# The Kragle - Battlecode 25 Postmortem

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# 1 Introduction

We believe our postmortem will be of unique help to future Battlecode participants. Often, teams who make postmortems are those who have consistently placed highly from the start. We have the perspective of a team that spent years with no consideration of qualifiers, who this year made the jump to be one of the higher level contenders, going toe-to-toe with the 2 seed in the US qualifiers.

Hopefully you find this useful!

## 1.1 Our Team - The Kragle

We are a team of three computer science students at Rensselaer Polytechnic Institute (RPI). In this year's competition, we made 'the leap' from being a team that couldn't submit a final bot to being a contender for the final tournament. We'll outline some tips and tricks for other aspiring teams to make 'the leap' themselves. We are historically terrible at coming up with team names, and ended up choosing "The Kragle" after Justin had a burst of inspiration working on a MicroManager class shortly after re-watching The Lego Movie. Just you wait for what we cook up next year. Some more information about us is below:

Andrew Bank has competed in Battlecode since 2021. He is a Computer Science and Computer Systems Engineer at RPI and will graduate in this spring 2025. Andrew has interned at Johns Hopkins Applied Physics Lab, he has created the Circuit Randomizer for Personalized Learning as an undergraduate research project under Professor Shayla Sawyer, and is currently working on another undergraduate research project: the MusicX project for Professor Sawyer's Mercer Xlab. In his free time, Andrew enjoys having more Strava followers than Instagram followers, and he enjoys treating Stardew Valley like an operating systems optimization problem while playing with his siblings.

Justin Ottesen has competed in Battlecode since 2022. He began his undergraduate degree in Computer Science in Fall 2021, and graduated early in Spring 2024. He stayed at RPI for his Master's, where he does research with the BRAINS Lab under Oshani Seneviratne with a focus on incentive design for smart contract protocols, and is on track to graduate in Fall 2025. Outside of classwork, he has worked as an intern at Nasuni since Summer 2023, and is on the D3 NCAA Cross Country and Track teams at RPI. In his free time, he enjoys hiking, running, programming, and playing Mario Kart Wii.

Matthew Voynovich started competing in Battlecode this year. He began his undergraduate degree majoring in both Computer Science and Information Technology and Web Science at RPI in Fall of 2023. He is scheduled to graduate in Spring 2026 and plans to pursue a masters under the Co-Terminal program at RPI. Currently, Matthew works as an undergrad researcher under Thomas Morgan studying quantum phase investigations, and in his free time enjoys long distance running with his friends in the Rensselaer Running Club.

### 1.2 Past Performance

Andrew was the only member to compete in 2021. In 2022, Justin joined Andrew, working together again in 2023 and 2024 along with some other teammates. We did not perform particularly well in any of these years, typically holding a 1200-1500 rating, with no notable performances in any of the tournaments. Unfortunately, the RPI spring semester starts very early, so we were always balancing classwork along with the competition. This year we went all in, and entered the US qualifier tournament with the 10 seed, rated at 1730. I guess you'll have read this to see how we did...

Although we aren't sure our mess of a repository will be useful to anyone else, our GitHub repositories can be found below:

```
2021 Fire Nation - Lost to the sands of time...
```

```
2022 Kernel Byters - https://github.com/justinottesen/Kernel-Byters
```

```
2023 PC Ghosts - https://github.com/andrewkbank/bc2023 (This is still private)
```

```
2024 Goat House - https://github.com/justinottesen/battlecode24
```

2025 The Kragle - https://github.com/justinottesen/battlecode25 (This is still private)

#### 1.3 Game Overview

This year's introduction message is shown below:

The bread and food of yore has begun to run out, forcing robot society to adapt. Gone are the jovial ducks, replaced by steampunk robot bunnies who have converted their need for nutrients into a reliance on paint. These bunnies have become territorial, forming clans and defense formations to protect the resource that keeps them running.

For the past two centuries, these bunnies have stayed within their own territory, but clans have begun to degrade their environment and need to start branching out. Will these clans be able to expand their territory and generate enough paint to protect their families? Or will they stray too close to other clans and be wiped out in conflict?

As hinted above, this year's game was a competition of territory control between paint-crazy bunnies. The first team to paint 70% of the map would win. Each team had two starting towers, which could produce robots. These robots had different abilities, but their goal was to work together to build more towers and paint as much territory as possible.

We saved copies of both the initial and final specs to our repository in case they are taken down in the future. For a full description of the game, see our repository.

# 2 Strategy & Implementation

## 2.1 Sprint 1

### 2.1.1 Setup

The first thing we did was decide how to structure our code. We took a lot of inspiration from XSquare's 2024 code https://github.com/IvanGeffner/BTC24. We created an abstract Robot class, and created a subclass for each robot type (Soldier, Splasher, Mopper, Tower). This heavily simplified our RobotPlayer class, and helped us split functionality between differing robot types<sup>1</sup>. Along with these classes, we also had utility classes, like MapData, Explore, Bugpath, and others.

<sup>&</sup>lt;sup>1</sup>In hindsight, it may have been even smarter to have another level to the hierarchy where we had Unit as the base class, Robot and Tower as subclasses, and other types inherit from these. Towers don't need to do pathfinding, and they can't really learn much about the map, so this was bad organization and wasted bytecode

Our first priority was getting a bot that could effectively explore and capture towers. This was our first priority since building towers would allow our team to increase their resource income and create more robots. We had each tower spawn a soldier and a mopper who would search for a ruin. We copied the capture logic from examplefunciplayer. The soldier was to paint the pattern around the tower, and the mopper was to clear any enemy paint in the way.

#### 2.1.2 Resources & Towers

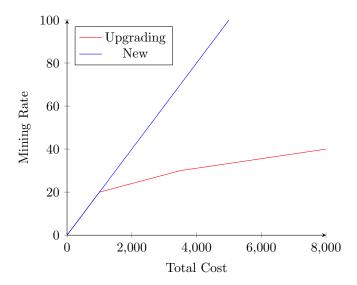
We made two observations very early on about resources:

- 1. It is not worth upgrading towers.
- 2. It Money towers are always better than Paint towers.

The first one is relatively obvious looking at the costs of new towers vs upgrades. For Money towers, the following table shows the cost associated with each level of the tower:

| Level | Upgrade Cost (Chips) | Mining Rate (Chips / Turn) |
|-------|----------------------|----------------------------|
| 1     | 1000                 | 20                         |
| 2     | 2500                 | 30                         |
| 3     | 5000                 | 40                         |

From this table, we can graph the cost and mining rates:

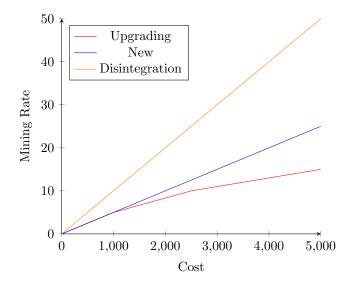


As is clearly shown here, when you have the choice between spending on a new tower, or upgrading an existing one, there is no reason to upgrade. The same can be said for Paint towers, but the next observation makes that redundant.

Even though paint is arguably the more important resource, Paint towers are useless<sup>2</sup> because money towers actually produce paint more effectively than paint towers. Each tower spawns with 500 paint. Since money towers produce 20 chips per turn and they cost 1,000 chips, they pay for themselves in 50 turns. So, if a Money tower calls rc.disintegrate(), we can trade 1,000 chips for 500 paint. If we do this every 50 turns, money towers have an effective paint mining rate of 10 paint per turn, which is double what paint towers are capable of.

Here is the corresponding graph to the above for paint production vs chip cost:

<sup>&</sup>lt;sup>2</sup>This is true until you factor in SRPs, but we are getting there



There are a few other big bonuses of this strategy:

- 1. We now have the choice of converting chips to paint. With paint towers, if we have more paint than we need or can use, we are stuck with it.
- 2. We no longer have to coordinate what tower types we are building, since we are always building money towers (we ignore defense towers for now).
- 3. Since chips are global and paint is local, we can use money towers that are far away from the active parts of the map to subsidize tower production where it is needed most.

## 2.1.3 Special Resource Patterns

However, once SRPs are introduced, this math changes. Each active SRP boosts paint production by 3 units per tower per turn. The following table shows how the number of SRPs affects the per turn production rates of different towers:

| SRPs | Money Tower Chips | Money Tower Paint <sup>3</sup> | Paint Tower Paint |
|------|-------------------|--------------------------------|-------------------|
| 0    | 20                | 10                             | 5                 |
| 1    | 23                | 11.5                           | 8                 |
| 2    | 26                | 13                             | 11                |
| 3    | 29                | 14.5                           | 14                |
| 4    | 32                | 16                             | 17                |
| 5    | 35                | 17.5                           | 20                |

As you can see, once there are 4 or more SRPs, paint towers are better suited for their intended purpose. We decided not to worry about this and stick with our money tower only strategy, to keep things simple for Sprint 1. Plus, we (regrettably) did not prioritize SRPs for Sprint 1, our soldiers simply checked if the current location could have an SRP and if so, they marked one.

## 2.1.4 Pathfinding & Map Representation

Units stored their knowledge of the map in memory, however we ended up rewriting this system, so this will be described further later. Units traversed the map using a BugNav algorithm. We reused an algorithm from a previous year, and it worked with minor tweaking. If you have questions about how this works, there was a lecture on it this year here.

#### 2.1.5 Sprint 1 Performance

We entered the tournament as the 55 seed out of 160 teams, with a rating of 1533, our first match was a 4-1 win against the 74 seed, "be right back" with a rating of 1483.