

Autonomous Learning of Ball Trapping in the Four-legged Robot League

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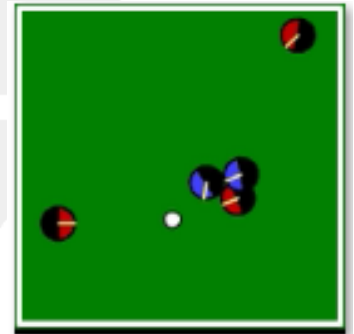
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Motivation

- Passwork in the four-legged robot league
 - KeepAway Soccer [Stone et al. 2001]
 - Benchmark of good passing abilities in the simulation league
 - Passing Challenge
 - Technical challenge in this year

It is too difficult for dogs

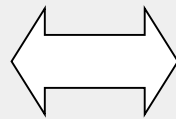


KeepAway Soccer

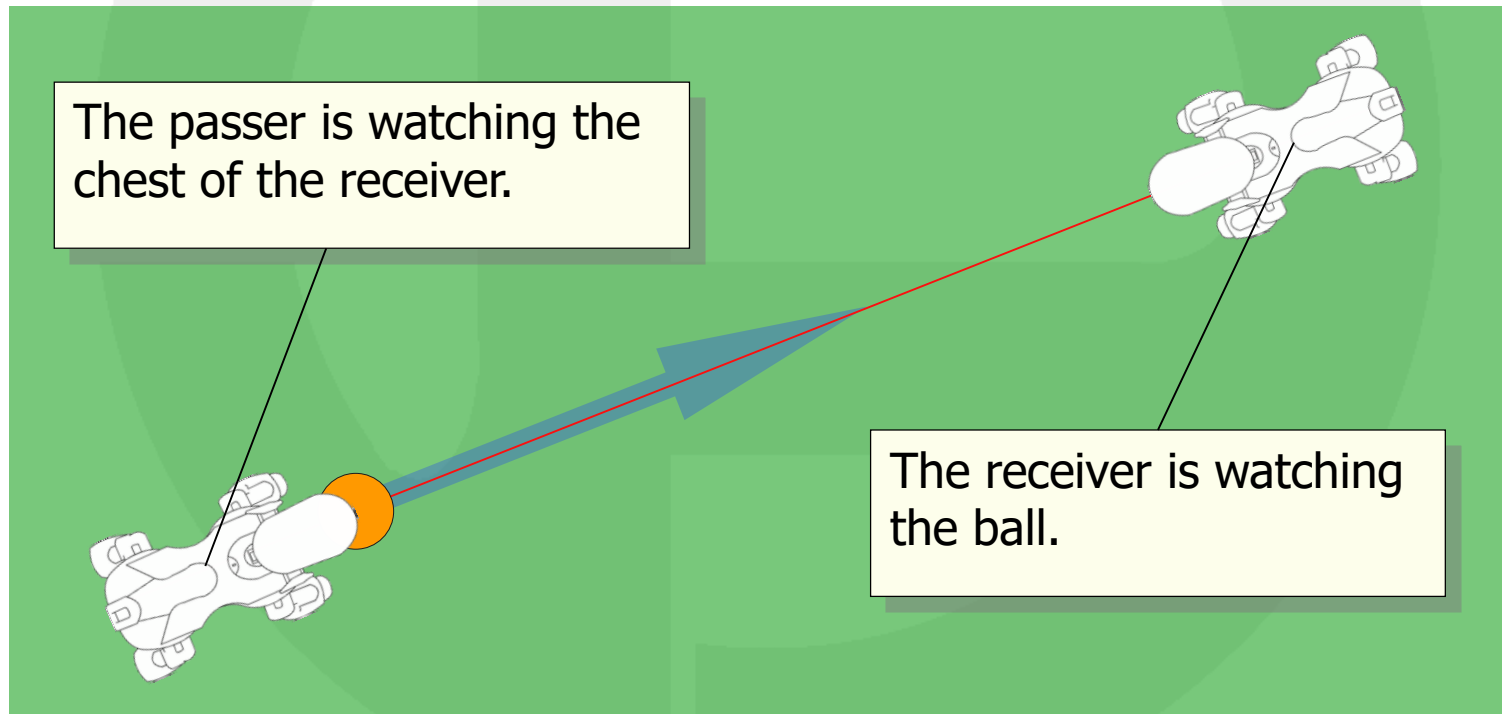
<http://www.cs.utexas.edu/~AustinVilla/sim/keepaway/>

Ball Trapping

- Stop and control an oncoming ball



One-dimensional Model



Autonomous Method

- Same way as diligent humans



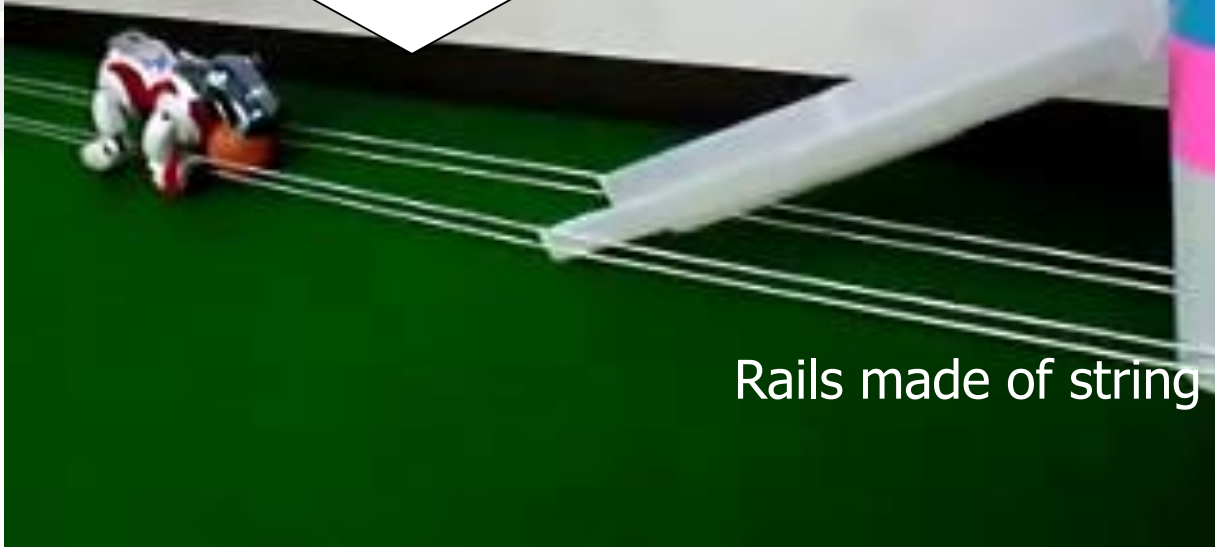
Kick Wall

Training Equipment

Limit ball's movement and robot's locomotion to one-dimension

Slope made of cardboard

Rails made of string

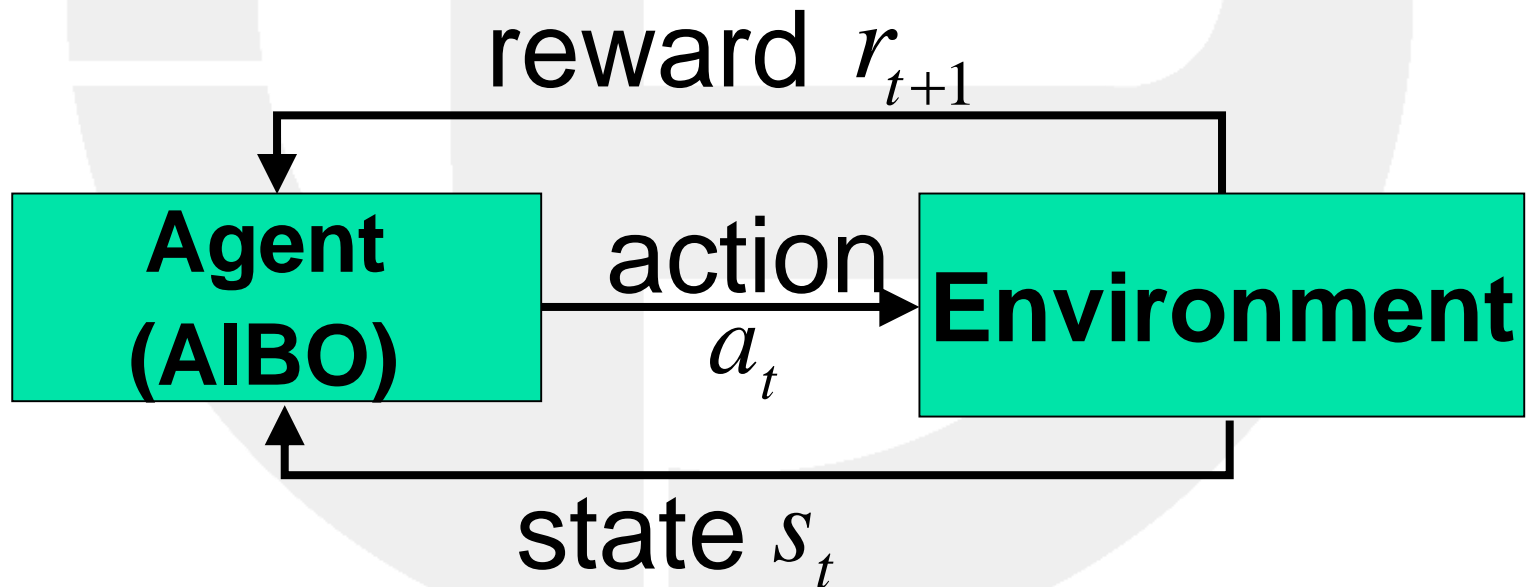


Learning Method

- Sarsa(λ) [Rummery and Niranjan 1994; Sutton 1996]
 - Reinforcement learning algorithm
- Tile-coding (aka CMACs [Albus 1975])
 - Linear function approximation
 - For speeding up their learning

Reinforcement Learning

- Acquire maps from state input to action output maximizing the sum of rewards



In our study, each time step $t = 0, 1, 2, \dots$ mean 0ms, 40ms, 80ms, ...

Implementation

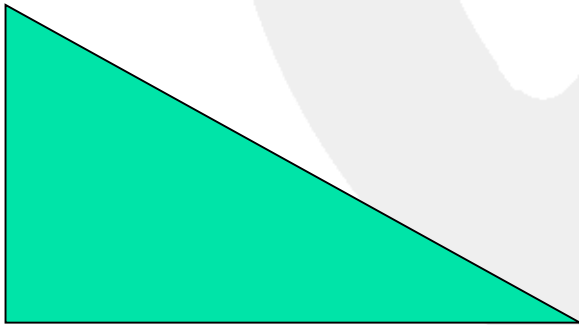
- State $s_t = (x_t, dx_t)$
 - x_t ··· The distance from the robot to the ball
[0,2000] (mm)
 - dx_t ··· The difference between the current x_t and the previous x_t of one time step before.
[-200,200] (mm)
- Action a_t
 - *ready* ··· Move its head to watch the ball
 - *trap* ··· Initiate the trapping motion

Implementation

- Reward r_{t+1}
 - Positive
 - If the ball was correctly captured between the chin and the chest after the *trap* action.
 - Negative
 - If the *trap* action failed, or
 - If the ball touches the chest PSD sensor before the *trap* action is performed.
 - Zero
 - Otherwise

Implementation

- Episode
 - The period from kicking the ball to receiving any reward other than zero

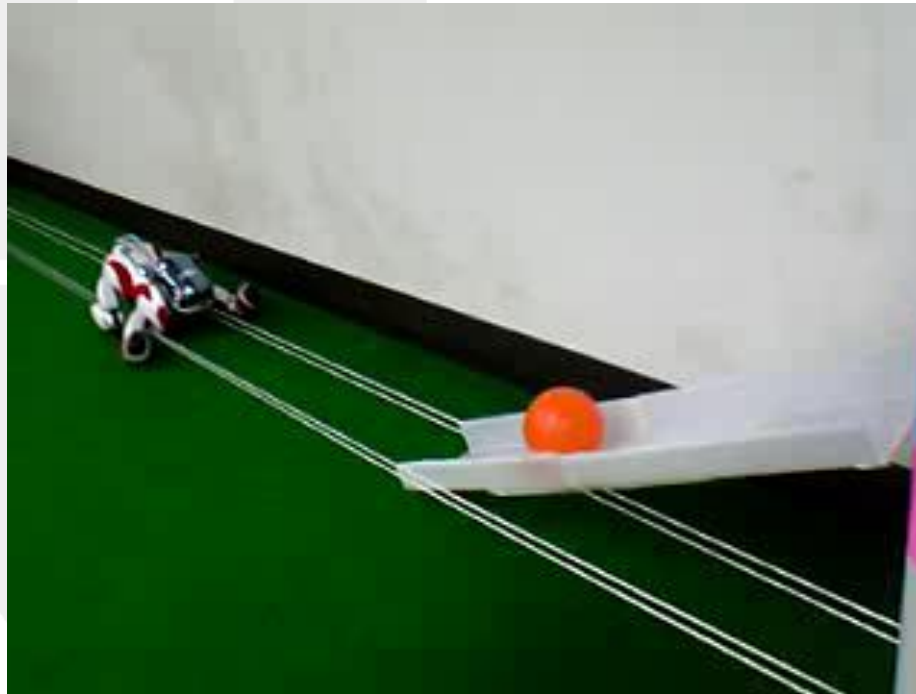


Experiments

- Using one robot
- Using two robots
 - without communication
 - with communication

Using One Robot

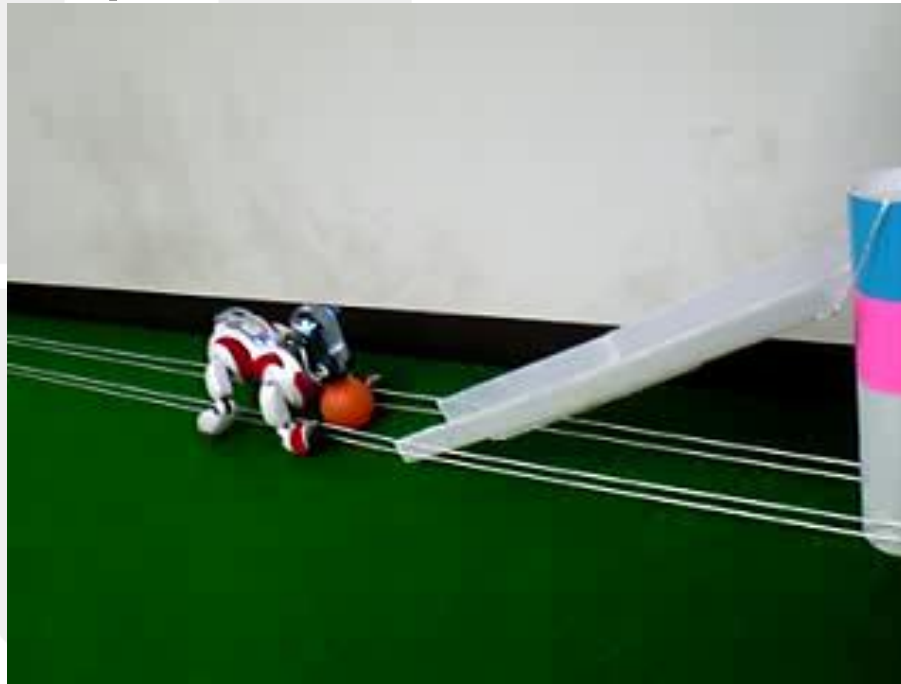
- Earlier phase



<https://youtu.be/hv1sgIZLpKA>

Using One Robot

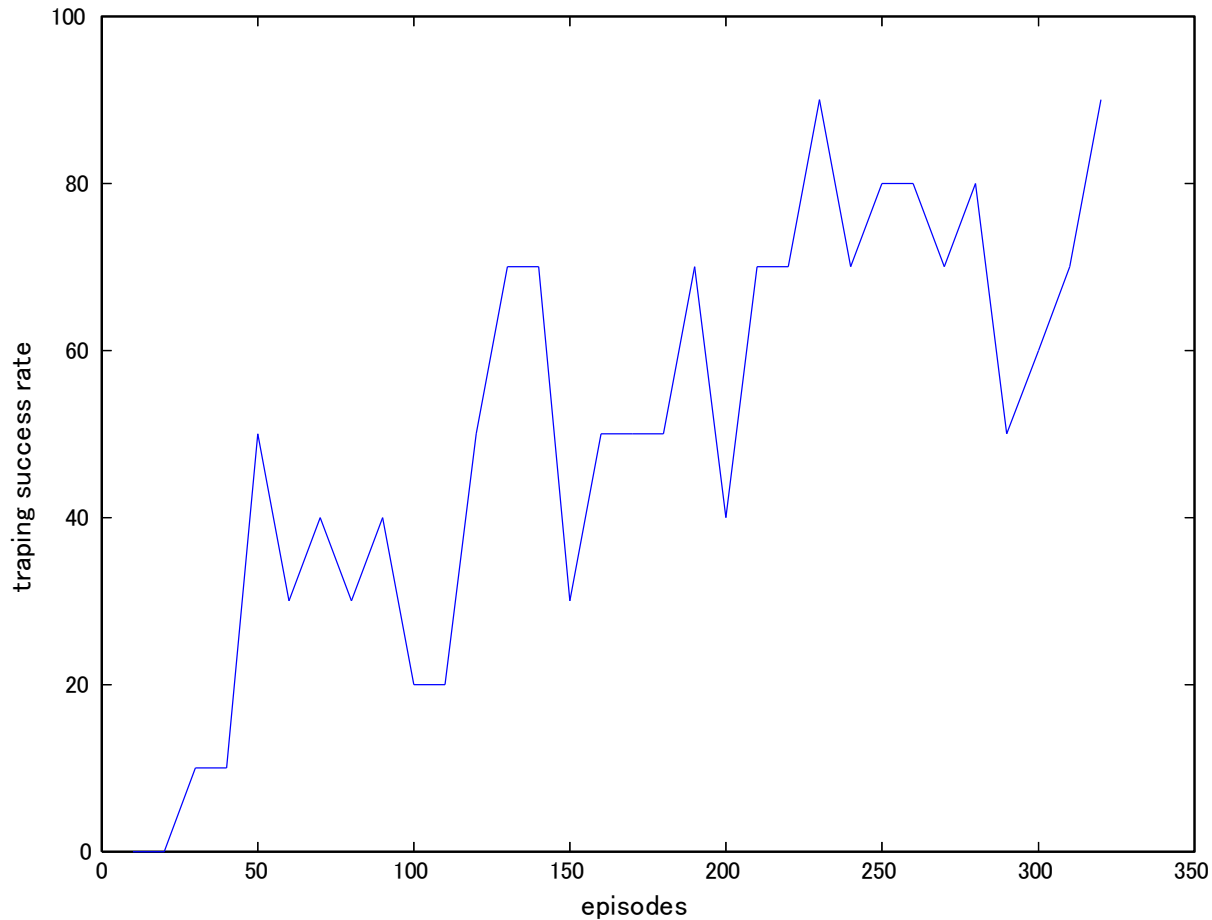
- Later phase



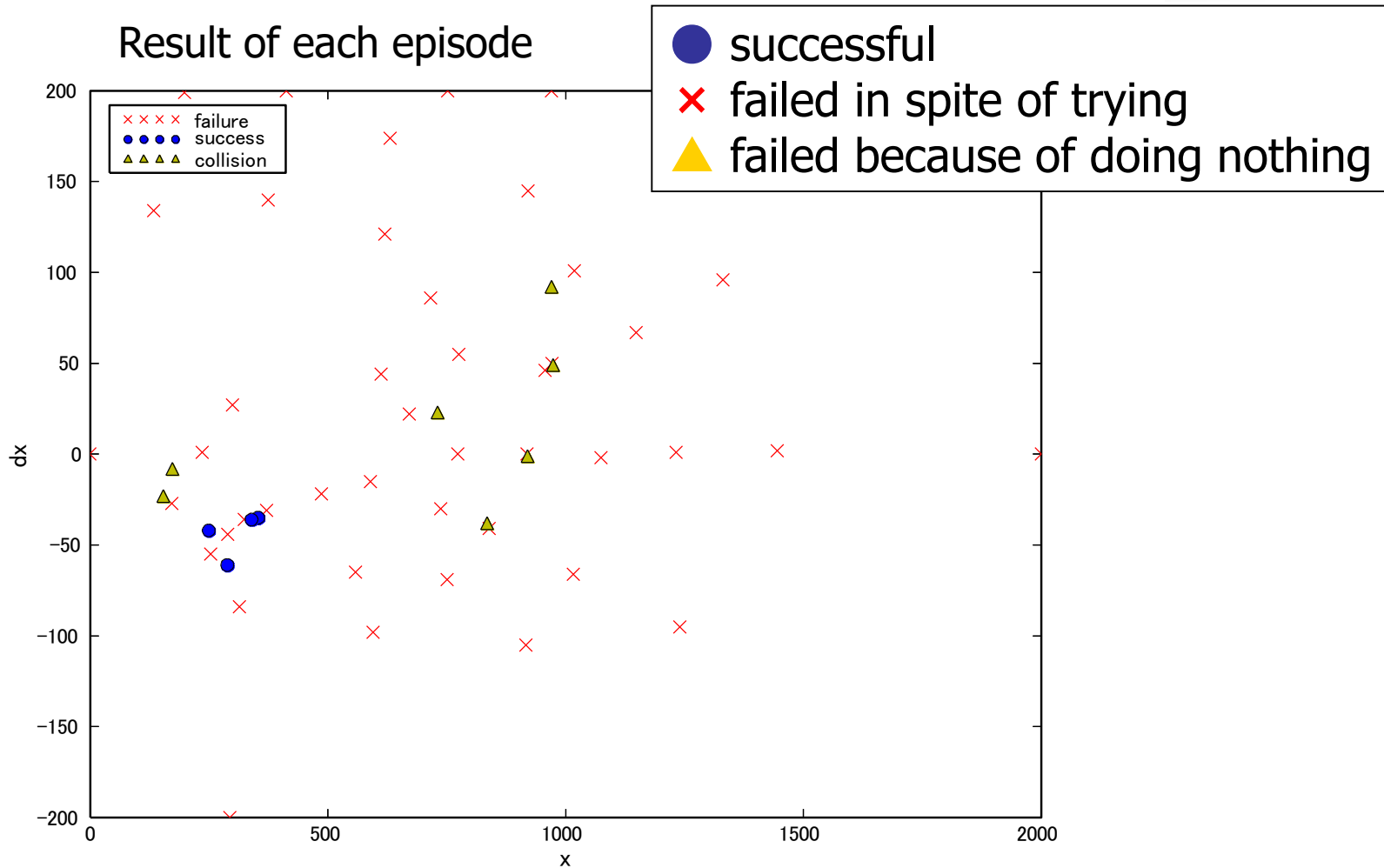
<https://youtu.be/XJBllv7wJXQ>

Result of Learning Using One Robot

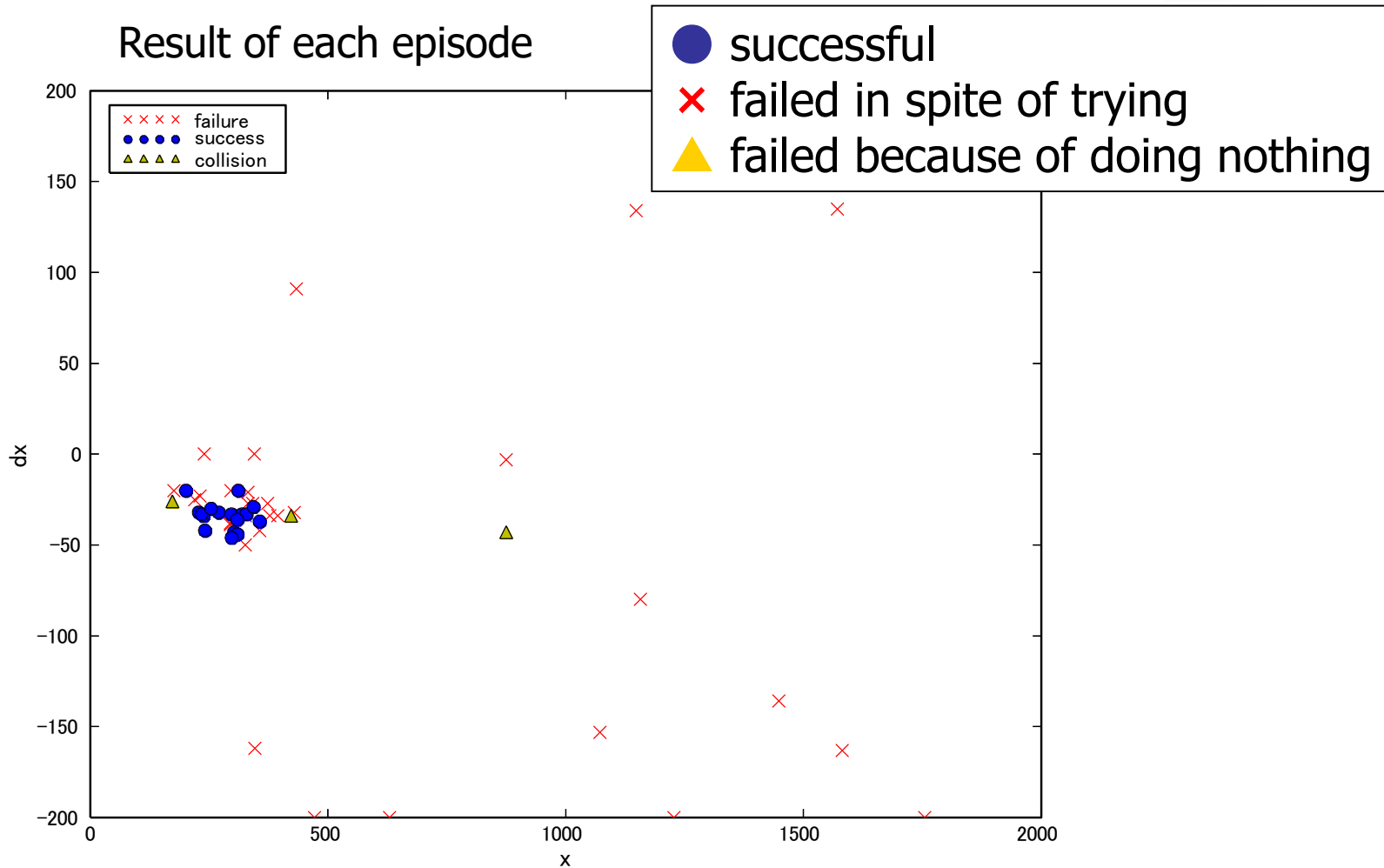
trapping success rate every 10 episodes



Episodes 1...50

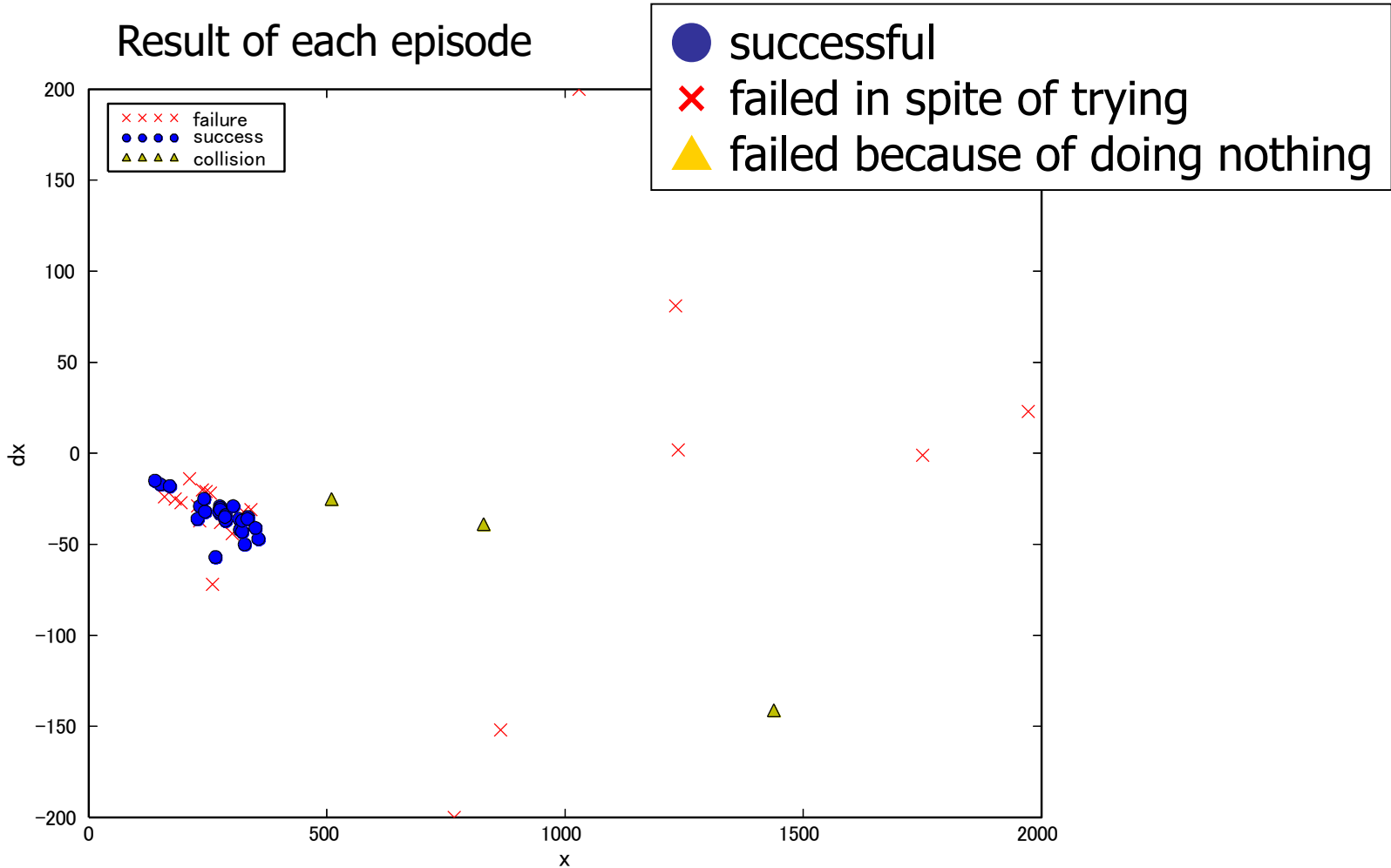


Episodes 51...100



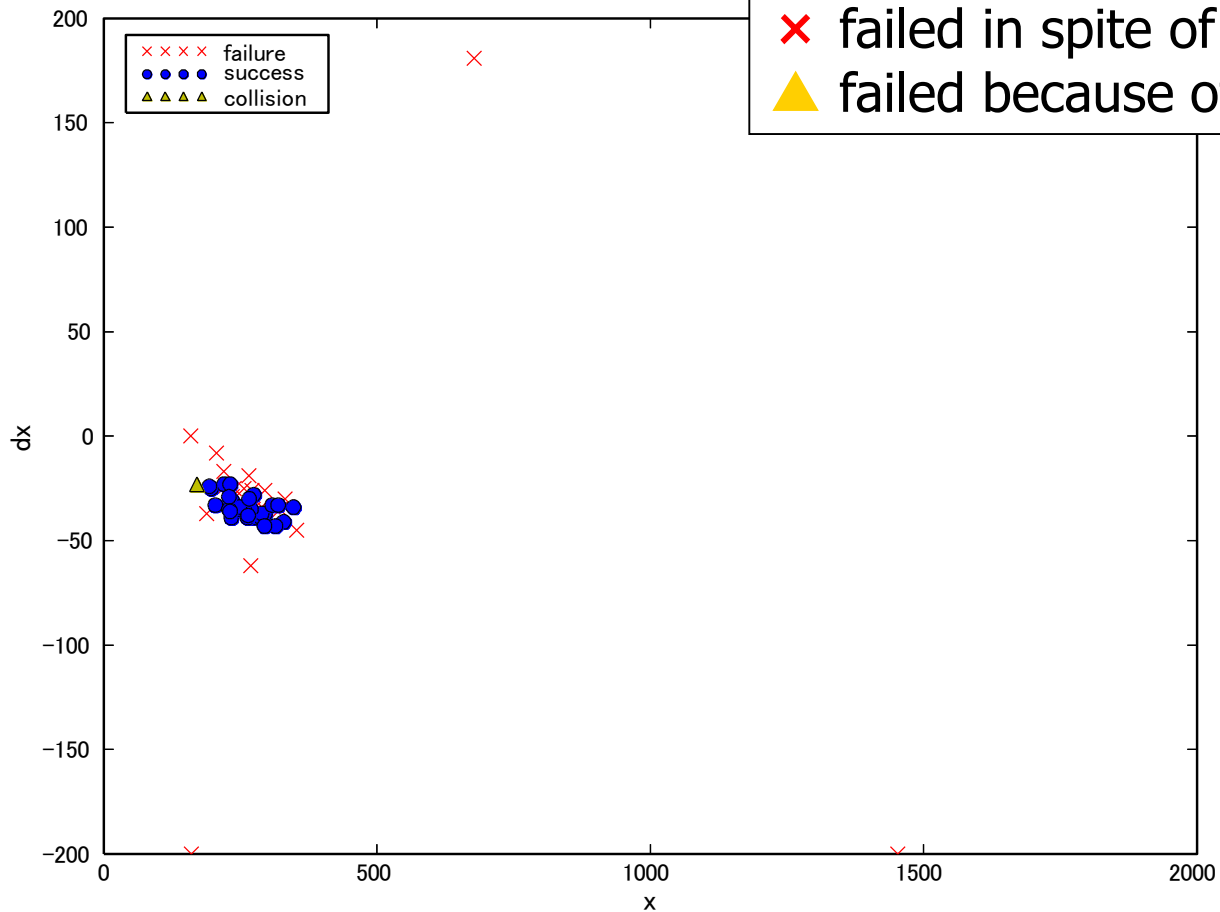
Episodes 101...150

Result of each episode



Episodes 151...200

Result of each episode

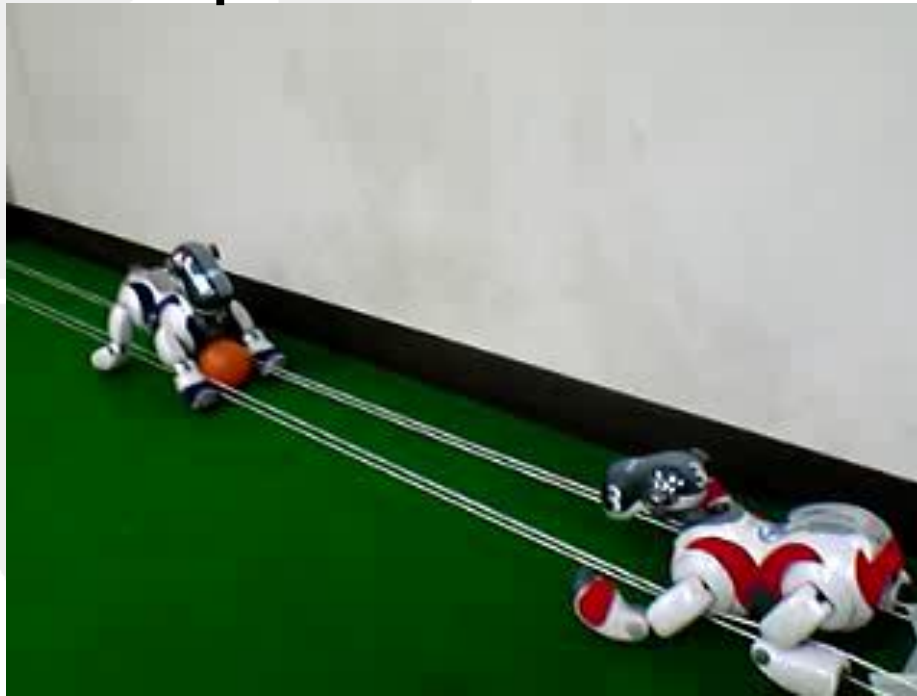


Using Two Robots

- Simply replace slope with another robot
 - Active Learner (AL)
 - Original robot
 - Same as in case of training using one robot
 - Passive Learner (PL)
 - Replaces slope
 - Does not approach the ball if the trapping failed

Using Two Robots

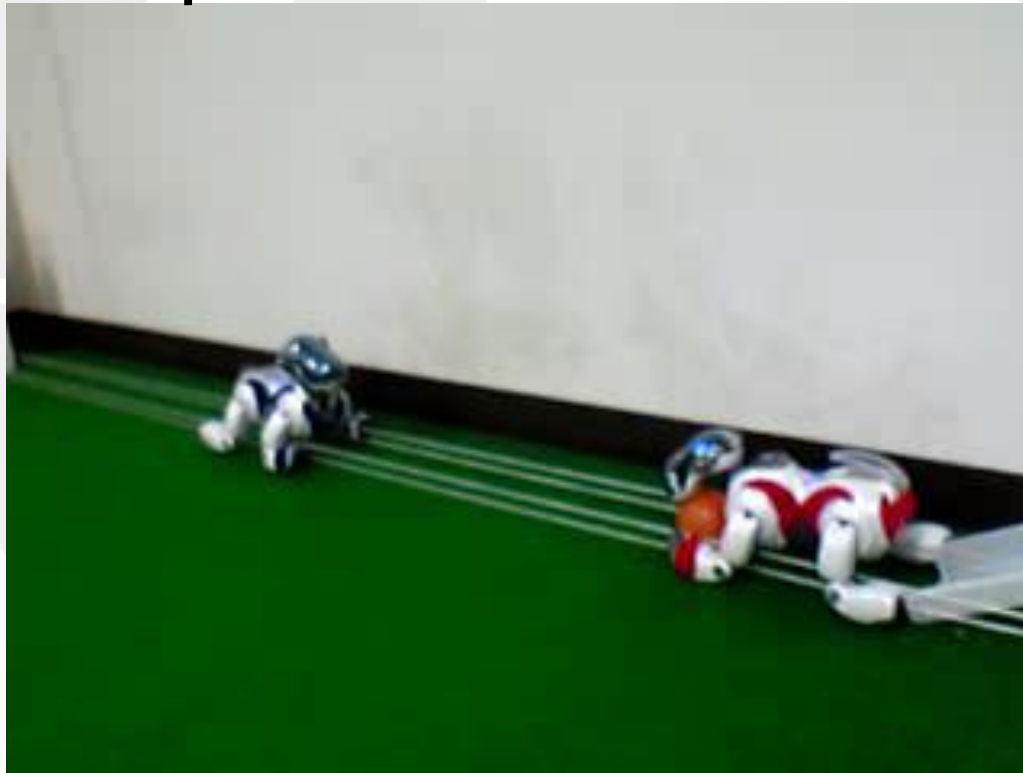
- Earlier phase



<https://youtu.be/sXkVYZjOzjg>

Using Two Robots

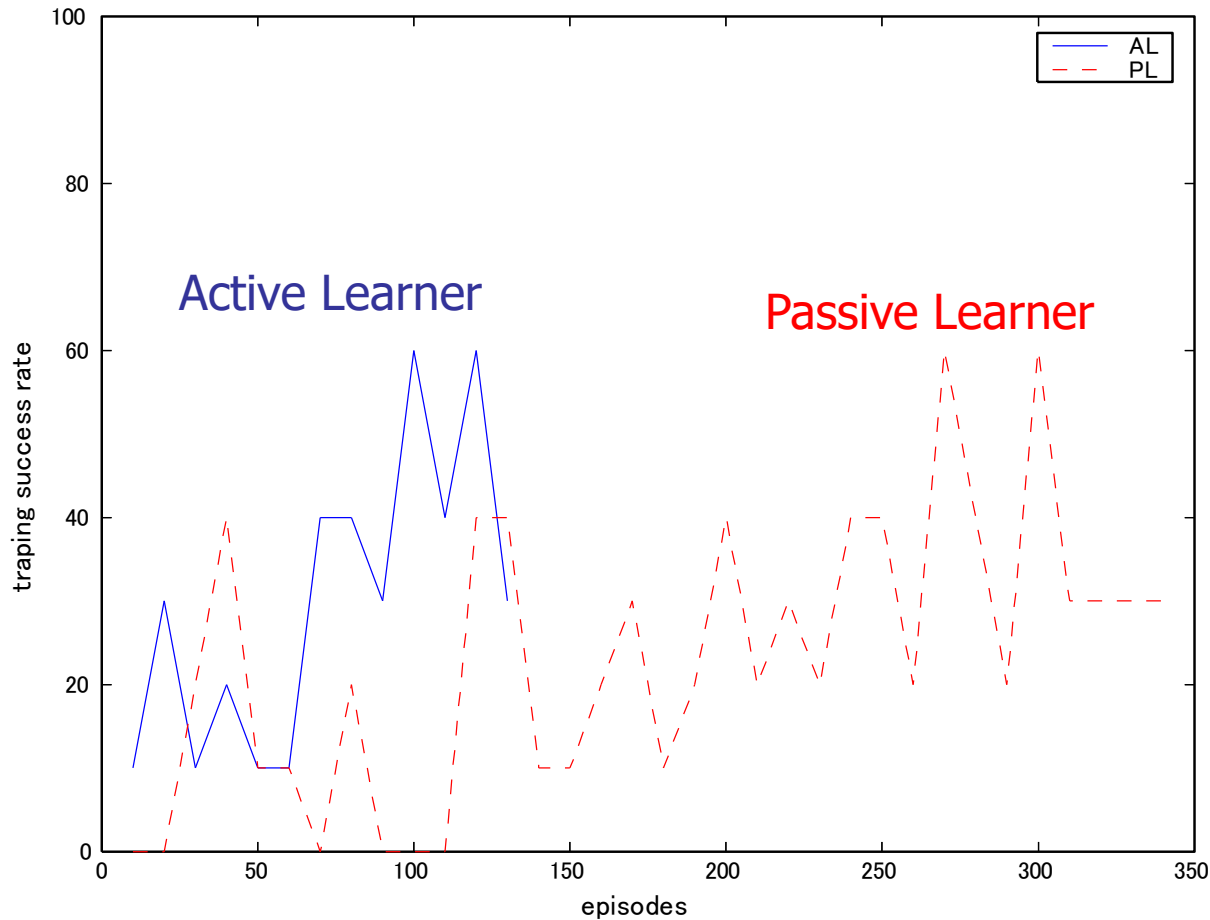
- Later phase



<https://youtu.be/opvoyv9h-GU>

Result of Learning Using Two Robots Without Communication

trapping success rate every 10 episodes



Problem of Using Two Robots

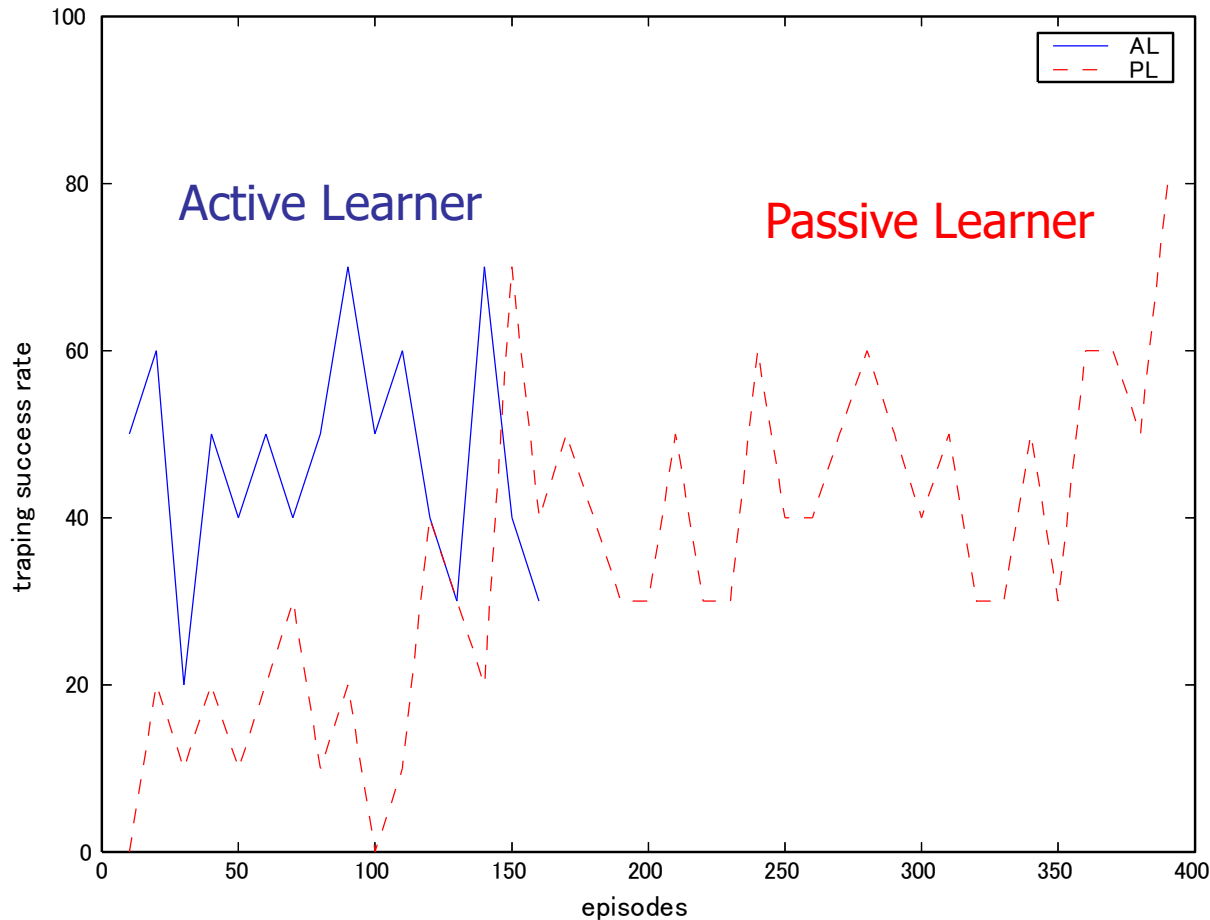
- Takes a long time to learn
 - AL can only learn when PL itself succeeds
 - Cannot learn if the ball is not returned
 - Even if we use only two ALs, the problem is not resolved
 - Just learn slowly, though simultaneously.

Solution

- Sharing their experiences
 - Their experiences include
 - Action a_t (*trap* or *ready*)
 - State variables $s_t = (x_t dx_t)$
 - Reward r_{t+1}

Result of Learning Using Two Robots With Communication

trapping success rate every 10 episodes



Conclusion

- The goal of pass-work is achieved in one-dimension
 - learned the skills without human intervention
 - learned more quickly by exchanging experiences with each other

Future Work

- Extend trapping skills to two-dimensions
 - Layered Learning [Stone 2000]
 - Make goalies stronger
- Make robots learn passing skills simultaneously

Thank you for your attention!



Bremen is a good town!

JollyPechie