# MovitIt

An HackaGames game.

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- 1. Applying Q-Learning
- 2. Model-Based Decision Making

## **Basic State Representation**

- ightharpoonup Robot position (6 imes 4)
- ightharpoonup Robot goal  $(6 \times 4)$
- ▶ Robot direction (6)
- $\triangleright$  Obstacles Positions (6  $\times$  4) (6 *obstacles*)
- $\blacktriangleright$  Humans' position (6  $\times$  4), direction (6) (2 humans)

**States:** 
$$24^{(2+6+2)} \times 6^3 = 1.3 \times 10^{16}$$

## **Relative State Representation**

- ▶ Robot goal direction (6), distance (16)
- $\blacktriangleright$  Distance-1 Cells: obstacle? (2<sup>6</sup>)
- $\blacktriangleright$  Humans' position-direction (6), position-distance (16), movement-direction (6)

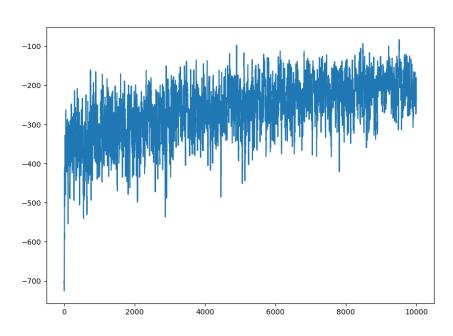
**States:** 
$$6^5 \times 16^3 \times 2^6 = 2.0 \times 10^9$$

A huge gain on the number of state + promising factorization.

However: not covering -> no guarantee

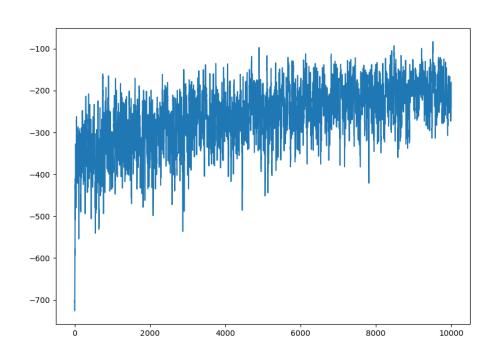
## **QLearning based on Relative State Representation**

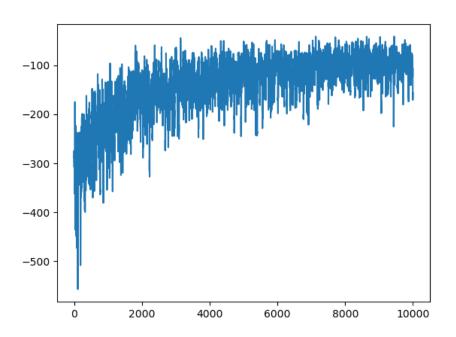
Average score over 10 000 games of 10 cycles.



### **QLearning based on Relative State Representation**

#### **Qlerning Relative-basic versus Relative-limited**





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## **Simulating Action MoveIt?**

Is it possible to model and simulate GameMoveIt?

## **Simulating Action MoveIt**

#### **Board builtin function:**

```
collisions= self.board().multiMoveHumans( humanMoves )
collisions+= self.board().multiMoveRobots( robotMoves )
```

**moves** list of start position and direction.

$$moves = [[x1, y1, dir1], [x2, y2, dir2]...]$$

#### **Simulation squeletom:**

```
def simulate(board, robotMoves):
    copiedBoard= copy(board)
    humanMoves= generateHumanMove(board)
    collisions= board.multiMoves...
    return collisions, copiedBoard
```