

# AROB Practical Work

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**Abstract—Practical work abstract. 200 words approx.**

## I. INTRODUCTION

Navigation is a core component of modern videogames, as it determines how non-player characters move through complex environments while avoiding obstacles and respecting design constraints. It also lies at the heart of behaviour AI, since characters must not only reach their goals but do so in ways that appear intentional, legible and consistent with the game's fiction. From enemy flanking in action games to crowd movement in open worlds or unit coordination in strategy titles, robust navigation directly impacts believability and player immersion. At the same time, navigation systems must remain controllable and expressive enough to support artistic and level-design goals, so that designers and artists can obtain plausible, cinematic movement without manually scripting every trajectory.

Classical pipelines rely on graph-based representations (figure 1), such as navigation meshes or waypoint graphs, and search algorithms like A\*, which require a carefully constructed graph discretizing the free space. In practice, these graphs are often built or curated by level designers, a tedious and error-prone process that is difficult to maintain in large or dynamic worlds. Automatic graph construction is possible but complex, hard to adapt in real time, and challenging to tune for artistic or gameplay requirements.

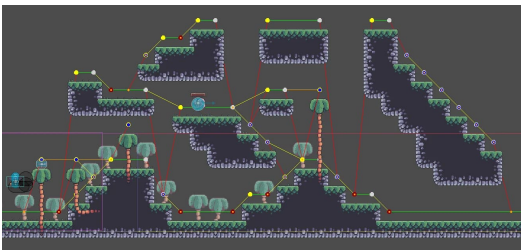


Fig. 1. Example of a graph navigation mesh in Godot Engine [1].

As an alternative, robotics research has developed plenty of methods that operate directly in continuous space, such as sampling-based planners and reactive navigation. However, many navigation pipelines designed for robotics are too computationally expensive to run in a videogame, since they are designed to work in a robot with its own dedicated resources. In contrast, videogame agents must share limited CPU time with other agents, rendering, physics, AI, and other systems, using frameworks that typically have very limited multithreading capabilities and need to run at a high

framerate (60+ FPS). Thus, the challenge is to design a navigation stack that is both efficient and effective enough to run in real-time within these constraints.

This practical work implements a complete navigation pipeline based on these ideas: a customized RRT\* algorithm acts as the global planner, computing collision-free paths in continuous space, while an efficient discretized holonomic ND controller serves as the local planner. Together, they provide a flexible, real-time navigation stack suitable for dynamic videogame scenarios.

### A. Bob in the candyland

To showcase the navigation pipeline, a simple videogame has been created (figure 2). It consists on a 2D platformer where Bob the triceratops (the player) has to navigate through a procedurally generated candy maze to reach the end. The maze has bat enemies (agents) that pursue the player using the navigation pipeline described in this report.

To make the challenge more interesting the player has a special ability that destroys walls to create new paths in the maze, making it a dynamic environment.

In addition, it also includes a debug level to used in this report to test and visualize the navigation pipeline. It is fully implemented in GDscript (a python-like language), allowing easy deployment and modification, and can be downloaded here [Insert url].

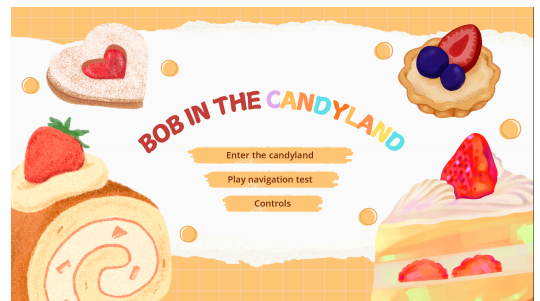


Fig. 2. Main menu of Bob in the candyland.

### B. The navigation pipeline

The system is built on top of the Godot game engine [2], which offers the physics and rendering capabilities needed for the simulation. The navigation pipeline consists on multiple components that interact with Godot to mimic the whole robotic navigation stack.

- **Global planner:** Computes a collision-free path from the agent's current position to a goal position using a custom implementation of the RRT\* algorithm.
- **Trajectory follower:** This component is responsible for following the path generated by the global planner. It selects intermediate waypoints along the path and provides target directions to the local planner.
- **Local planner:** A discretized holonomic ND controller serves as the local planner. It takes the path generated by the global planner and computes movement commands to follow the path while avoiding dynamic obstacles.
- **Low-level controller:** It receives the commands from the local planner and executes them. Here the designer can implement movement mechanics like acceleration, jumping, etc.

## II. GLOBAL PLANNER: RRT\*

The Rapidly-exploring Random Tree (RRT) [3] family provides efficient sampling-based planning in continuous, high-dimensional spaces. We employ a customized RRT\* variant tailored for rapid replanning in dynamic environments.

The key design choice is goal-agnostic tree growth: each new sample is inserted by selecting the parent that minimizes cumulative cost from the root (the agent's current start), and only this root cost is stored. No heuristic or explicit goal-directed information is maintained during expansion, reducing bias and simplifying rewiring.

To obtain a path to any reachable target, the tree locates the best parent for that target and backtracks parent links up to the root, yielding the current minimum-cost route known currently. This allows to quickly extract paths to arbitrary goals without re-growing the tree, which is crucial to track dynamic targets.

During execution the agent's position diverges from the tree root, so when an unreachable path emerges or dynamic obstacles invalidate part of the current route the tree needs to be rebuilt from the current agent's position.

[Figure of the RRT\* display]

### A. Dynamic obstacles

In dynamic environments, continuous sampling allows to naturally discover new paths that were not previously available. To allow doing that even with a filled tree, the planner allows to temporarily increase the tree size limit when a dynamic environment update is detected.

If the path becomes invalid due to a moving obstacle, the local planner will need to adjust or query a full global replan.

### B. Collision checking

The physics simulation introduces some challenges that need to be addressed. To allow the algorithms to work with points and rays, the obstacles are usually enlarged by the

agent's radius to prevent it from getting stuck at corners. However that is impossible to implement in the physics engine, since it would require to manage two independent collision shapes per object that may change dynamically.

A work around could be to use Godot's *shape casting* feature, which allows to cast full shapes instead of rays. This would allow to check collisions of the agent's full shape along the edge, but comes with a significant performance cost in an already expensive algorithm. Instead, the paths are checked using simple ray-casts, and the final refinement is left to the local planner.

In this section, additional constraints could be added to only allow points that meet the kinematic constraints of the agent.

### C. Real-time optimization

To ensure real-time performance, several optimizations are implemented:

- **K-nearest rewiring:** When a new node is added, only the k-nearest neighbors are considered for rewiring, limiting computational overhead to a fixed amount. However this directly limits the quality of the path found, in comparison to considering all nodes within a radius. It is a direct trade-off between performance and optimality.
- **Sparse sampling:** When a sample is generated too close to an existing node, it is discarded and a new sample is generated. This prevents the tree from growing too dense in small areas, which would increase computation time without improving path quality. However, this threshold must be small enough to allow the tree to explore narrow passages. This feature may reduce the bias introduced by K-nearest rewiring, since a less dense tree will favor wider (and better) rewirings.
- **Path sampling:** When a path already exists, a configurable percentage of samples are generated around the existing path instead of the global disk. This focuses exploration on improving the existing path, which is likely to be close to optimal already. The rest of the samples are still generated in the global disk to allow discovering new paths.
- **Efficient nearest neighbor search:** A k-d tree structure is used to allow fast nearest neighbor queries during tree expansion. It was adapted from [4] to work with Godot's Vector2 class.
- **Adaptive sampling:** Samples are generated uniformly within a disk around the agent and the current goal, focusing exploration on relevant areas. The size of the disk is configurable, which directly affects the convergence speed. If the path goes outside the disk it will never be found, so the disk could be automatically increased if no path is found after some time.
- **Quick start:** Each time the tree is rebuilt, a large number of samples is generated to quickly find an initial path. Each new sample tests the goal, so the process

stops as soon as a path is found. Then, subsequent refinements will use fewer samples to optimize the existing tree in a goal-agnostic way. This helps to reduce the frame-time budget used by the planner while still providing fast responsiveness of the agent’s movement.

- **Tree size limit:** To reduce the tree-search space, there are multiple maximum point amounts depending on whether a path has been found or not. When there are no paths, the tree can grow larger to explore more space. When the map changes this amount is also increased temporarily to allow finding new paths.

### III. TRAJECTORY FOLLOWING

Once the global planner provides a path, the agent needs to follow it. For that, the trajectory follower iterates over all path segments to find the closest one to the agent, and selects its second endpoint as the target to reach.

To prevent the agent from getting stuck at the endpoints, if the next segment endpoint has no obstructions the agent skips to it directly.

[Figure of a trajectory, the selected segment and the target point]

### IV. LOCAL PLANNER: DISCRETIZED NEARNESS DIAGRAM

The local planner is responsible for directly going towards the target provided by the trajectory follower, while avoiding immediate obstacles that may not be considered by the global planner.

It implements a discretized Nearness Diagram (ND) controller that reacts to nearby obstacles, offering local adjustments that prevent collisions with corners and small obstacles undetected by the global planner without triggering costly replanning. Thus, it needs to be as lightweight as possible while providing thorough detection of immediate obstacles.

Instead of maintaining a continuous angular clearance map, the implementation samples the  $360^\circ$  space around the agent with eight evenly spaced proximity rays (every  $45^\circ$ ). Each ray reports whether its direction is blocked by geometry, and those that are not obstructed are considered *valleys* (free sectors). When a valley is being considered for navigation, a full shape-cast with the agent’s shape (typically a circle) is performed along the ray to refine the clearance. This refinement ensures the agent does not choose a direction that is free for a point but invalid for its actual footprint.

Given the goal direction, valley selection then finds the closest free valley to that heading by performing a linear search over them. To prevent jitter when several valleys are alternately opening and closing (e.g., at doorway edges), the controller applies a *hysteresis* rule. It stores the previously chosen valley and only switches to a non-consecutive valley if the angular improvement toward the goal exceeds a threshold ( $\approx 135^\circ$ ). Immediate switching is allowed between

consecutive valleys (including wrap-around between the first and last sectors), which preserves responsiveness while suppressing large, unstable oscillations.

Obstacle avoidance and goal pursuit are combined additively. If the direct line toward the target is occluded (ray or shape-cast collision), the controller computes a repulsion vector by summing per-ray wall avoidance forces: each colliding ray contributes an outward normal scaled by how deep the collision point lies inside the sensing radius. This produces a smooth fleeing force stronger near closer walls. The final steering direction is the normalized sum of the valley heading unit vector and the accumulated repulsion. The result is a smooth, goal-oriented motion that moves around immediate obstacles without abrupt changes in direction.

[Figure with ND rays + goal, an obstacle, the selected valley, the repulsion vectors and the final direction]

This reactive layer complements the global RRT\* planner: the global module supplies a path in continuous space; the agent’s controller decides which sub-point to follow, and the ND controller locally perturbs the motion to surround immediate obstacles without triggering full-tree rebuilds. By combining discretized angular sensing, hysteresis for stability, and additive wall repulsion, the approach yields smooth, goal-oriented motion suitable for small dynamic adjustments, while still remaining fairly inexpensive.

#### A. Performance considerations

Usually, physics queries are the most expensive operations in these algorithms. Thus, most effort is put into minimizing the number of queries performed each frame.

First, the agent has a circular area around itself that detects obstacles, permanently registered to the physics engine. This is an extremely efficient way to detect obstacles, but it only provides binary information (is there an obstacle inside or not). Once an obstacle is detected, the ND controller activates the 8 proximity rays (like a LIDAR) to get more detailed information about the obstacle’s position. If there are no obstacles nearby, the ND controller is completely disabled to save computation time and the agent simply moves toward the goal.

Before performing the valley search, they rays are sorted by their angular distance to the goal direction. This favors testing the most promising valleys first, allowing to early-exit the search as soon as a valid valley is found. This optimization is particularly effective in cluttered scenarios where many valleys may be blocked, and directly reduces the number of shape-casts needed for clearance refinement. Ray-casts however are always performed for all 8 rays, since it is more performant to use Godot’s dedicated raycast node.

### V. LOW-LEVEL CONTROLLER

The low-level controller is responsible for executing the movement commands provided by the local planner. In this

case, it simply moves the agent in the desired direction at a constant speed. This is the place where the designer can implement more complex movement mechanics like acceleration, jumping, etc, although many of those would require to modify the global and local planners to take them into account.

## VI. TESTING

### A. Navigation testing

To test the navigation pipeline, a test maze (figure 3) has been hand-created to feature narrow passages with a complex path.

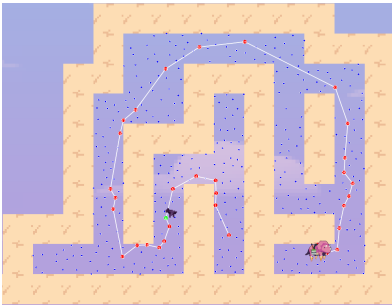


Fig. 3. Path generated in the demo level.

By just adding 10 new points per frame,  $RRT^* + ND$  is able to find a path through the maze in an average of 3.7 seconds. While the path has not been found the local planner temporarily traverses the maze following any free passage, with the possibility of getting stuck in dead ends.

After opening a new dynamic path on the bottom of the maze (figure 4), the agent finds a new path in 0.3 seconds (which highly depends on the sampling to randomly find the new opened spot).

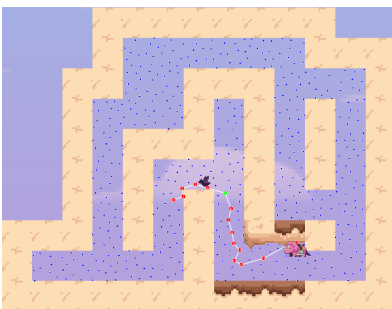


Fig. 4. Path generated in the demo level after opening a new path.

### B. Performance testing

The navigation pipeline takes around 0.6ms per frame, mostly inside the get-k-nearest function. This is a heavily algorithmic function with a significant penalty coming from GDScript's interpreted nature.

Even with that, in the test scene completely remains stable even when new paths are opened dynamically, without

significant latency spikes and always under 1ms of total script execution per frame.

### C. Load testing

To test the navigation pipeline under load, a scene with 50 agents inside a procedurally generated 30x20 maze is used. The number of points generated per frame by  $RRT^*$  is reduced to 3 to maintain performance, at the cost of increasing the time needed to find a path.

The scene runs mostly at 144fps, although the frame time (figure 5) remains unstable due to the high number of agents querying the navigation pipeline simultaneously. Most of the time is spent again inside the get-k-nearest function, so a C++ implementation would significantly improve stability. Even with that, the frame time consistently remains around or under 16ms.

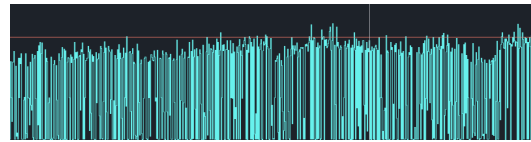


Fig. 5. Frame time of the execution with 50 agents navigating simultaneously. The red line marks 16ms, the budget for 60fps.

In any case, 50 agents are a lot for this kind of videogames, and videogames with significantly more agents (1000+) usually use much simpler systems due to other limitations.

Because of that the navigation pipeline is considered to perform well under load, with even more optimizations possible if needed.

## VII. CONCLUSIONS

In this work we have implemented a complete navigation stack for 2D videogame agents, combining a goal-agnostic  $RRT^*$  global planner, a trajectory follower, a discretized Nearness Diagram local controller, and a simple low-level movement layer. The system is fully integrated in Godot and drives non-player characters in a small platformer, where agents must navigate a maze-like environment that can change over time as the player opens new passages.

The proposed  $RRT^*$  variant grows a single tree in continuous space, supports rapid replanning, and can extract paths to arbitrary targets without rebuilding the structure from scratch. The trajectory follower turns these paths into short-horizon targets, while the ND controller refines motion locally using a small set of proximity rays, hysteresis for stability, and repulsion forces to slide around nearby obstacles. Together, these components keep the global planner lightweight, offloading fine collision avoidance and corner cases to the local layer.

Experiments in hand-crafted and procedurally generated mazes show that the system finds valid routes within a

few seconds while staying within a tight per-frame budget, even with dozens of agents running simultaneously. The combination of sampling-based planning, discrete ND control and simple low-level actuation yields motion that is responsive, smooth and robust to dynamic changes, providing a usable navigation solution for videogame scenarios without sacrificing performance or designer control.

## APPENDIX

### A. Project structure

The navigation pipeline has two main folders:

- **Scripts/Actors:** Contains all the scripts related to the agents in the scene. `Bat.gd` contains the trajectory following logic and the low-level controller, which have been combined under the main agent controller due to their simplicity. `Player.gd` controls the player input and movement.
- **Scripts/Navigation:** Contains all the scripts related to the navigation pipeline. `RRTStarNavigationGD.gd` implements the global planner, `HolonomicND.gd` implements the local planner, and `RTTtree.gd` implements a helper KD-tree structure for efficient nearest neighbor search. All of them are combined inside `Bat.gd`, which queries a global path, computes a target from the trajectory, feeds it to the local planner and finally executes its movement commands and handles the animations.

The rest of the project contains an `Art` folder with the sprites used, `World.gd` which provides global convenience logic, `Scenes` which contains Godot scene files, and `Scripts/Navigation/Display` which contains the classes used to visualize the RRT\* tree for debugging purposes.

## REFERENCES

- [1] Godot Engine, “Navigation in godot,” [https://docs.godotengine.org/en/stable/tutorials/navigation/navigation\\_introduction\\_2d.html](https://docs.godotengine.org/en/stable/tutorials/navigation/navigation_introduction_2d.html).
- [2] —, “Godot engine web page,” <https://godotengine.org/>.
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