BDI for Intelligent Agents in Computer Games

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ABSTRACT
With the emergence of complex computer games and advanced gaming hardware, possibilities for overcoming some of the deficiencies in traditional game AI are becoming feasible. These deficiencies include repetitive, predictable, and inhuman behaviour are caused by the reliance on simple reactive AI techniques. By using more sophisticated AI and agent techniques, we intend to overcome some of these problem areas. The aim of our research is to create new forms of intelligent characters (agents) that will exhibit human-like intelligence and provide more challenging and entertaining virtual opponents and team mates for computer games. We present here our prototype application that implements a BDI agent system within the 3D computer game Unreal Tournament via GameBots and JavaBots technology.

INTRODUCTION
With this continued popularity of computer games, game players are expecting new challenges with more sophisticated games and game playing experiences. Increases in processing power are now giving game developers opportunities to develop novel techniques to incorporate into their games. With graphics capabilities now reaching the point where game environments are becoming almost photorealistic, some of this power must become available for AI systems. Currently developers are looking for new and inventive ways to keep the game players entertained. The challenge is to produce artificial intelligence for computer games characters that can utilise the increased power afforded by improvements in games hardware, and make AI agents appear as human-like as possible so as to improve the game playing experience.

To produce agents capable of this behaviour in complex computer game environments we are using the Belief-Desire-Intention (BDI) model of agency (Bratman, 1987). In this model, agents are constructed using humanistic concepts such as goals to achieve, beliefs about the environment, and plans to achieve goals. Using BDI, we can simulate the decision making processes performed by humans, using the same information available to humans, in order to make the agent act in a human like way. It is expected that this will make computer game characters appear more realistic, and therefore, improve the experience of playing against artificial game characters. The paper is constructed as follows. In next section we outline the design of our system for integrating a BDI reasoning engine into the computer game Unreal Tournament, and detail the three layers of the system including deliberation, communication, and virtual environment. We follow this with our experimental results, including the initial implementation where agents interact with the game environment, exploring, attacking enemies, producing paths through the environment, following path, and building health by locating health packs. We conclude with our future aims, including the addition of a multi-agent layer for common goals amongst agents.

SYSTEM DESIGN
Our framework, (Figure 1), is constructed using three layers that integrate several tools available as open source projects and commercial applications. The layers are:

<table>
<thead>
<tr>
<th>Layer</th>
<th>Tool</th>
</tr>
</thead>
<tbody>
<tr>
<td>Intelligent Agent</td>
<td>Jadex</td>
</tr>
<tr>
<td>Game Environment</td>
<td>Unreal Tournament</td>
</tr>
<tr>
<td>Communication Layer</td>
<td>JavaBots/GameBots</td>
</tr>
</tbody>
</table>

Each layer is described in more detail below. It should be noted that our implementation is client-server based, and as such, requires an extra layer for external communication with the game engine. This means the agent is a combination of two separate entities. The first entity is situated within the Unreal Tournament game, and can be considered an avatar for our intelligent agent. The intelligent agent guides the avatar by sending commands over the network, and builds up a view of the environment by receiving perception messages. There are several reasons the system is implemented in this way; not least is the desire to implement our system in a modern commercial computer game. It is not possible to incorporate our AI directly within the game due to limitations in access to the engines source code. However, this architecture has the benefit of forcing the
agent to play the game in the same way as human players; based on sensor information. In addition, this type of architecture allows experimentation with human / agent teams in an extensible framework, and allows many agents to connect to the game server simultaneously.

**Intelligent Agent**

The intelligent agent layer consists of a reasoning system, a knowledge base of plans, and data structures for storing environment information. The reasoning system is developed using Jadex (Braubach et al, 2004). This is an agent platform that allows the creation of BDI agents in the Java programming environment i.e. it allows the creation of agents that use the mental attitudes of belief, desire, and intention to model human like reasoning processes. Agents are created via the specification of beliefs, goals, plans, events, and capabilities. Goals relate to the state an agent would like to achieve and can be of several types; achieve, maintenance, and perform. Achieve goals are used to perform an action, such as move to a waypoint. This type of goal can either succeed or fail. Maintain goals are used to monitor the agent, e.g. make sure health stays above 50. If the agent’s health drops below this level, then some action is triggered to rectify the situation. Perform goals are used to perform actions that are consistent with a state that don’t have a target state to reach e.g. ‘explore’ where an agent will continue to search an environment until some other event causes it to change behaviour. To accomplish goals, relevant plans are used. Plans are created via extending an abstract Plan class that allows messages to be sent between plans and agents. Agents are defined in an XML based Agent Definition File (ADF) were beliefs, goals and plans are linked by specifying their applicability via trigger statements. There may be many plans applicable to specific goals, therefore the plans applicability can be further reduced with clarifiers such as context conditions, that state that plans can only be created if certain belief conditions are met. Beliefs specify agents’ knowledge of the world, and are used to trigger goals, and plans success or failure. Beliefs can be any Java object. In our system, we have created belief objects that store the agent’s status, enemy locations, navigation info etc. Using this system, we have developed AI agents that have the ability to respond to environmental events, identify appropriate plans to handle the events, and execute those plans in a timely manner. While executing plans, the agents monitor the environment, and internal belief structures, in order to ensure plans are still relevant, and identify new opportunities.

**Virtual Environment**

Our intelligent agent connects to a game environment, and interacts with it through perceptions and performing actions. We have chosen to develop our system using the game engine Unreal Tournament, a three-dimensional, networked FPS computer game. The game includes several game types. These include Death Match; a free-for-all match with the winner decided by the highest number of frags achieved over a certain time period, Domination; where players compete to capture and defend domination points for a specified amount of time, Capture the flag; where players have to retrieve the opposition’s flag and bring it back to their base, while defending their own flag and Team Death Match; where teams work together to achieve the highest frag rate in a set time period. The game can be modified in several ways via an editor and a scripting language. New levels can be created in the graphical designer Unreal Edit. Game rules and physics can be modified via the scripting language Unreal Script. Lewis and Jacobson (2002) point out the benefits of this solution. The engine is inexpensive. The graphics rendering capability is superior to anything that can feasibly be created by small research groups. Also, the game logic is fully implemented i.e. the game comes complete with standard functions such as collision detection, physics systems, game maintenance etc. Other benefits include the large user base of players of the game. This gives access to domain experts for knowledge elicitation, and groups for evaluation of the completed system. The use of a game engine is not an ideal solution however. It would be better to develop a complete computer game where all functions are accessible at a source code level; a level that is unavailable through the use of game engines. Game engines impose limitations that are created by the game engine developers, which can cause problems, but is a neat solution to our requirements.

**Communication Layer**

To facilitate communication between the intelligent agent and the game world we have adopted the use of the dual middleware product of GameBots/JavaBots (Marshall et al, 2006). GameBots is an extension to Unreal Tournament that resides upon the game server. It is written in the scripting language provided by the Unreal developers; Unreal Script, and extends the basic AI and networking components shipped with the game.
GameBots allows external processes to access internal game AI functions through a network socket connection via the exchange of messages. For every game loop, agent perception data is sent as a synchronous message packet across the network to a JavaBots client. This information consists of currently visible navigation points, inventory items, and other visible agents. Status information is also sent including the agent’s health, current location, current weapon etc. Event messages, such as collision information, damage reports etc, are sent whenever they occur in the game via asynchronous messages. The communication is a two-way process, and GameBots also receives messages from the JavaBots client that consist of actions for the agent to perform. Actions include rotate, walk, run, shoot etc, and also more complex operations such as find path, which queries the navigation system of Unreal Tournament, and sends a list of navigation nodes back in the form of an asynchronous message. The client portion used by the intelligent agent is called JavaBots; a Java based system developed to connect to GameBots. It is an extensible API that contains example bots, visualisation applications, and a Bot Runner application that allows agents to be connected and visualised via a GUI interface. We have taken the original JavaBots project and removed the extra functionality to produce a very simple API that simply listens for messages, and sends instructions back to Unreal Tournament. We have removed the sample bots and Bot Runner classes, and instead use functions within Jadex to maintain the connection to the game.

![Figure 2: Navigation messaging system](image)

**Figure 2: Navigation messaging system**

Figure 2 shows an example of the messaging system in which GameBots sends a message block containing navigation point information to the intelligent agent, where it is recorded in a belief structure. In Unreal Tournament, the game agent has a set view cone of around 45 degrees. At any point in the game, the game agent is capable of observing a discrete portion of the game environment contained within this view cone, which is not occluded by walls. Figure 3 shows a typical Unreal Tournament view, the game agent can see inventory items (Heath Packs), and waypoint nodes. The waypoint nodes are either reachable, or unreachable. In the illustration, there is a gap between the game agent, and a platform containing the health packs; it can therefore see them, but cannot reach them directly. The navigation node to the right of the illustration is visible and reachable directly. Therefore, the agent can run directly to it. This information is grouped into a single message block, and sent to the intelligent BDI agent. The intelligent agent receives this message, parses it, and populates its belief sets. New nodes (nodes an agent has not seen before) are added to the list of known nodes. Nodes at the position of the agent are marked as ‘visited’ to indicate the agent has explored the position. Two other data sets are populated: visible nodes and reachable nodes. At each frame, these sets are cleared, and populated with the new data contained in the message i.e. only nodes that the agent can currently see are stored.

![Figure 3: Navigation messaging system](image)

**Figure 3: Navigation messaging system**

**EXPERIMENTAL RESULTS**

A prototype application has been developed that shows the potential of the architecture. The intelligent agents are able to connect to Unreal Tournament, build a 3D view of the environment, and navigate the world. The agents are also capable of some basic behaviour in the game. This includes exploring, navigating, hunting and escaping. The overriding goal of the agent is to maintain health above 50. Once health drops below this level, the agent attempts to disengage from a combat situation, and find health until its health has reached 90 or above. This behaviour is achieved via a maintenance goal that inhibits the ‘explore’ and ‘attack’ goals. When the agent is attempting to build its health, it will observe the environment to see if any health packs are currently visible and reachable. If it can see a health pack, it will move towards it and pick it up. If there are no visible health packs, it checks its memory to recall the location of the nearest health pack to its current location. At this point, it queries Unreal Tournament to find a path to the identified health pack from its current location, and then follow this path. If, on route, it spots another health pack, it will temporarily drop the goal of following the path in favour of collecting the new health pack. It will then
resume the goal of following the path (assuming it still requires health). If the agent encounters an enemy agent while following the path, it will retreat, and choose a path to an alternate health pack. Following paths and collecting health behaviour is created using achieving goals. At each stage of the goal, the agent can either succeed or fail. If a section of the plan fails, the agent is capable of retrying the goal, or dropping the goal and starting again at any stage. If the agent health is adequate, the agent will revert to the default behaviour of exploring the environment. It will query its belief set to find a list of reachable nodes, and check if any of them have not been explored before. If it finds an unexplored node, it will move towards it. If it cannot see an unexplored node, it will run to a random reachable node. This behaviour is created using a perform goal, and the agent will continue with this behaviour until some event causes it to drop the behaviour. An example of a condition where the agent will drop the goal is if the agent spots an enemy. When an enemy is spotted, the agent will engage in an attacking behaviour that includes firing its weapon, running towards the enemy, and jumping left and right until it has either killed the enemy, its health drops to a point where the maintenance condition forces it to try to escape, or it is killed. The behaviour of the agent is presently for illustrative purposes, and is not intended as a sophisticated behaviour system. However, it does prove that developing a more complex agent is possible with a combination of goals and plans within the BDI framework.

CONCLUSIONS AND FUTURE WORK

It has been proposed that the goal based, deliberative architecture, BDI, has the potential to produce more human like behaviour in computer game characters that will exhibit more realistic behaviour than can be accomplished in lower levels, and high level cooperative goals can be accomplished at higher levels.

REFERENCES


AUTHOR BIOGRAPHIES

NICHOLAS P. DAVIES was born in Wolverhampton, UK, where he studied Computer Science at the University of Wolverhampton, and obtained a First Class Degree in 2003. He is currently researching AI and Computer Games, and has completed the first two years of his PhD.