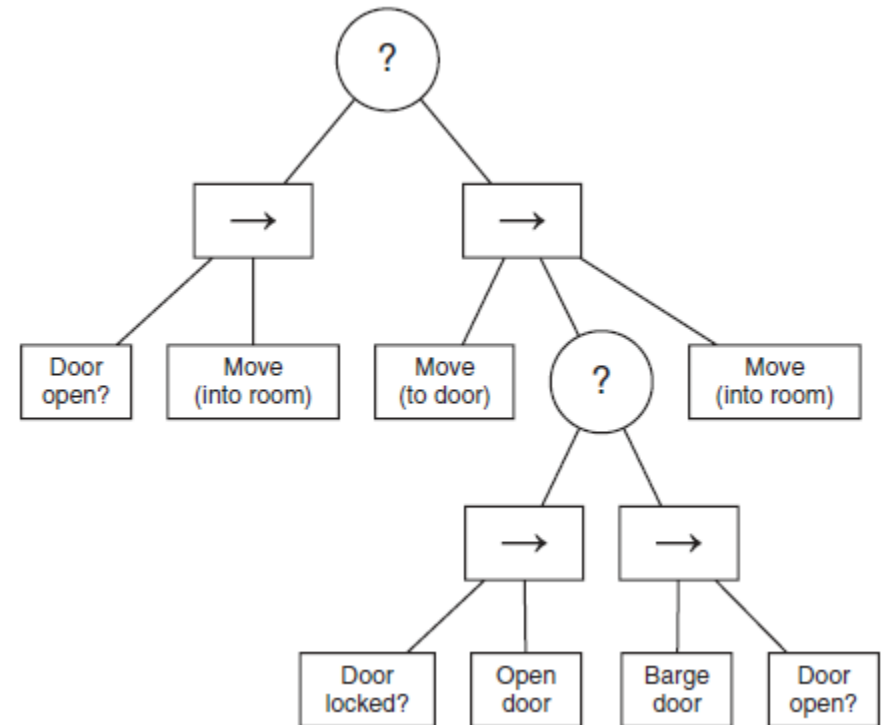
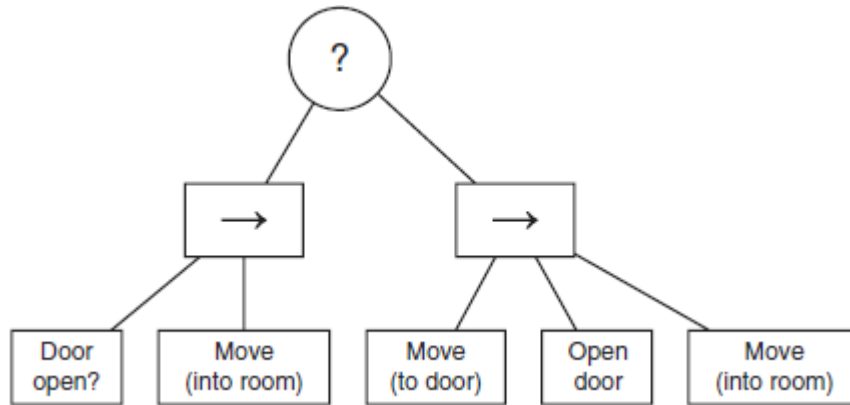
- First description in Halo 2 (2004)

Tasks instead of states

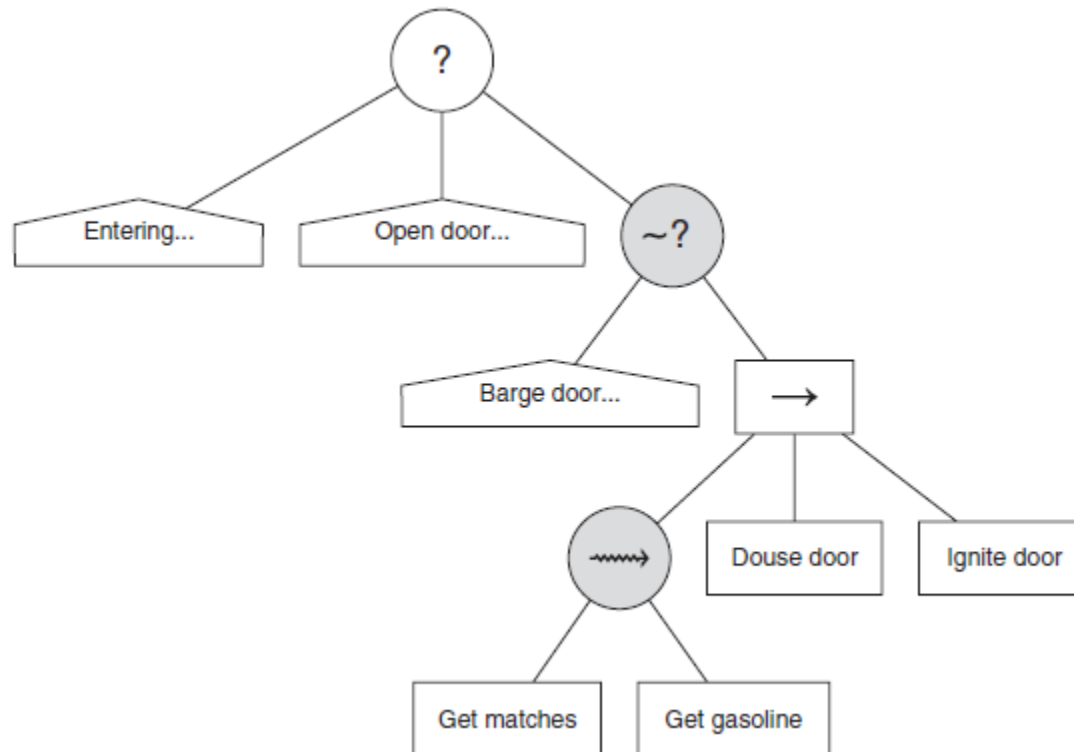
- Conditions
- Actions
- Composites
 - Selector
 - Sequence



Behaviour trees



Non-determinism



Game states can be ambiguous

- How much health is „injured“?
- How close is „close“?

Fuzzy logic

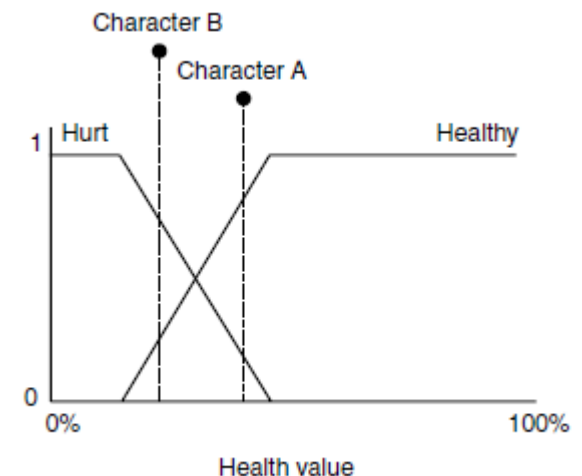
- Instead of being or not being in a set, fuzzy logic states that an object has a set degree of membership in the range $[0,1]$

Fuzzyfication

- Turning game data into set degree of membership

Defuzzyfication

- Mapping fuzzy data to game data



Fuzzy operators and rules

Operators: Similar to Boolean operators

Expression	Equivalent	Fuzzy Equation
NOT A		$1 - m_A$
A AND B		$\min(m_A, m_B)$
A OR B		$\max(m_A, m_B)$
A XOR B	NOT(B) AND A	$\min(m_A, 1 - m_B)$
	NOT(A) AND B	$\min(1 - m_A, m_B)$
A NOR B	NOT(A OR B)	$1 - \max(m_A, m_B)$
A NAND B	NOT(A AND B)	$1 - \min(m_A, m_B)$

Rules

- Define membership in fuzzy sets based on membership in other sets

$$m_{(\text{Should Brake})} = \min(m_{(\text{Close to the Corner})}, m_{(\text{Traveling Quickly})}).$$

Example



corner-entry AND going-fast THEN brake
corner-exit AND going-fast THEN accelerate
corner-entry AND going-slow THEN accelerate
corner-exit AND going-slow THEN accelerate

Corner-entry: 0.1

Corner-exit: 0.9

Going-fast: 0.4

Going-slow: 0.6

Brake = $\min(0.1, 0.4) = 0.1$

Accelerate = $\min(0.9, 0.4) = 0.4$

Accelerate = $\min(0.1, 0.6) = 0.1$

Accelerate = $\min(0.9, 0.6) = 0.6$

Take maximum:
Accelerate = 0.6

Goal-oriented behaviour

Goals

- A.k.a. motives or insistence
- Something the character wants to achieve
- E.g. (Sims) Hunger, bathroom distress, ...

Actions

- Anything the character can do
- Depends on the game state (e.g. is the food it wants to eat raw or cooked?)
- Changes the goal levels (eating → less hunger)



Selection



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Just choose the current optimum

- Can be sufficient
- But will not lead to longer strategies

Goal: Eat = 4 Goal: Sleep = 3

Action: Get-Raw-Food (Eat – 3)

Action: Get-Snack (Eat – 2)

Action: Sleep-In-Bed (Sleep – 4)

Action: Sleep-On-Sofa (Sleep – 2)

Overall utility

- Add a distress/energy metric
- Choose the option that leaves the character with the best overall result

Timing

- Also keep time in the view
- Short, ineffective vs. Long and effective actions

Planning

Goal-Oriented Action Planning (GOAP)

- Look at sequences of actions
- Keep a world model and extend it into the hypothetical future
- Rank the actions by the overall utility they produce in the long run

Use A* (variants)

- Instead of depth-first search, use a heuristic
- Use A* variants that can cope with infinite graphs
 - E.g. IDA* (Iterative Deepening A*)

Horizon effect

- How many actions do we look ahead?
- What if n actions lead make only slow progress, but $n+1$ (beyond our horizon) is the best option?

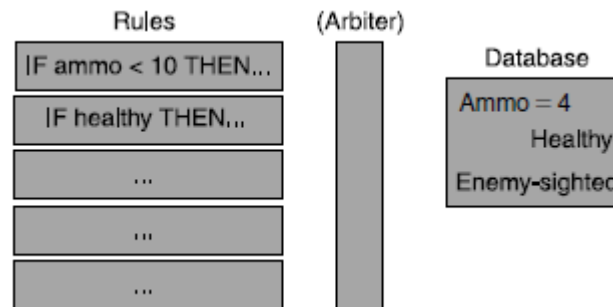
Aka. Expert systems

Set of rules in the form IF condition THEN action

Created by a designer

Arbitration and variable matching

- Which rule to use if several fit?
- Variable matching: When rules have (connected) variables

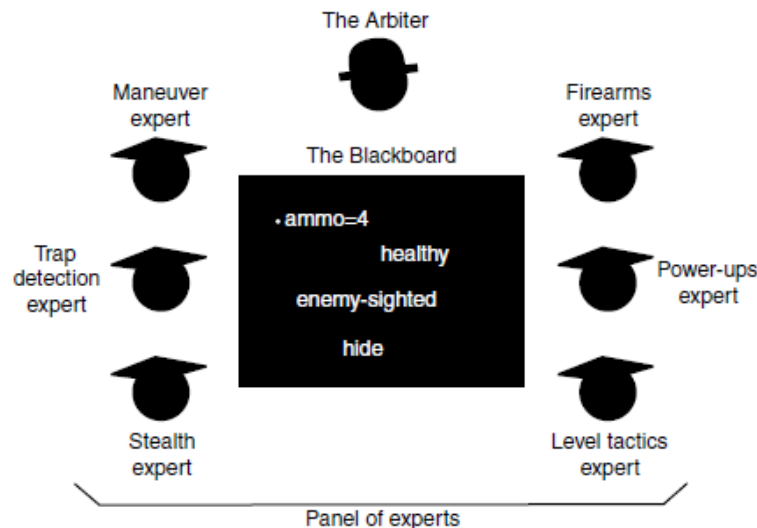


Blackboard architectures

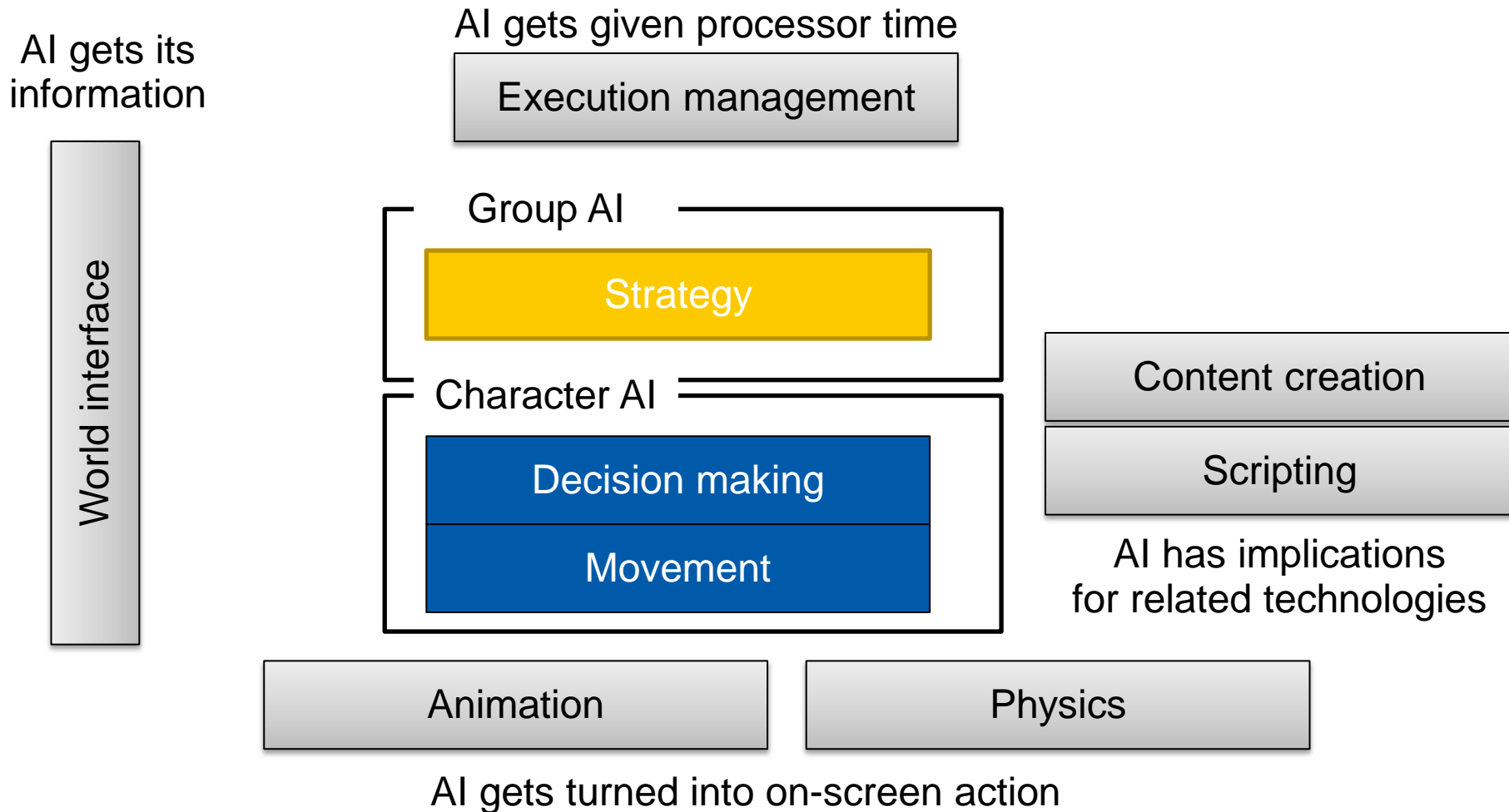
Every AI can write to and read from the blackboard
Efficient way of communicating with other AIs

For decision making

- AI components estimate if they can act on the current blackboard state
- When given control by an arbiter, they control the character



Tactical and Strategical AI



Tactical and Strategical AI



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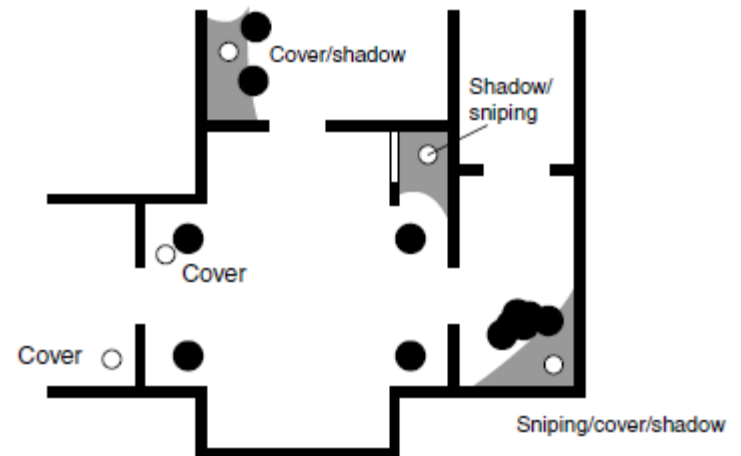
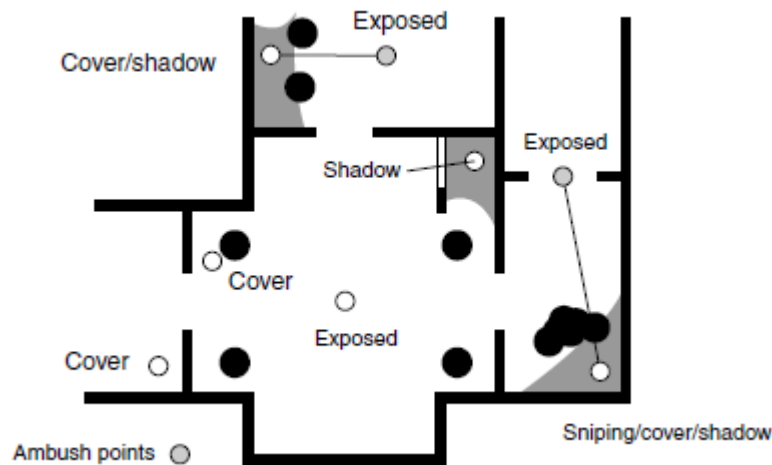


VS



Waypoint Tactics

Find points that are suited for different actions

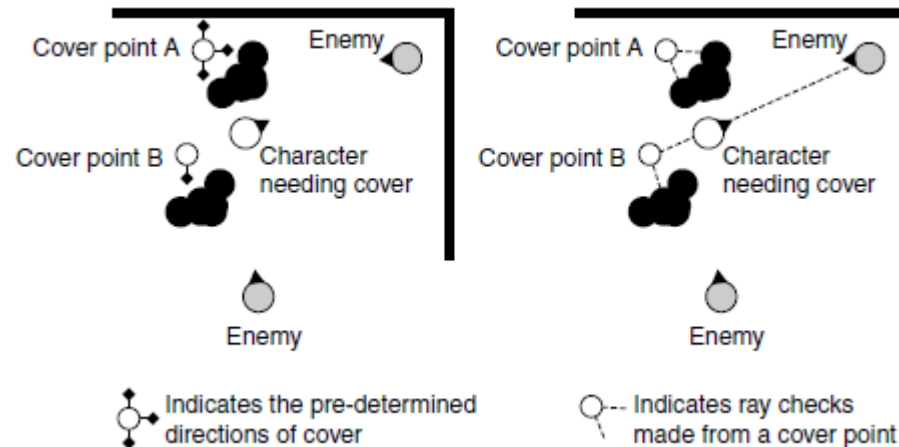


A tactical waypoint can depend on the actions of other AIs

- Example: Cover depends on where the enemy is

Precomputation step: Find directions that the point provides cover for

At runtime: Find out if the currently relevant direction is among them



Generating the points

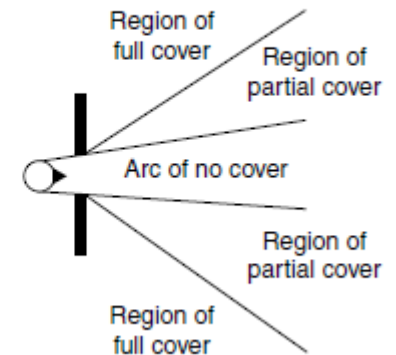
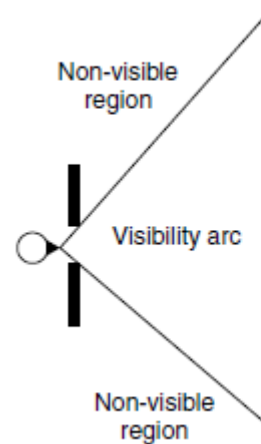
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Using visibility checks

- Visibility: How much space can be overlooked?
- Cover: How well is a character covered?



Shadow points

- Cast rays to light sources
- Use an existing global lighting solution

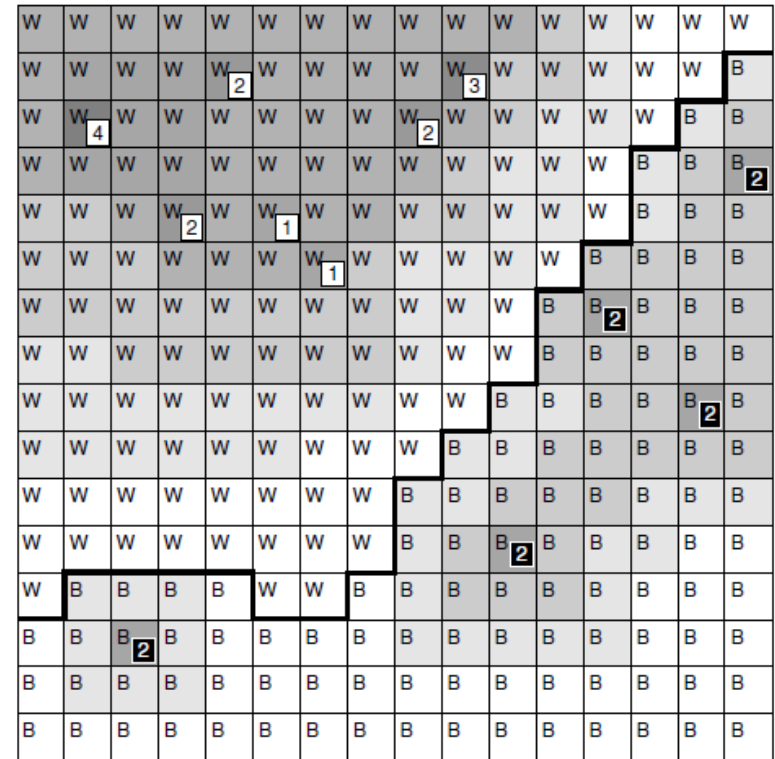


Tactical Analyses

Analyze the game world

Influence maps

- How strong is the influence of each player?
- Based on unit and building strength values
- Distance attenuation



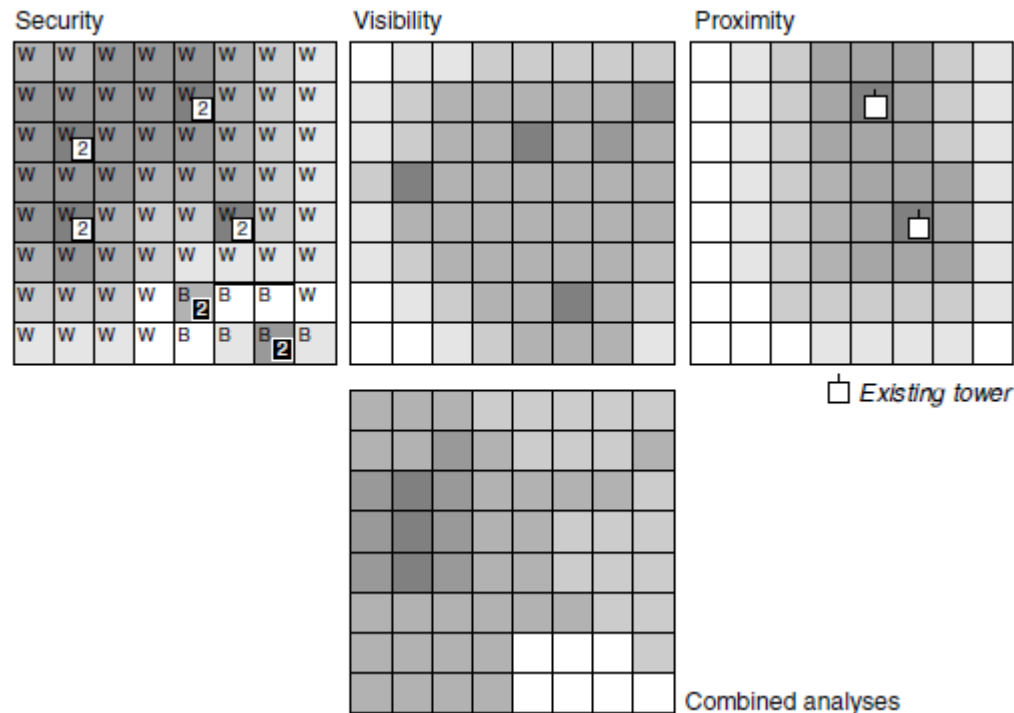
Learning influence maps

Frag Maps



Multiple Layer Analysis

Combine different tactical analyses into one



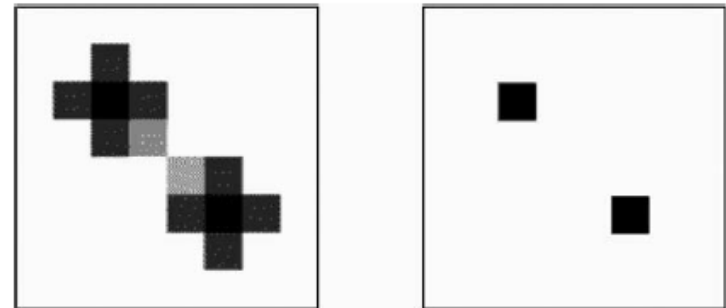
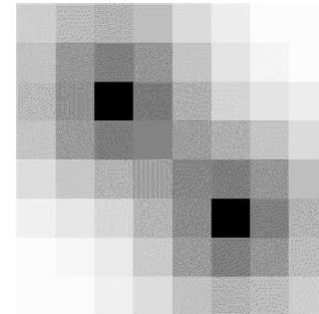
Generating influence maps

Map Flooding

- Use Dijkstra Algorithm variant
- If influences overlap, the stronger influence wins
- Leads to Voronoi Diagram-like map

Filters

- Blur
- Sharpen
 - E.g. to find the most important location



Generating influence maps

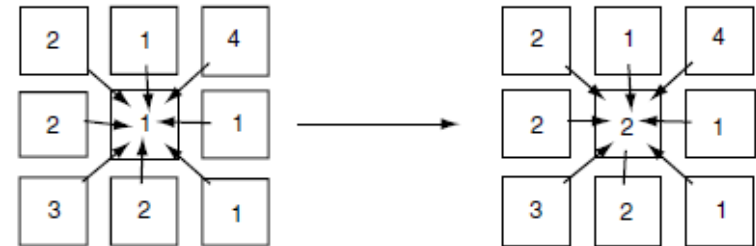
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Cellular automata

- Define rules how tiles are filled based on neighbors
- Used in Sim City



IF two or more neighbors with higher values,
THEN increment

IF no neighbors with as high a value,
THEN decrement



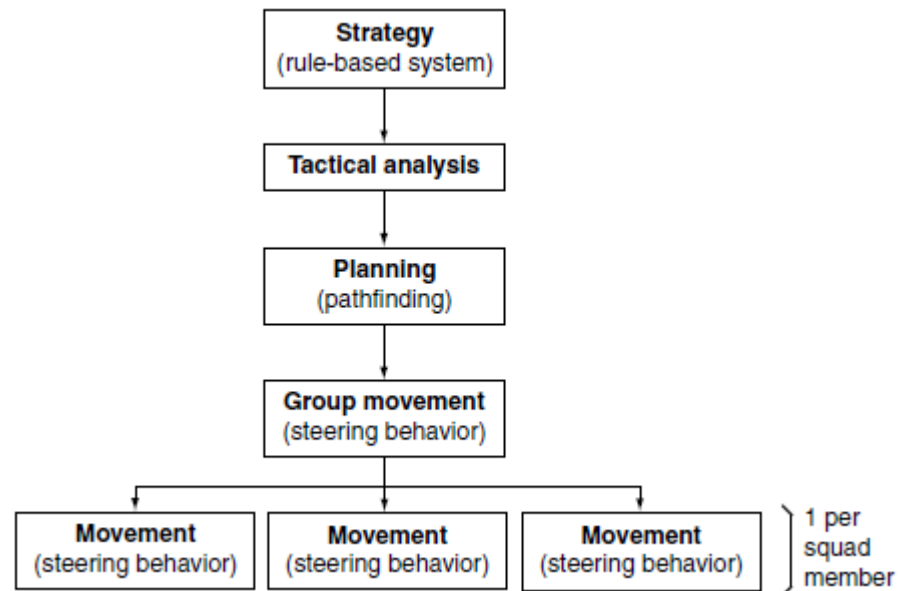
Coordinated Action

Bottom up

- Als query higher-level systems for tactical data

Top-down

- High-level commands passed downwards



AI gets its
information

AI gets given processor time

Execution management

World interface

Group AI

Strategy

Learning

Character AI

Decision making

Movement

Content creation

Scripting

AI has implications
for related technologies

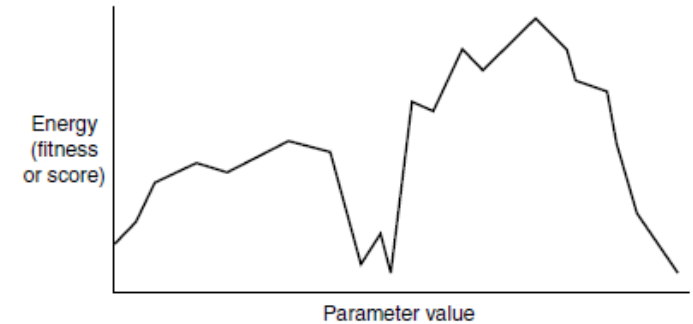
Animation

Physics

AI gets turned into on-screen action

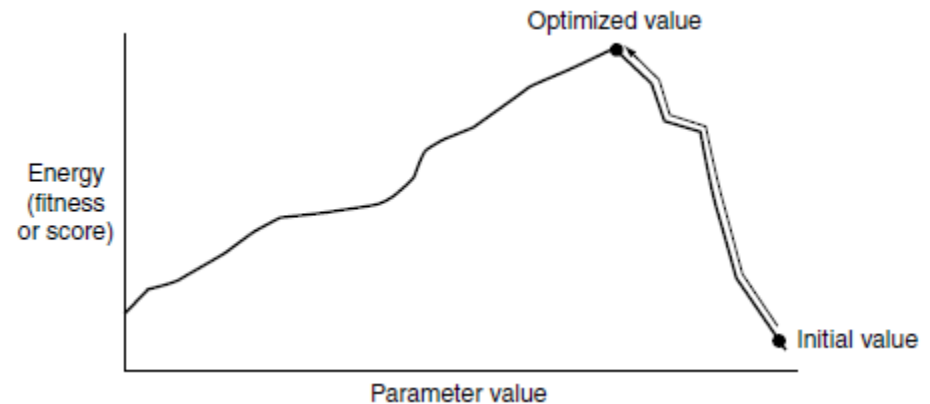
Learn the optimal value of a parameter

- E.g. optimal driving speed, number of zergs in a rush, ...



Hill climbing

- Move towards ascending gradient
- Stop when a maximum is reached



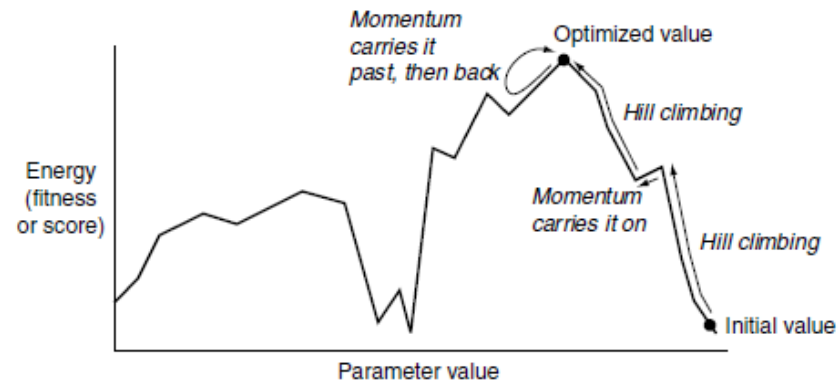
Finding global maxima

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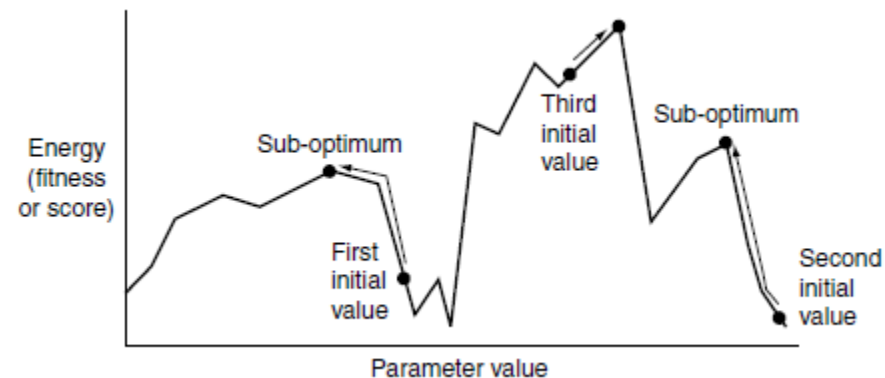


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Momentum



Multiple trials



Action Prediction

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Raw probability

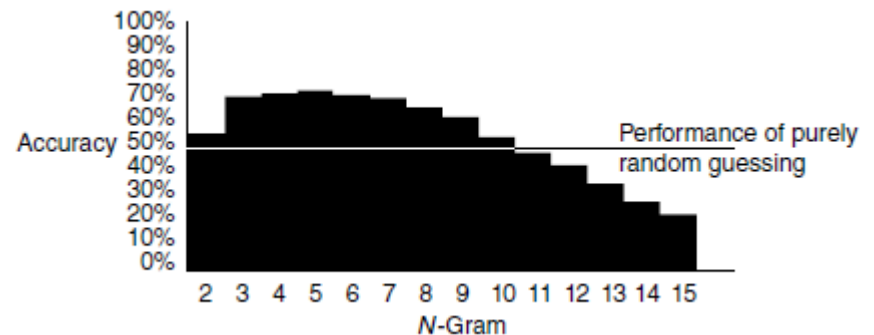
- Based on counting previous examples

N-Grams

- Take strings of several characters
- E.g. 2-gram for buttons A and B
 - AA
 - AB
 - BA
 - BB

Window Size

- Quality highly dependent on N



Further learning topics

Skipped



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Naive Bayes

- Combine probabilities

Learn decision trees

- ID3
 - Entropy and information gain
- ID4
 - Incremental learning

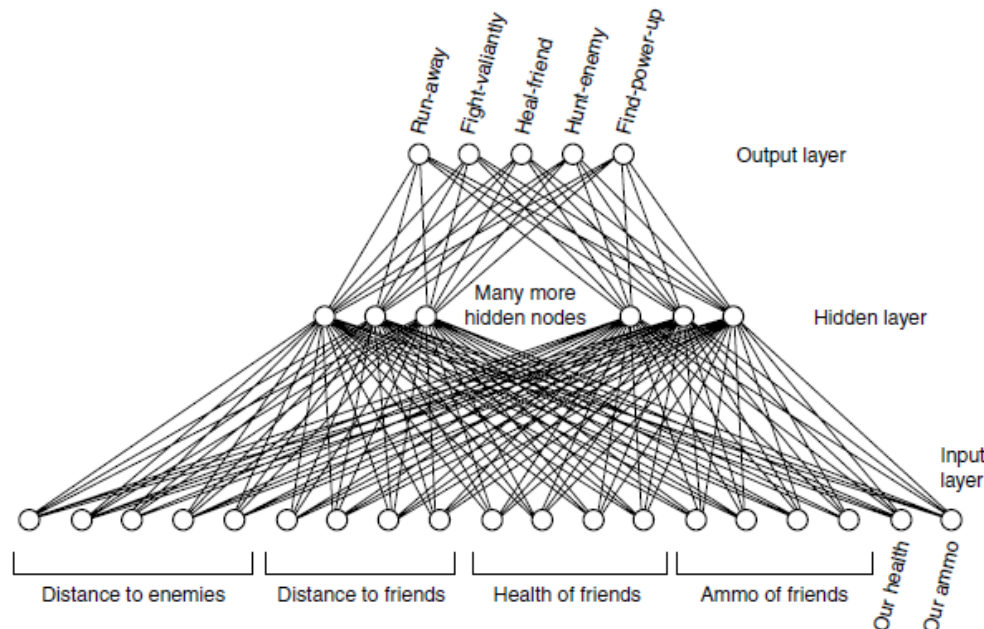
Reinforcement learning

- Learn qualities (Q-Learning) for each state and action

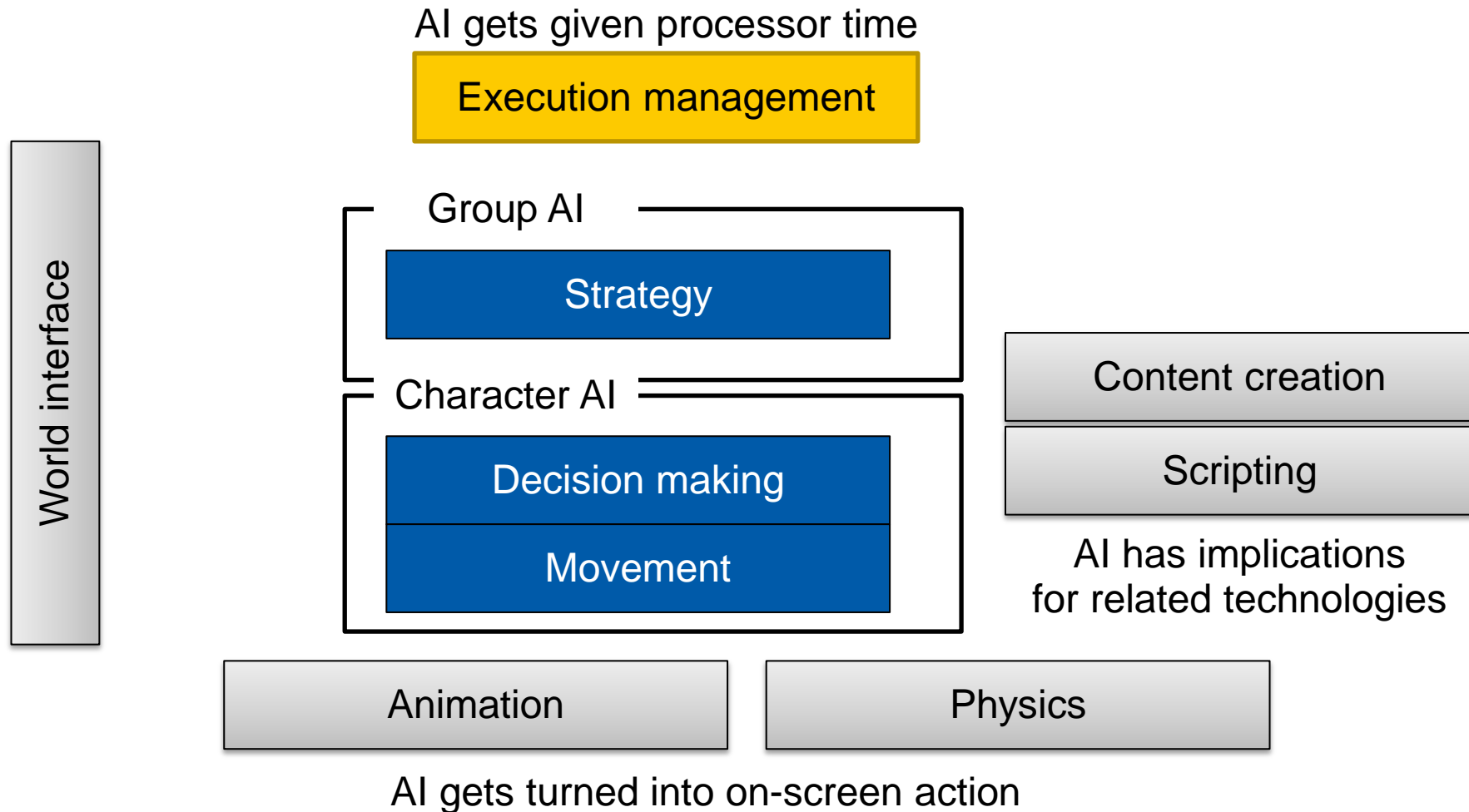
Model networks of neurons that each have specific values determining the output as a function of the inputs

Decision-making: Feed a state's variables in and observe the output

Learning: Feed an example state in, compare the output to the intended output, and adjust the values accordingly



Execution management



Scheduling



Frequency

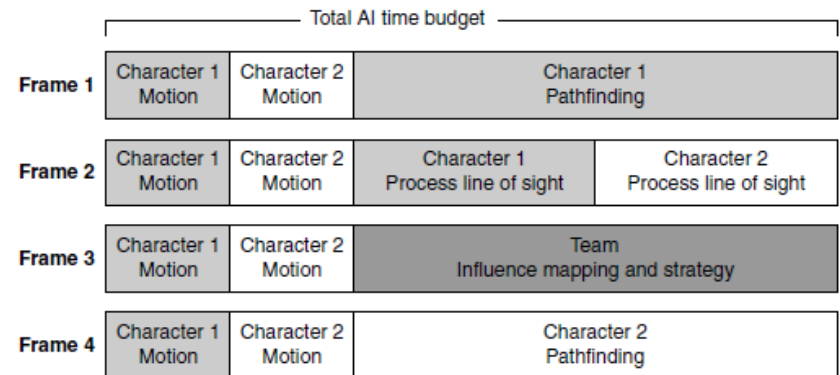
- How often should the AI task be called?

Phase

- Offset the task's execution by n frames
- If several tasks have the same phase, spread them out by changing phases

Choose a good phasing

- Wright's method
 - Simulate (only counting) ahead N frames
 - Choose the frame with the lowest number of tasks and set the phasing accordingly



Interruptible tasks

Allow AI tasks to run over several frames

E.g. pathfinding across the whole map

Continue where it stopped the last time



Anytime Algorithms

Be interruptible

Always provide a solution

- Can be used as a good starting point
- Will be refined the longer the algorithm runs in total

Also remember hacks

- Just start going into the direction of the target
 - Can always backtrack later
 - Always better than not moving for some seconds
- Carry out a „order received“ animation

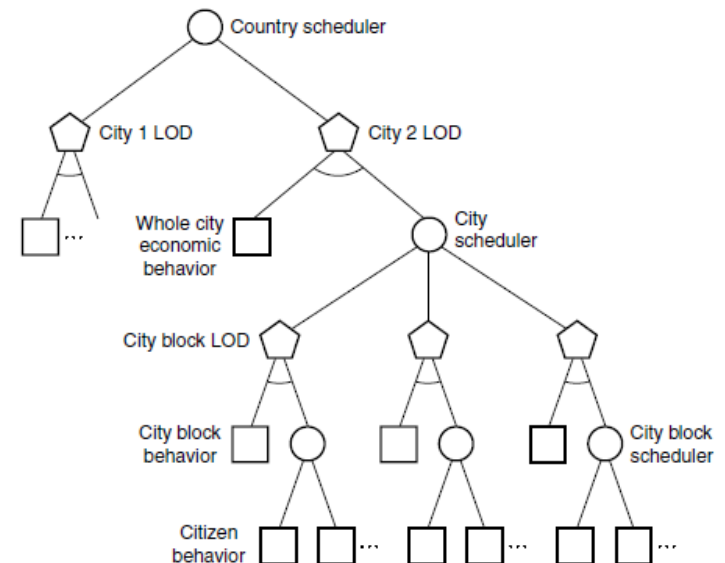
AI Level of Detail

AI that is imperceptible gets less time and simulates less

Hysteresis

- Characters moving in and out of the radius

Phase out areas that are not nearby



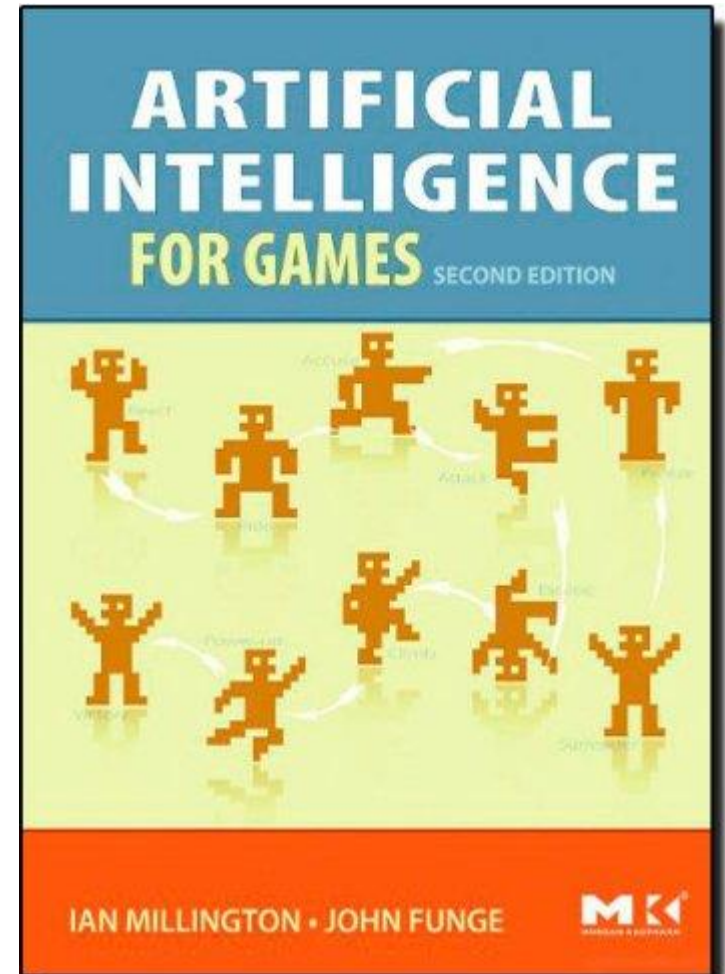
Literature

Artificial Intelligence for Games

- Ian Millington, John Funge

AI Game Programming Wisdom Series

<http://aigamedev.com/>



Exercise 13 - Sound

2.1 Doppler Effect

Consider a car driving with 150 km/h and a person running away from the car with 15 km/h. The car emits lots of different sound effects. How much does the frequency of those sound effects change for the person when the car passes him?

$$f_1 = f * (c - 15 \text{ km/h}) / (c - 150 \text{ km/h})$$

$$f_2 = f * (c + 15 \text{ km/h}) / (c + 150 \text{ km/h})$$

$$f_{\text{dif}} = \text{abs}(f_1 - f_2)$$

Exercise 13 - Sound

2.2 Sound location simulation without headphones

Directional sound can be simulated effectively using headphones. Can this also be done using regular speakers? What are the expected limitations?

Only for one person.

Exercise 13 - Sound

2.3 Sound reflection data

Considering the data available to a physically based rendering engine – what data can be reused to simulate realistic sound reflections?

Geometry, normal maps, roughness maps

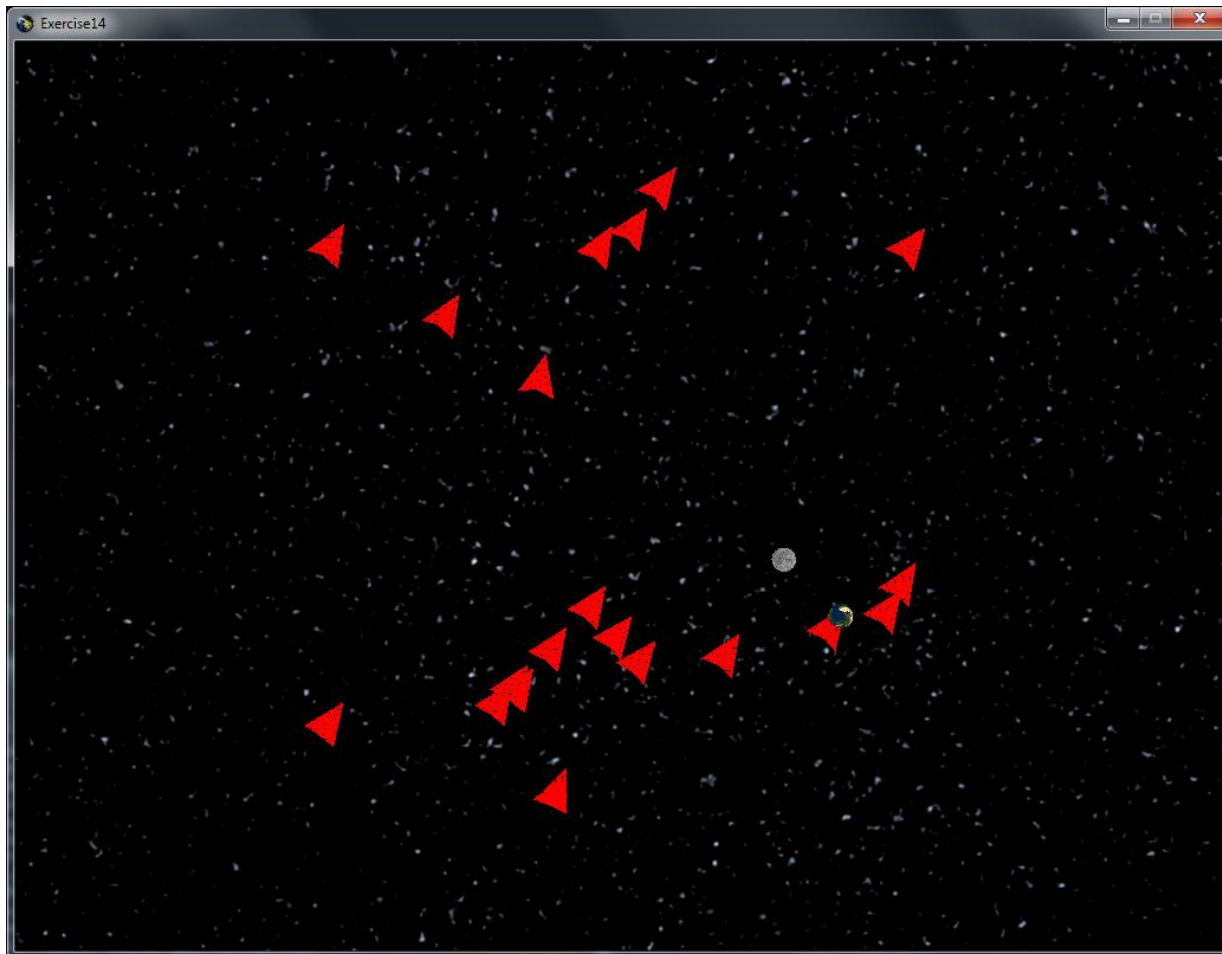
Exercise 14 - AI



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Exercise 14 – AI – Practical Tasks



Exercise 14 – AI – Practical Tasks

Flocking

- Implement Flocking as a blended steering behaviour involving
 - Separation
 - Cohesion
 - Alignment of velocity

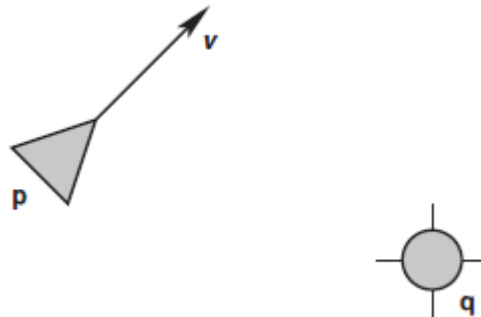
Decision Making

- Implement a finite state machine for the moon
 - If it is not close to the earth: Wander
 - If it is close to the earth: Seek

Exercise 14 – Theoretical Tasks

Task 2.1

- For a given configuration, give the steering values for
 - Seek
 - Flee
 - Pursuit
 - Evade



Exercise 14 – Theoretical Tasks

Task 2.2

- a) Can A^* be interrupted during its execution and continued at a later time? If yes, describe why. If not, give a counterexample.
- b) Are there problems to be expected if the next time the pathfinding continues is a certain amount of time (e.g. 1 second) later?
- c) If a) is true, does this make A^* an Anytime algorithm?

Exercise 14 – Theoretical Tasks

Task 2.3

Imagine a scene with many characters walking in random directions. In this scene, an AI LOD system is used. Nearby characters use collision avoidance, far away characters may freely interpenetrate.

What problems could arise if characters exploded when they collided?

Questions & Contact



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