

Faculty of Science and Technology

BSc (Hons) Games Programming

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Shop Simulator Focused on AI

by

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**Contents i**

Faculty of Science and Technology ………………………………………………………1

Dissertation Declaration…………………………………………………………………….2

Acknowledgements………………………………………………………………………....3

Abstract……………………………………………………………………………………….5

Chapter 1: Introduction and Aims………………………………………………………….6

Introduction…………………………………………………………………………..7

Aims…………………………………………………………………………………..8

Objectives…………………………………………………………………………….9

Methodology………………………………………………………………………..10

Chapter 2: Background Theory…………………………………………………………..12

Chapter 3: Design, Development and Testing………………………………………….15

Initial coding………………………………………………………………………...16

Pathfinding………………………………………………………………………….16

AI Development……………………………………………………………………21

Coding and Development…………………………………………………………24

Testing………………………………………………………………………………27

Chapter 4: Results of Evaluation ………………………………………………………...37

Chapter 5: Discussion, Future Work and Final Conclusion …………………………..40

Comparison…………………………………………………………………………41

Future Work ………………………………………………………………………..42

Conclusion ………………………………………………………………………....43

Bibliography ……………………………………………………..…………………………44

Appendices ……………………………………………………………………………...…48

**Abstract i**

There are currently very few shop management games, and training simulators. Games currently available are very simplistic and hard to navigate. Working as a Customer Service Assistant (CSA) and working as a manager, or supervisor, are very different tasks and sometimes the jump from one to the other is too large. A big problem is the lack of ways to show how managing a shop works and how to be good at organizing and running a shop.

Creating a realistic, in-depth simulator/game, to show off the shop environment would be a great addition to a shop’s training system, and can be a great way for people to spend their time, while also developing their skills in important real-life ways, such as money management, customer satisfaction, team management etc.

The program created during this project will not be a full game, it will be a stripped down, simpler version to simulate and show off the AI systems.

Chapter 1:

I Introduction and Aims

**Introduction I**

Working in a shop, of any kind, can be overwhelming at first. There is such a huge range of personalities that cashiers and supervisors need to deal with daily. Usually, it is a learn-as-you-work experience since there is no manual, or cut and paste solution for dealing with difficult customers. The program developed throughout this project will be the baseline for a complex, and AI intensive simulation which can be used by shop employers to help train their employees by manufacturing complex and involved scenarios. The program will provide intelligent AI, which reacts realistically to the employees in the simulation, and a user can control an employee, so that they can try different solutions for dealing with difficult customers. The employees can be taught, and practice dealing with complicated customers and situations without it happening in real-life, which means that if they perform badly, the shop’s reputation does not get damaged and customers do not get upset or angry.

Upon initial research shrouding this type of educational program, it was discovered that there are very few, if not any, exceptional simulations that can be used in the way described above. One 3D version of a shop management game [1] is made for mobile only, and at first glance does not look too advanced or developed. Due to the lack of good research material being available for shop management games specifically, the next best place to take inspiration from would be simulation games with highly intelligent AI, such as RimWorld [2] and Prison Architect [3].



Figure 1: RimWorld Pathfinding Example

Using management games with intelligent AI as main points of inspiration, this project will be focused upon creating a basic program, with basic user interactions, but with a complex and advanced AI, which will show off the potential educational simulation that will be developed using this program as its baseline. The program will have a top-down/bird’s eye view camera set-up, as this allows the most amount of visibility of the entire shop, and helps the user get the best assessment of the AI systems.

**Aims I**

The final program that will be developed by the end of this project, will be used as an example, and proof of concept item, of a potential simulation full of complex and realistic AI decisions, which can be manipulated by the user to create scenarios which then can be used to train and teach employees about how to deal with an enormous amount of different shop interactions and scenarios.

This project’s AI systems will include:

* Pathfinding – Research will be done to find the best pathfinding algorithms, which will then be implemented to create a realistic looking pathfinding system.
* Job Decisions – The employees will need to decide which job should be done depending upon criteria such as customers in the store, and the time of day. Research will be done on games already using this system, such as RimWorld [2]
* Shopping Choices – The customers will need to know what products they are after and then look for them. If the customer finds the items they want, they need to decide on whether to purchase them based upon price, quality etc.
* Relationships – The characters in the program, employees and customers, will have different relationship levels. These levels will be used to determine if two characters want to engage in a conversation, and then their relationship will increase or decrease based upon that interaction, and the characters’ traits.

The base program system/UI implementation needs to be completed so that the user can interact with the world. The user should be able to play the program and get a feel for what a final, fully developed version of this program will be like. The main systems that will be developed and implemented are as follows:

* World/Tile Interactions.
* General Character Pathfinding and Interactions
* Employee Job Decisions
* Customer Decisions

All these features will be tested and feedback will be given in Alpha and Beta testing.

**Objectives I**

Most the project time will be on developing the AI for the characters, employees and customers, and will be the key focus of the project. Some time, but a limited amount, will be used to develop the backbone/outline of the game to demonstrate the AI in a realistic environment and so that users can get a good sense of the final program once it has been fully developed beyond this project’s deadline.

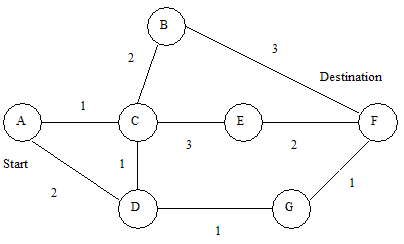
 There shall be investigations and research on different pathfinding techniques to find out which is best for the situation and to create the most realistic looking simulation. The A\* algorithm would be a good place to start but research will also be conducted on others such as Dijkstra’s algorithm.

Figure 2: Dijkstra’s Pathfinding Algorithm.

The backbone/outline of the program will be a top-down/bird’s-eye-view tile-based world which can be interacted with, and the player will be able to click on tiles/furniture/characters so that they can receive a realistic simulation of what the full game experience shall be like.

The player will not have a physical character to represent them in-game, but they will be able to interact with the world. Due to it strictly being a simulation, a pre-made scenario will be used as the demonstration, and will contain two employees, which will have randomly assigned traits, and the shop will be set up so the layout is the same each time. After that the spawning of the characters, their traits, and their shopping lists will all be randomised from a pre-built selection. This is so that if the user runs the simulation more than once, there is a very high chance that the way the simulation plays will be very different.

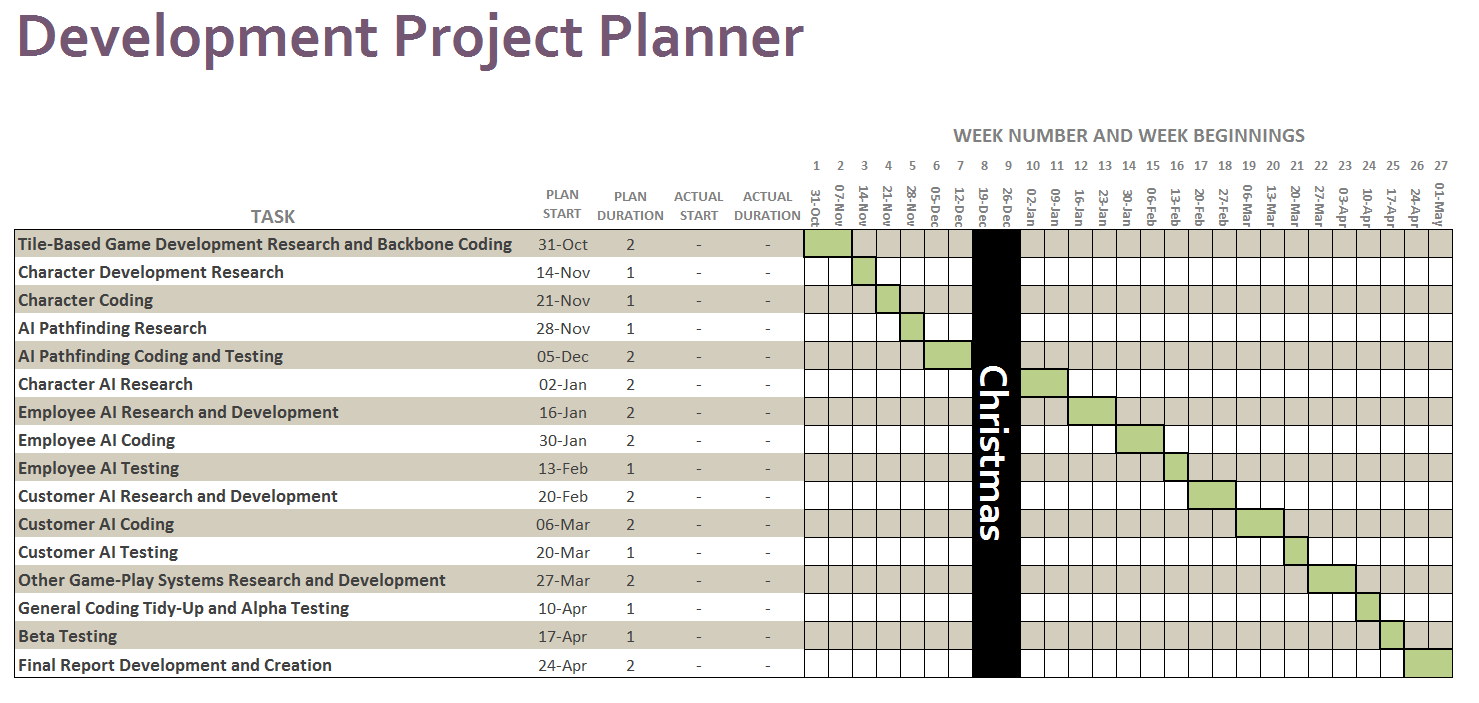
**Methodology I**

The main stages will be carried out using the waterfall method. The base/outline of the program will need to be the first thing that is developed. Investigations will be carried out to implement world/tile interactivity and then development of the tile features such as furniture will be conducted.

The second part will be the character development, and will be where the characters are made and given traits. This whole step is required second due to the world needing to exist first, but the character’s existence is needed before any AI can be developed.

The third stage is the largest and most important part. The AI will be the core of the project and will require the most amount of time and research. Within the AI section, 3 sub-sections are required: Pathfinding, Jobs, and Customer Decisions. The pathfinding will need to be done at the start of the AI system, due to this being essential for every character regardless of whether they are an employee or a customer. The employee AI and the customer AI are interchangeable as they are independent from each other. For this reason, the decision has been made to develop the employee AI second and the customer AI third. The employee AI is more important than the customer AI, and more complex, this is because the employee AI better demonstrates the final version of the game’s AI than the customer AI.

Research will be done throughout the development through finding documents about already developed management games, and tile-base games, and using their systems and information as a basis for the project’s development. General game-creation documents and reports will also be used when developing the pathfinding, and the other AI systems. Examples of good topics to focus on are fuzzy logic, finite-state-machines, and decision trees.

 The final stage of the project will be testing. There will of course be brief testing for each section throughout the project and on completion of each stage, which will all be documented appropriately, but also a final alpha testing stage at the end of the program development based upon the proposal, and other initial planning stages created throughout the project. After this, beta testing will be conducted where other people outside the project will be asked to play the game and give feedback on their thoughts and opinions. The game will be sent off to selected persons who are believed to be experienced in real-life shop environments, ideally in positions of management. Upon receiving the feedback and testing results, a final report will be created based upon them and comparisons to the initial predictions and alpha testing will be conducted.

Here is the initial project gantt chart. These deadlines will be kept to as much as possible. All work will be documented in the project journal and so the dates can be compared.

Chapter 2:

I Background Theory

Due to this program being a tile-based game, research will need to be done on general tile-based game development, as well as research on successful ways to create a backbone structure which will provide a secure and strong foundation for the rest of the more complex systems’ developments.

A peer reviewed literature paper Spuy (2010) [4] has been found and will be looked at as an example of a way to begin the development. It explains general tile creation, and how to use those tiles efficiently such as by creating an array of tiles to represent the world, and collision detection should be done on a tile by tile basis rather than object to object. Through this research, ideas have arisen about using tile sheets for the sprites used within the program, this would save memory and processing power due to one large sprite needed to be accessed instead of smaller sprites.

C# Unity3D Code was discovered (Project Porcupine, 2015) [5], which links to a set of tutorials for a tile-based base building game. After looking through the code, the baseline code is very useful, and its logic is similar to what is required for this project.

The project’s general set-up is that the visual aspects of the game are separate from the hidden game logic. It uses controllers to link the two together. These controllers are the only classes derived from Monobehaviour and so are the only ones that can use Unity’s GameObject functions and methods.

A class called WorldController is the center point of the code, and all the other classes, and instances can be reached via this class, either directly or through other classes, and it is the only class in the project that has been developed using a singleton system.

Other controllers such as the FurnitureSpriteController and the CharacterSpriteController are used to link the furniture and character logic with Unity’s GameObject logic, respectively. These classes collect data from standard C# classes and use that data to place and move GameObjects in Unity to where they need to be. There are other controllers but their logic will not be used in this project so will be skipped.

The rest of the classes are not derived from monobehaviour, and therefore they represent the hidden game logic. The main class is named World and contains all the general game data such as the world’s tiles, lists that contain references to all the characters and furniture in the world, and the jobQueue.

Some of the classes are models and represent templates for different things in the game such as characters, furniture, tiles etc.

There is a set of classes that are used for the program’s pathfinding system, which uses the A\* pathfinding algorithm. This algorithm is generally known as the best algorithm as proven in a DePaul University technical report (Krishnaswamy, 2009) [6]. Research will be done at a later point to compare A\* with other algorithms and a decision will be made on the best algorithm for this project. If A\* is used for the development, then the A\* logic shown will be used as it is already integrated with the rest of the logic.

The Project Porcupine code is licensed as shown in Appendix 2.

A major influence on this project is the tile-based colony management game RimWorld (2013) [2], because of this, here is an insight into its systems and similarities with this project.

RimWorld allows the player to indirectly control characters in-game known as colonists. The player assigns jobs and the colonists fulfil those jobs based upon their own personal job priority lists, also set by the player. The colonists have moods, relationships, and thoughts. They also have traits which affect their speed, moods, relationships, and thoughts. The player can place a range of furniture, which must be built by the colonists using inventory. Due to the player not controlling the colonists directly, RimWorld has a very complex and integrated AI system which is used to make the characters seem life-like by changing their moods and thoughts.

The similarities with RimWorld and this project are as follows: uncontrollable characters, traits, relationships, and conversations. In RimWorld these systems are much more advanced and developed than what will be created in this project, however, the general idea and backbone of them will be used as inspiration throughout this project.

Chapter 3:

Design, Development and Testing

I g

**Initial Coding I**

Using the Project Porcupine code as influence, the program began development with a WorldController, a World class, and a Tile class. The WorldController and World class are the center points of the rest of the code. After this the MouseController was developed, along with the Furniture class, FurnitureSpriteController, and the Furniture Actions class. The furniture actions class represents the Update function for furniture, however, due to not all furniture needing an update function, this class allows specific types of updates and only when the furniture needs it, such as when a door opens, or a piece of furniture moves.

All this coding was backbone, basic coding and put into place the basic functions needed to create these in-game objects which were developed further into the project.

Research was done on general character coding, however not many useful publications were found. Research was then done on RimWorld’s [2] character systems, such as their traits, moods and thoughts. Using this system, and the Project Porcupine character code, this project’s character code was started, and the Character class and the CharacterSpriteController class were created. These were again developed in such a way that the basics were completed, with lots of room and help for future development in several ways, such as pathfinding which was the next step in development.

**Pathfinding I**

Many pathfinding literatures were reviewed and a few of them were focused on. By briefly explaining the major literatures found, reasons for the pathfinding choice made should become apparent.

*Hybrid Pathfinding in StarCraft (2015)*

Hybrid Pathfinding in StarCraft (2015) [7] describes the use of A\* algorithms for long-range pathfinding, but also the use of potential fields for short-range and combat based pathfinding. It evaluates the possibility of replacing the potential fields part of the hybrid pathfinding with a system based upon flocking algorithms. The work of Olssen (P. M. Olsson, 2008) [8] addresses the issue of changes in the pathfinding graph due to the construction and deconstruction of buildings. Koenig and Likachev (S. Koenig, 2004) [9] (S. Koenig and M. Likachev, 2006) [10] have made contributions to the field with their work on real-time A\* pathfinding. Both works were considered useful for this project since some of the furniture can move.

*Direction Based Heuristic Pathfinding in Video Games (2015)*

Direction Based Heuristic Pathfinding in Video Games (2015) [11] explains a general pathfinding concept and then talks about the two primary problems which are to find a path between two nodes in a graph, and to find the optimal shortest path [12].

It then explains A\* pathfinding. A\* is a generic search algorithm that can be used to find solutions to many problems, pathfinding is just one of them. A\* is the most popular and widely used AI pathfinding algorithm proposed by Hart, Nilsson and Raphael in 1967. Due to its simplicity, A\* is almost always the search method of choice. This is because A\* is guaranteed to find the shortest path on a graph.

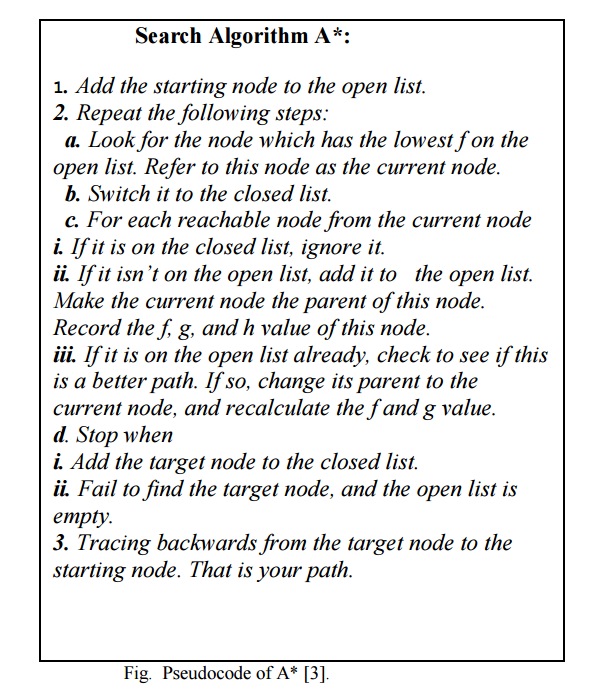
A problem with A\* is that a shortest path on a graph is not equivalent to the shortest path in the continuous environment. Another issue is when the map size is significantly large, it cannot find a minimum path to the goal state in a limited amount of time. For larger maps, A\* uses memory extensively. A\* uses a heuristic to improve on the behaviour to Dijkstra’s algorithm.

Figure 3: A\*’s Pseudo Code.

Next the publication goes on to talk about general heuristic features. The trade-off between speed and accuracy can be exploited to create a good balance. One way to construct an exact heuristic is to precompute the length of the shortest path between every pair of nodes. This is not feasible for most game maps. However, there any many ways to approximate this heuristic:

1. Fit a coarse grid on top of the fine grid. Precompute the shortest path between any pair of coarse grid locations.
2. Precompute the shortest path between any pair of waypoints. This is a generalization of the coarse grid approach.

Links between the publication and the project were then made. The downside of the large maps being slow does not matter in this situation due to the map in the program only being 2030 tiles high and wide. Also, only a maximum of perhaps 10 characters will be present in the world at any time, so the high memory demand is not a problem unless the world is expanded and more characters were present on the map, at which point A\* may become less viable

*Uninformed Multigoal Pathfinding on Grid Maps (2014)*

Uninformed Multigoal Pathfinding on Grid Maps (2014) [13] is a publication focused on implementing multigoal logic into standard pathfinding algorithms.

There are two classifications of pathfinding algorithms; informed and uninformed. Informed involves the use of a heuristic function [14] to estimate the location of the goal. The direction of the pathfinding is guided towards the estimate, making informed searches typically faster than uninformed searches, but can be less optimal. Uninformed algorithms have been developed using different pathfinding models, such as iterative-deepening searches [15], boundary searches [16], bidirectional searches [17] and multigoal searches [18] [19].

The publication tested two algorithms: Dijkstra’s Shortest Path Multigoal Algorithm with Multigoal Boundary Iterative-Deepening Depth-First Search. The results showed that when multiply goals are required in the game, using the multigoal algorithm significantly decreased pathfinding times. There is an exponential increase in pathfinding times recorded by single-goal algorithms to search for multiple goals on open maps. This project does not require multigoal pathfinding due to the nature of the AI system that was implemented, however, this is very interesting and will defiantly be considered for the future if a more complex AI is developed with the ability to stack destinations for more efficient pathfinding.

*Pathfinding in Partially Explored Games Environments (2014)*

Pathfinding in Partially Explored Games Environments (2014) [20] talks about a very big problem with pathfinding and the fact that usually the characters in the game know everything about the layout of the map, and because of this, they can create perfect routes to their destination which is often unrealistic. It proposes a system using a hybrid approach [21] that allows characters to path-find as normal, but they do not know everything about the map. This system would require some way for the characters to ‘learn’ about the environment with, for example, line-of-sight. The character will detect changes in their environment as they move, and adapt and change their pathing accordingly. For example, they could turn a corner assuming they can walk down it, and as they carried on walking they realised that it was a dead end and be forced to turn around and come back, which is much more realistic than them knowing the dead end was there without seeing it.

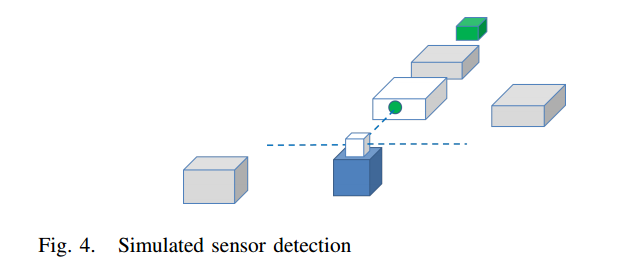
 The publication tests this system in Unity3D using ray casting as a sensor to ‘see’ the environment. The raycast works in a similar way to a sonar or LIDAR based system in that a single point is projected outwards. If the raycast intersects with an object, the system records the collision point. The character will be able to see any object the raycast intersects with.

Figure 4: Raycast ‘Sight’ Example.

This publication is very interesting and very relevant for this project. Having our characters not be able to see the entire map and require them to learn as they move would increase the realism and allow a move in-depth and advanced AI. The problem with this however, is the requirement for line-of-sight, as well as each character having their own perception of the world. Both these things require more coding, and additional systems be put in place for them to work successfully. This partially explored environments system was not implemented due to time constraints but the idea behind it was a huge possibility and will be one of the first additions for future work.

After pathfinding research was completed, the A\* pathfinding algorithm was chosen as the best for this project. Thought was given to the partially explored environments system, and it was decided that if enough time was available near the end of the project, then that system would be implemented along with line-of-sight system. Due to Project Porcupine [5] already having a successful A\* pathfinding system, it was looked at and its basic outline was used. The pathfinding system was tweaked later in the project when moveable furniture was fully implemented.

The pathfinding uses a weighted system which allows furniture, and others things if required, to have a weight, and moving through a tile with furniture will change the time it takes to move through it, and therefore sometimes moving around the tile can be quicker.

Tiles contain one of three ‘enterability’ states: YES, NO, and SOON. YES means that the character can walk onto the tile, NO means that the tile is now occupied and the character will re-evaluate their path, SOON means that currently the tile cannot be entered but if they character waits then it should be able to be moved soon, such as with a tile with a door. The character will wait, inform the tile they are waiting, and then furniture or character on that tile will perform their own logic for when a character wants to enter their tile.

Once this code was developed it needed to be tested. A basic character was created and a ‘GoTo’ button was created to allow the user to click on a tile, which then set the character’s destination tile to the clicked tile. Simple walking with no obstacles was tested, as well as testing with a single line of walls, and finally a closed off room with only a door to leave by. The character correctly stood next to the door until the door opened using its furniture actions, and when it was open the tile had an ‘enterability’ of YES instead of soon, and the character carried on their path. When the character left the tile, the ‘enterability’ changed back to SOON and the door closed.

**AI Development I**

The next section of the research and development was the most intensive and important part. In a similar way to the pathfinding, several publications were looked at, but only a few were focused on.

*Emotion-Based Synthetic Characters in Games (2008)*

Emotion-Based Synthetic Characters in Games (2008) [22] talks about emotions for characters in video games, and proposes a model using fuzzy logic to create facial expressions and body positions to represent emotions, and other characters used those emotions to learn about the character. It used basic emotions such as happiness, anger, fear, sadness etc. It talks also about keeping these moods around for a given amount of time, and so these moods can stack, or oppose each other if they happen at the same time.

The use of facial expressions and body positions is not relevant for this project; however, the idea of fuzzy logic is something that needs to be thought about and is used slightly in the moods and conversation responses in the characters in the program. The idea of keeping moods around was not relevant in this project as the characters don’t exactly have moods, they instead change their relationships with each other. The idea of them having moods, and the mood changes sticking around for a time is interesting and will be considered for future work. Having a bad conversation with someone could cause a character to be angry and may cause them to be more irritable to other characters, which is realistic.

*Towards the Design of Human-Like FPS NPC using Pheromone Maps (2013)*

This publication [23] begins by talking about different ways to model decisions for NPCs such as state machines, fuzzy logic, behaviour trees etc. and talks about their advantages and disadvantages [24]. The paper goes on to talk about different ways that characters can choose where to go or what to do. The example given was Quake 3. It talked about how if a certain location has had a lot of fire, and was being attacked more than other locations, then that logic should be worked into the AI to allow more solders to go to that location to defend, as they should know it is more likely to get overrun.

This logic can easily be used in a shop environment, but was not implemented into this project due to it not needing to be due to the very short space of time the simulation would run for. However, the idea behind it would be if it should be a normal day, with a normal number of customers, but today for some reason it was much busier due to some kind of event, then employees should use that information and perhaps not go off and do their normal jobs and instead stay near the tills or around the shop front to help the increased number of customers.

*Adaptive Behaviour Control Model of Non Player Character (2013)*

This paper [25] talks about a model using fuzzy logic to help define the behaviour in a non-player character. It takes the basic fuzzy logic system and creates an adaptive version. This allows the character to learn in an independent way from the other characters, meaning that the same event will cause different characters to react differently even if other things such as traits are the same, as the fuzzy logic values changes the more the character interacts with the world.

This logic is not useful in this project due to the limited time the simulation is run but is something very advanced and will create a very interesting AI for future work on this project.

*Component-Based Hierarchical State Machine – A Reusable and Flexible Game AI Technology (2011)*

This publication [26] begins by talking about general finite state machines (FSMs) and talks about how these are the most common in video games [27] [28]. These are used to model the behaviour of computer-controlled game objects to make the NPCs react to game events in the most natural and intelligent way possible. FSMs consist of a set of states which represent actions or behaviours and a collection of transitions from state to state.

It then proposes a new technique called Component-Based Hierarchical State Machine (CSHSM) which introduces software component techniques to the implementation of hierarchical state machines. It overcomes the limitations of Object Oriented Hierarchical State Machines (OOHSM) and has three significant advantages:

* Compile Time Composability. At compile time, new high-level and complex states are created.
* Design Time Configurability. CBHSMs are no longer completely fixed at compile time by programmers, and so can be configured at design time according to the game’s high-level design.
* Run Time Flexibility. A run time, CBHSMs can be reconfigured as needed. This feature frees CBHSMs from fixed hierarchical structure and greatly improves their flexibility and adaptability to the changing game environment.

The paper goes on to talk about standard OOHSMs and their limitations such as the defective ‘white-box reuse’ [29], and talks about decoupling the state and the context, but that requires an Asynchronous Event-Driven System (AEDS). It explains the process of establishing the AEDS and the implementation of the CBHSM system.

This paper, although useful for considering an advanced FSM system, is not relevant for this project. The complexity is unnecessary due to the simple simulation design the of program and so was not used for the development of the AI. For future development, where the scenarios are not predetermined, this may be something that should be implemented.

*Game Coding Complete (2013)*

This book contains a huge amount of content about all aspects of game creation. Looking in the AI section provided lots of insight into the best and most known techniques such as: Finite State Machines, Decision Trees and Fuzzy Logic. Each one is shown and explained in detail and allowed a good source of knowledge to make an informed and correct decision into which approach was best.

With this book’s large amount of information and the rest of the AI research, the chosen system was the use of a finite state machine (FSM) for the employee job code and the customer code. The different jobs that the employee needs to do works well with the different states needed. The FSM has primary states and secondary states. The primary states represent the major jobs required, and the secondary states are smaller tasks that the major tasks are made up of, for examples moving position or emptying a stockcage.

The characters’ relationships, interactions and traits will use simple fuzzy logic to add randomness to the simulation so that it isn’t the same each time, which would allow the user to keep repeating the simulation with different results to get a good grasp of the realism.

**Coding and Development I**

The AI development began with creating employees that inherit from the Character class. These will use certain conditions to decide on the job they need to do. The FSM first begin in the Job class, but this was quickly abandoned as the employee didn’t have complete control over the machine, and the FSM now is implemented into the employee class. This means the machine acts as the employee’s ‘brain’ which is realistic. All characters have a DoThink() function but the logic will be overridden for the employees and the customers independently since they are in the shop for different reasons. The employee has a DoJob() function which is different from the DoThink() function as the DoJob() function only runs when the employee is not moving and needs to perform an action on the tile where they are.

After there were simple jobs, and the employees had simple logic, stock was added to the game. Stock can be added to furniture and characters can pick up stock. Character stock logic was then developed so that characters can try to pick up stock from furniture. They can try to pick up any stock from the furniture, a specific type of stock based upon the stock ID, or an exact piece of stock. Stock is a model class that acts as a template just like the furniture class or the character class.

Next was the advancement of the selection logic, which allows the user to select furniture and see what stock is on that furniture, this was done at this point due to the need for it when it came to testing the employee’s take stock code.

Once that was working successfully, then all the other furniture was implemented. The only moveable furniture is the Trolley, so that the employees can use the Trolley to take stock to and from the warehouse.

A very complex and important part of the development process was the moveable furniture. This caused the most amount of problems due to furniture needing to be moved by a character rather than independently, also the way the furniture is implemented it does not work well if it needs to be moved from tile to tile.

Characters can set a piece of furniture as their own piece, and then move it if they require it to be moved. This was simple when just moving to another tile, but became complicated when a Trolley was in their way and needed to move it just to go around it. This links closely with AI since there are several ways the character could choose to move the trolley. Initially, the character would just push the Trolley until they reached their destination, but this quickly because unrealistic as it may make more sense for the character to move the trolley to the side and walk through the tile where the trolley used to be. This was when a small bit of pathfinding was implemented. The character would perform a floodfill from the trolley’s tile. This would return the nearest empty tile away from the trolley that is not in the character’s path to get to their task’s location. The character would then move to that tile with the trolley, once there they would return to the task they were performing.

After testing, this again became unrealistic as often the floodfill would return the tile behind the character, at which point the character would reverse and then once they moved back, the trolley would be again in their way. This would repeat until the character literally trapped themselves in the corner of the store. The problem is that the character wasn’t thinking about where they were going to be once they did move. This sparked the creation of a complex function which used recursion to predict where the character and trolley would be if the character moved to the tile found by the floodfill. If the trolley would still be in the character’s way, the process would repeat, and a new tile would be found, once again reclusively checking where the character and trolley would be. This is all being done without the character moving anywhere. The recursion stops when either the trolley will reach a tile that will not be in the character’s path, or the character gets trapped. If the character won’t get trapped, then the character proceeds as normal, but if they would get trapped, then that tile they would have moved to gets flagged as an invalid tile, and the process repeats until a valid tile is found. After testing this, it was found that it is a very realistic AI which successfully moves a blocking trolley somewhere logically and allows the character to pass.

Once the trolley movement logic was in place, the rest of the employee AI logic was completed. Buttons were developed to change the speed of the simulation so that already tested parts could be moved through quickly, and other parts could be tested without the wait. Once the employee code was close to complete, the customer code was added.

The customer FSM is much simpler and more straightforward than the employee one due to the simpler tasks the customers need to complete. The FSM still contains primary and secondary states, but they are not as plentiful, and did not take as much time to develop and plan. The customers spawn, and then are given a list of items they require, which they will then go and find. They begin from the start of the store and look in each shelf to find the item they need. Once they have found all their items, they floodfill from their current tile, until they find a tile marked as queue. All queue tiles have a number; the number is how far away that queue tile is from the checkout. If a customer is in the queue, and there is a queue tile adjacent to them, and the queue number is less than theirs, they check to see if it is free, and if it is they move there. Once at queue tile 1, they put their stock on the checkout, and the employee scans the stock, and the customer then picks the scanned items back up and pays and leaves when they have them all.

This is the point at which the line-of-sight/partially explored environment code would make the most sense. The customers may not have been to the store before and it would be realistic if they got confused and move around the store a lot, and sometimes missed shelves that they couldn’t see.

After the customer logic was completed, the employee logic was completed. It had to happen in this order as some employee logic, such as scanning and finishing a transaction, relied upon the customer AI being completed.

After this, general character AI code was added which caused the characters to not occupy the same tile, and this meant that when moving into a tile, in the same way a door causing the character to stop and wait, the character will stop and wait if the tile contains a character already. This would not always work however, as if two characters wanted to move to each other’s tiles, they would stop forever, and would never be able to move. This caused some logic to be added which meant that the characters would wait for five seconds, after which the character would try to find another way to their destination without going through the tile that is causing the problem. Sometimes there is no other way, and so they need to ask the other character to move so they can pass. This is the first character interaction required and so before this was done, the interaction logic needed to be developed.

Characters can request interactions with characters next to them, and could be rejected if the character receiving the request is busy, or doesn’t want to talk. This allowed the asking to move logic to be completed. At which point relationships and traits were added. Traits change the way that characters respond to other characters and there are positive, such as understanding, and negative traits, such as lazy. After this, character selection was added to allow the user to select characters and see the stock they are carrying, and the traits they have. If a character is interacting then that can also be seen, along with the character they are interacting with, and the relationship level of that character.

The rest of the development was small additions such as a message pop-up. The scenario used in the testing was developed, with two employees, and the beginning and finish screens were created.

**Testing I**

Once the scenario was created, and the program was ready, testing began to find bugs in the game. The simulation was repeated numerous times and various bugs were found and dealt with. Unfortunately, a couple of significant bugs were found, and due to the short time remaining on the project, these could not be fixed fully. Instead, to make sure the simulation was ready for release to additional testers, the bug needed to be fixed using bad coding practises, which will be the first things looked at in the future to make sure the bad coding practises do not exist.

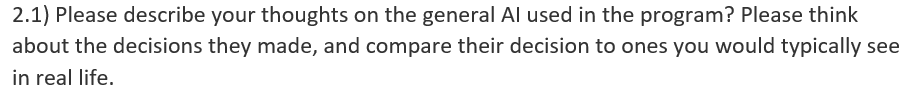
Once the simulation was run more than a dozen times, and no additional bugs were found, the alpha testing began. This was done by an in-house tester, James Jamieson, who ran the simulation and completed the same set of questions which would be given to the beta testers.

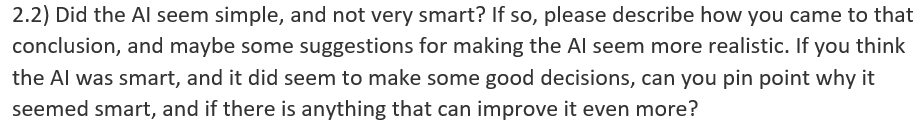
The questions were broken down into different categories. This helps the tester think about how to answer the question be thinking only about a certain aspect of the AI, and not the AI as a whole unless the question asked for it.

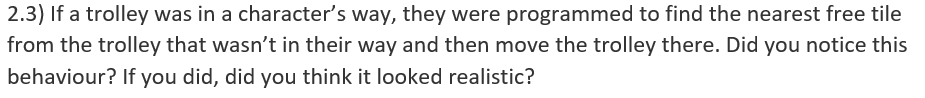
The first section was about the general simulation concept. This section allowed the tester to give their feedback on their thoughts about the general idea of having a program to help train employees. This means that later, if their answers seemed odd it might be down to them not agreeing with the concept in the first place, which of course if good and helpful feedback.



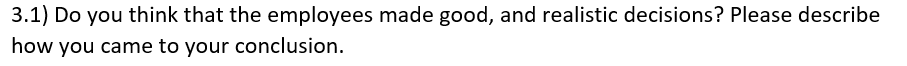
The second section is about the general AI as a whole. This allowed the tester to give their opinion on the general realism of the simulation and not think about individual characters yet. This was the basic idea behind question 2.1. Question 2.2 was looking at the intelligence of the AI in their decision making, and allowed the tester to perhaps explain why they thought it didn’t seem smart, if that’s what they thought. They were also able to give opinions on improvements which can be taken in the future. Question 2.3 talks about a general interaction with the trolley. Since the moving trolley logic links so closely with the general AI logic, this question allowed the tester to give their opinion on the interaction and talk about whether it was realistic or unrealistic.

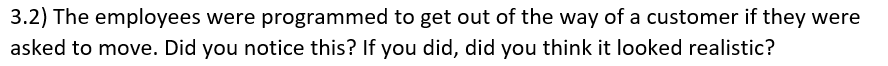






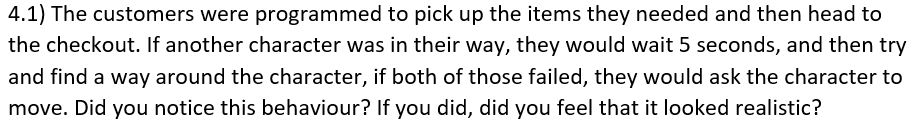
The third section is the employee specific AI. Question 3.1 talks about the general intelligence of the employee AI, and asked the tester to give their opinion and an example of how they came to their conclusion. Question 3.2 talked about the interaction when an employee is in a customer’s way, and asks the tester if they noticed the interaction and if they did to comment on whether it looked realistic. This is important since this is a realistic part of a real-life shop environment. The final question, 3.3, asks for their feedback on what they thought could be added to make the employees more intelligent.





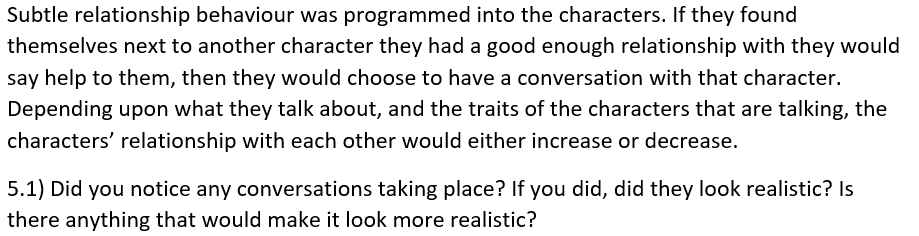


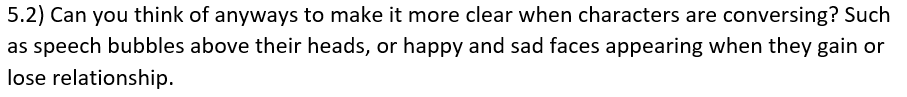
Section four was on the customer AI. Question 4.1 asks the tester about the customer AI logic of waiting for another character to move, then finding a way around, and then asking for that character to move and whether they though it looked realistic. Question 4.2 was a general question about the tester opinions in ways to improve the customer AI.

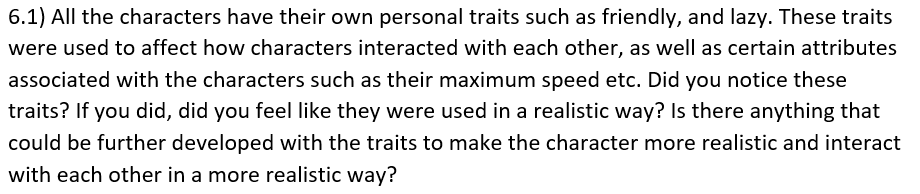


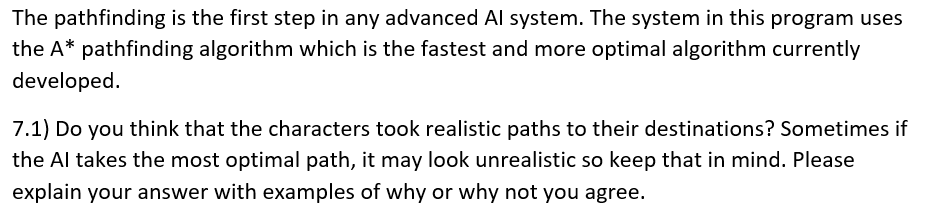


Section five was focused on the relationship AI logic. Question 5.1 asks the tester if they noticed any conversations taking place, and then if it looked realistic. This is important since this could be a UI issue, and this needs to be addressed in the future if the interactions weren’t noticed. The next question, 5.2, asks if there are any ways to make it more clear when characters are conversing. The question lists two ways so that the tester can get an idea of what sort of things might be good.

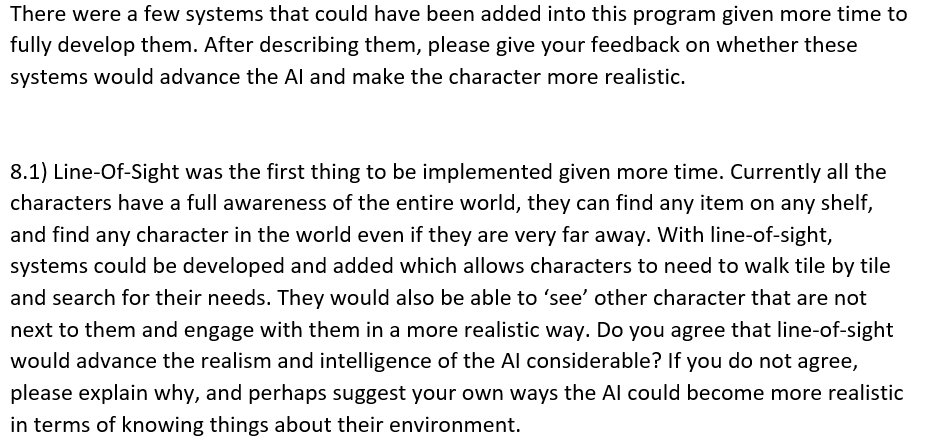


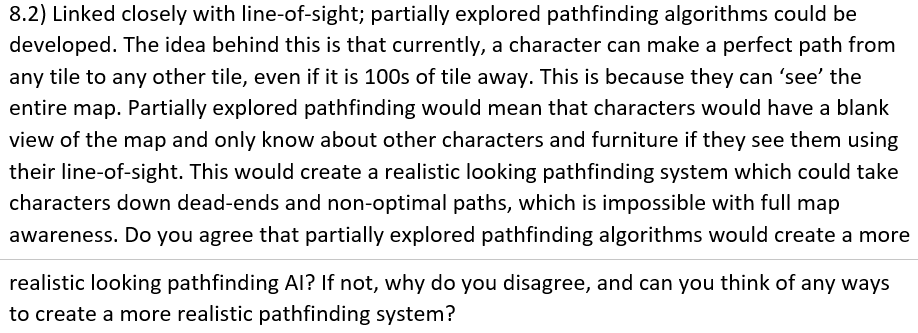


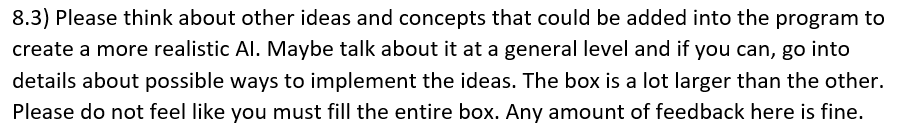
 The sixth section was on the characters’ traits. 6.1 asks if the tester if they noticed these traits being used, and if it seemed realistic. The traits were the less obvious part of the AI and so this question allowed the tester to realise it was there, and to give an opinion of whether it added anything to the simulation.

 The next section was about the pathfinding. If explains the system and then 7.1 asked the tester if the pathfinding looked realistic, it also said to keep in mind that perfect paths are not always realistic.

The final section was added to allow the tester to give their opinions on additional features and ways of increasing the AI’s realism. Questions 8.1 and 8.2 described already planned additions, and asks for the tester’s opinions. This is important as if the testers in general think that idea won’t add anything then it will probably not be added in the future. The final question, 8.3, is a general and very open question where the tester can give their ideas on additions to the program. The box is larger and encourages the tester to go into detail about their idea. This is a vital part of the analysis since this can potential give some great ideas for future work that were not thought.







The alpha testing analysis will be gone through during chapter 4 since that analysis will be compared to the beta analysis and the results will be joined to give a final conclusion. The general response from the beta feedback will be summed up now, but will still be gone through question by question. Each response can be seen in full as appendices.

* 1. There was a mixed response in regards to the concept as a training tool, however, most responses agreed that with more development and additions this program can be used as a baseline for future development of a training tool. An issue raised was the fact that this simulation does not help to teach employees about general interactions such as not being shy, and talking confidently to customers. Another point was that this simulation is good since it will help teach employees without the risk of doing something wrong, and not hurting the shop’s reputation; which was one of the originals aims of the program. A suggestion of a VR version was made to immerse the user into the simulation. This was very interested and something not considered before.
  2. A response suggested this be used as a tool for general job training, and will help people learn about other job environments such as offices.
  3. General thoughts were that the AI was realistic. There were some good and bad points. Good points:
  + The characters chose a wide variety of products instead of lots of the same product.
  + The connection between the customers and the employees was excellent and the cashier never stopped serving to keep customer satisfaction.
  + The employees got out of the way and showed courtesy to the customers.
  + The negotiations between characters when bunched around a trolley worked well.

Bad points:

* + The AI seemed confusing and was hard to judge.
  + The customers seemed to keep choosing the same items, but this may be down to the lack of items available in the simulation.
  + The customers all seemed to follow the same path.
  + The customers never forgot what they needed and never went back on themselves.
  1. The response was split. Some thought it looked good, and others thought it looked robotic. The customers seemed simpler than the employees. The customers should skip shelves they know are not required, and maybe forget things and go back on themselves to get missed items.
  2. The interaction was noticed for the most part. The customers should get annoyed if they need to move a trolley that is in their way. For the most part, characters never got stuck, and moved the trolley to realistic places. Some interactions took too long.

3.1 The employee AI looked fairly realistic. They restocked large bulks of items instead of small amounts which is good. The interactions between employees and customers worked well. The important parts of the employees’ jobs were highlighted which is good for a training tool. Sometimes another employee was required at the shop front but did not go and help, and a few times it was hard to notice what the employees were doing. Occasionally the employees did not interact with a customer when they should have done.

3.2 The general opinion was that it was noticed and seemed realistic. Sometimes it was hard to notice, but the shop always flowed well and so this interaction may have just been missed. Employees would often take too long to move, and should be much quicker.

3.3 Here is a list of different responses:

* Customers could ask employees to show them around the store to find items.
* Adding damages to stock, and the employees need to sort of damaged stock as it cannot be sold.
* Duties off the shop floor such as phone calls.
* Employees could greet customers into the store if they are near the entrance.
* Not enough use of interactions and traits.
* The employees could help on the checkout more.
* Check shelves before loading trolleys.

4.1 In general the AI seemed realistic. The employees should move out of the way of customers without being asked instead of the customer asking for the employee to move. Sometimes people won’t wait and barge through, which was not shown. The time for the interactions was too long. A few of the responses didn’t notice the interaction.

4.2 Here is a list of the responses:

* The customers should have a budget and react to high prices or if they run out of budget for the items.
* The customers should not always take the same path.
* Ask employees for stock not on the shop floor.

5.1 There was a mixed bag between the responses. Some saw the interaction and thought it looked realistic, some did not think it was realistic and some did not notice the interaction. The cashier and customer interaction seemed good especially with the relationship level going up and down. A suggestion was made for good and bad moods that would change depending on recent interactions. Another suggestion was made which said that if two customers know each other they may move around the shop together while talking.

5.2 Both these additions were responded to as positive, perhaps with the addition of an emoji. Other suggestions were coloured numbers representing the relationship change such as “+10” or “-5”, and the addition of voices when characters were talking.

6.1 the general response was that the traits were not noticed, or didn’t do enough to warranty being in the simulation. Someone wrote that some traits should be visible as they may be seen by others, but some shouldn’t be and been hidden traits. The traits should affect the employees’ work ethics.

7.1 The general pathfinding was good as the shortest routes were taken, but sometimes characters were too robotic. Suggestions included characters backtracking to find other items they missed, and that the most optimal route isn’t always the most realistic.

8.1 Line-of-sight would be a great addition. It would allow older and newer customers to act differently. Customers could walk past a shelf they need but not see the item they need, which is realistic. Characters could have conversations with people not right next to them.

8.2 This would be a great addition, and would allow much more room for advanced development which is good if this should become a training tool. Old customers would already know the layout of the store which is realistic. Having dead-ends that the customers walk down would look realistic.

8.3 Here is listed all the suggestions from the responses:

* More interaction logic with traits.
* Events occurring such as spillages and breakages which the employees will need to go and sort out.
* Someone wrote that proper shop skills cannot be gained using this simulation and so they couldn’t make any suggestions.
* More situations such as phone calls, and angry customers.
* Employees becoming tired.
* Generally, more input from the user to train them.
* Shop lifters.
* Customers returning items.
* Health and safety, like when there are spillages.
* Fire alarms.
* Disabled customer help.
* Deliveries.
* Rush of customers.

There was a mixed bag of responses which was great since it gave the opportunity for all kinds of responses and suggestions. The beta responses will be looked at and compared to the alpha response in the next chapter.

Chapter 4:

I Results of Evaluation

The range of responses were positive and allowed plenty of room for constructive criticism and suggestions for improvements. The responses in general matched the alpha testing, which confirms that the AI was fairly smart, however it did prove to be a highly intelligent AI that would be required for a fully completed simulation. The program can be used as a good baseline since there are hints of all kinds of AI which can be developed further and worked on to create a very powerful simulation. The point of real-life skills such as people skills not being learnt in this simulation was brought up and is a great point to make. Regardless of the level of intelligence in the characters, the user can never actually worry about the decisions they make and so their skills can never fully be taught just in the program. The suggestion of adding VR into the program was very interested and was not considered before. This would allow a much more real feeling for the user, and may help to increase the maximum help the program can provide when training customers.

It was agreed that the employees were much more realistic than the customers, which makes sense since their system is a lot more advanced and they need to make more complex decisions, which seemed to add to the realism. The employees should be given more jobs, such as phone calls to answer, as well as more customer interactions such as complaints and customers not finding items. These were great suggestions and would help the advancement of the AI and the realism. Often the employees seemed too robotic and too set on their tasks, which could be good as a training tool to help future employees work harder when they do start in real-life, but was also bad in terms of the realism.

The customers were generally talked about as robotic and seemed to follow the same path. They should sometimes forget what they need and backtrack, but also know exactly what they want and where to get it, and so go straight to the correct location in the shop to pick up the items. More interactions could be made between customers to allow them to move while talking for example. A more completed program with more items and shopping lists may add to the realism by itself. The trolley interactions seemed good but sometimes took too long and the customer moved the trolley to odd places.

Character interactions could be developed more to make it easier to notice, but in general were good and realistic. The addition of moods which linger with the character would be a nice development. The problem with the interactions was the lack of messages to the user that there were interactions, which is a shame, but a good point for development. The suggestion of numbers to represent relationship changes is a good idea, and the addition of voices, since this would cause busy shops to be louder, which is realistic.

The consensus on the traits were that they added basically nothing to the simulation, except the point that they were there. They should do more things, or be less of a ‘big’ feature. They are not important for training except perhaps to train employees to read the mood of a customer which is sometimes important.

The pathfinding in general seems robotic due to the characters never backtracking or changing their mind. The characters should skip shelves they know won’t have their stock, as previously stated. This moves on nicely to the next point.

The addition of line of sight was agreed to be a positive feature and would add a lot of realism, by allowing customers to talk to each other while further away, and to not notice some items which is realistic. The partially explored pathfinding addition was agreed to also be a great improvement. It would allow different behaviour between old and new customers, and open room for a huge amount of development which is required to change the program into a full training tool. This would also allow customers to walk down dead-ends which is realistic too.

Most of the given suggestions were great, with good potential. They were thought about before development but was not added to keep the project within the defined scope, and due to the time constraints. Some ideas were not thought about before or during development, and would add great training situations such as disabled customer help, and fire alarms. Both these situations can be tough for someone not experienced and the additional help may be vital for customer satisfaction and safety.

Chapter 5:

Discussion, Future Work and Final Conclusion

I n

**Comparison I**

Here will be the comparisons between the initial aims, goals and methodology and the finish project and program.

Firstly, the pathfinding was successfully completed. After research, A\* was found to the be best choice, and so that was used to create the pathfinding. The pathfinding looked mostly realistic, however in the testing it was decided that sometimes the optimal routes made the characters look less realistic. This was a good observation and was a concern throughout the project.

The job decisions were developed well. The employees correctly chose the jobs to complete based upon different conditions such as empty stockcages. Rimworld’s system was not used in the end, since the user could not change priorities, but this was a correct choice due to this program being a simulation rather than the final training tool.

The shopping choices were randomly assigned and the customer did not make any attempt to change the list based upon price and quality. This was a fail from the aims first discussed, but was noted and picked up on by the beta testers, and will be a good addition to future work.

The relationships worked in the correct way based on the original aims. The interactions got affected by character traits and already existing relationship levels, but the beta testers found they were not obvious enough and so that change will be made in future work.

The objectives were all reached, with most of the time being spent on the AI, and the character development. Different AI techniques were considered and the best was chosen. The scenario was created correctly with two employees, and customers spawning randomly, and running the simulation more than once would cause different situations to come up, which was proven in the beta testers’ responses.

The Methodology was stuck to, with the ordering being mostly the same. The only change would be the final stages of the employee AI needing to wait for the customer AI, although this was not too crucial was they were interchangeable anyway as stated in the methodology.

**Future Work I**

This program is very clearly far from complete, which was known to be the case even before the project had begun. The aims that were not completed will make good additions to future work, as well as ideas created throughout the project, and the final suggestions by the beta testing team.

Firstly, the bad programming practises that were added to get the program ready for testing will be removed. It is never good to have these in the code, and so these must get fixed as a priority.

Next, customer item changes. The customers should care about the price and quality of the stock. This should affect their purchasing decisions and whether they want to spend money at the store.

The addition of line of sight is another huge piece of work which would add a large amount of realism, and of course this links with the partially explored environment code, which would allow the program to open up more and help development in a lot of different directions. Line of sight would allow signs to be added, which helps customers find different parts of the store.

Sales could be added to the game, which would mean signs for the sales could be added. This in turn, would allow the possibility of a bad looking sign to cause a customer to believe an item is a different price than it actually is. The customer may buy the item, and then realise the price was wrong which would then cause the customer to perhaps come back for a refund. All of these additions would add to the realism and be great for future work.

Other smaller pieces of work which would add to the improvements would be breakages, and spillages, this would add more jobs for employees, and more issues such as health and safety. Other additional jobs for employees could be phone calls, and deliveries of stock.

Fire alarms and disabled customer help would be good additions and will be worked on in the future.

To allow this program to be used as a training tool, which is the main point of the program, the world would need to be able to be fully customizable which means work must be done to allow furniture to be placed, time of day and year to change. Customers with certain traits and needs will need to be able to be added. This all means an entire scenario editor would need to be developed which would be a huge job, but would excel the program into an advanced and powerful piece of software that would be able to be used by employers for training staff.

**Conclusion I**

This project has been long and has contained lots of problems and obstacles. Time was a big factor in the development of the program, with a few features not being able to be developed as planned due to the time constraints of the project. Despite this, the final program showed a large amount of the originals aims and objectives, and these were the most vital ones. It was easy to keep to the bounds and scope of the project, since the AI was the main focus. Some of the AI features, such as the traits and conversations, could not be fully shown without UI, which was unforeseen at the beginning of the project, but did not cause too much trouble in terms of the time it took to complete this.

The testing process has proved the program to be a success at being a good baseline for future development of a full integrated training tool. Feedback has shown that the AI is not as developed and sophisticated as originally planned, but was still good enough to show off the basic idea behind the project. Because of this, the project has been a success. The employee AI was shown to be more advanced that the customer AI, which was the plan since the employee AI did a better job at showing off the best features of a fully functional program.

The program was finished at a very good point, with all the systems that were begun, being finished to a standard that could be tested by outsiders. The program can also be restarted at any point, with any of its systems to carry on the development of the training tool.

In conclusion, the program, and therefore the project, is a success. With the program being able to demonstrate the best features of a training tool once development has created a more in-depth and advanced set of systems. All the objectives were met, and most of the aims were met with the ones not being met being the smaller and less important features. This program can be used successfully as a great baseline for future work and a potential training tool.

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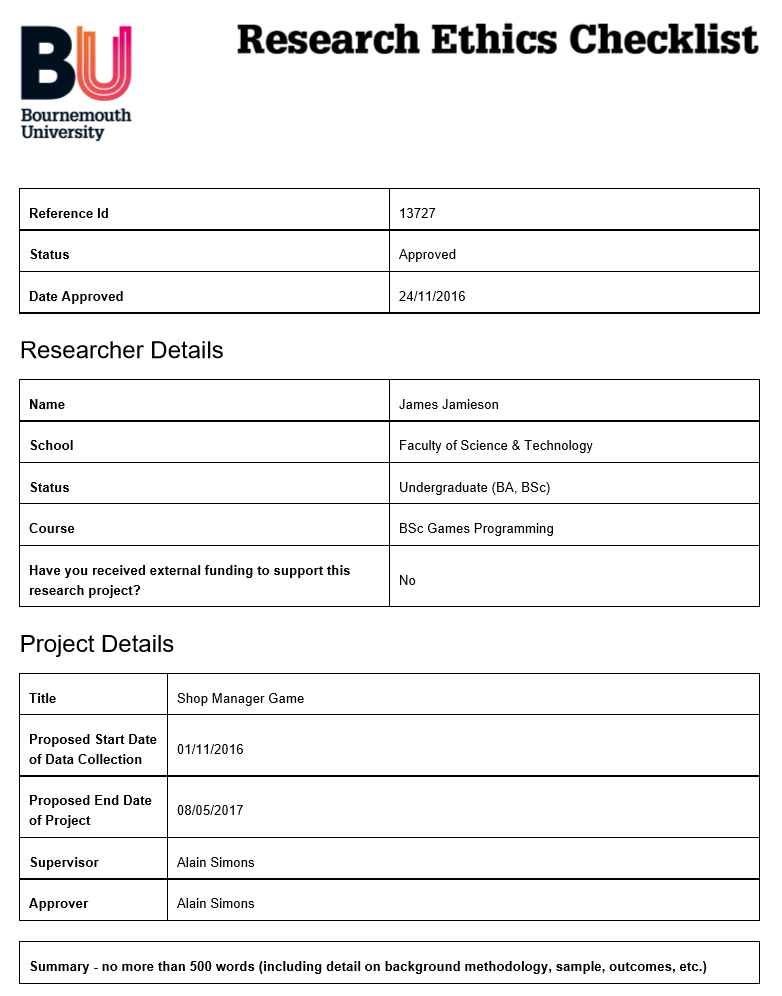
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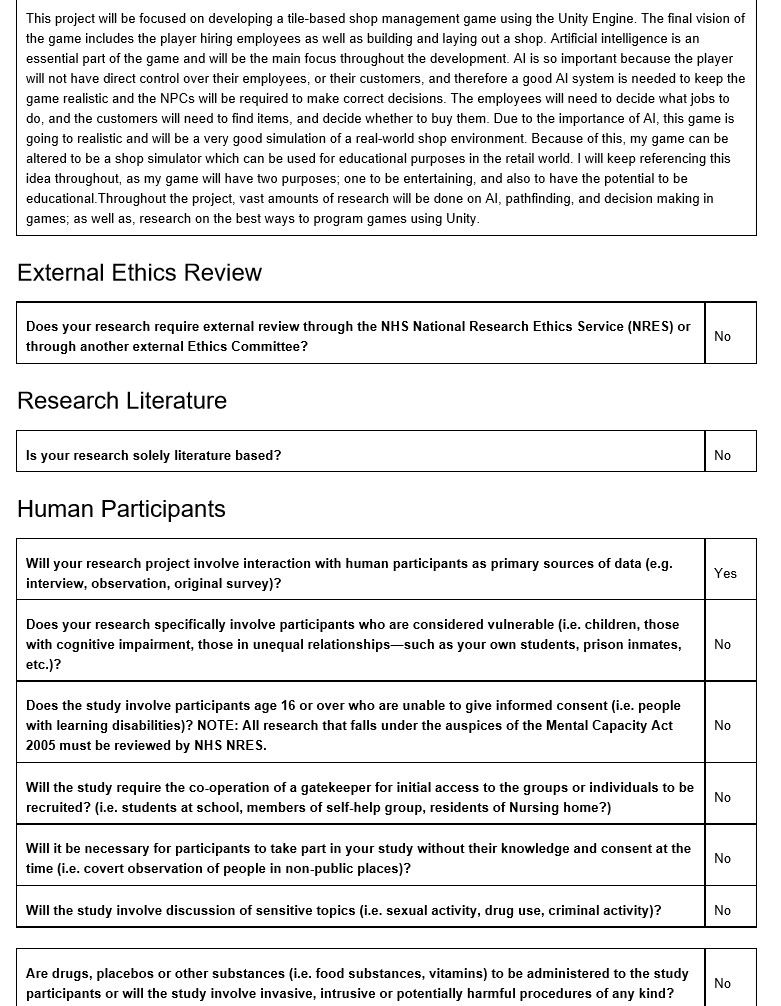
Figure 2: dotstarmoney, 2015. *Generalized Platformer AI Pathfinding (A Guide!)* [online]. Available at: <http://www.reviewmylife.co.uk/data/2008/0715/dijkstras-graph.gif> [Accessed November 2016]

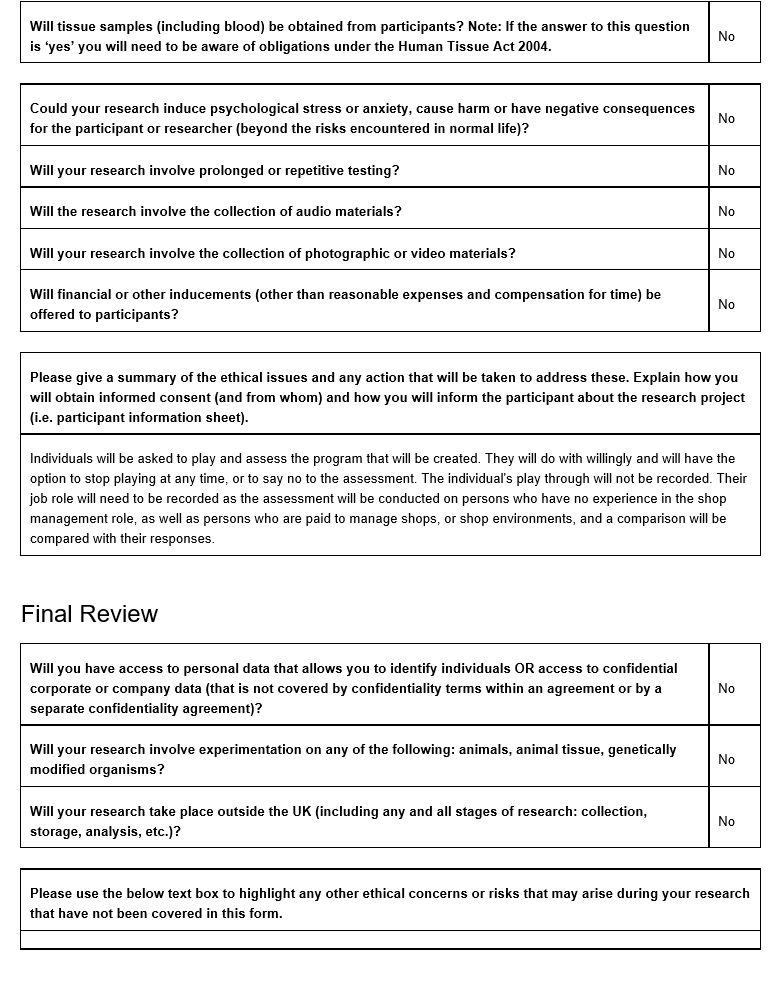
Figure 3: Björnsson, Yngvi; Enzenberger, Markus; Holte, Robert C. *Fringe Search: Beating A\* at Pathfinding Game Maps*; IEEE 2005 Symposium on Computational Intelligence and Games, 2005, 128.

Figure 4: Stamford, J., Khuman, A. S., Carter, J. and Ahmadi, S. 2014. Pathfinding in Partially Explored Games Environments: The Application of the A\* Algorithm with Occupy Grids in Unity3D [online]. *In 14th UK Workshop on Computational Intelligence (UKCI)*. Bradford 8-10 September 2014. IEEE. Available at: <https://ieeexplore.ieee.org/document/6930151/> [Accessed 3rd December 2016]

I Appendices

**Appendix 1 I** 





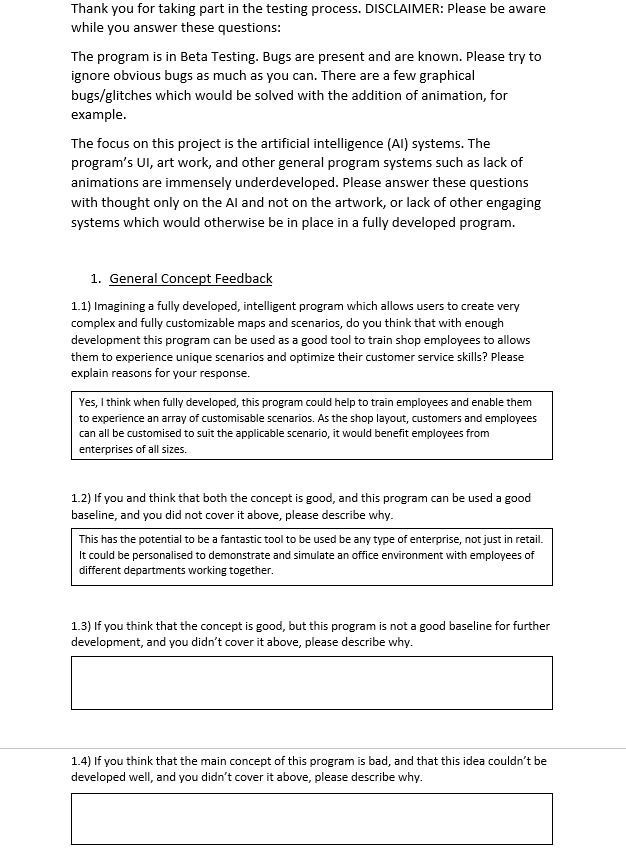
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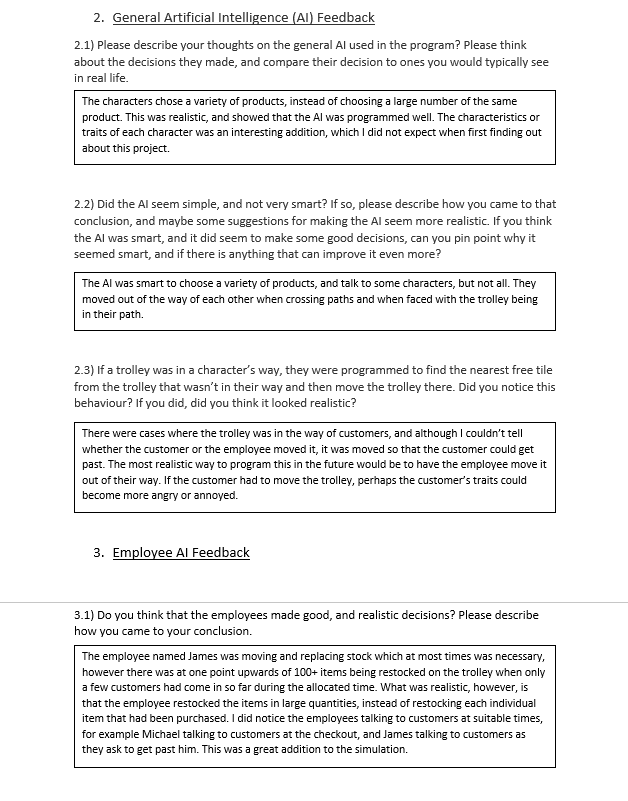
Copyright (C) 2015 Martin "quill18" Glaude

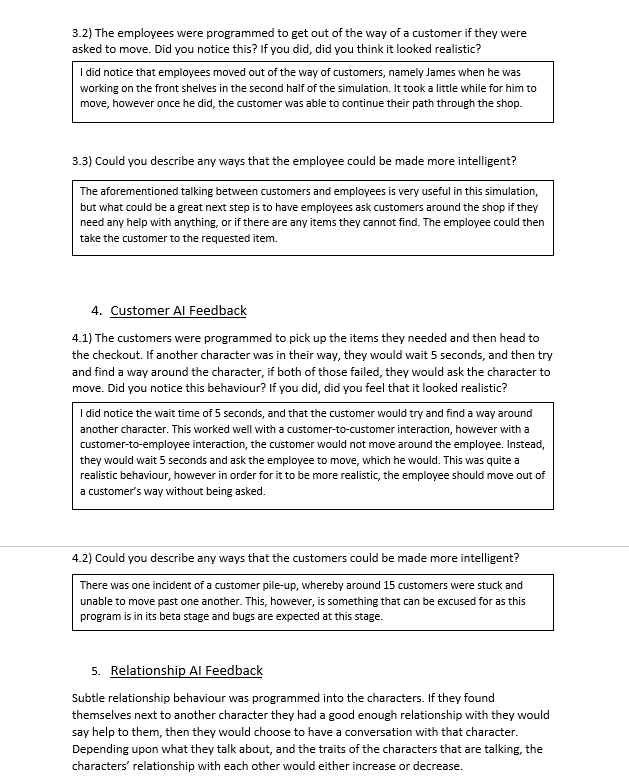
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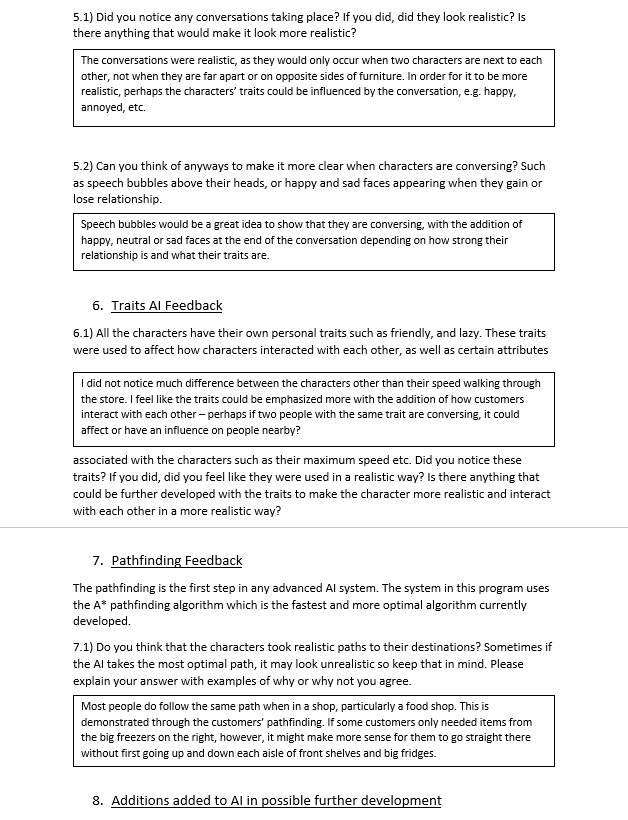
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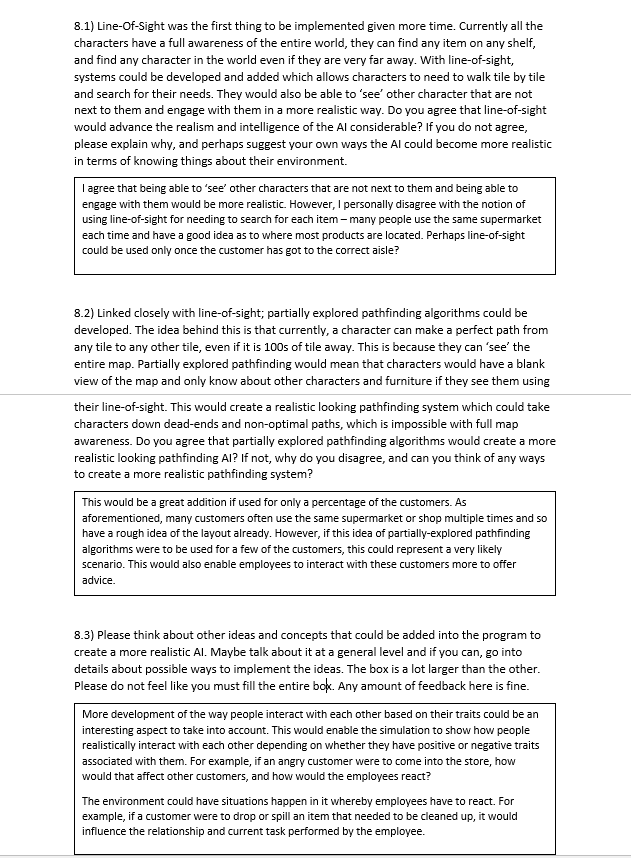
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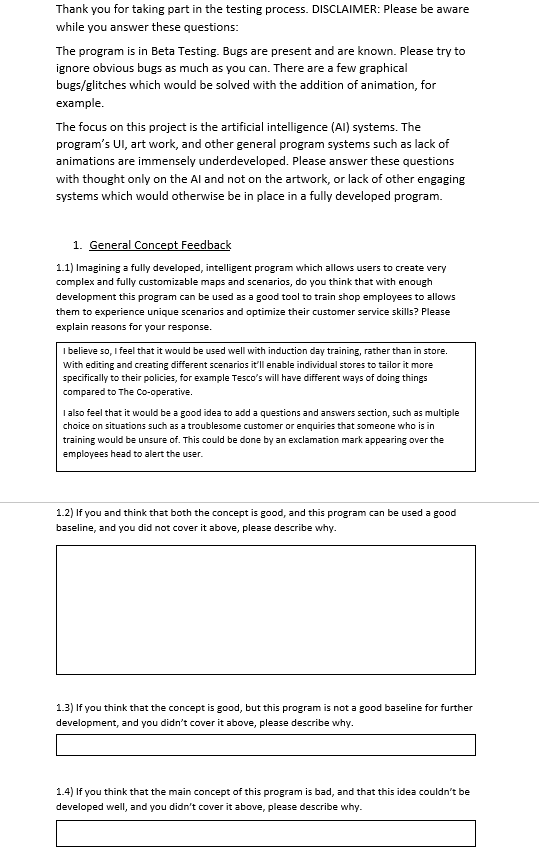
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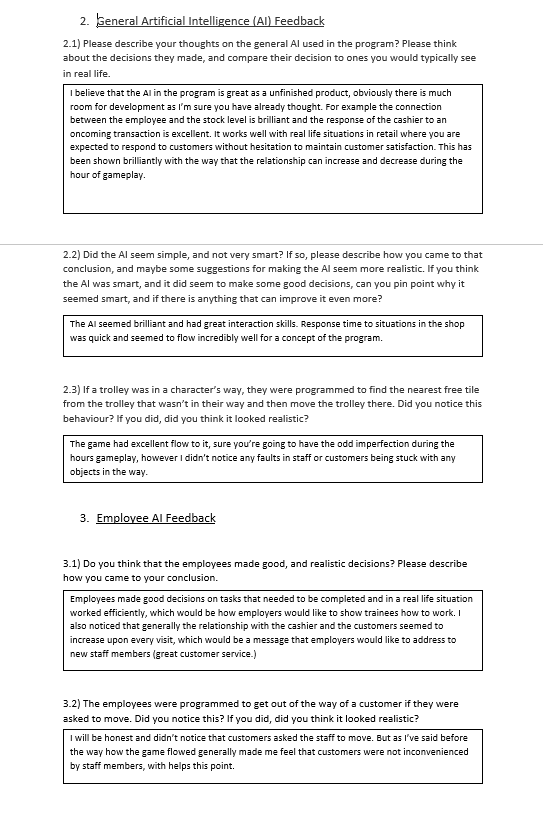
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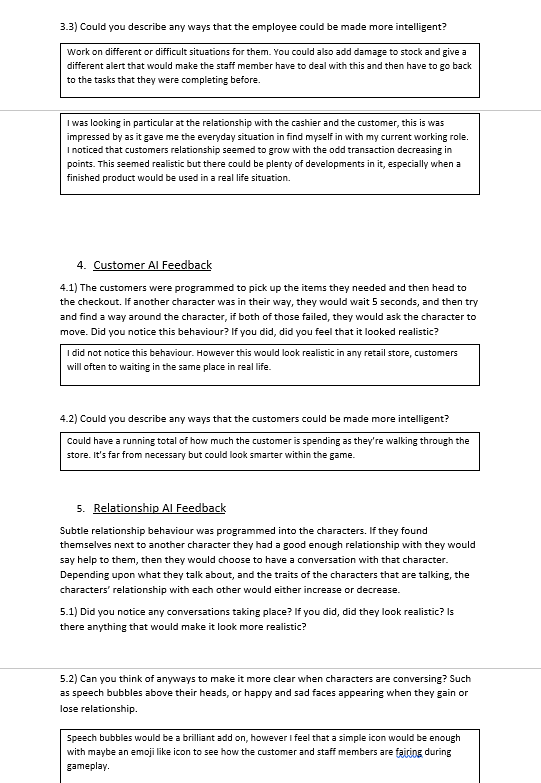


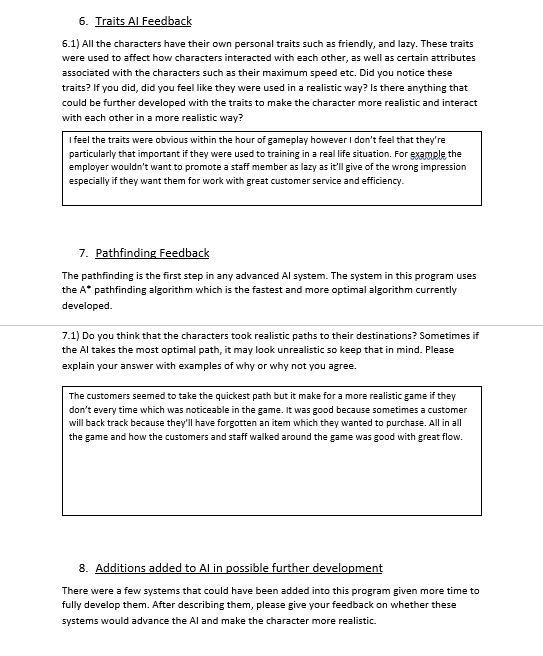


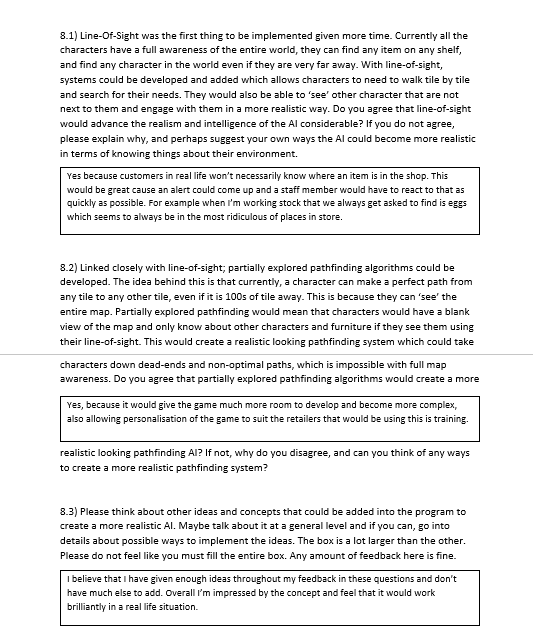


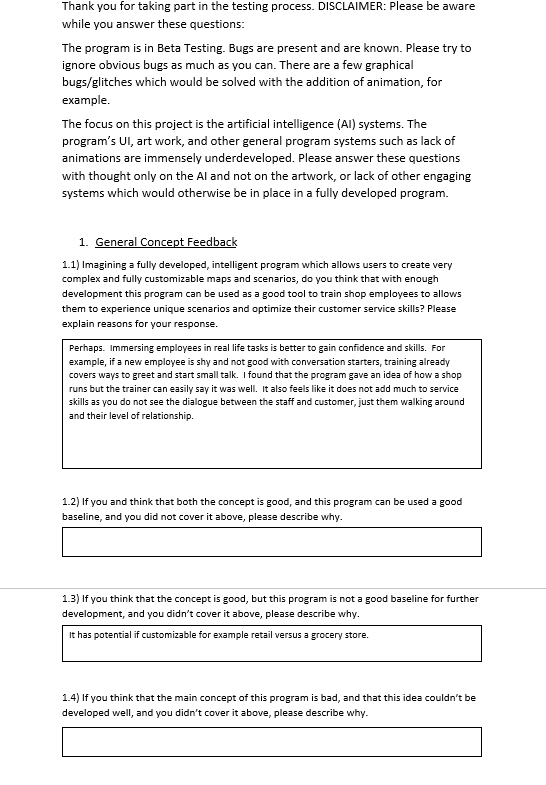
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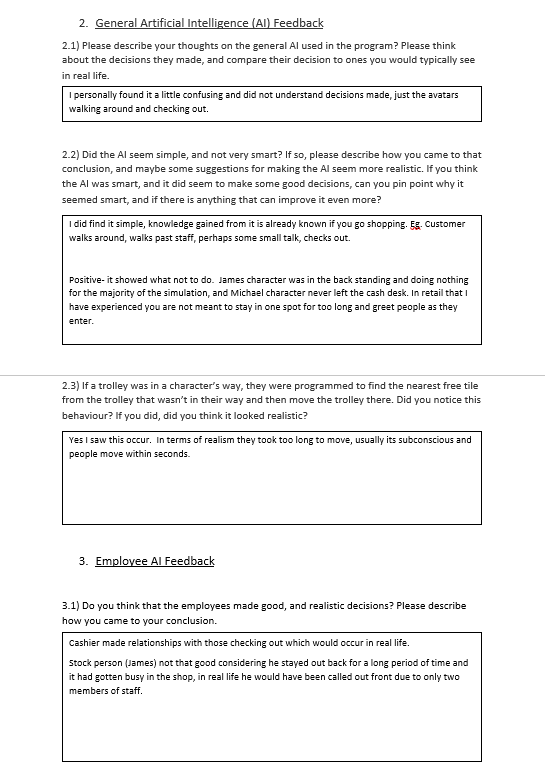


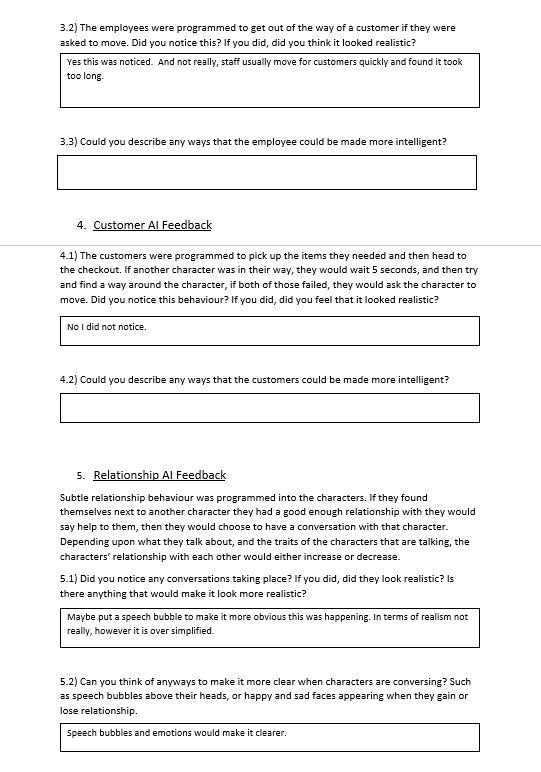


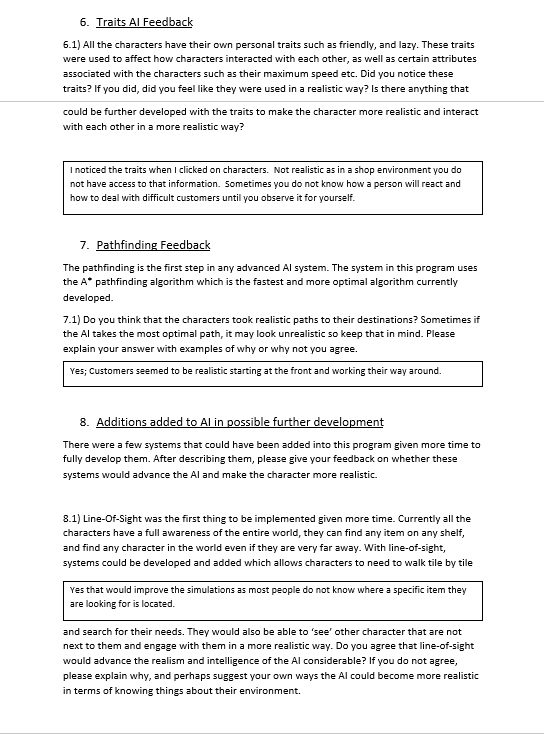


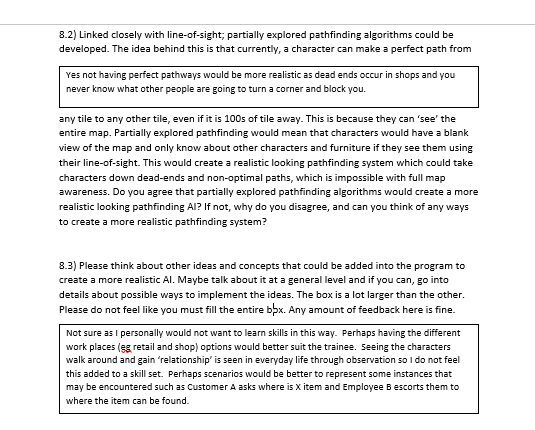


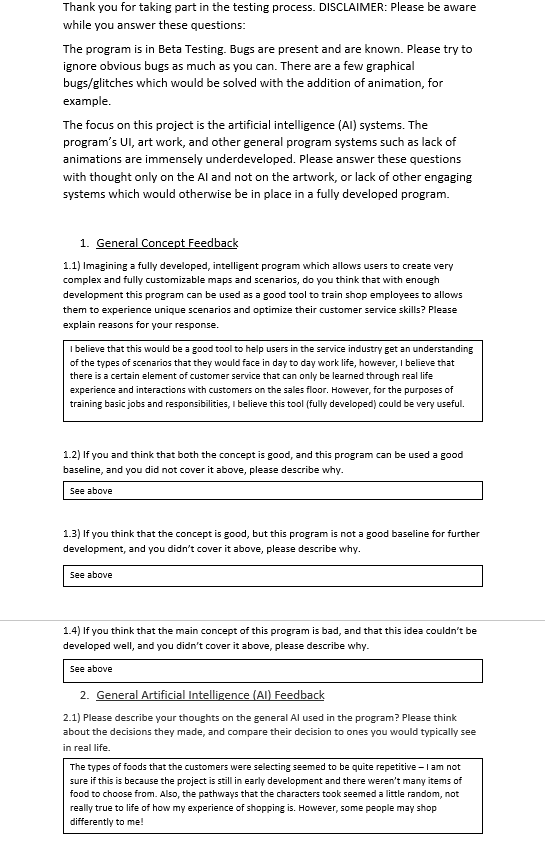
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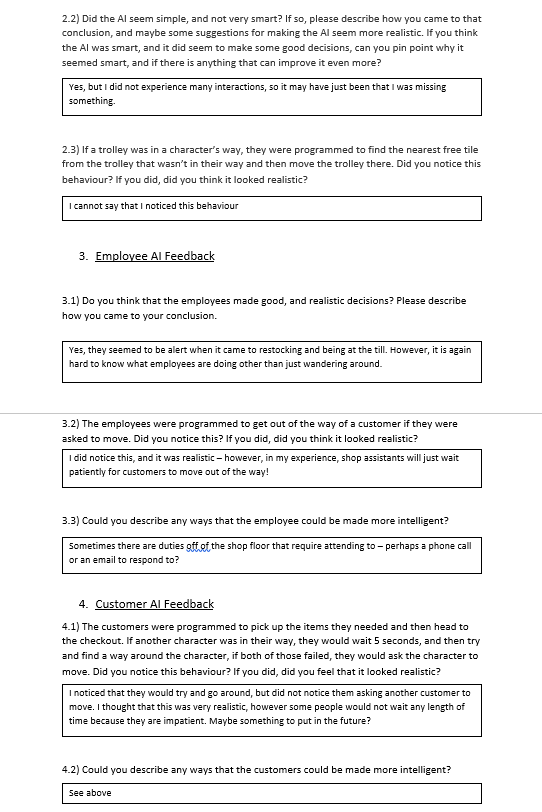


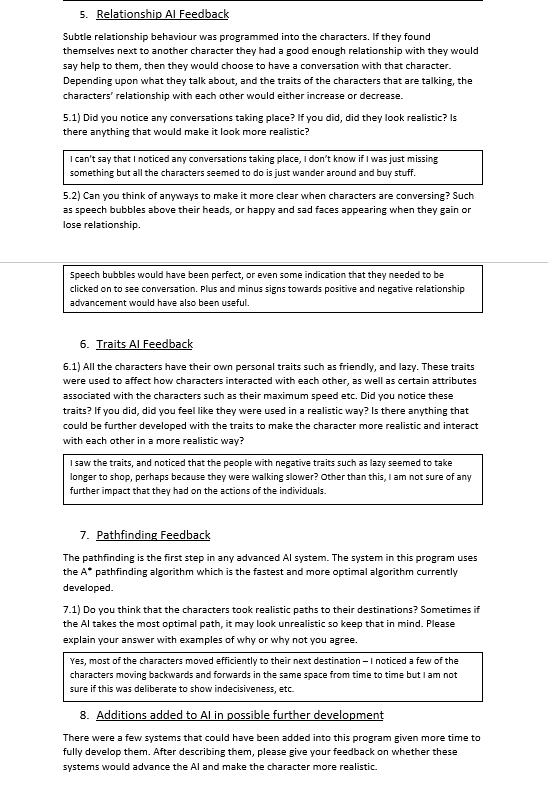




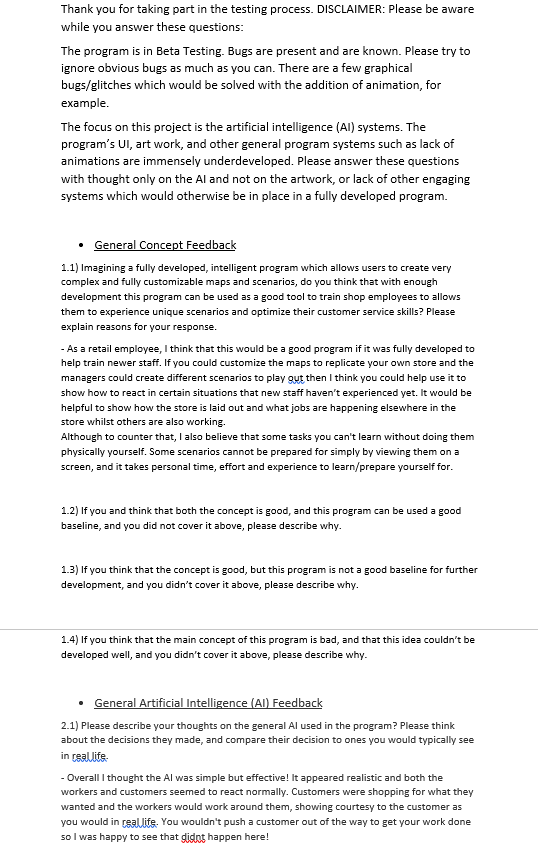


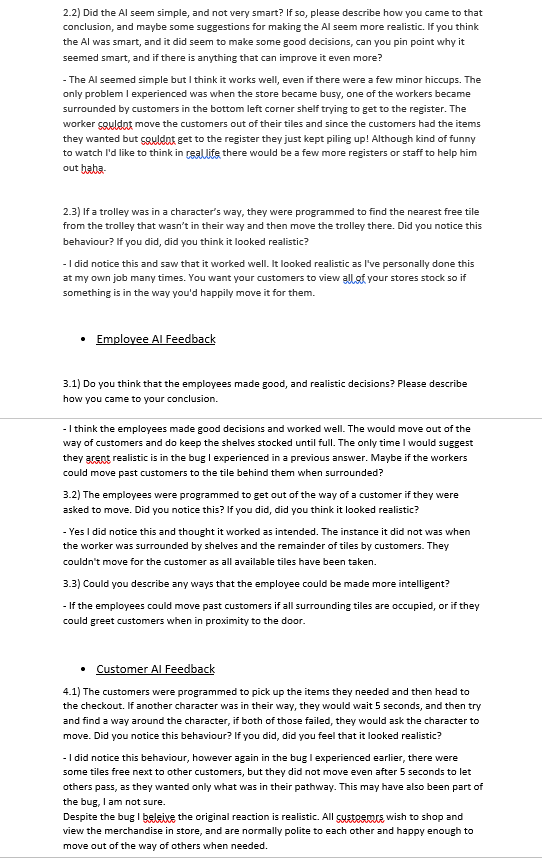
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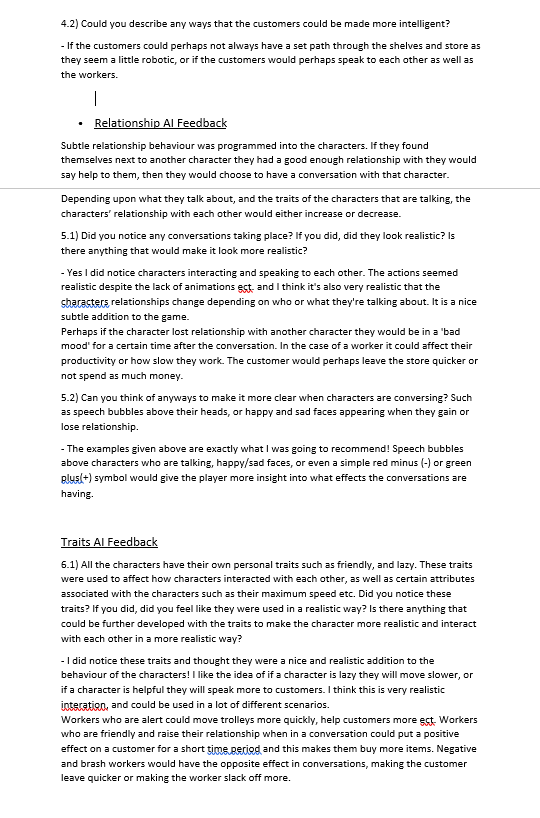


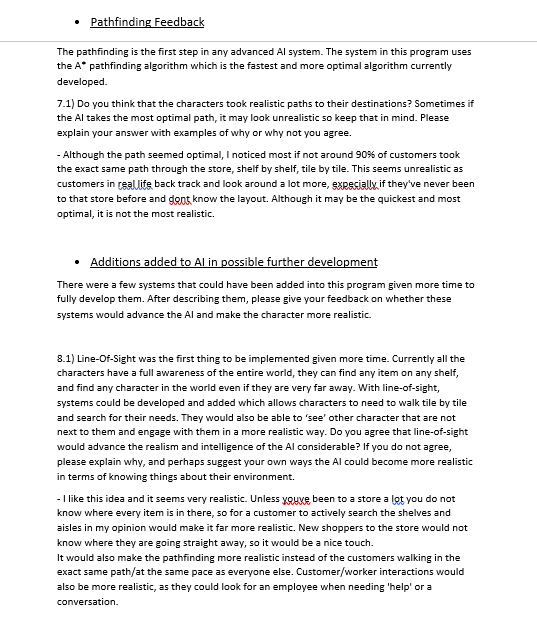


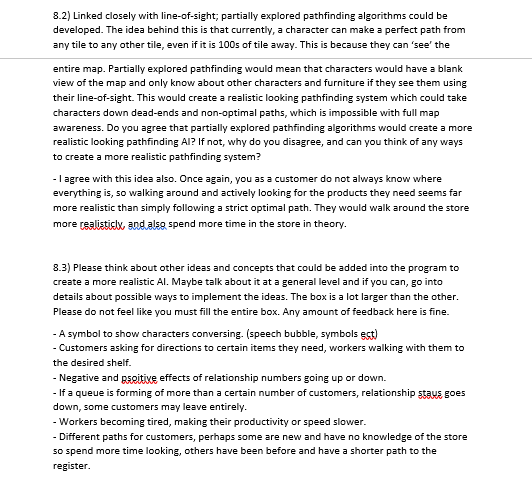


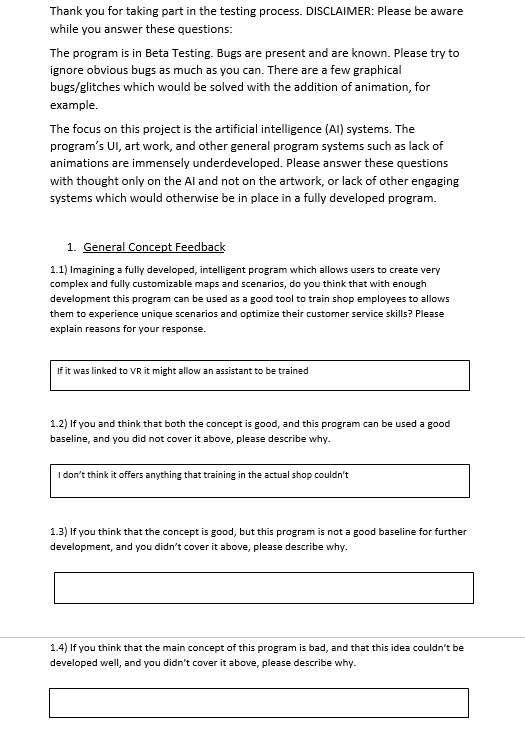
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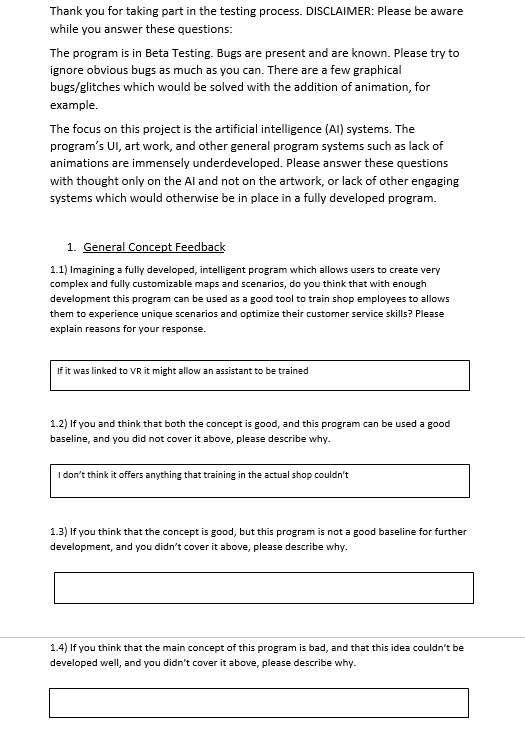


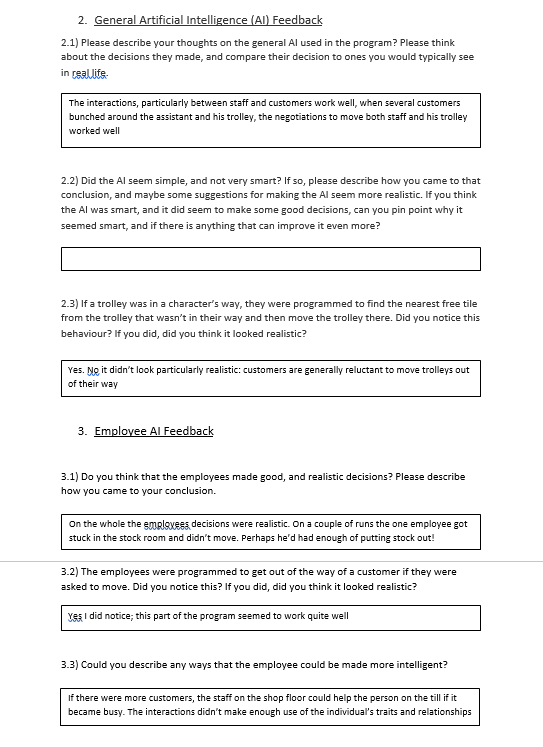


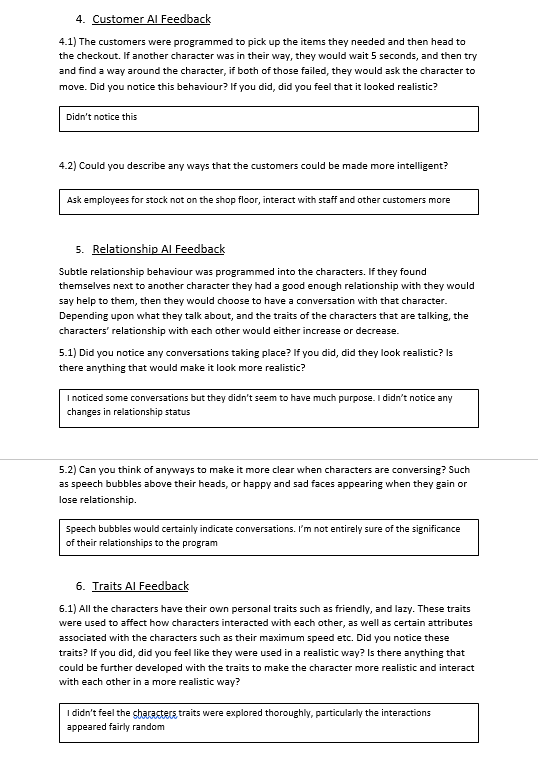


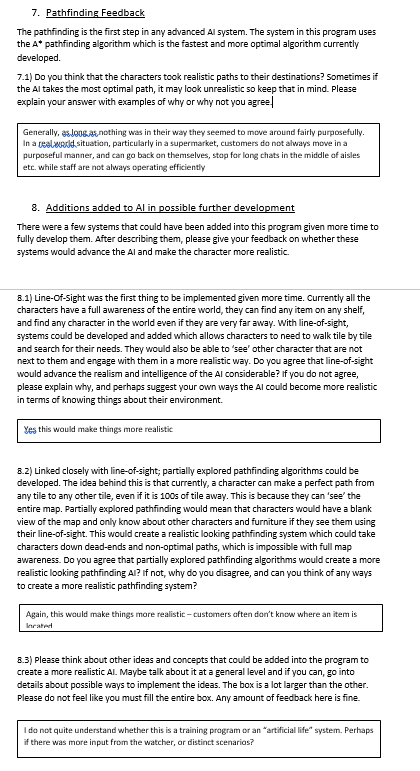


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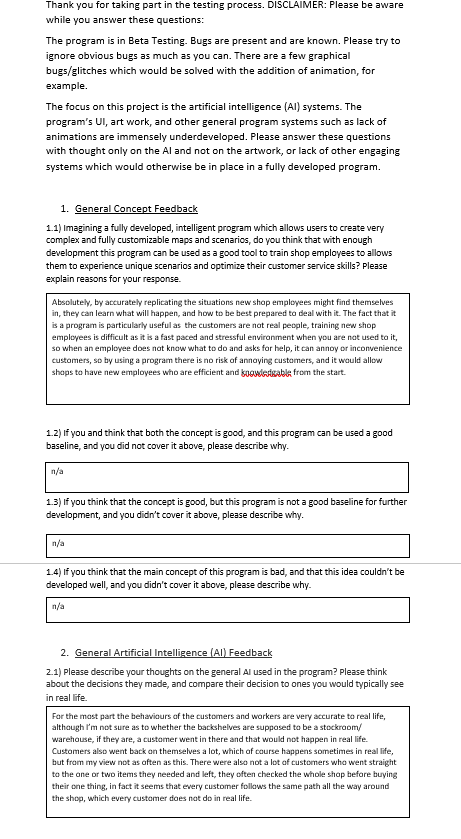


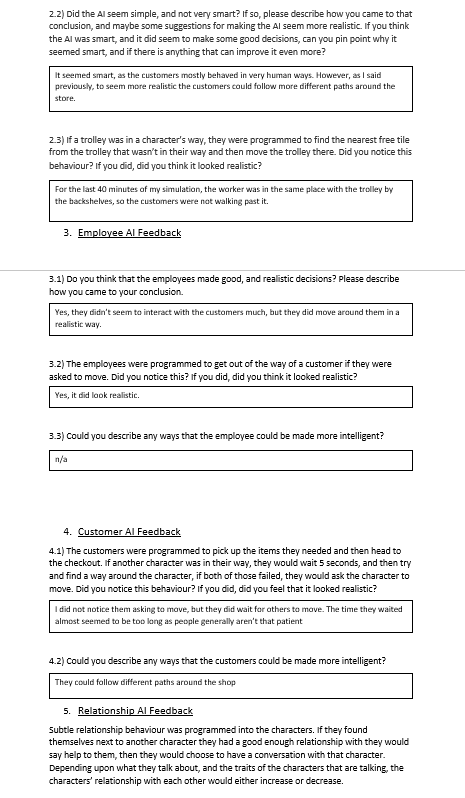


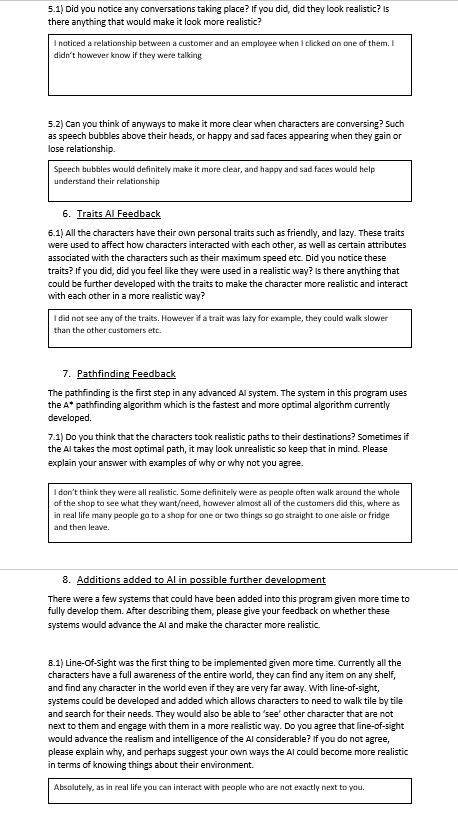


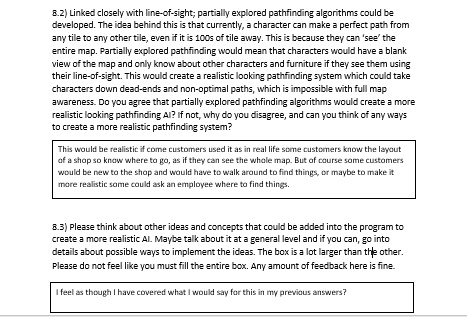


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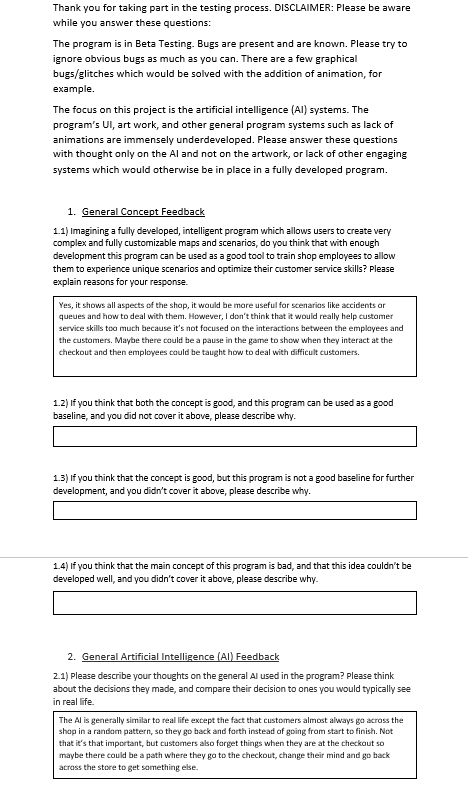


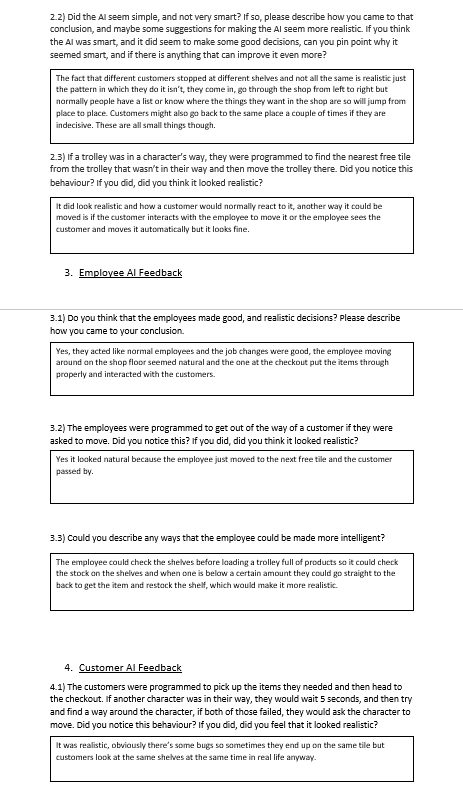


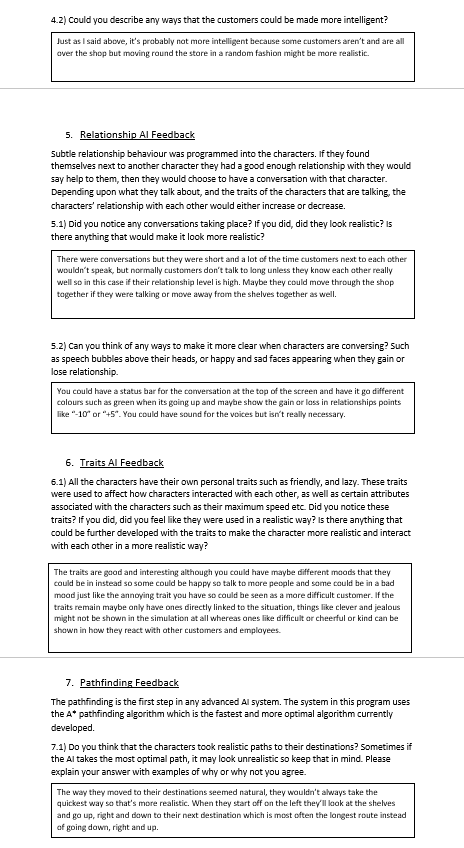


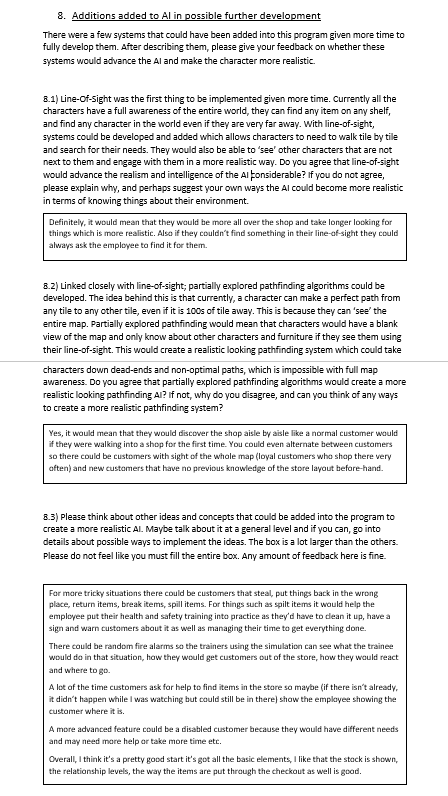


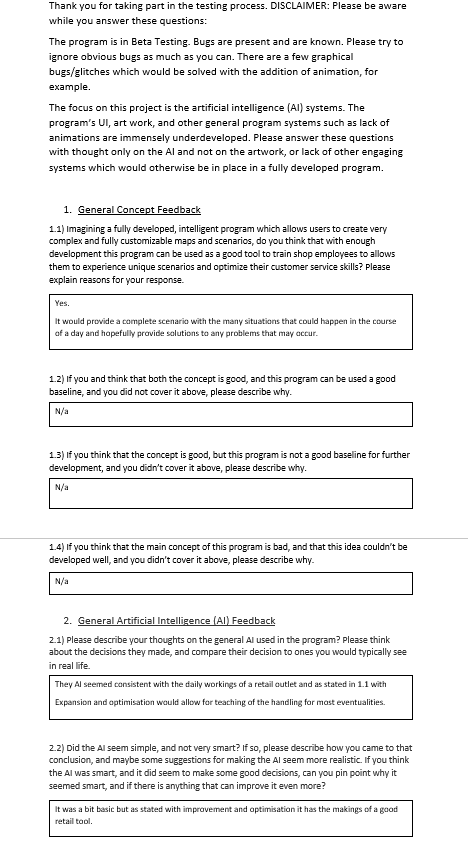
**Appendix 10 I**









 **Appendix 11 I**

