

# MINI-DOOM

**Computer Graphics (E016712A)**

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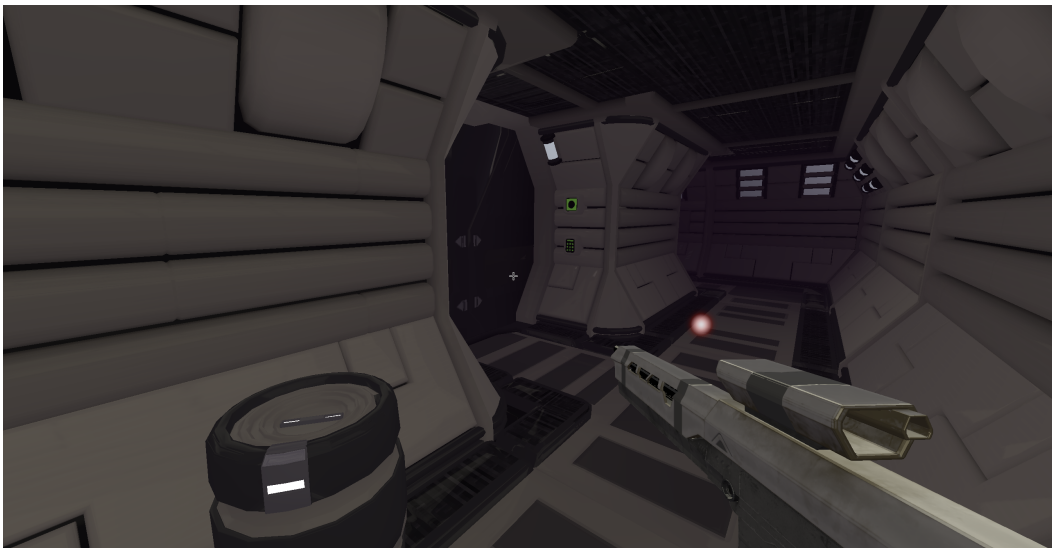


Figure 1: Gameplay screenshot of our Doom-inspired game implemented in Three.js.

## 1 Introduction

### 1.1 Tech Choices

Our project uses Three.js, a lightweight JavaScript library that abstracts WebGL for accessible 3D graphics programming in browsers. We chose it for its robust ecosystem, documentation, and optimized web rendering performance. Development is powered by Vite for fast server capabilities, and we implemented the codebase in TypeScript to benefit from static typing, which improves code reliability and facilitates better team collaboration.

### 1.2 Gameplay Overview

Our game is a first-person shooter inspired by the classic Doom, implemented using Three.js. Players navigate through procedurally generated mazes, interact with doors, follow path markers, and encounter various props while trying to reach their destination. Players can also shoot their gun to open doors, adding a combat element to the navigation mechanics.



Figure 2: Gameplay screenshot showing key elements: (A) Debug menu for adjusting game parameters, (B) Closed door with its control panel (C), (D) Path markers guiding the player through an open door, (E) Mini-map displaying player position and path to destination, (F) Barrel spawned as a random prop, and (G) Weapon in gun mode.

Figure 2 illustrates the main gameplay elements:

- **A:** Debug menu that allows changing camera settings, enabling debug mode, and displays game engine statistics.
- **B:** A door that was previously opened, allowing passage through the maze.
- **C:** Door control panel that can be activated by shooting it with the player's weapon.
- **D:** Path markers showing the optimal route to the destination, guiding the player through an open door.
- **E:** Mini-map displaying the player's current position in red, and the path to the destination, marked in green.
- **F:** Barrel that was procedurally spawned as a random prop in the environment.
- **G:** The player's weapon in gun mode.

## 2 Game Engine

Before diving into specific features, we'll provide an overview of our custom game engine architecture that powers the entire application.

### 2.1 Core Architecture

Our game engine follows a component-based design pattern, with a central state management system that coordinates all game elements. The engine is structured around these key components:

- **GameObject:** An abstract base class that serves as the foundation for all entities in the game world. Each game object automatically registers itself with the state management system upon creation and implements lifecycle methods such as **animate** (called every frame), **cleanup**, and **destroy**.
- **State:** The central management class that maintains references to all active game objects, handles scene configuration, manages the camera system, and coordinates the physics simulation. It provides methods for registering and unregistering game objects, finding objects by type, toggling debug visualization, ...
- **Physics:** A collision detection and resolution system that uses axis-aligned bounding boxes (AABBs) to represent object boundaries. The physics system calculates collision corrections and applies them to maintain proper object separation. For efficient collision detection, we implemented an octree-based spatial partitioning system (explained in detail in the Space Partitioning section), which significantly reduces the number of collision checks by organizing objects hierarchically based on their spatial location.

### 2.2 Game Loop

The engine implements a standard game loop that:

1. Calls the **animate** method on the state, which in turn calls all registered game objects.
2. Updates the physics system and rebuilds the dynamic octree.
3. Renders the scene.

This architecture provides a solid foundation for implementing the game-specific features described in the following sections.

## 3 Maze Generation

First, we will start by explaining our maze generation algorithm, which forms the foundation of our game's level design.

### 3.1 Implementation

The maze generation system is implemented as a modular component in the `utils` directory. It provides function for creating the abstract maze structure but not for converting it to

the physical 3D environment. This separation of concerns allows the maze logic to be tested independently from the game rendering system while maintaining a clean integration between the two.

### 3.1.1 Data Structures

The maze is represented by a grid of cells, where each cell contains information about its walls and visited state:

- **Cell:** A type representing a single cell in the maze with properties for walls (north, east, south, west) and a visited flag.
- **Grid:** A type containing an array of cells and dimensions (number of rows and columns).
- **Pos:** A type alias for a position in the grid, represented as [row, column].

### 3.1.2 Generation Algorithm

The maze generation follows these steps:

1. Initialize a grid where all cells have all four walls intact and are marked as unvisited.
2. Start at a cell (typically [0,0]) and mark it as visited.
3. Push the starting cell onto a stack to track the path.
4. While the stack is not empty:
  - (a) Get the current cell from the top of the stack.
  - (b) If the current cell has any unvisited neighbors:
    - i. Choose one randomly.
    - ii. Remove the wall between the current cell and the chosen neighbor.
    - iii. Mark the neighbor as visited.
    - iv. Push the neighbor onto the stack.
  - (c) If there are no unvisited neighbors, pop the current cell from the stack (back-track).

This algorithm ensures that every cell in the maze is reachable from any other cell, creating a perfect maze with exactly one path between any two points.

### 3.1.3 Upscaling

The implementation includes an optional upscaling feature that doubles the effective resolution of the maze by inserting buffer cells between the original cells. This creates a more visually appealing maze with wider corridors while maintaining the logical structure of the original maze. Additionally, the upscaling process leaves room between parallel hallways, providing space for game elements such as doors to animate into when opened, enhancing the interactive experience without causing clipping or collision issues. Figure 3 illustrates

the difference between a maze without spacing and one with spacing, clearly showing how the upscaling creates buffer zones between parallel corridors.

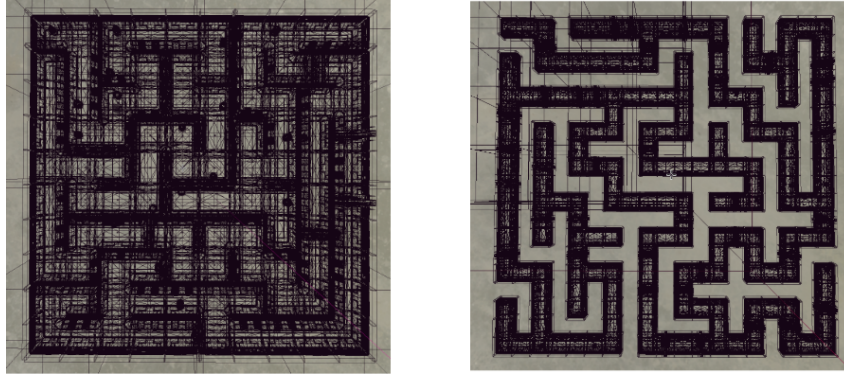


Figure 3: Comparison of a 10x10 maze: left without spacing, right with spacing, viewed from the top in debug mode. Notice the extra spacing making sure that parallel hallways do not touch.

#### 3.1.4 Pathfinding

The implementation features an efficient A\* pathfinding algorithm that calculates optimal routes between any two points in the maze. This algorithm employs a Manhattan distance heuristic—particularly suitable for grid-based movement—and accounts for walls when evaluating potential paths. The pathfinding system provides visual guidance for players through path markers that highlight the shortest route to objectives. The algorithm maintains separate data structures for tracking both the cost of the path so far (g-score) and the estimated total cost to the destination (f-score), ensuring optimal path discovery even in complex maze configurations.

## 4 3D World Organization

Our 3D world uses a modular approach with standardized hallway units as building blocks, optimizing both performance and visual quality.

The world is built on a grid of uniform hallway segments, each occupying one grid cell with consistent dimensions. Hallways have identical outer dimensions with internal variations (straight, corner, junction) and are positioned using precise grid coordinates. While hallway elements are hierarchically organized in the Three.js scene, our GameObject system maintains a flat structure for efficient game logic.

Our system loads all hallway variants from a single GLB file and clones meshes rather than creating new geometry for each instance. Materials and environment maps are cached to prevent redundant creation. This approach significantly reduces memory usage and draw calls, enabling efficient rendering of large mazes.

The player exists as a separate entity from the environment structure, comprising a first-person camera at eye level, a weapon model in the lower view portion, collision geometry for

physical interaction, and independent movement and interaction logic. The player navigates through the hallway grid using keyboard and mouse controls, with collision detection preventing movement through walls and other obstacles. The camera’s position and orientation update in real-time based on player input, creating a smooth first-person experience.

Our world features several dynamic elements: sliding doors that update collision geometry in real-time; interactive door control panels; sprite-based path markers; procedurally distributed props; and temporary bullet impact markers that fade over time to maintain performance while providing visual feedback (see Figure 4). All these elements—doors, control panels, props, and other interactive objects—are implemented as children of their respective hallway parts in the scene hierarchy, maintaining a clean organizational structure while preserving the flat `GameObject` system for game logic.



Figure 4: Bullet impact markers on walls that gradually fade over time and are eventually deleted to conserve resources, providing visual feedback of weapon impacts while maintaining performance.

## 5 User Controls and Movement

Our Doom clone implements a classic first-person shooter control scheme. Player movement is controlled using the WASD keys, with W and S for forward and backward movement, and A and D for strafing left and right. Mouse input controls the camera rotation, allowing players to look around the environment with configurable sensitivity parameters.

The movement system applies the player’s directional input to a normalized vector, which is then rotated according to the player’s current view direction. This creates the expected behavior where pressing W always moves the player forward relative to their view. Movement velocity is controlled by adjustable speed parameters and is framerate-independent thanks to delta time scaling. Collision detection prevents the player from moving through walls, this is explained later.

The game also features immersive visual feedback during movement, with a weapon bob animation that simulates the natural motion of carrying a weapon while walking. This is



achieved by applying sinusoidal oscillations to the weapon position based on the player's movement state. Additionally, the weapon model automatically lowers when the player approaches a wall, providing a realistic response to close-proximity obstacles and preventing clipping issues.

## 6 Lighting, Materials and Texture Mapping

TODO

## 7 Collision Detection Using Octree

For efficient collision detection in our 3D game environment, we implemented an octree-based spatial partitioning system that significantly reduces the computational complexity of detecting object interactions.

### 7.1 Octree Implementation

Our implementation uses a standard octree data structure with several optimizations for performance and memory efficiency. The octree stores colliders directly in a list, with the nodes of the octree itself storing indices into this list rather than the actual collider objects. This approach allows multiple octree nodes to reference the same collider without duplication.

When inserting a collider into the octree, its indices are spread over all partitions that overlap with the bounding box of the element, ensuring that queries will find colliders that span multiple octants. For efficient octant determination, we use bit patterns to index the octree, leveraging the isomorphism between 3 bits and the 8 octants of the tree. This binary representation allows for quick computation of which octant a point belongs to.

To prevent excessive subdivision in dense areas, our implementation enforces a maximum depth limit for the octree. This constraint ensures that the tree doesn't become too deep in areas with many colliders, which would otherwise lead to diminishing returns in performance and increased memory usage. The maximum depth was determined experimentally to balance the tradeoff between query performance and memory consumption, with a value that provides optimal performance for our game's typical scenes.

For memory management, we optimize removal operations by marking elements as dead in the backing list rather than physically removing them. This approach avoids costly array operations and simplifies the tree maintenance logic.

#### 7.1.1 Computational Complexity

The octree provides significant performance advantages over naive collision detection approaches, with the following complexity characteristics:

- **Construction:** Building the octree has a complexity of  $O(n \log n)$  in the average case, where  $n$  is the number of colliders. Each insertion operation takes  $O(\log n)$  time as it traverses down the tree to place the collider in the appropriate nodes.

- **Query:** Collision queries achieve an average-case complexity of  $O(\log n + k)$ , where  $k$  is the number of colliders in the vicinity of the query. This is substantially better than the  $O(n)$  complexity of brute-force approaches that check against every collider in the scene.
- **Memory:** The octree's space complexity is  $O(n)$  in the average case, but can approach  $O(n^2)$  in the worst case when many colliders span multiple octants, requiring multiple references in the tree.

In practice, our implementation maintains near-logarithmic performance as objects are evenly distributed in the grid. The practical benefits of our octree implementation are quantified in the benchmarks presented in Section 8.

## 7.2 Engine Integration

Our game engine maintains two separate octrees: one for static objects (walls, floors, stationary props) and another for dynamic objects (player, moving doors, projectiles). This separation optimizes performance since the static tree never changes after level initialization and thus never needs to be rebuilt, while the dynamic tree is rebuilt each frame to account for moving objects.

Game objects are responsible for managing their own colliders and must register or unregister them with the state management system. The registration process includes a flag to specify whether the collider should be added to the static or dynamic tree, giving developers fine-grained control over collision behavior.

For debugging purposes, we implemented a visualization system that renders the octree structure with color coding based on depth. Deeper nodes are displayed in blue while shallower nodes are displayed in red, with varying opacity to enhance visibility. This visualization has been invaluable for debugging spatial partitioning issues and understanding collision behavior in complex scenes. This is visualised in Figure 5.

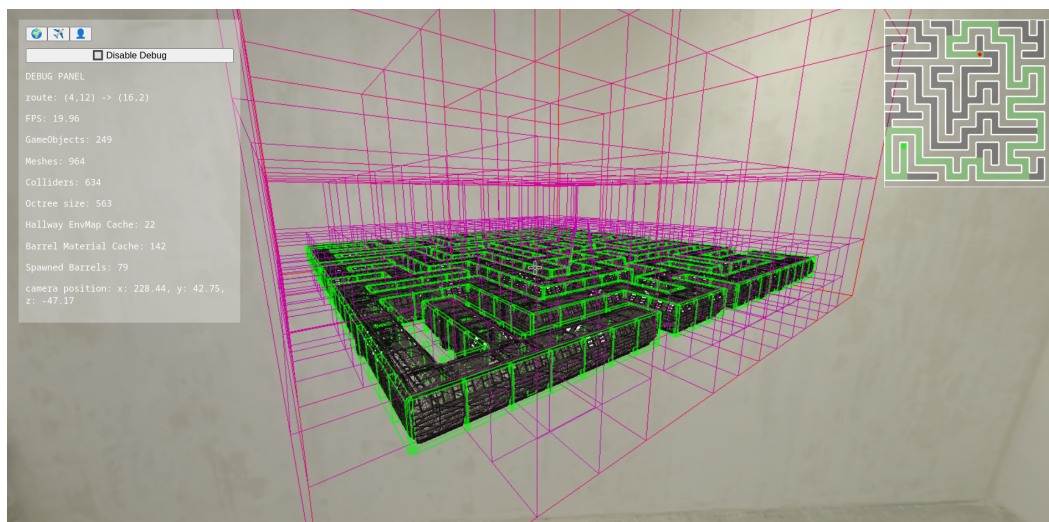


Figure 5: Screenshot of our debugging view showing the octree.

### 7.3 Collision Resolution

Since our game world is organized on a grid, we use axis-aligned bounding boxes (AABBs) as colliders for all game objects. When checking for collisions, we query both the static and dynamic octrees to retrieve all colliders that share the same spatial partition as the object being tested.

Our collision resolution algorithm implements several sophisticated techniques:

First, we determine the amount of overlap between colliding objects in the direction of movement. The system then implements minimum penetration axis selection by comparing overlaps in the X, Y, and Z directions and only applying correction along the axis with the smallest overlap. This approach, based on the separating axis theorem, ensures that objects are pushed apart along the most efficient path.

The algorithm employs direction-aware displacement, determining the sign of the correction based on the relative position of collider centers. This ensures that objects are always pushed in the correct direction to separate them, regardless of their relative positions or movement directions.

For handling multiple simultaneous collisions, we accumulate displacement vectors from all collisions and track how many meaningful collisions occurred (ignoring zero-displacement cases). The accumulated vector is then scaled by a factor of 1.1 to add a small push-out buffer that prevents objects from getting stuck at the exact collision boundary. Finally, the displacement is averaged by dividing by the number of collisions, creating a balanced response that prevents erratic movement when multiple collisions occur. A different way this could also be approached is by taking maximum displacement along each component. However, this would take a lot of comparisons, which is why we opted for a simple normalized sum instead.

## 8 Benchmarking

To assess the performance and efficiency of our implementation, we conducted a series of controlled tests focusing on key areas that affect gameplay and user experience. These benchmarks provide quantitative data on how our design choices impact performance as scale increases, helping to identify optimization opportunities and hardware requirements.

### 8.1 Maze Size Performance Analysis

To evaluate the scalability of our implementation, we conducted performance tests with mazes of increasing size. All tests were performed on a MacBook Air (2022) with an Apple M2 chip, 8-core CPU, 8-core GPU, and 24GB of unified memory. The performance was measured in frames per second (FPS) while maintaining all graphical settings and game features at consistent levels.

Table 1 presents the relationship between maze dimensions, scene complexity (measured by the number of game objects and meshes), and rendering performance. The maze dimensions represent the logical grid size before upscaling, while the total tiles indicate the actual number of navigable cells in the generated maze.

Maze Size	Total Tiles	Game Objects	Mesh Count	FPS
10×10	100	204	697	100
20×20	400	904	2,844	100
40×40	1,600	3,204	11,595	50
80×80	6,400	12,804	46,477	10
160×160	25,600	51,204	187,113	2.5

Table 1: Performance metrics for different maze sizes showing the relationship between maze dimensions, scene complexity, and frame rate.

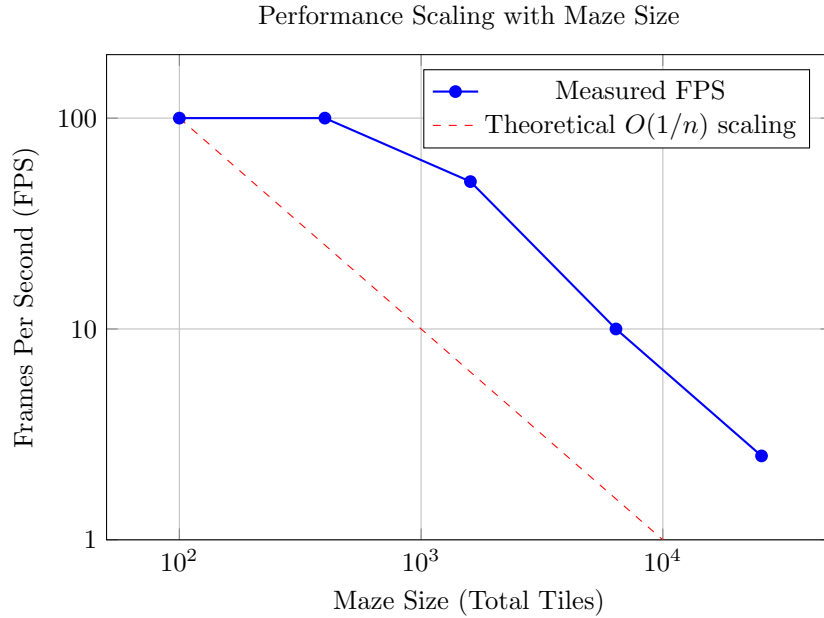


Figure 6: Log-log plot showing the relationship between maze size (total tiles) and performance (FPS). The dashed red line represents theoretical inverse scaling, demonstrating how our implementation closely follows expected performance characteristics as scene complexity increases.

As evident from the data, our implementation maintains an optimal frame rate of 100 FPS for mazes up to 20×20 (400 tiles), demonstrating that the game is well-optimized for small to medium-sized environments. Note that the 100 FPS ceiling is due to the testing monitor’s maximum refresh rate; the actual performance for smaller mazes may exceed this value. Performance begins to degrade at 40×40 (1,600 tiles), dropping to 50 FPS, which is still playable but shows the increasing computational demands.

At the extreme end, the 160×160 maze (containing 25,600 tiles and over 187,000 meshes) pushes the hardware to its limits, resulting in only 5 FPS. While not playable, this test demonstrates the upper bounds of our implementation on consumer hardware and provides valuable insights for future optimization efforts.

## 8.2 Octree Performance Analysis

To evaluate the efficiency of our octree-based spatial partitioning system, we conducted a series of benchmarks to measure different aspects of the octree’s performance: initialization time, insertion time, query performance, rebuild operations, and direct comparison with linear search. Our benchmark methodology involved several controlled tests with elements distributed uniformly throughout a cubic world space of 1000 units per side. Colliders of varying sizes (between 1 and 5 units) were randomly positioned within this space, ensuring a realistic distribution of objects similar to what might be encountered in a game environment. Each experiment was performed multiple times with different octree configurations (varying node capacities and maximum depths) to determine optimal settings.

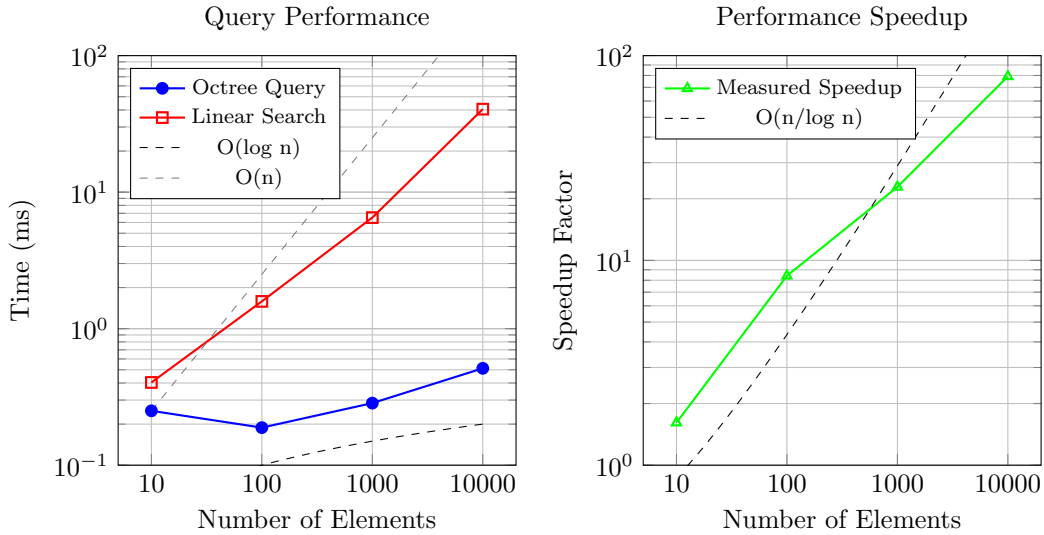


Figure 7: Octree performance comparison. Left: Log-log plot of query time for octree vs. linear search, showing how octree performance follows logarithmic scaling while linear search scales linearly. Right: Speedup factor of octree-based queries compared to linear search, demonstrating efficiency gains that grow with scene complexity.

Our benchmarks reveal that the octree provides significant performance advantages for collision detection in large scenes. As shown in Figure 7, octree query time scales approximately logarithmically with the number of elements, while linear search exhibits the expected linear scaling. For collision queries, we generated 100 random query regions with varying sizes (between 20 and 50 units) and measured how long it took to find all colliding objects using both octree and linear search approaches. With 10,000 elements, octree queries are approximately 79 times faster than linear search.

Elements	Octree Query (ms)	Linear Search (ms)	Speedup Factor
10	0.250	0.404	1.61×
100	0.188	1.585	8.41×
1,000	0.285	6.509	22.84×
10,000	0.512	40.542	79.13×

Table 2: Performance comparison between octree-based queries and linear search across different numbers of elements. Each test performed 100 queries, and times represent the average execution time in milliseconds.

These performance characteristics enable our game to maintain interactive frame rates even in complex environments with numerous dynamic objects, confirming that the octree implementation successfully addresses the computational challenges of collision detection at scale.

## 9 Conclusion

In this project, we successfully developed a Doom-inspired game using Three.js that implements a wide range of computer graphics techniques. Our maze generation system creates complex, playable environments with optimal pathfinding capabilities, while our octree-based spatial partitioning system efficiently handles collision detection even in large scenes. Performance benchmarks demonstrate that our implementation achieves interactive frame rates for reasonably sized mazes, with the octree providing up to 79× speedup compared to naive collision detection approaches.

The modular design of our game engine allows for easy extension and modification, with a clear separation between game logic and rendering components. By implementing advanced features such as dynamic lighting, material properties, and physics-based interactions, we’ve created an immersive first-person experience that captures the essence of classic FPS games while showcasing modern web-based rendering capabilities.

Throughout development, we faced and overcame several challenges, particularly in optimizing performance for larger environments and ensuring consistent physics behavior. Future work could focus on further optimizing the rendering pipeline, implementing more advanced lighting techniques, and expanding gameplay elements with additional enemy types and weapons.

## 10 Art Credits

While the focus of this project was on the technical implementation of computer graphics concepts, we utilized several third-party 3D models and textures to enhance the visual quality of our game. All assets were used under appropriate licenses, and modifications were made to better suit our game's aesthetic and technical requirements.

- **Hallway Modules:** The modular hallway system was based on assets from "Spaceship Modules" by ThisIsBranden, available at <https://thisisbranden.itch.io/spaceship-modules>.
- **Environmental Props:** Various props such as barrels and crates were sourced from the "Sci-Fi Assets Pack" available on TurboSquid at <https://www.turbosquid.com/3d-models/sci-fi-assets-model-1876664>.
- **Weapon Model:** The player's gun was adapted from the "Sci-Fi DMR" model available on TurboSquid at <https://www.turbosquid.com/3d-models/3d-scifi-dmr-model-1983451>.

Small modifications were made to these models using Blender. Texture adjustments were performed using Krita. This was done to maintain a consistent visual style throughout the game environment.

## 11 Acknowledgment

During this project we made use of AI tools to help us write the report and more specifically to help us formulate our arguments in proper academic English.