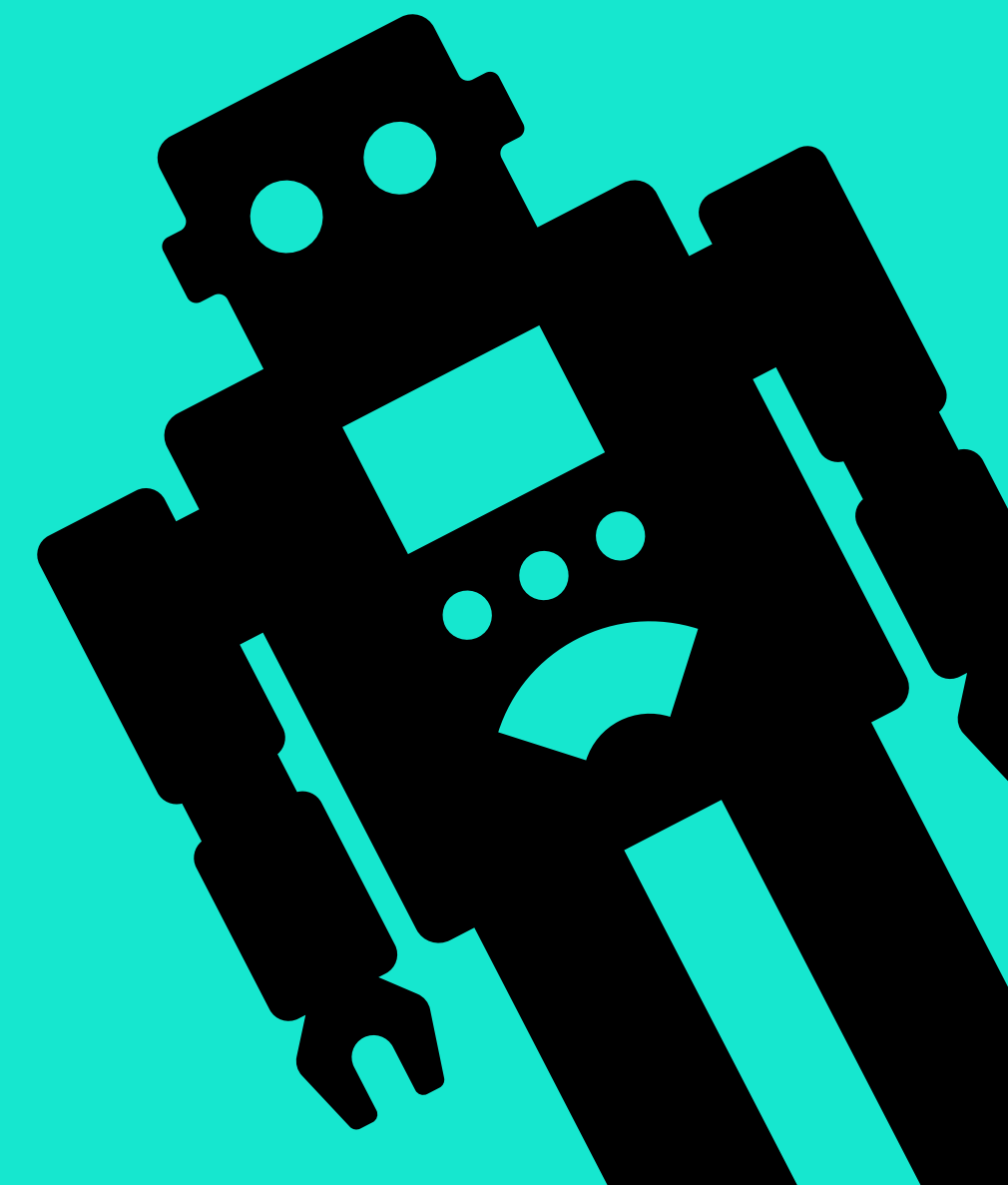


Co-evolution of Robots

Optimization for AI exam

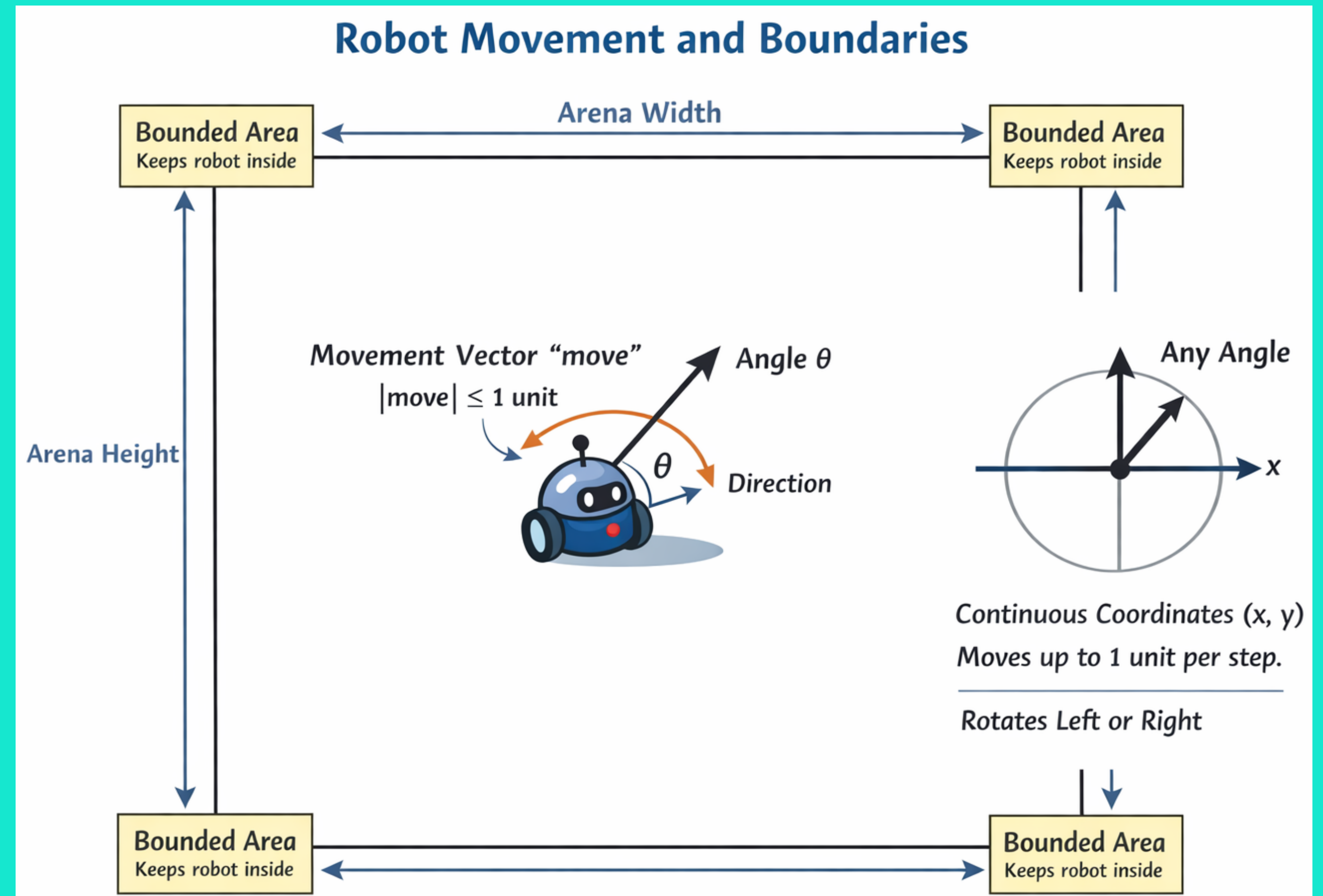
https://github.com/Gabriele-tomai00/Co-evolution_of_robots

Gabriele Tomai 27/01/2026



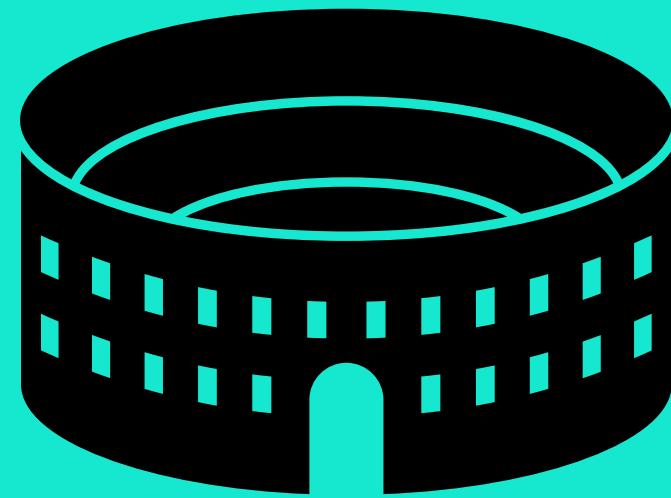
Simulation environment

Like “robot wars”,
two robots are fighting
against each other and the
winner is the surviving one



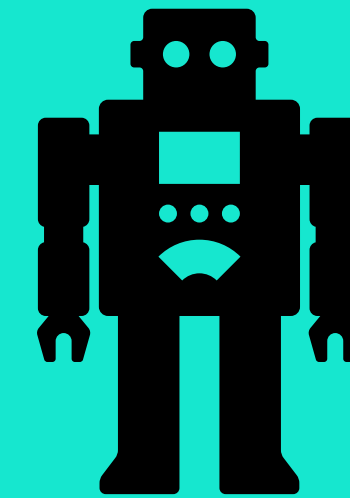
Main classes (1/3)

Arena



- Width
- Hight
- Robots
- Max steps
- Current step
- Damage information

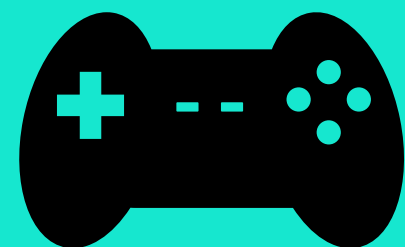
Robot



- Position
- Health
- Damage inflicted
- Controller

Main classes (2/3)

Baseline Controller



- Steering
- Throttle
- Shoot
- get(robot, arena)
- get_opponent(robot, arena)

RandomController

steering, throttle = uniform; shoot = random 0.1

StaticController

steering = fast towards opponent, throttle = 0, shoot = true

AggressiveChaser

steering = fast towards opponent, throttle = 1, shoot = if aligned

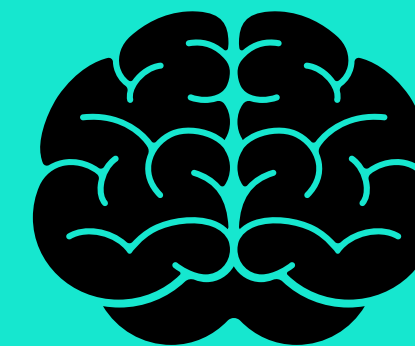
Main classes (3/3)

Sensor



- `get(robot, arena)`
- `get_opponent(robot, arena)`

neat.nn.FeedForwardNetwork



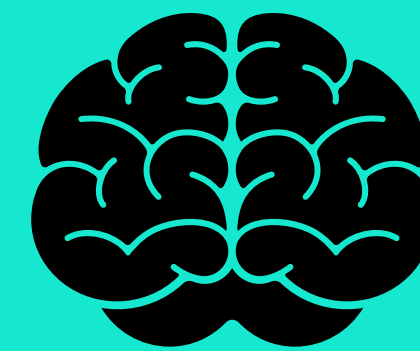
- Library class
- The brain of the evolve network
- Receive 7 inputs
- Produces 3 outputs: raw float values

Sensor and NN

Valued of sensor
are normalized

INPUT (7)

- Distance to opponent
- Angle to opponent
- Health
- 4 Wall distances



OUTPUT (3)

- Move
- Turn
- Shoot

Fitness calculation

Hybrid Fitness (Internal + External)

One-population competitive coevolution



Fitness function —————→

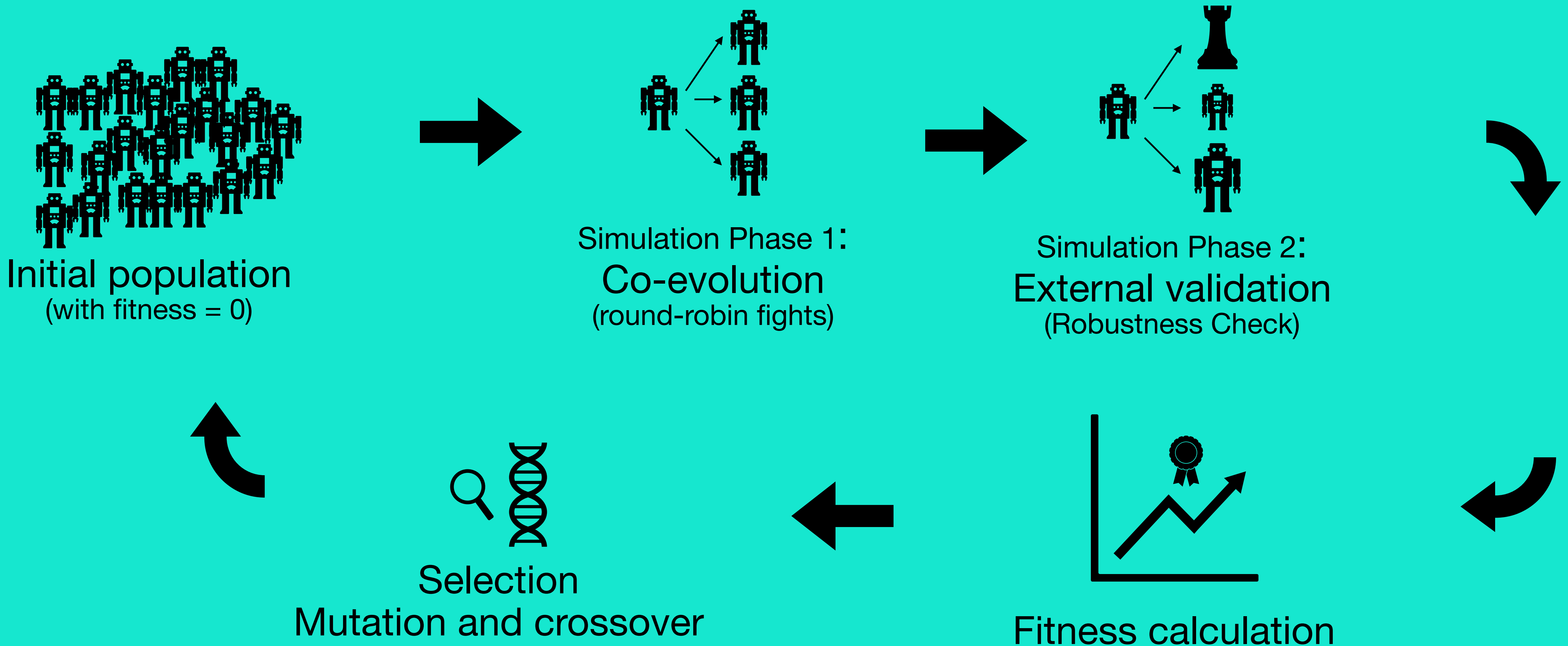
- +100 in case of victory
- +10 in case of aggression
- +0.1 to every time step survived
(up to 30 points)

Designed to reward three main behaviors:

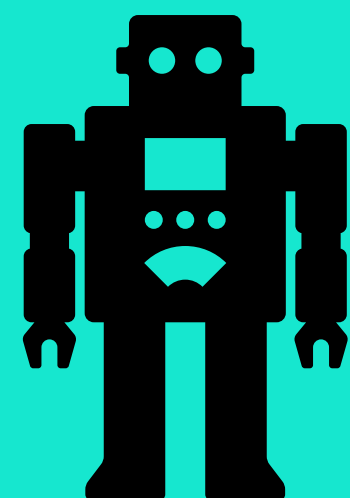
- Effectiveness (Victory)
- Aggression
- Survival

Evolutional NEAT cycle

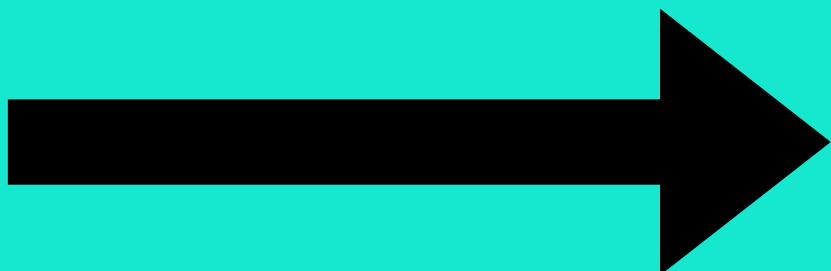
For N generations



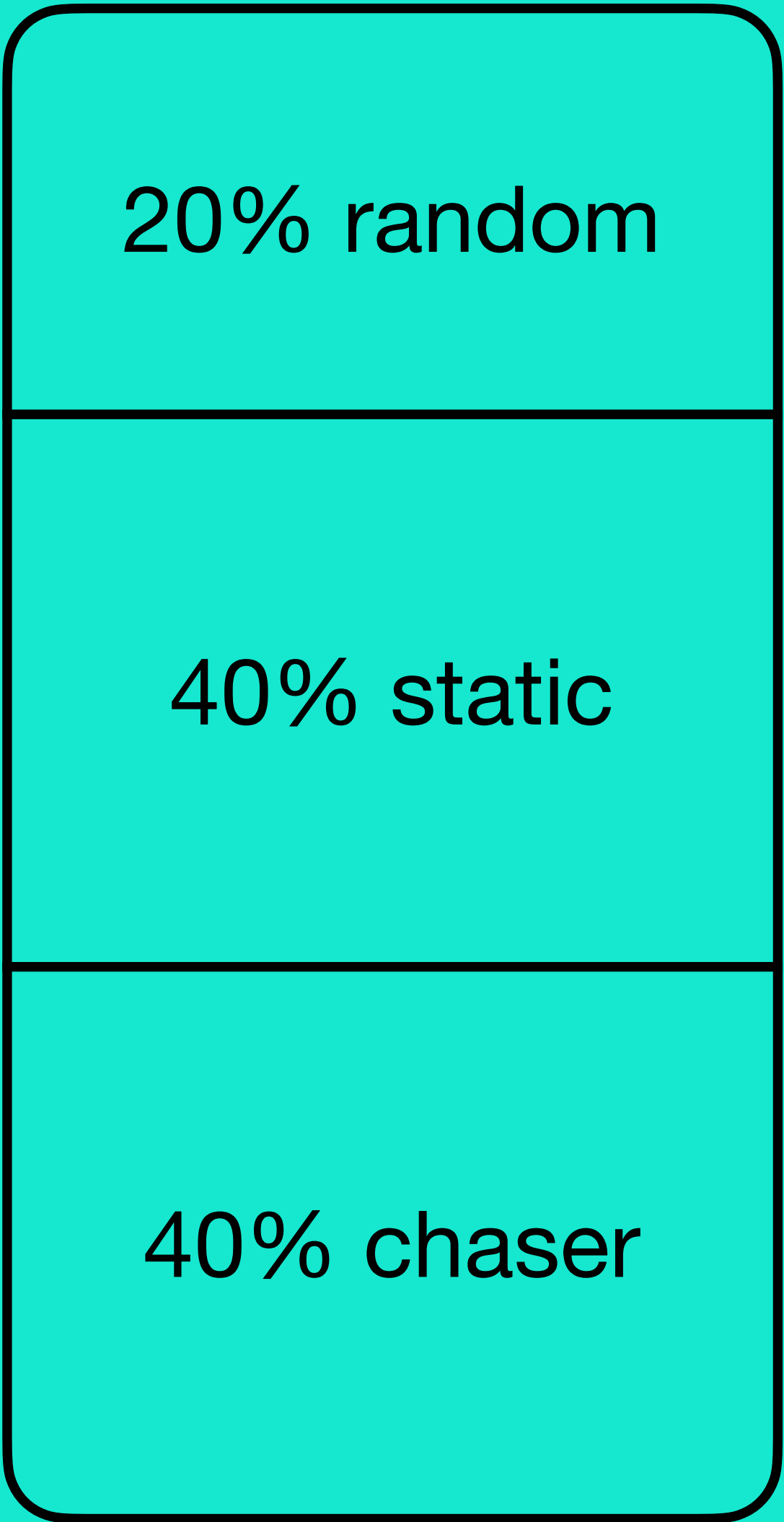
Final test



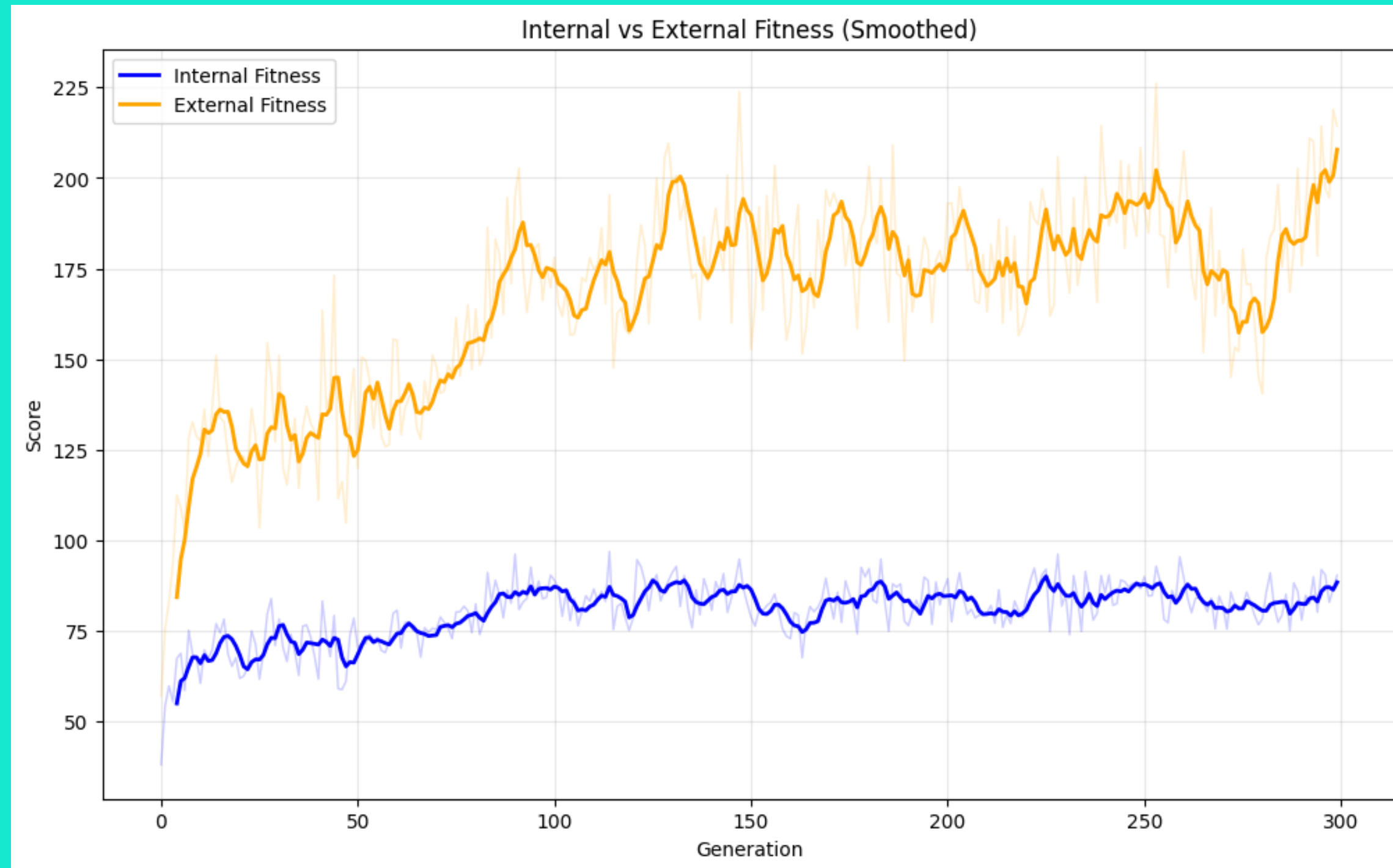
The selected
(after the evolution)



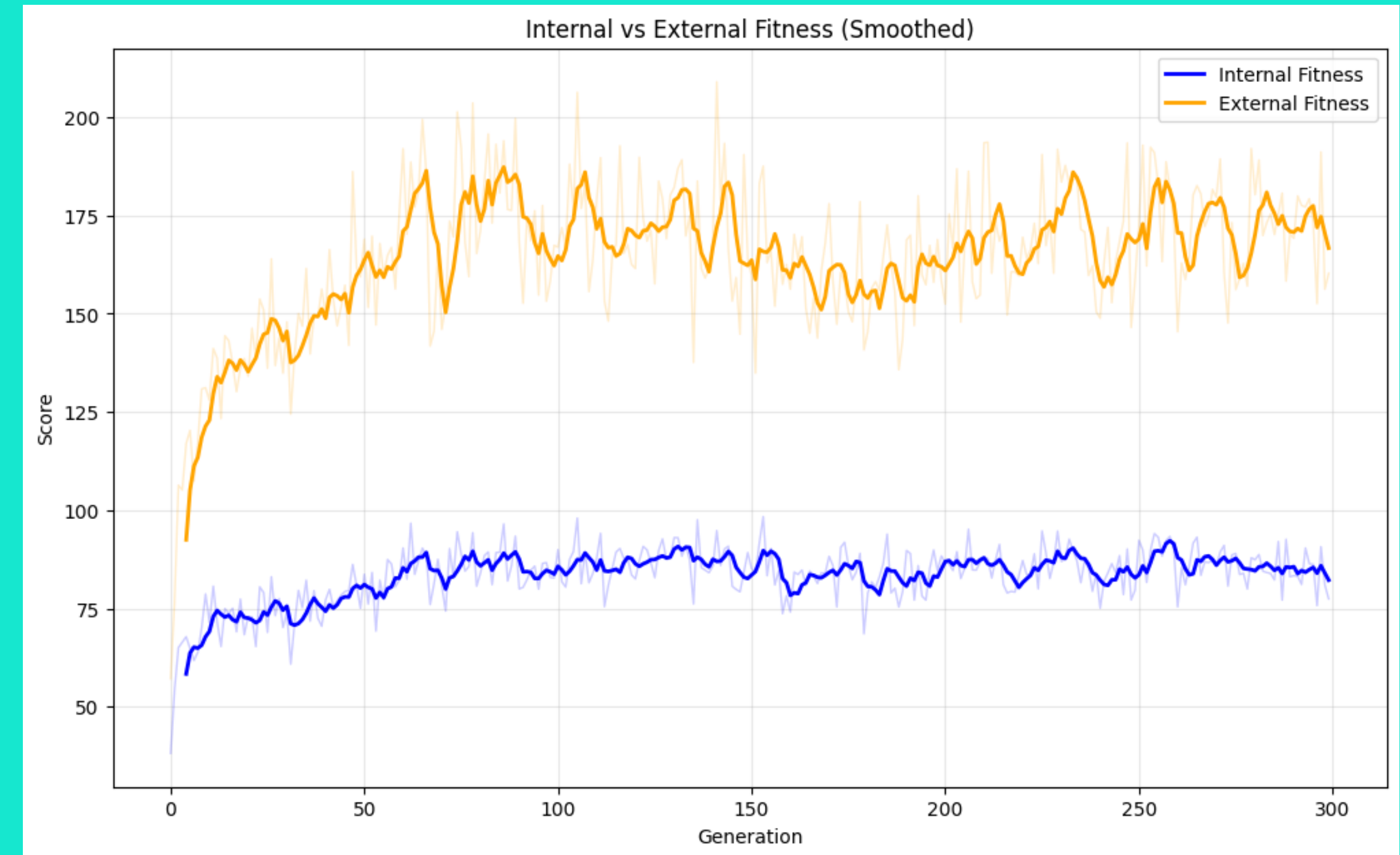
Random robots
(they don't change during the evolution)



Some results (1/2)



- Population: 50
- Generations: 300
- Elitism: 2



- Population: 50
- Generations: 300
- Elitism: 4