



Reinforcement Learning in Action: Creating Arena Battle Al for 'Blade and Soul'

Jinyun Chung, Seungeun Rho Reinforcement Learning Team / Game Al Lab / Al Center



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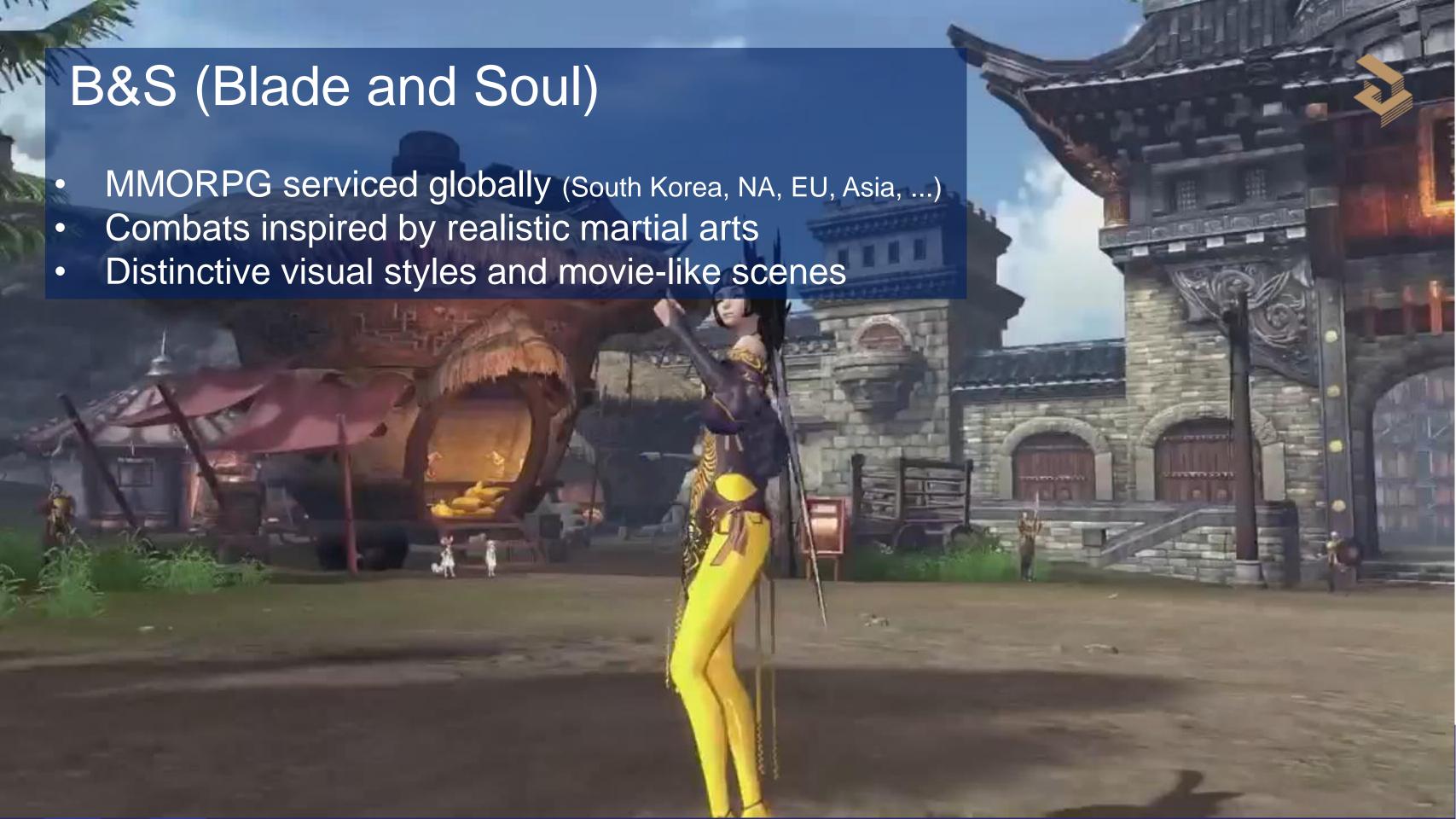
Contents



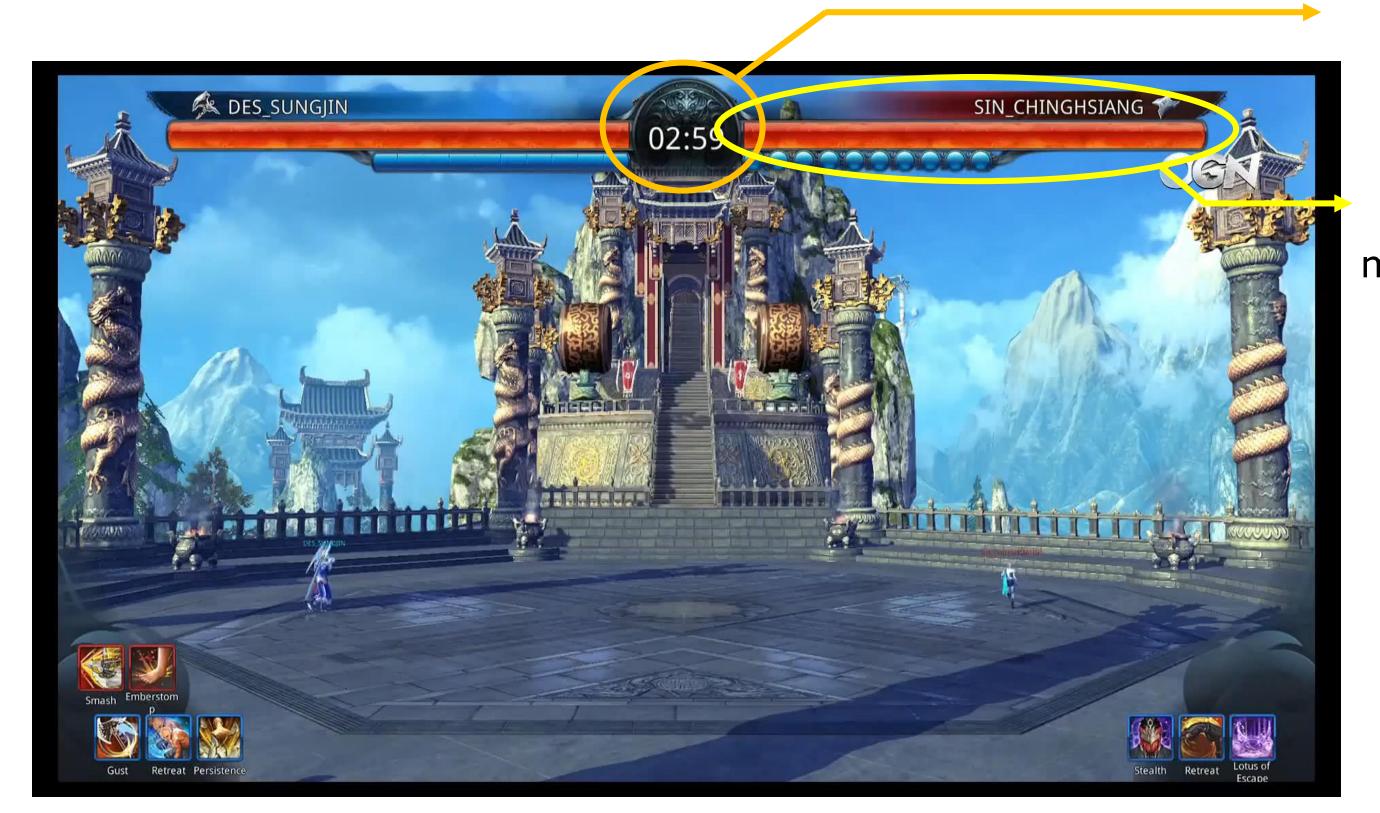
- The Problem B&S Arena Battle
- Methods & Tips
- Experiments
- Demo & Blind Match







B&S Arena Battle



Time limit 3 min

Goal make opponent HP to 0

Research Goal



Pro-Level AI in 1v1 Arena Battle with RL

AI with Hand-written rule

B1 or Tree

Print of Brit profile

B1 or Tree

Print of Brit profile

B2 or Tree

B3 or Tree

B4 or Tree

B5 or Tree

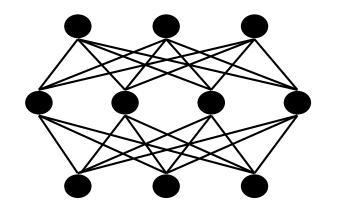
B6 or Tree

B7 or Tree

B8 or Tr



RL-based AI of various styles







Research Goal



Pro-Level AI in 1v1 Arena Battle with RL

- With 3 different styles
 - Aggressive, Balanced, Defensive
- Conditions
 - Class: Destroyer vs Destroyer
 - Skill Setting : Equal



What does our agent do?



Observation

- HP, SP
- Distance
- Skill cool time
- ..



Environment (Game)

Action

- Skill
- Move
- Target



Skill Systems of B&S



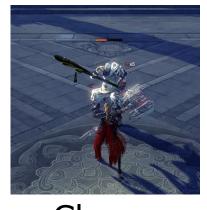




Damage Dealing



Judgement



Cleave

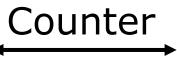








Typhoon







Persistence



Retreat

High cost

High

• **- -**V

Challenge 1) High Complexity





The game of Go



B&S Battle Arena

- Action Space ^ Game Length
 - Go ≈ 10^170
 - B&S $\approx 10^{1800} = (10(skill) * 5(move) * 2(target)) ^ 900 (game length)$

Solution: Action Space Reduction



Challenge 2) Real Time Response

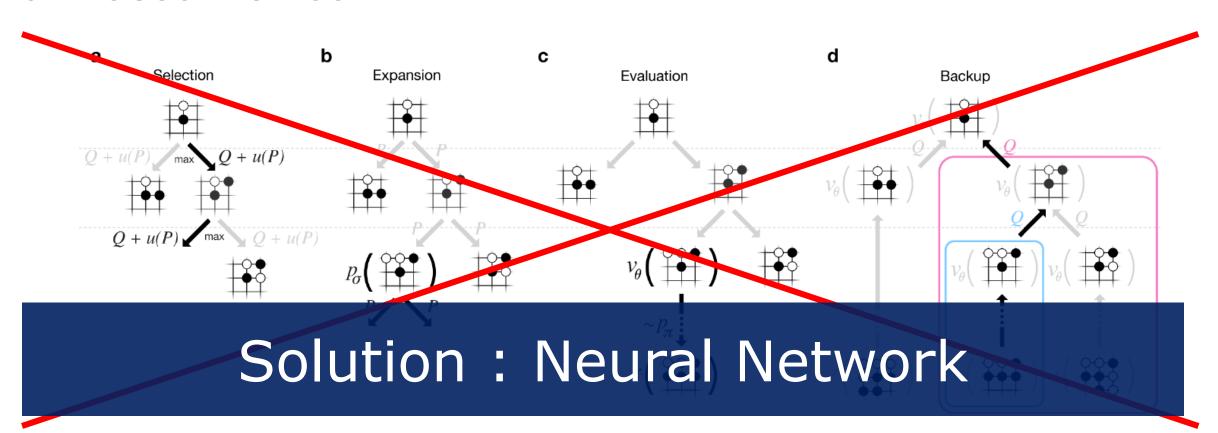


Decision Time Limitation

- Go: 60 sec

- B&S: 0.1 sec

No Search Based Method!





Challenge 3) Generalization



Best response differs for each player

- Pro-gamers' play styles are diverse.
- We don't know who will be the opponent
- The agent should be generalized to response optimally against unknown players



Solution: Who will be the opponent?

Opponent Pool of Various Selves



Challenge 4) Guiding Fighting Style



Aggressive

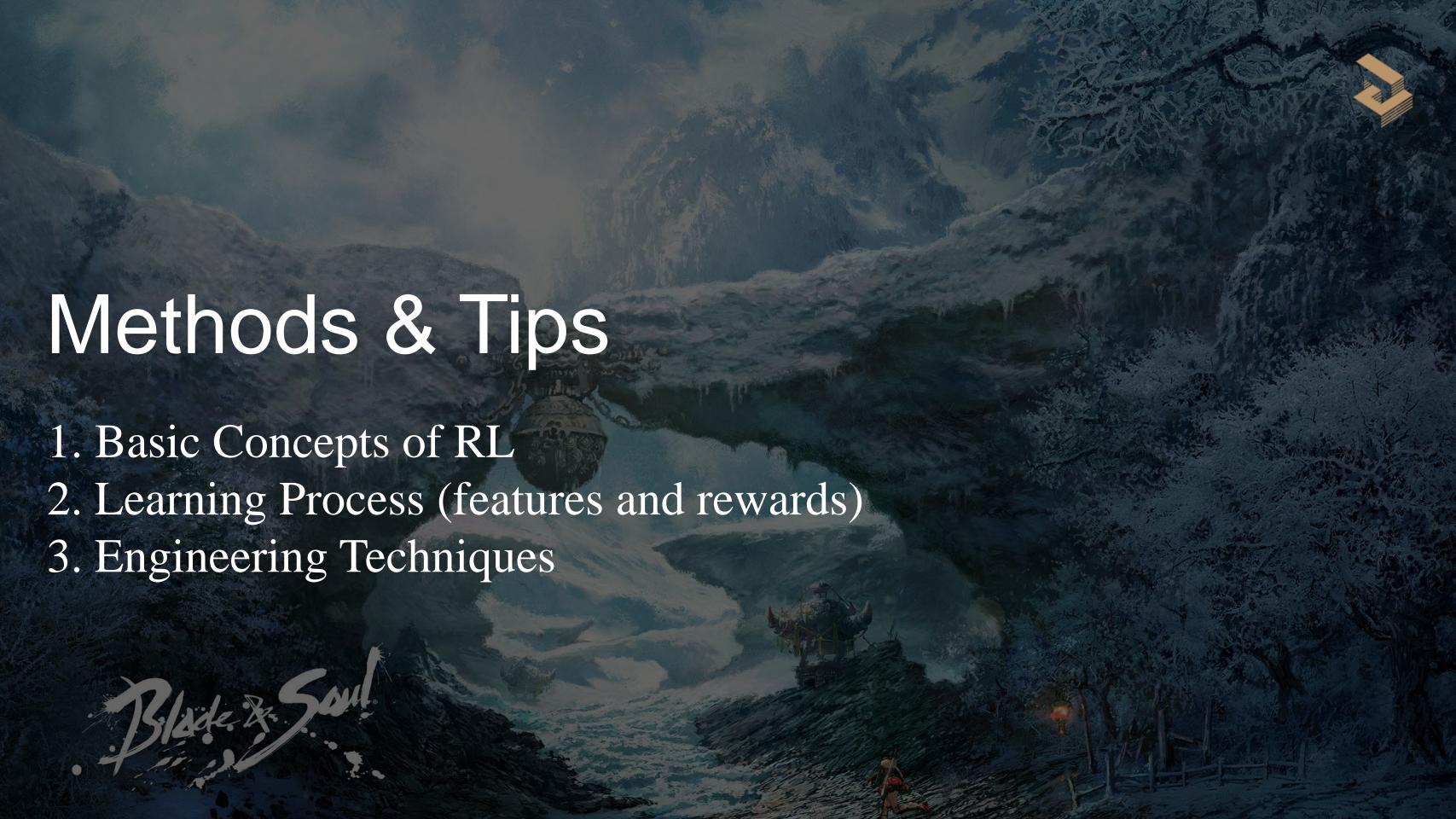
Balanced

Defensive

- 3 different fighting styles
 - Aggressive, Balanced, Defensive
- w/o hand-written rule

Solution: Reward Shaping





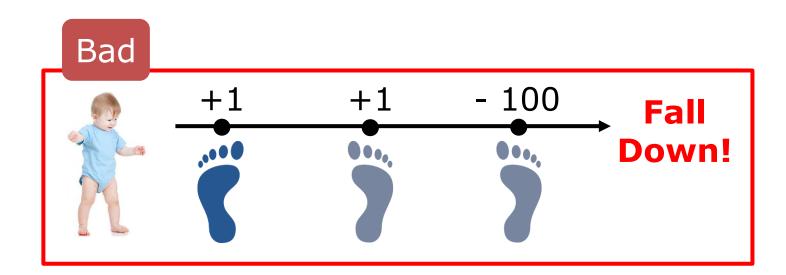
Learning how to walk

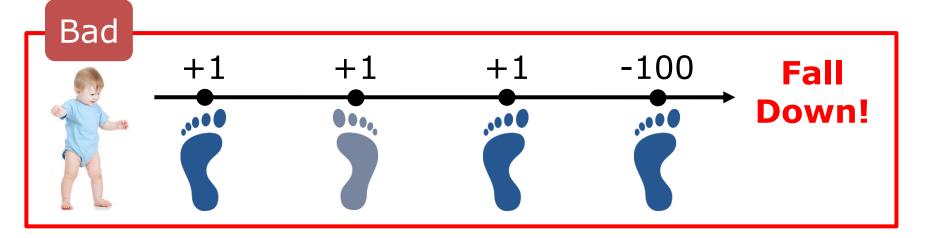


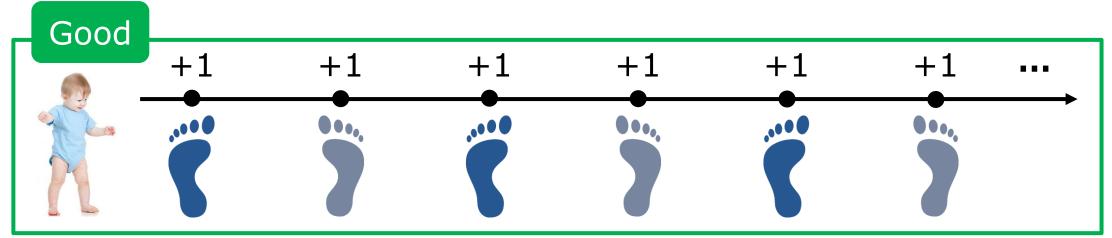


image ref: https://www.bundoo.com/articles/signs-your-baby-is-ready-to-walk/









- Reward
 - 1. Every timestep: +1
 - 2. Fall down: -100
- Objective
 Maximizing cumulative reward
- Agent modify policy
 - 1. Reinforce actions in good sequences
 - 2. Suppress actions in bad sequences



Additional Reward for Guiding Battle Style

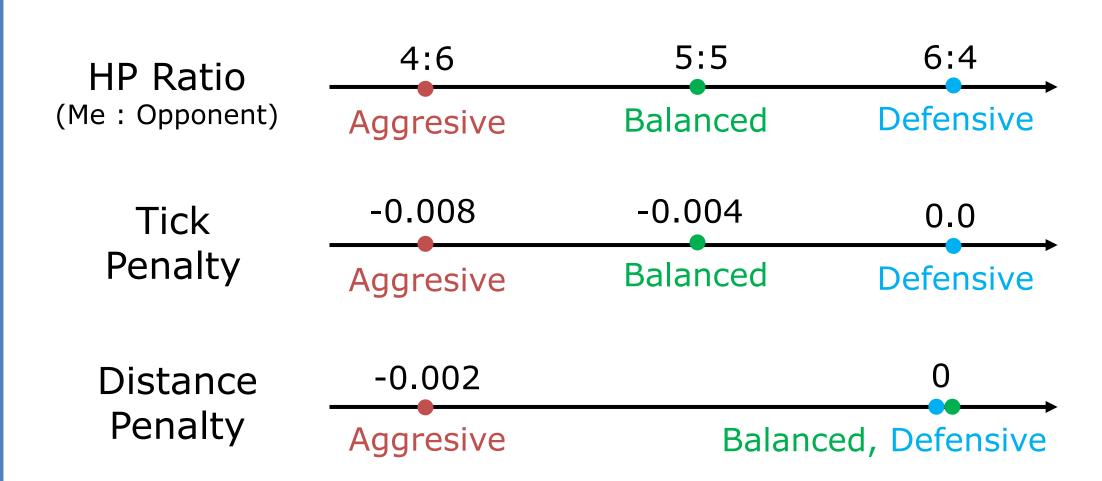


Basic

Win / Lose

HP Gap

Style





Agent-Environment Plot





(=Feature)

Represents how the agent sees the world



Action

(=Skill, Move)

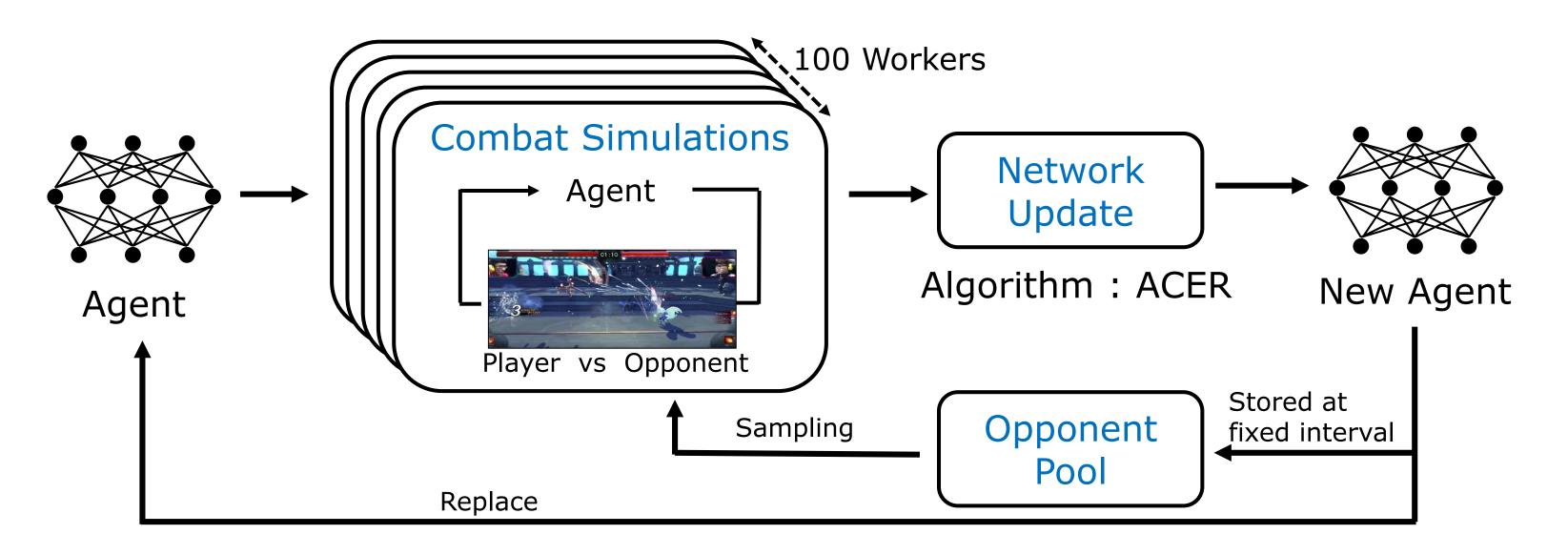
Environment

(Game Simulator)



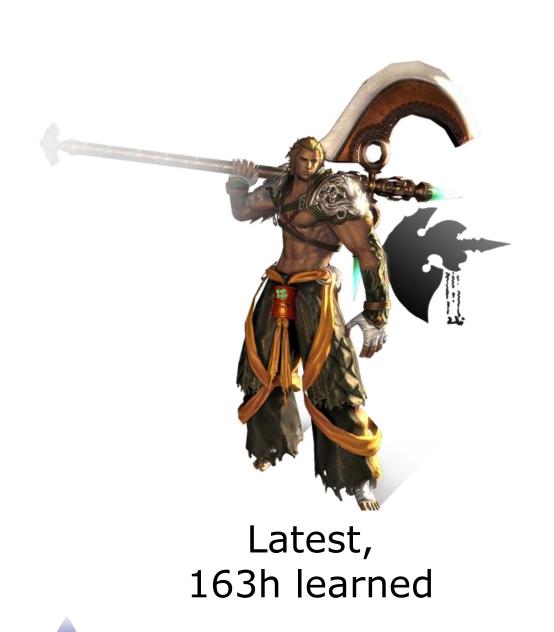
Learning Process



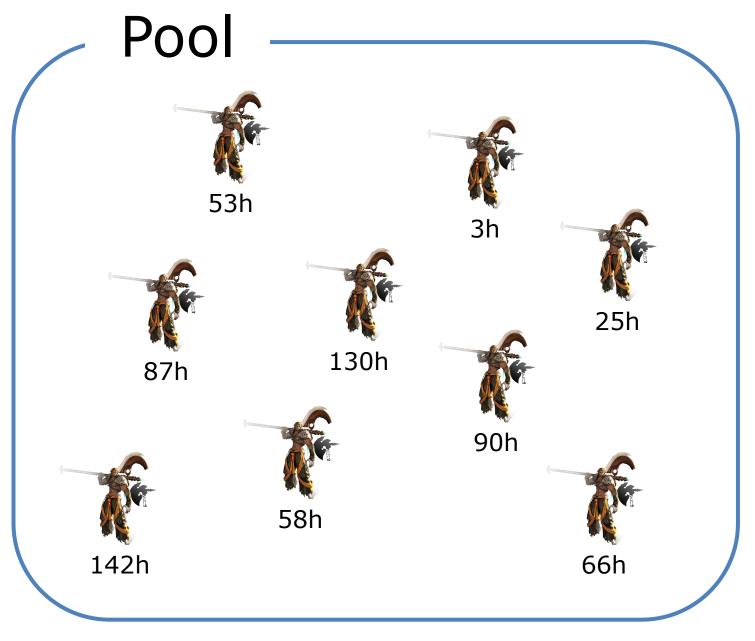


Who is the opponent? – The pool of past selves!





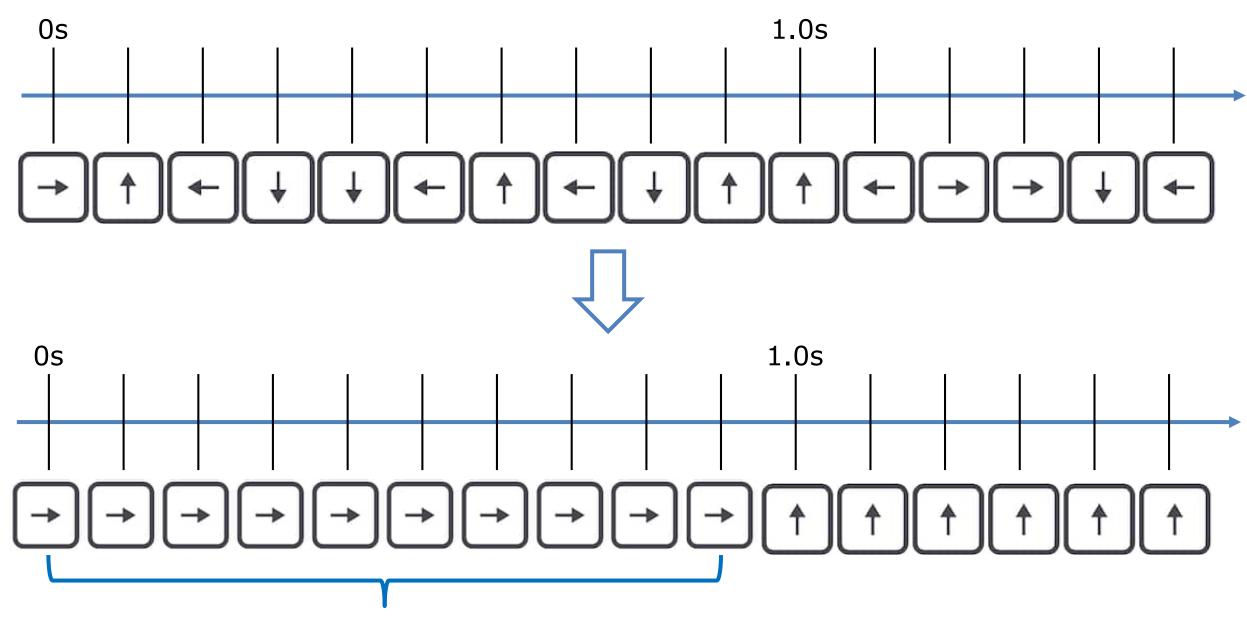
VS





Move Space Reduction – Decision Frequency





Maintain Decision for 1 sec



Skill Space Reduction





- When the opponent is too far, most of skills are blocked
- Boosts the learning process in early phase



Feature Engineering





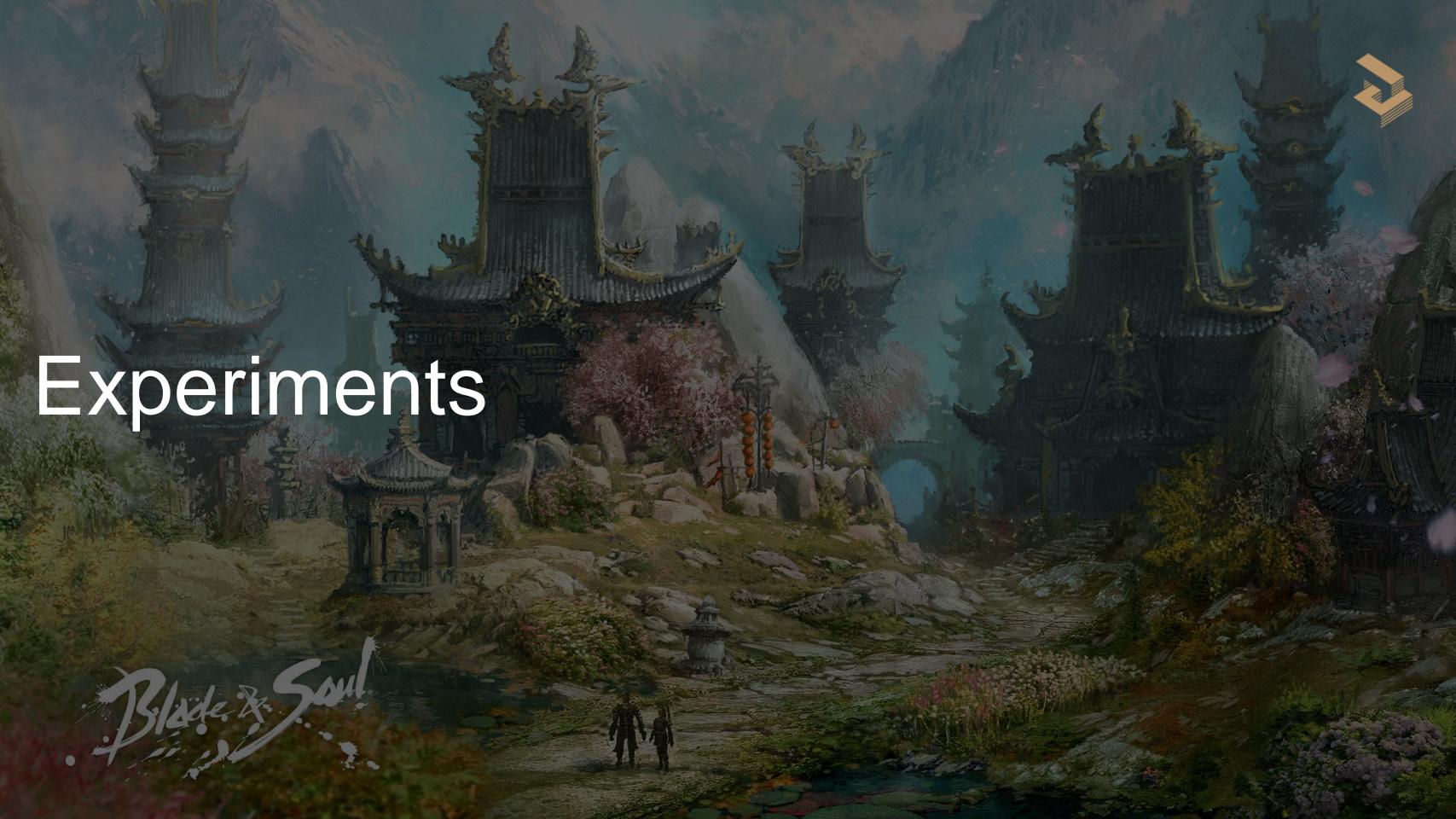
DRAG

- Discretization of continuous features
- Helps to distinguish fine skill range

<Distance feature>

$$0 \sim 1$$
 $0 \sim 3m$
 $0 \sim 3m$
 $1,0,0,0,0]$
 $3 \sim 5m$
 $[0,1,0,0,0]$
 $5 \sim 8m$
 $[0,0,1,0,0]$
 $8 \sim 16m$
 $[0,0,0,0,1,0]$
 $16m \sim$
 $[0,0,0,0,1]$





Sparring Partner – Built-in Al





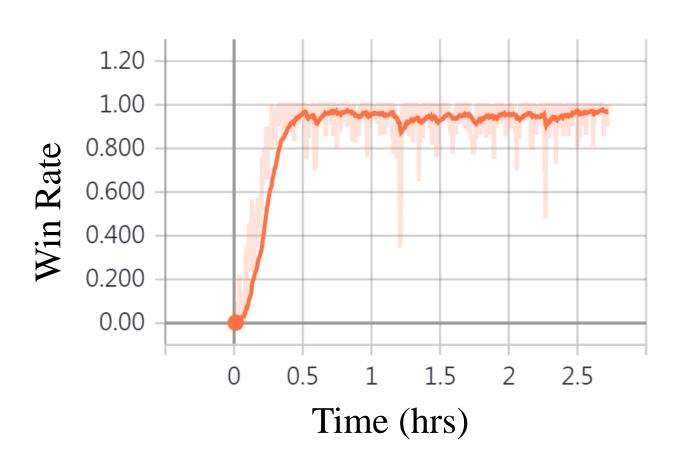
VS

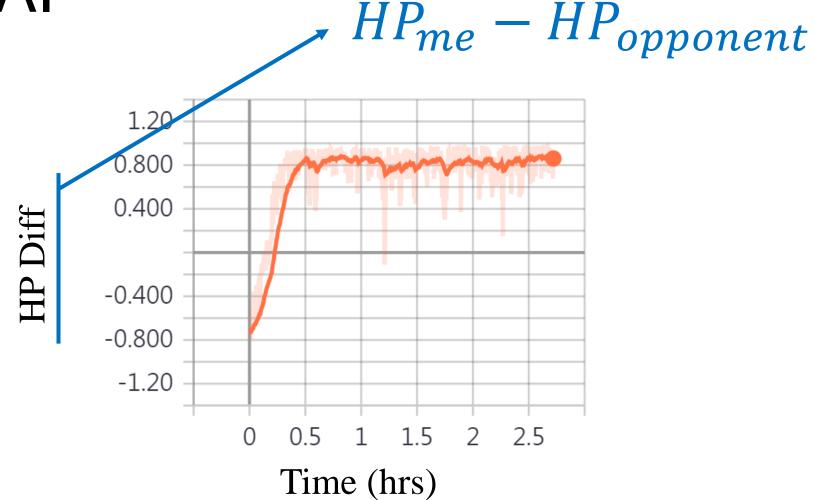


Built-in Al (from Infinity Tower)



Sparring Partner – Built-in Al





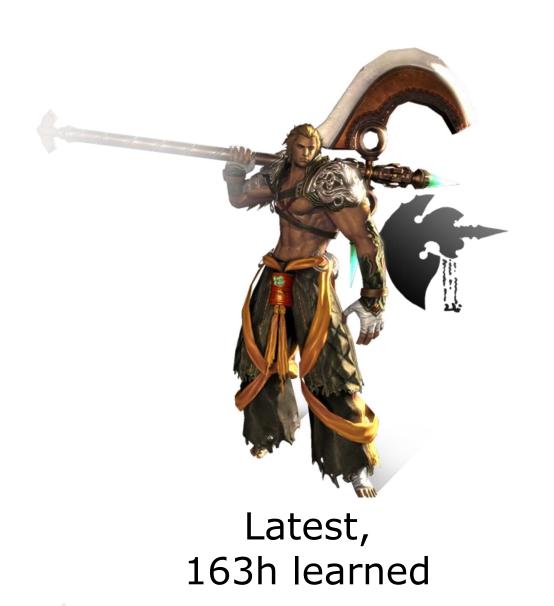
- 90% win rate in 30 min Level of difficulity against human :
- 100% win rate in 5 hours

 Easy

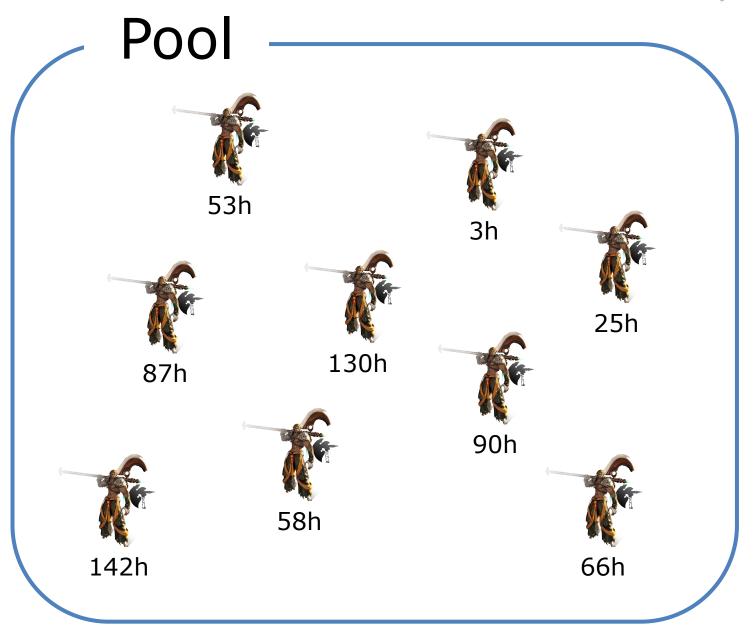


Sparring Partner – Self play





VS







Sparring Partner – Self play



Measuring generalization performance

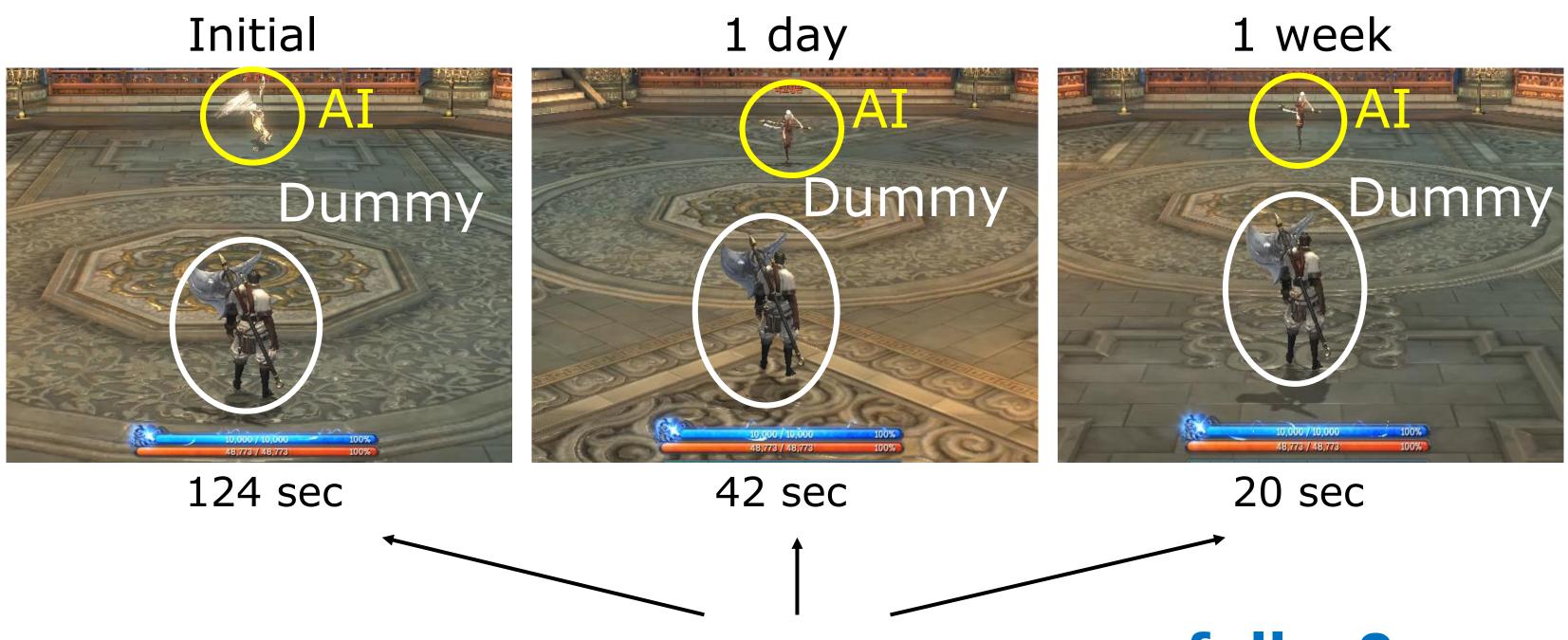
- vs Built-in AI (Infinity Tower AI)
 - Training time 1 day : 84% win rate
 - Training time 2 days: 96% win rate
- vs Skillful human player (rating: top 10%)
 - 5 matches results : Al 5 0 Human
 - However ... after about a hundred games of play, human adapt and exploit weak point of Al.
 - Training longer, weak points disappeared





Learning Progress



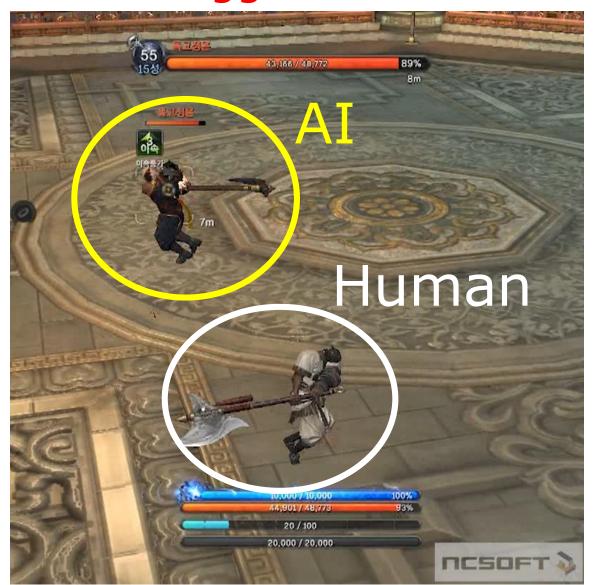


Time consumption for reduce dummy's HP from full to 0

Different Fighting Styles



Aggressive



Defensive





Pretest with professional gamers



AI - Human

	Aggressive	Balanced	Defensive
Pro-gamer 1	5-1	2-1	1-2
Pro-gamer 2	4-0	2-4	4-1
Total	9-1 (90%)	4-5 (44%)	5-3 (63%)

Pro-gamer 1: Yuntae Son (The winner of 2017 B&S World Championship)

Pro-gamer 2: Shingyeom Kim (The winner of 2015 and 2016 B&S World

Championship)



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NCSOFT

Results



AI - Human

	Aggressive	Balanced	Defensive
Pro-gamer 1	5-1	2-1	1-2
Pro-gamer 2	4-0	2-4	4-1
Blind Match	2-0	1-2	0-2
Total	11-1 (92%)	5-7 (42%)	5-5 (50%)



Conclusion



- RL actually works!
- Can guide fighting styles w/o hand-written rules
- Pool of opponents help generalization
- Space reduction helps training



