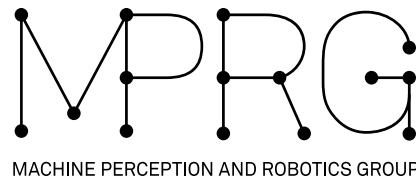


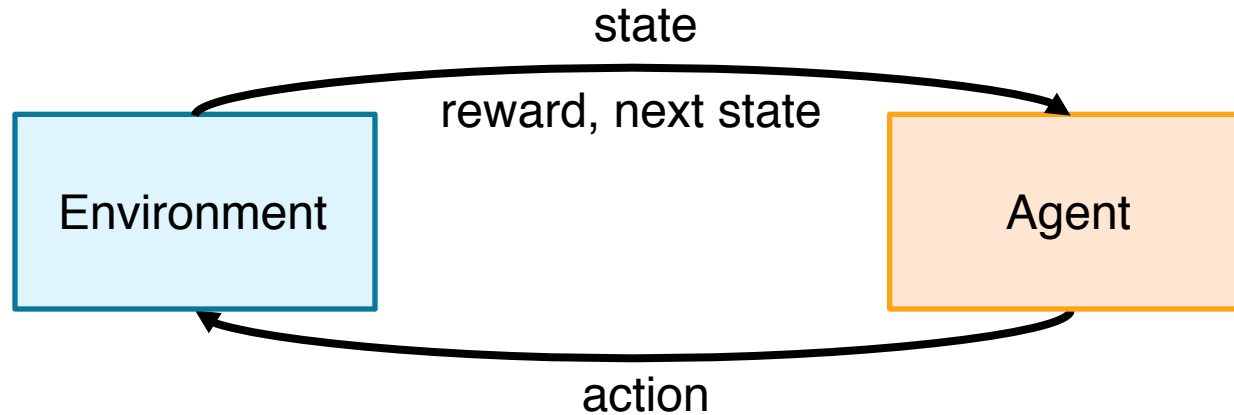
IJCAI2019 Scaling-Up Reinforcement Learning (SURL) Workshop

Adaptive Selection of Auxiliary Tasks in UNREAL

Hidenori Itaya, Tsubasa Hirakawa
Takayoshi Yamashita, Hironobu Fujiyoshi
(Chubu University)



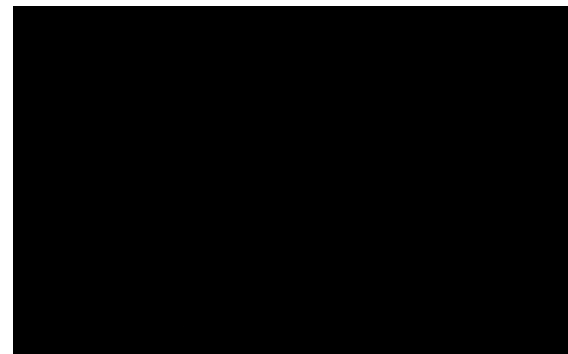
- Problems involving an agent interacting with an environment



- Application example of RL



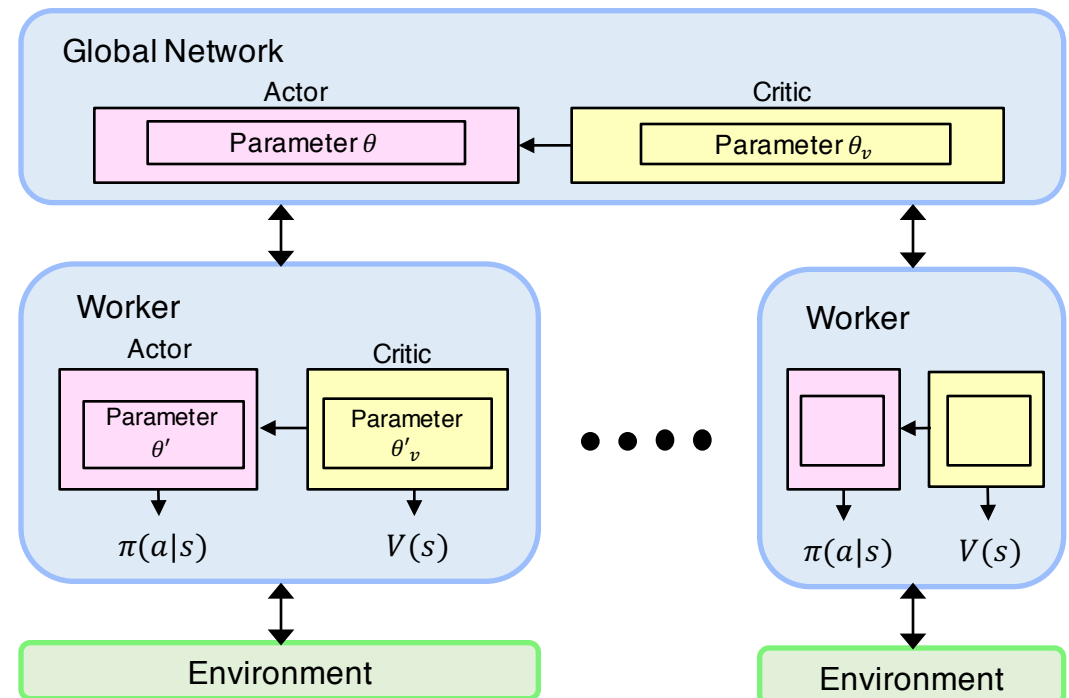
[Gu+, ICRA2016]



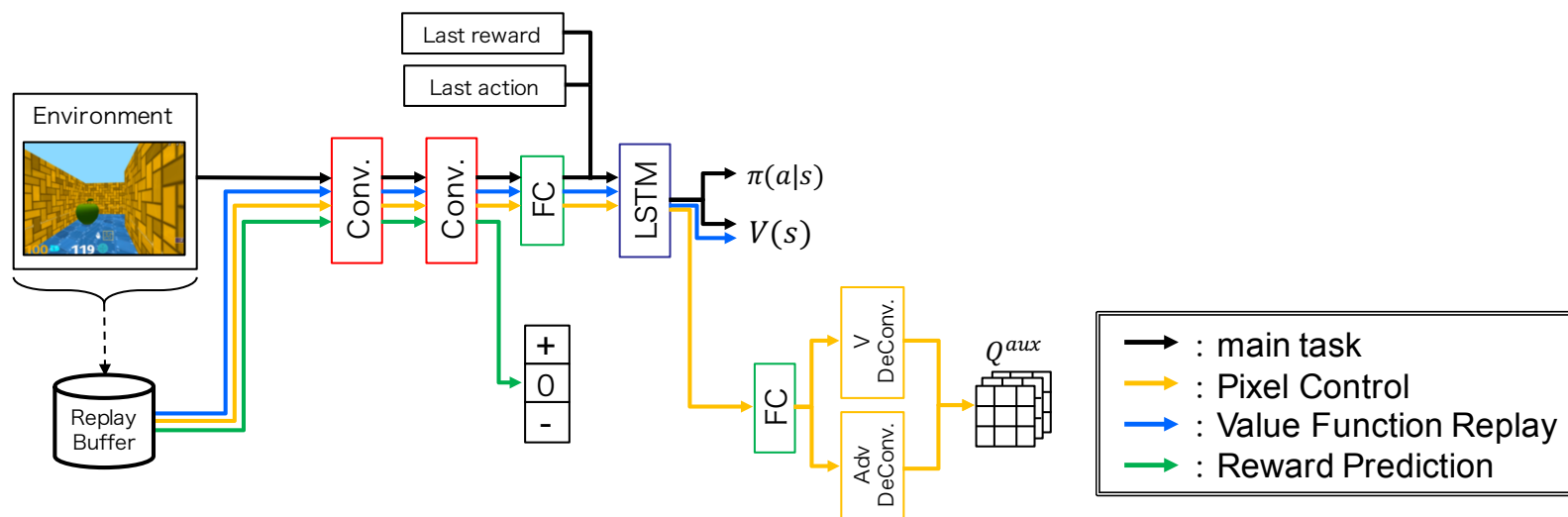
[Mnih+, Nature2015]

Asynchronous Advantage Actor-Critic [Mnih+, 2016]

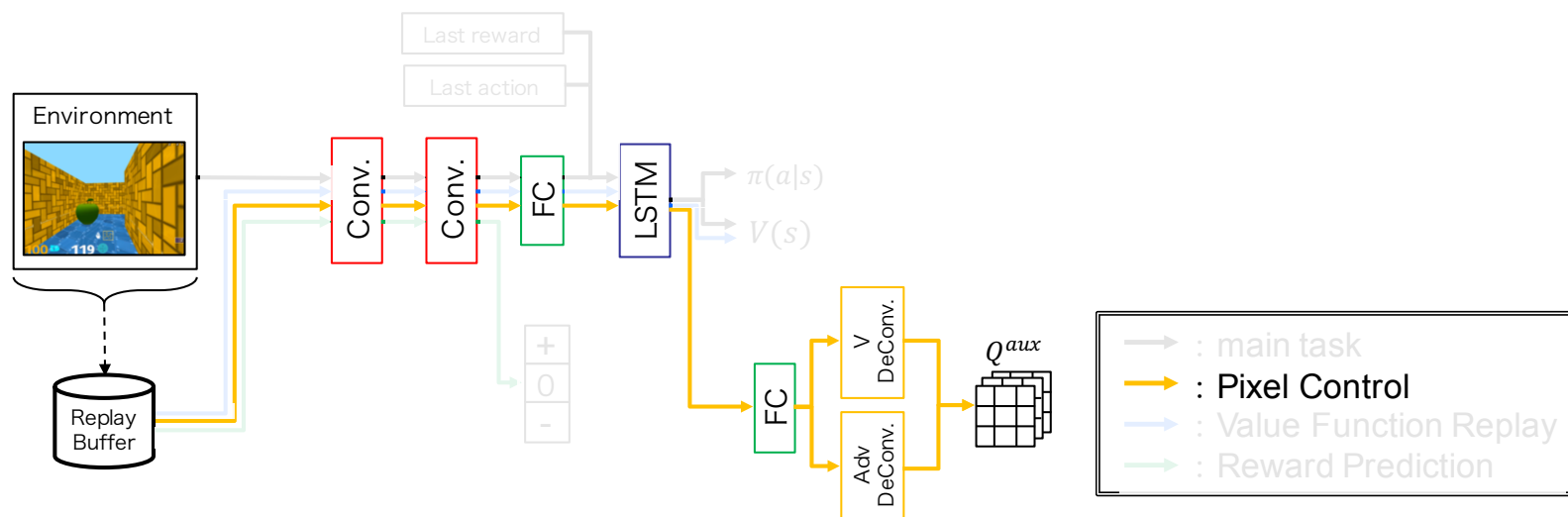
- Asynchronous
 - Each workers updates parameters asynchronously
- Advantage
 - Target error is calculate considering the reward more than 2 steps ahead in each worker
- Actor-Critic
 - Estimate
 - policy
 - State-value function



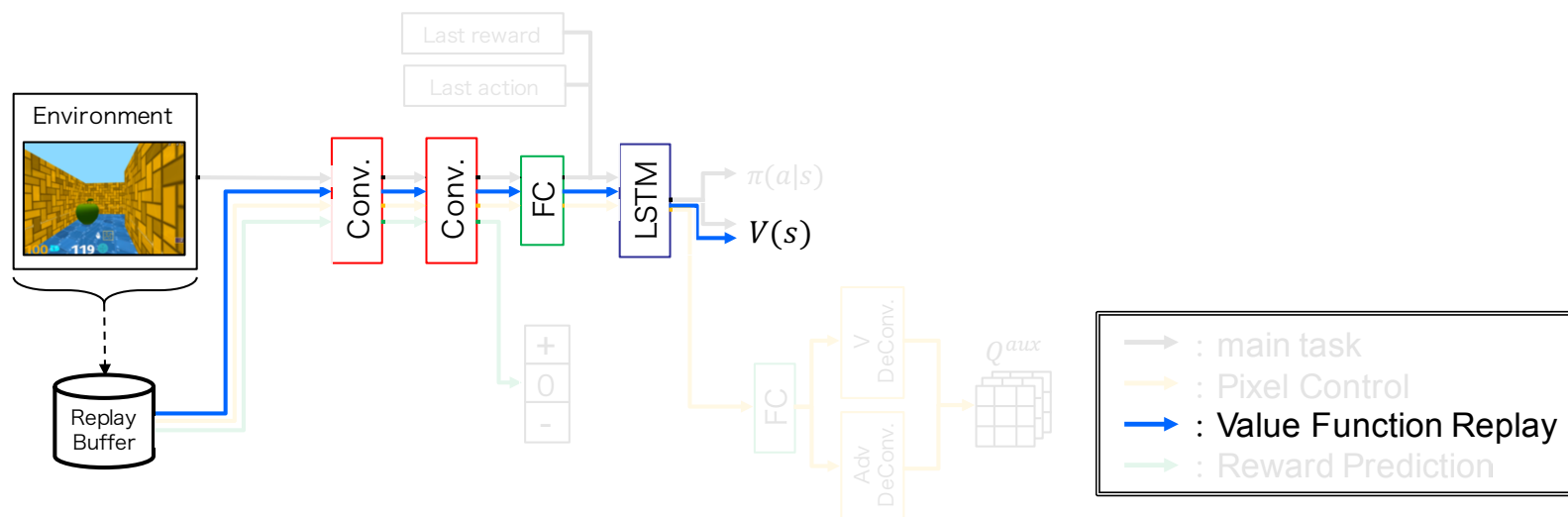
- Introducing three auxiliary tasks into the A3C
 - Pixel Control
 - Train actions that large changes in pixel values
 - Value Function Replay
 - Shuffle past experiences and train state-value functions
 - Reward Prediction
 - Predict future rewards



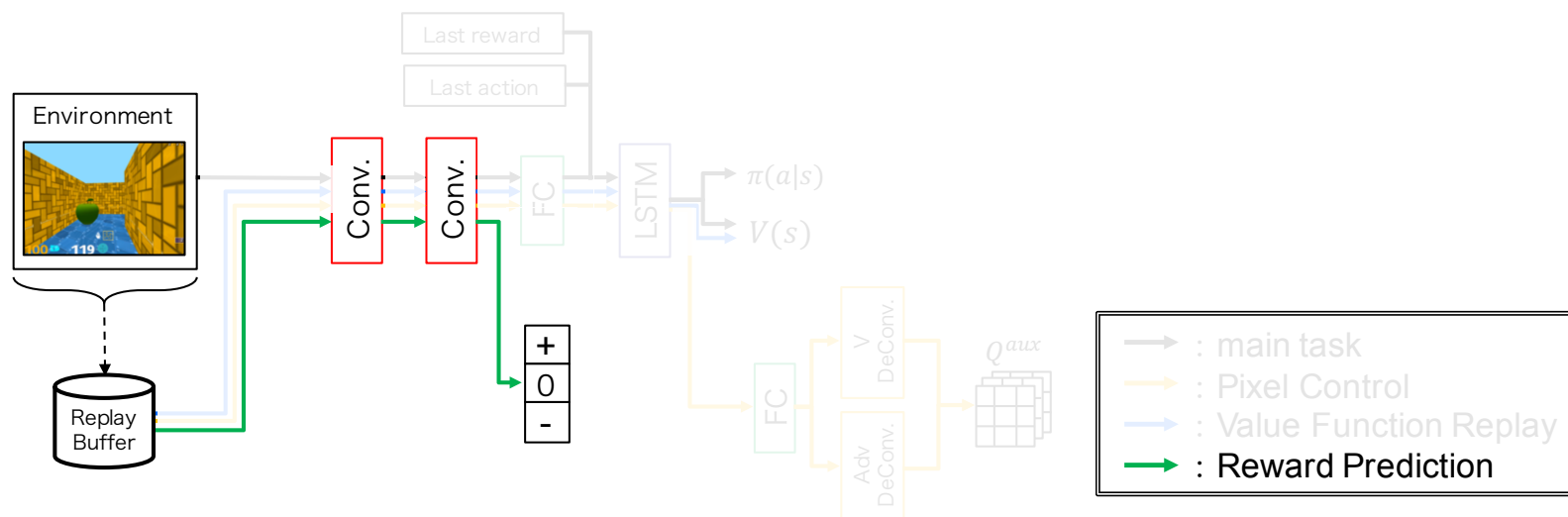
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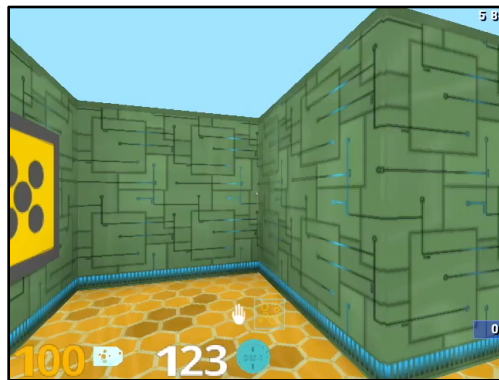


- The sum of main task loss and auxiliary tasks loss
 - L_{main} : Main task loss
 - $L_Q^{(c)}$: Pixel Control loss
 - L_{VR} : Value Function Replay loss
 - L_{RP} : Reward Prediction loss

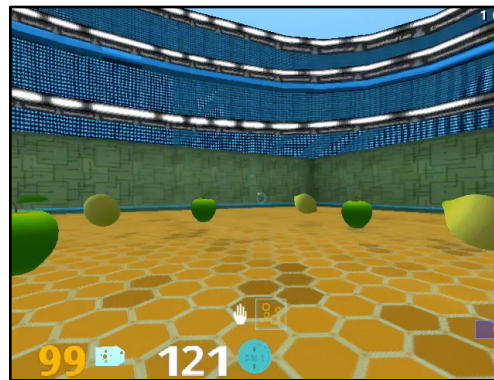
$$L_{\text{UNREAL}} = \underbrace{L_{\text{main}}}_{\text{Main task}} + \underbrace{\sum_c L_Q^{(c)} + L_{\text{VR}} + L_{\text{RP}}}_{\text{Auxiliary Tasks}}$$

Preliminary experiment

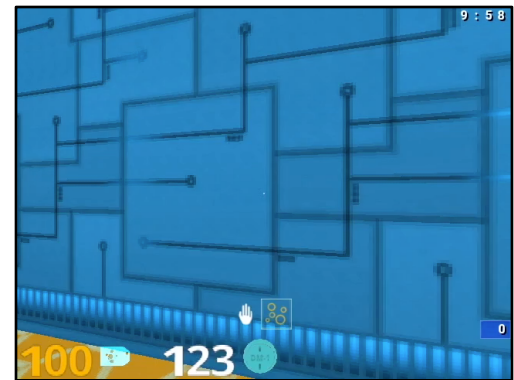
- Investigate whether each auxiliary task is effective or not
- Environment : DeepMind Lab [Beattie+, arXiv2016]
- Investigation functions
 - Pixel Control (PC)
 - Value Function Replay (VR)
 - Reward Prediction (RP)
 - Three auxiliary tasks (UNREAL)



nav_maze_static_01

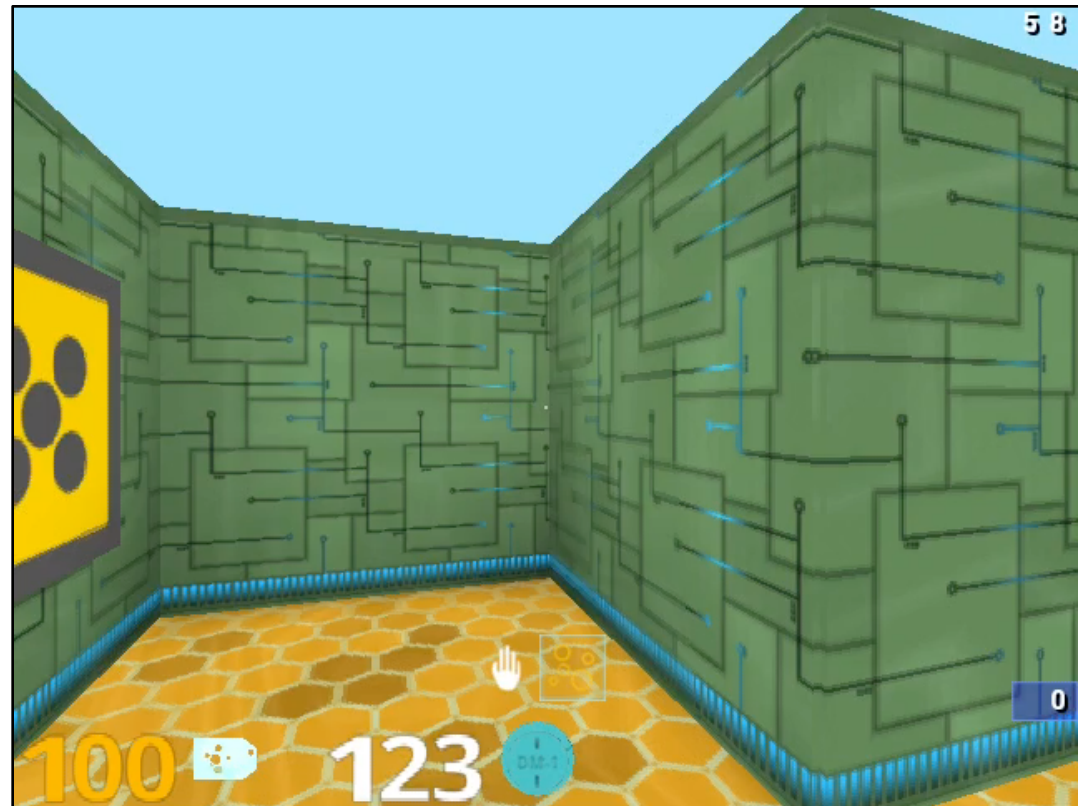


seekavoid_arena_01

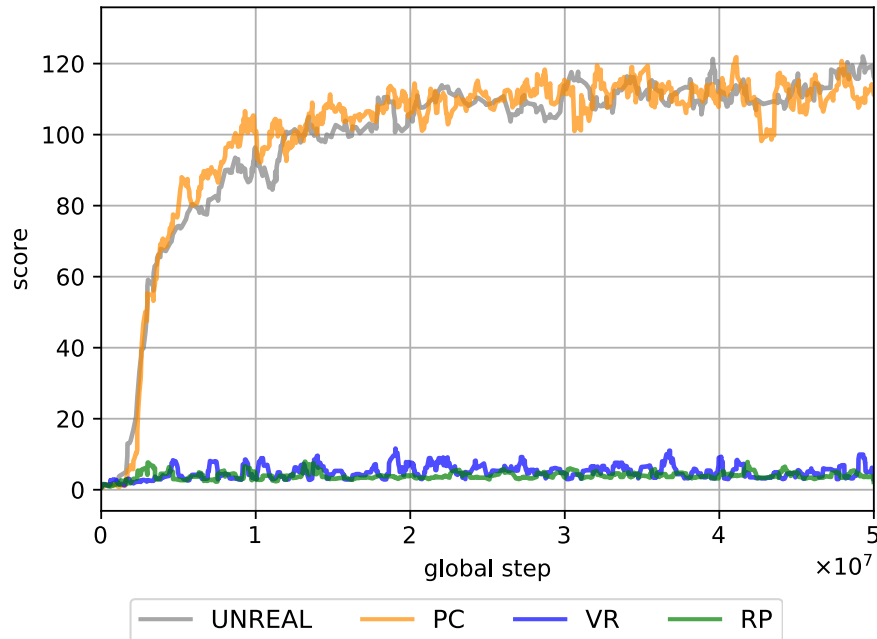


lt_horseshoe_color

- A First-person viewpoint maze game
- Action
 - Look left
 - Look right
 - Forward
 - Backward
 - Strafe left
 - Strafe right
- Reward
 - Apple : +1
 - Goal : +10

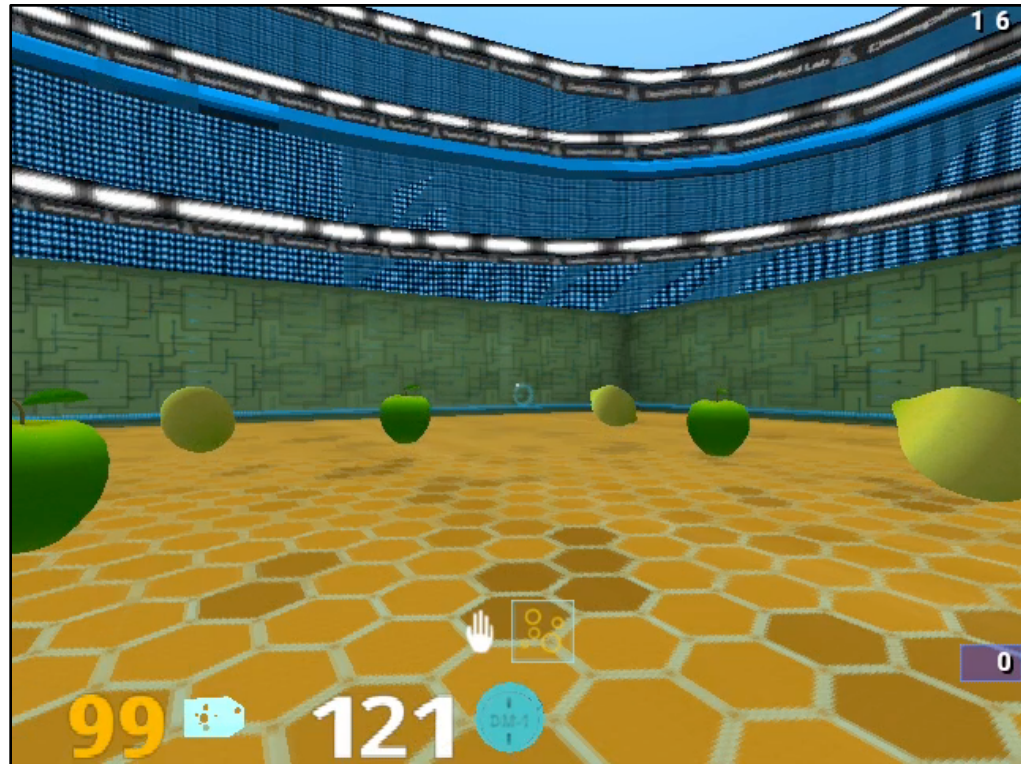


Result (nav_maze_static_01)

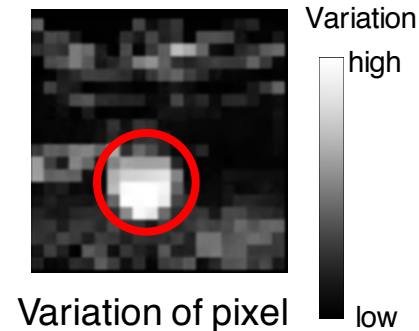
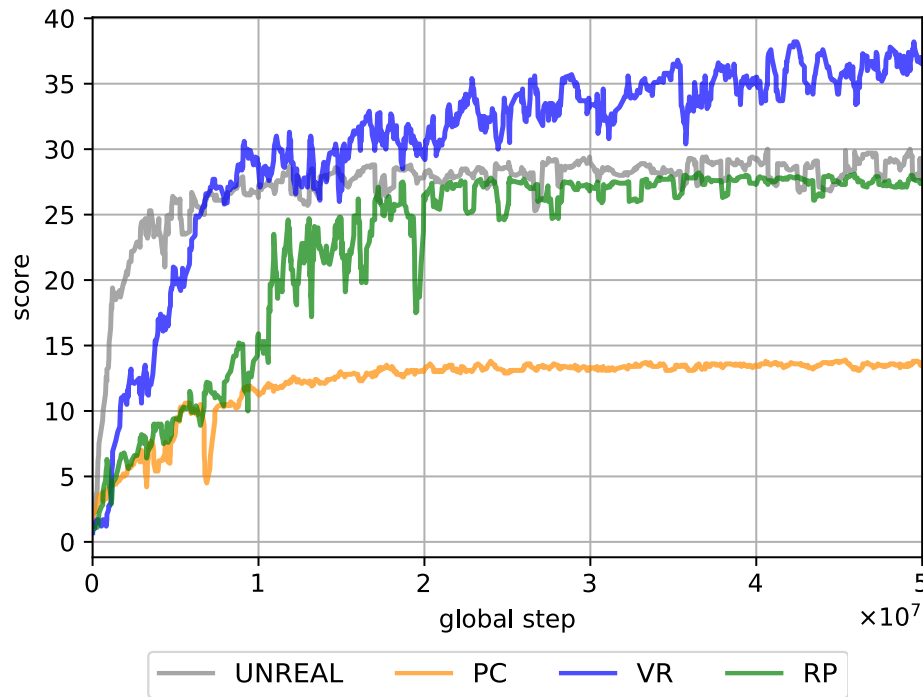


- **Pixel Control** is effective
 - Action changing pixel values promote movement

- Avoid lemons and earn apples game
- Action
 - Look left
 - Look right
 - Forward
 - Backward
 - Strafe left
 - Strafe right
- Reward
 - Apple : +1
 - Lemon : -1

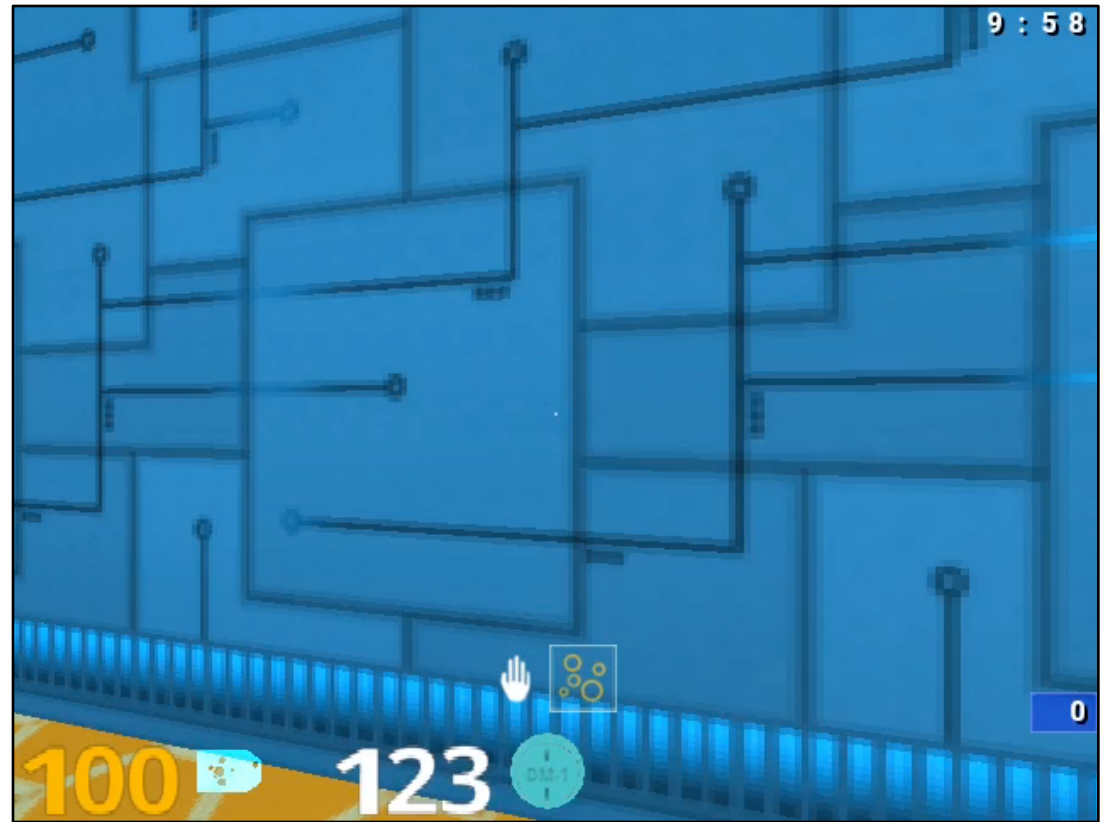


Result (seekavoid_arena_01)

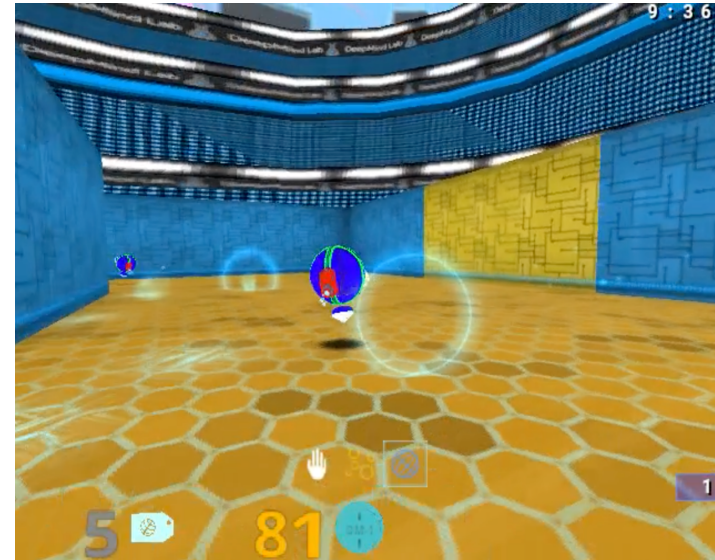
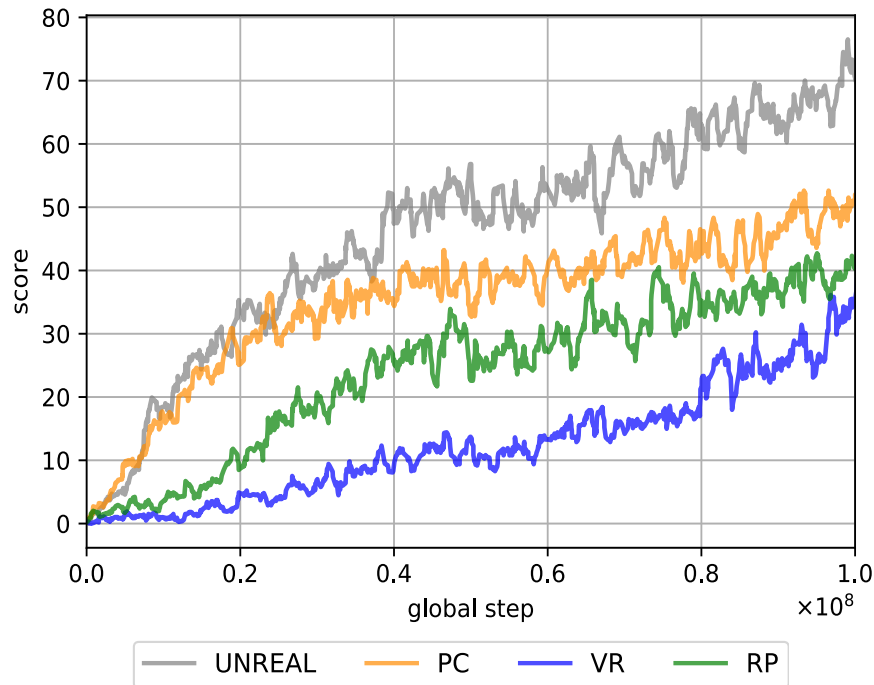


- Value Function Replay is effective
 - Actions changing pixel values are not suitable
 - Seekavoid obtains reward, frequently

- First person shooting game
- Action
 - Look left
 - Look right
 - Forward
 - Backward
 - Strafe left
 - Strafe right
 - Attack
- Reward
 - Kill the enemy : +1

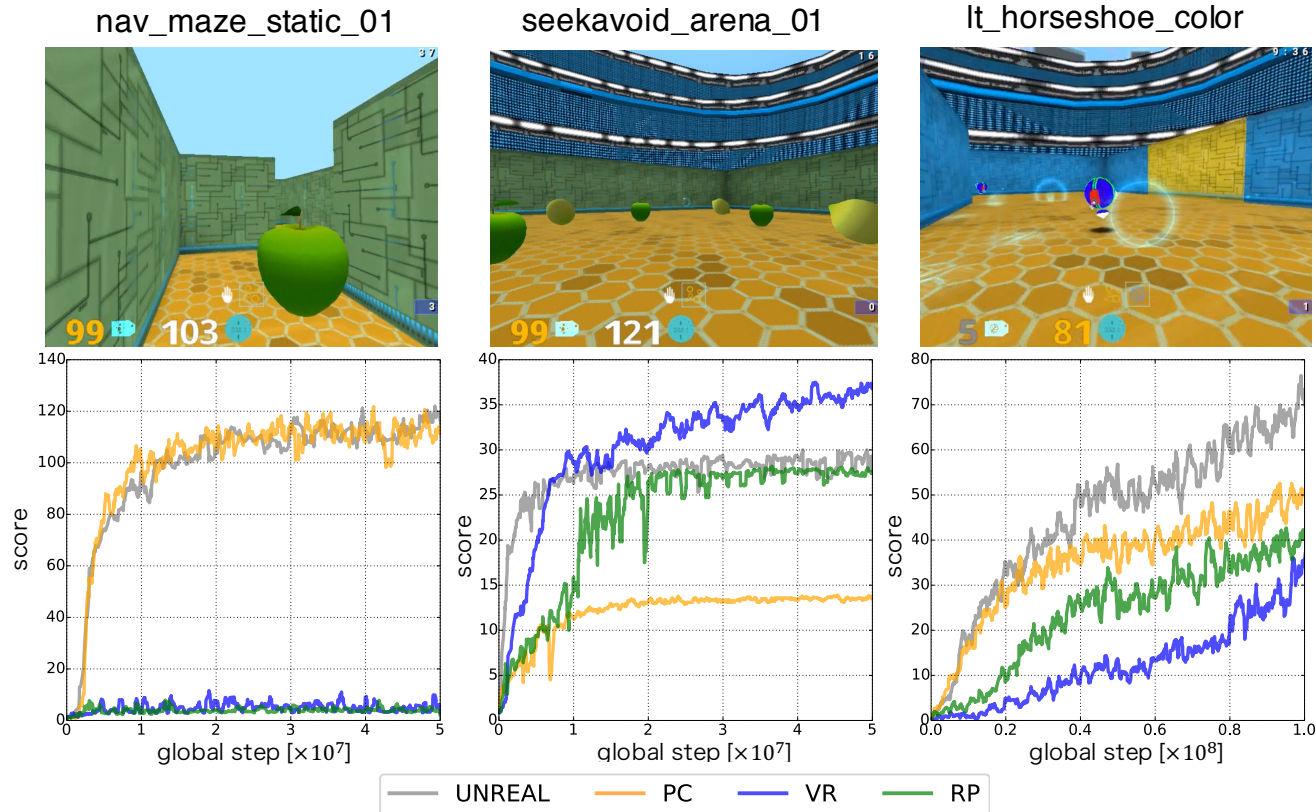


Result (lt_horseshoe_color)



- All auxiliary tasks are effective
 - Kill the enemy = Actions change pixel values
 - Reward (kill the enemy) acquired less frequent

Summary of pre-experiment



Optimal
auxiliary task

Pixel Control
UNREAL

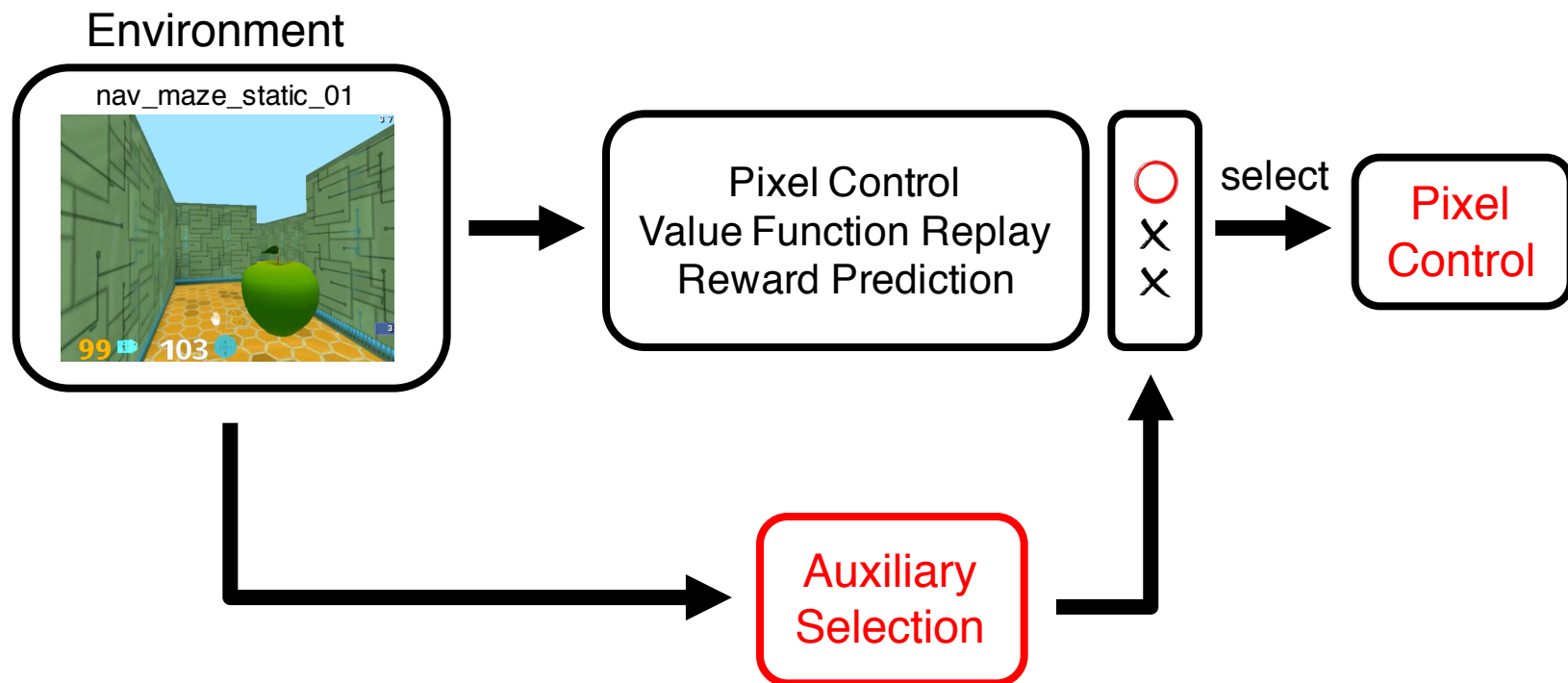
Value Function
Replay

UNREAL

→ Need to select suitable auxiliary tasks for game

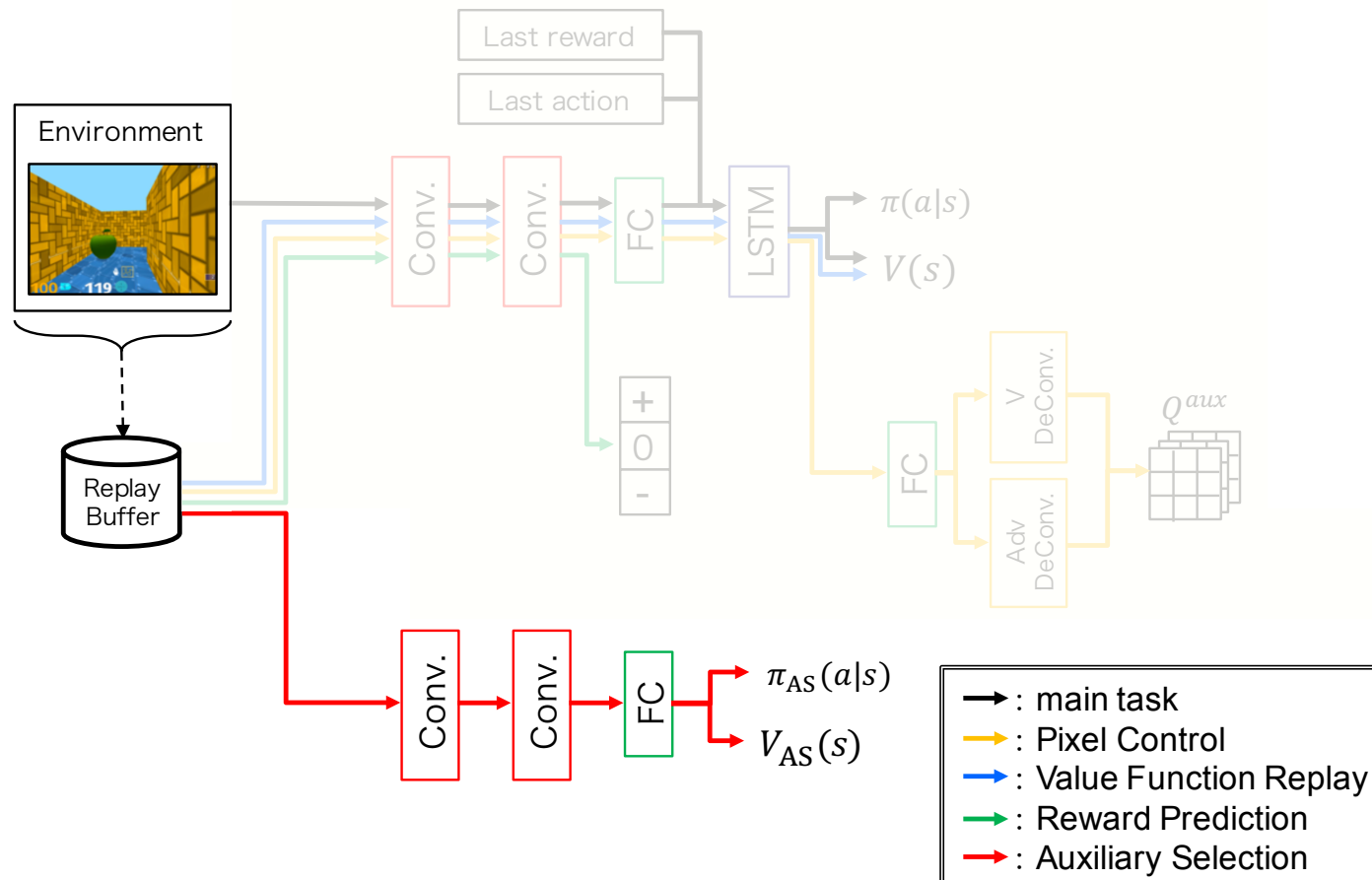
Purpose of proposed method

- Using only suitable auxiliary task for environment
 - Automatically select for suitable auxiliary tasks
- Proposed method
 - Auxiliary Selection
 - Adaptively selection of optimal auxiliary tasks



Auxiliary Selection

- A novel task to select the suitable auxiliary task for environment
 - Network build independent network from the main task



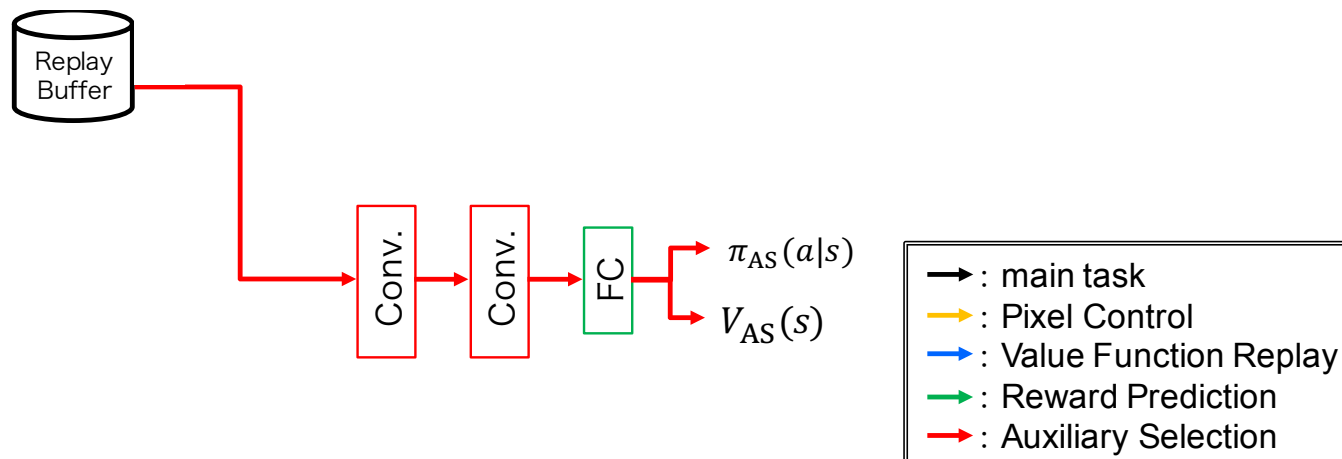
Action of Auxiliary Selection

- Weight of each auxiliary task C_{PC}, C_{VR}, C_{RP}

$$(C_{PC}, C_{VR}, C_{RP}) = (\{0, 1\}, \{0, 1\}, \{0, 1\})$$

- Actions of Auxiliary Selection

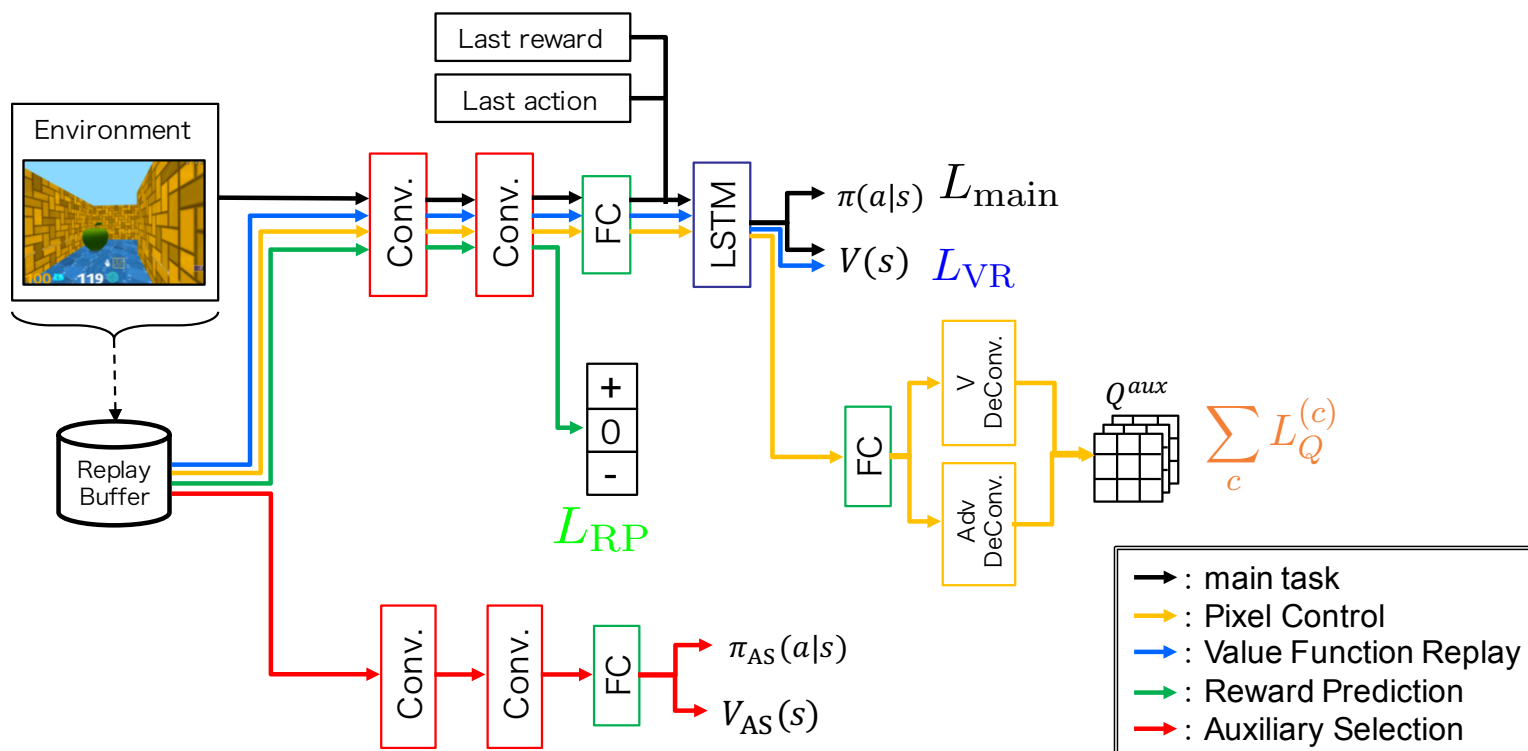
$$\arg \max_{\pi_{AS}} a = \{C_{PC}, C_{VR}, C_{RP}\} = \frac{\{0, 0, 0\} \sim \{1, 1, 1\}}{8 \text{ patterns}}$$



Loss of main and auxiliary tasks

- Multiply Auxiliary Selection outputs and loss of auxiliary tasks

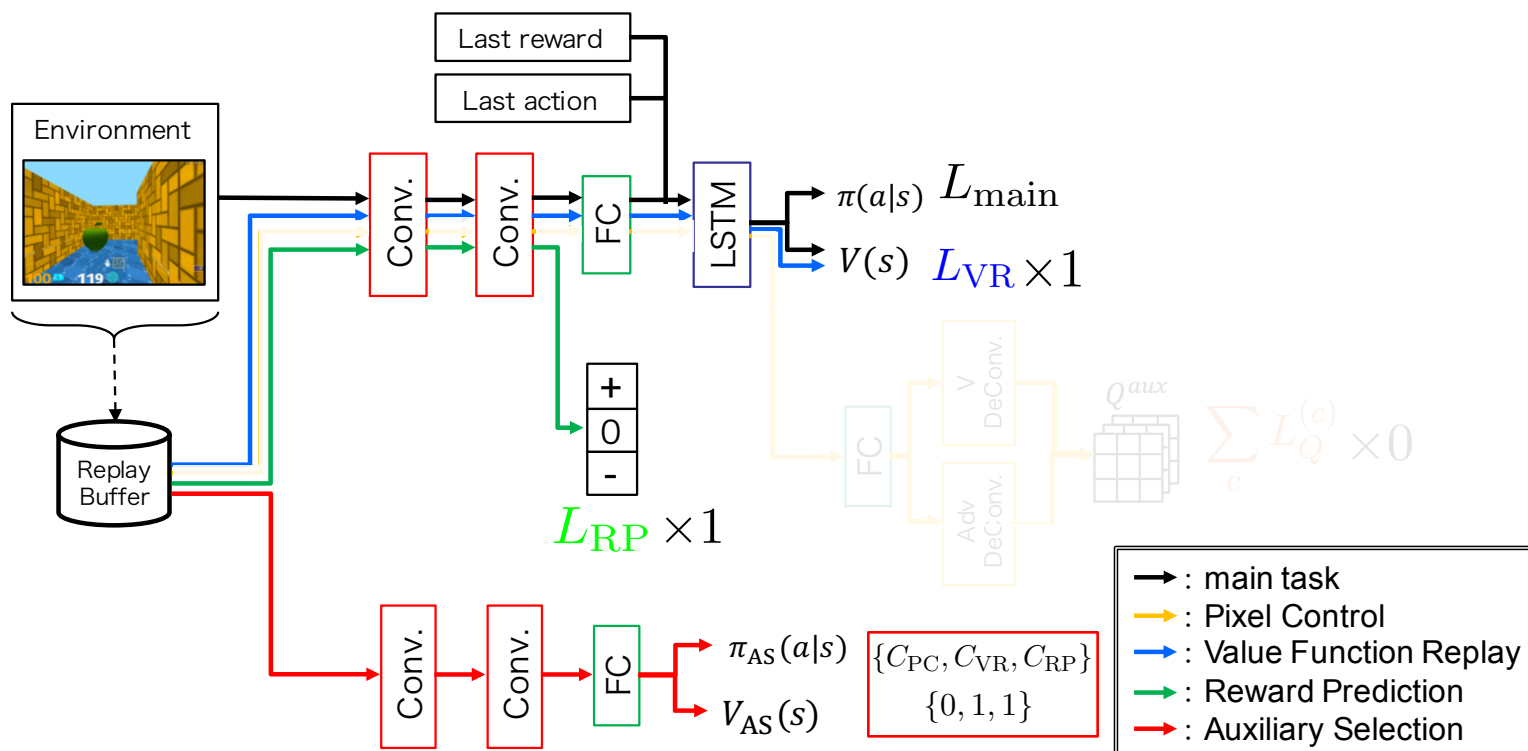
$$L_{\text{UNREAL}} = \underbrace{L_{\text{main}}}_{\text{black}} + \underbrace{C_{\text{PC}} \sum_c L_Q^{(c)}}_{\text{yellow}} + \underbrace{C_{\text{VR}} L_{\text{VR}}}_{\text{blue}} + \underbrace{C_{\text{RP}} L_{\text{RP}}}_{\text{green}}$$



Loss of main and auxiliary tasks

- Multiply Auxiliary Selection outputs and loss of auxiliary tasks

$$L_{\text{UNREAL}} = \underbrace{L_{\text{main}}}_{\{0\}} + \underbrace{C_{\text{PC}}}_{\{0\}} \underbrace{\sum_c L_Q^{(c)}}_{\{1\}} + \underbrace{C_{\text{VR}}}_{\{1\}} \underbrace{L_{\text{VR}}}_{\{1\}} + \underbrace{C_{\text{RP}}}_{\{1\}} \underbrace{L_{\text{RP}}}_{\{1\}}$$



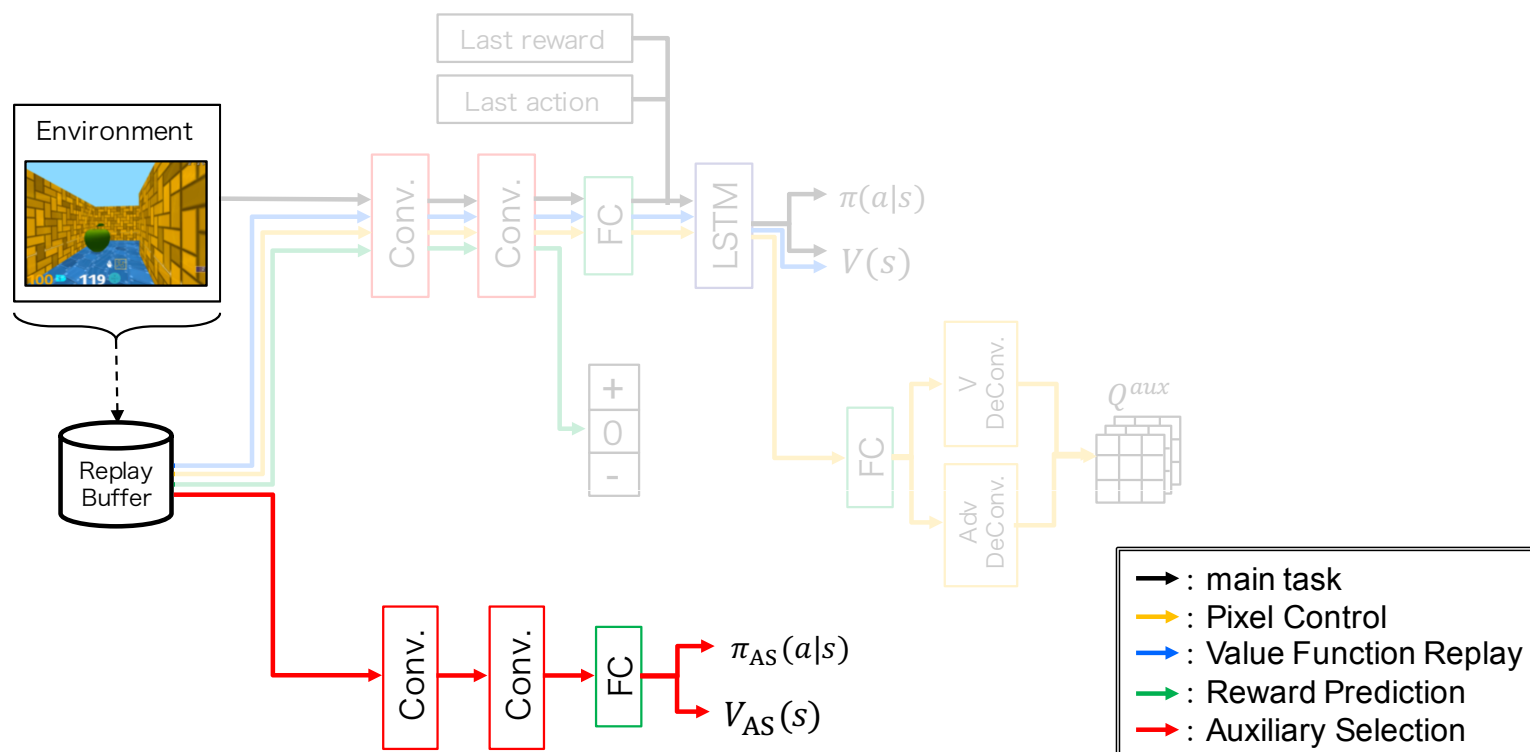
Loss function of Auxiliary Selection

- Adding losses of policy and state-value function

$$L_{AS} = \underbrace{L(\pi_{AS})}_{\text{Loss of policy}} + \underbrace{L(V_{AS})}_{\text{Loss of state-value function}}$$

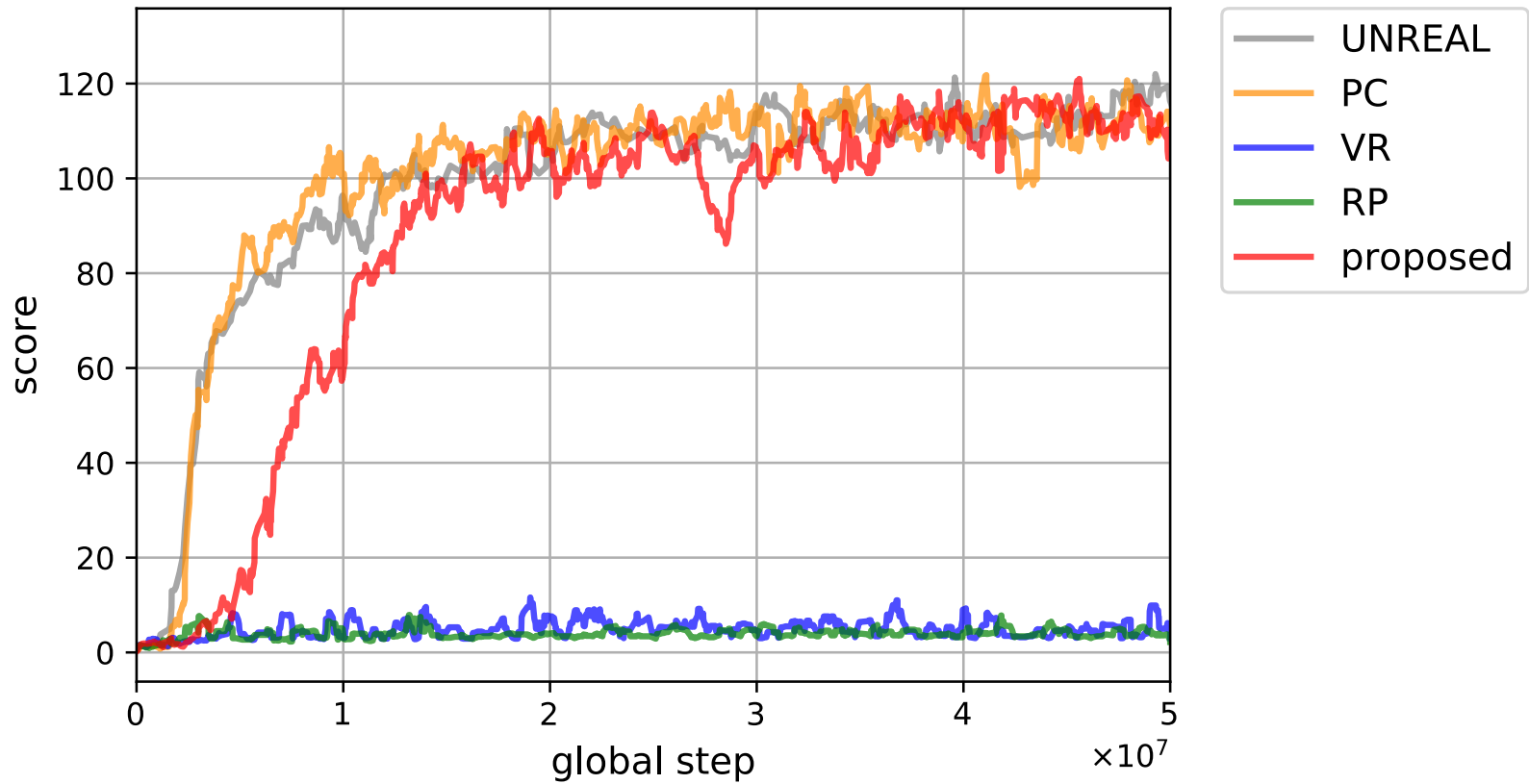
Loss of policy

Loss of state-value function

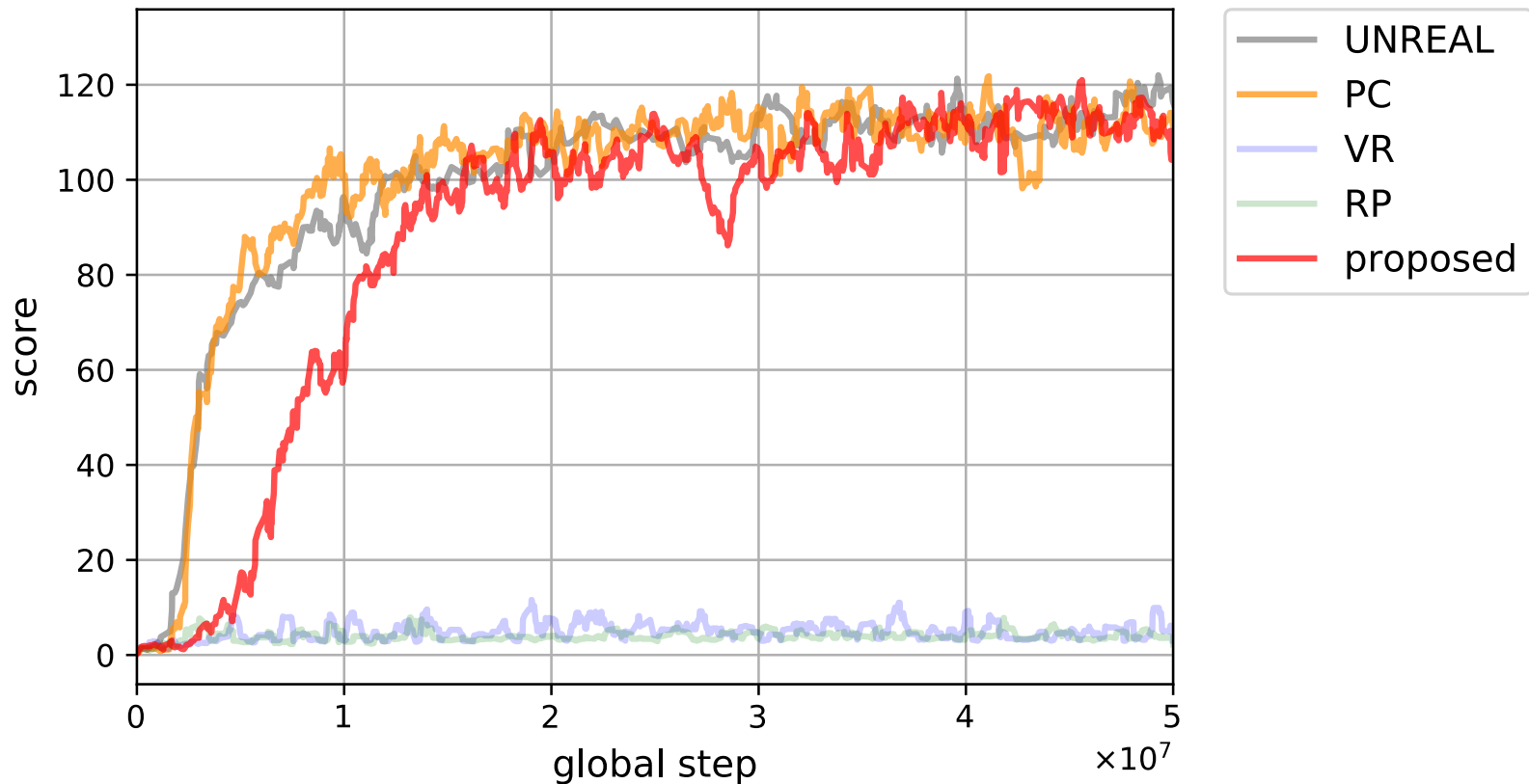


- Environment : DeepMind Lab [Beattie+, arXiv2016]
- Training setting
 - # of steps
 - 1.0×10^7 steps (maze and seekavoid)
 - 1.0×10^8 steps (horseshoe)
 - # of workers
 - 8
- Comparison
 - Only auxiliary task (PC, VR, RP)
 - Three auxiliary tasks (UNREAL)
 - Proposed method (proposed)

Result (nav_maze_static_01)

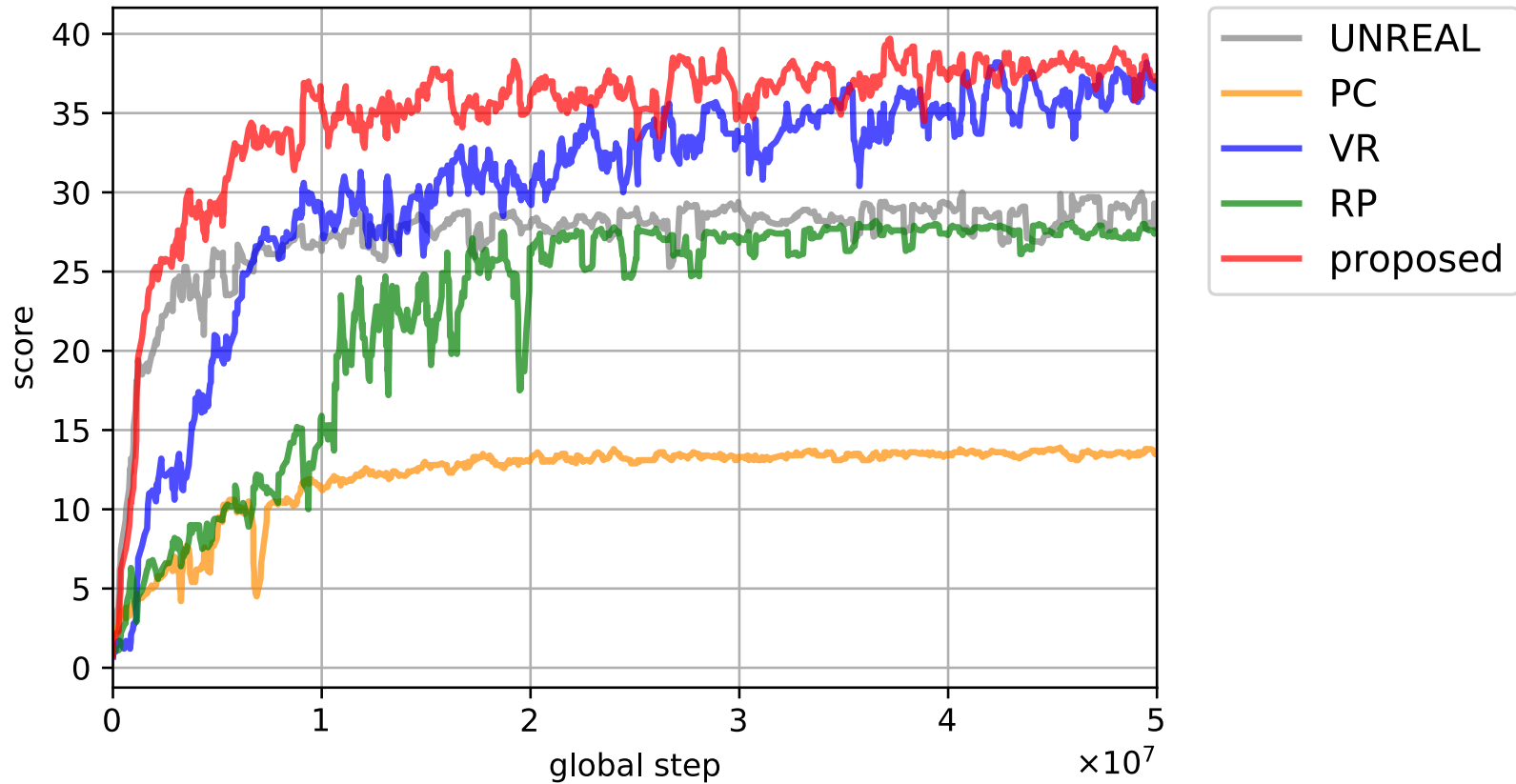


Result (nav_maze_static_01)

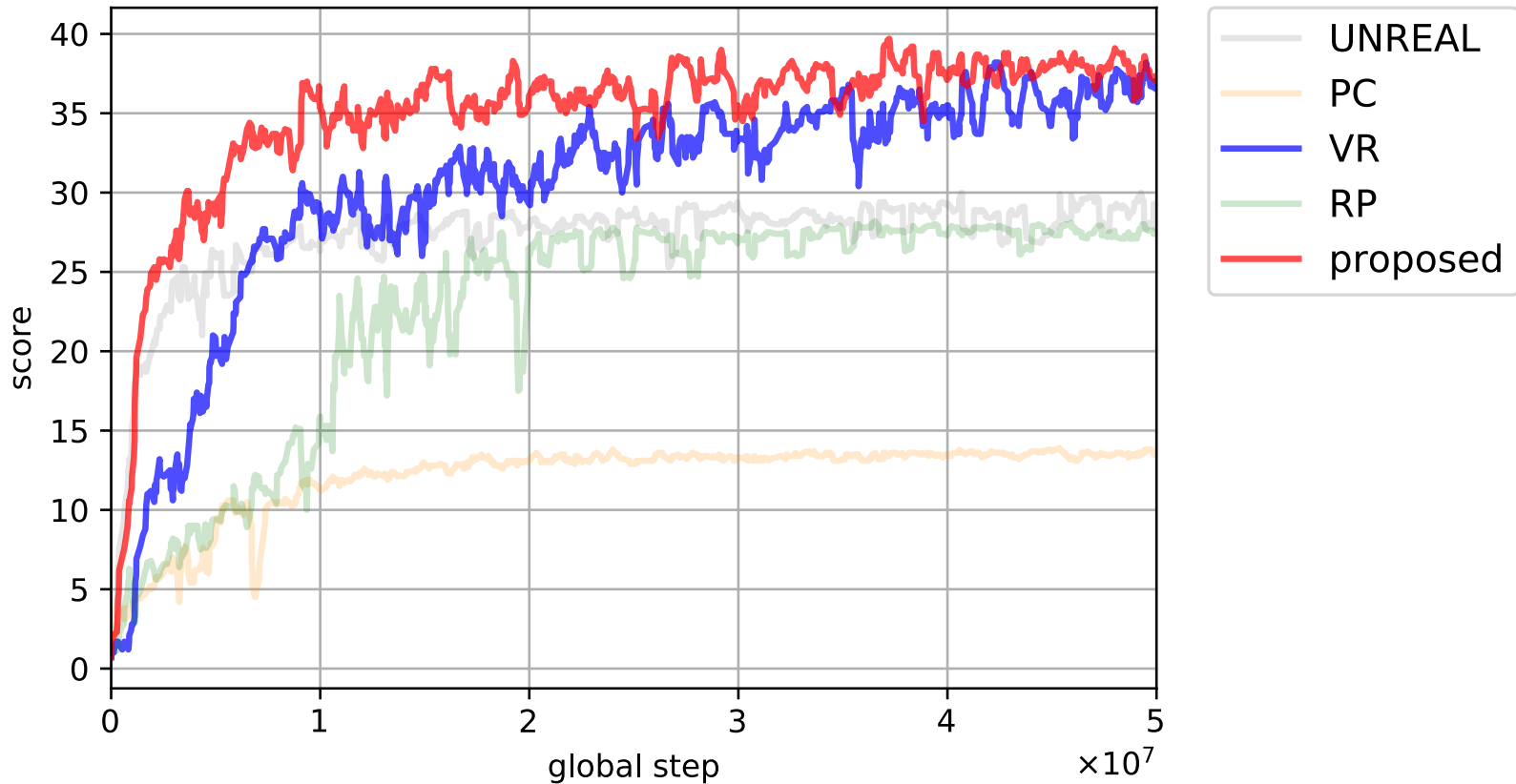


→ **Proposed method** achieve high score as same as UNREAL or PC

Result (seekavoid_arena_01)

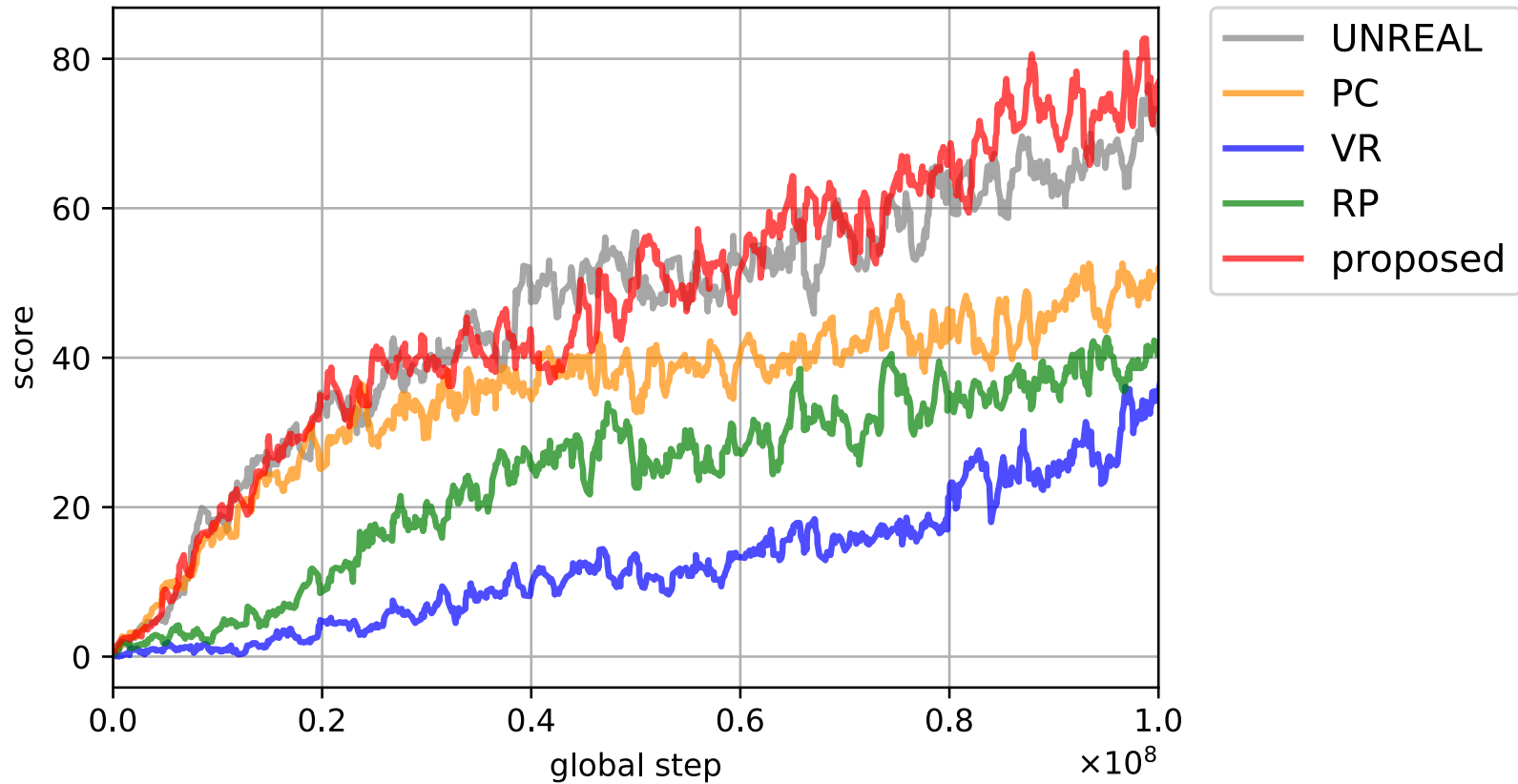


Result (seekavoid_arena_01)

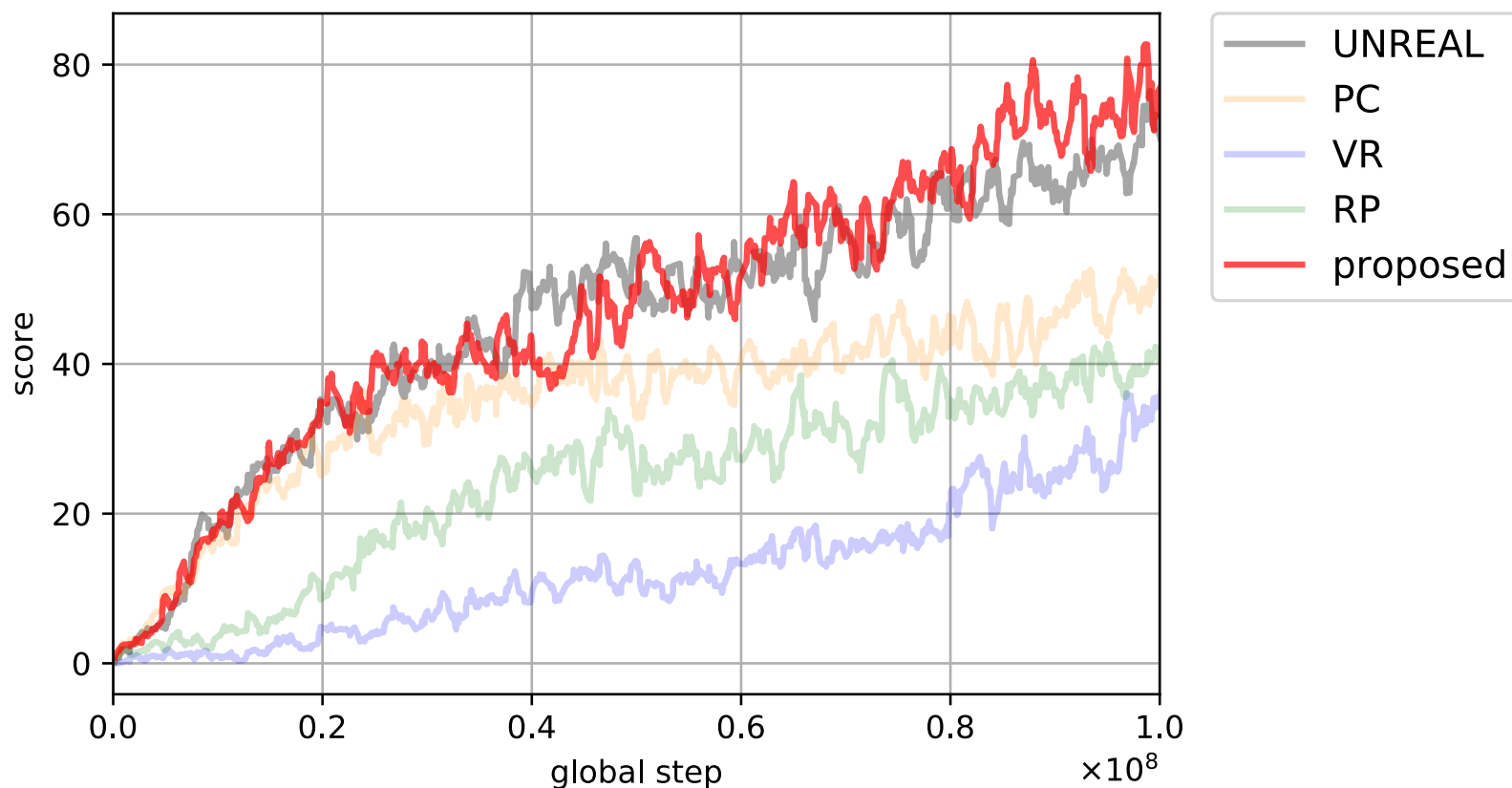


→ **Proposed method** achieve high score as same as **VR**

Result (lt_horseshoe_color)

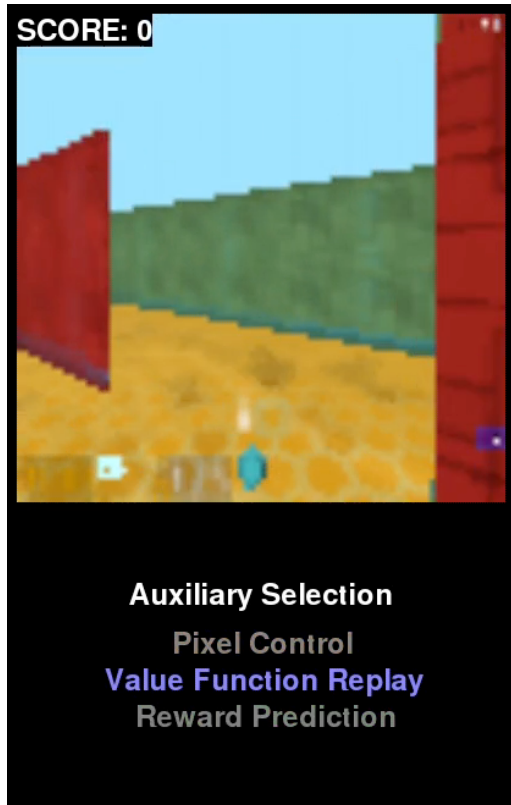


Result (lt_horseshoe_color)



→ **Proposed method** achieve high score as same as UNREAL

Analysis of the selected auxiliary tasks (nav_maze_static_01)



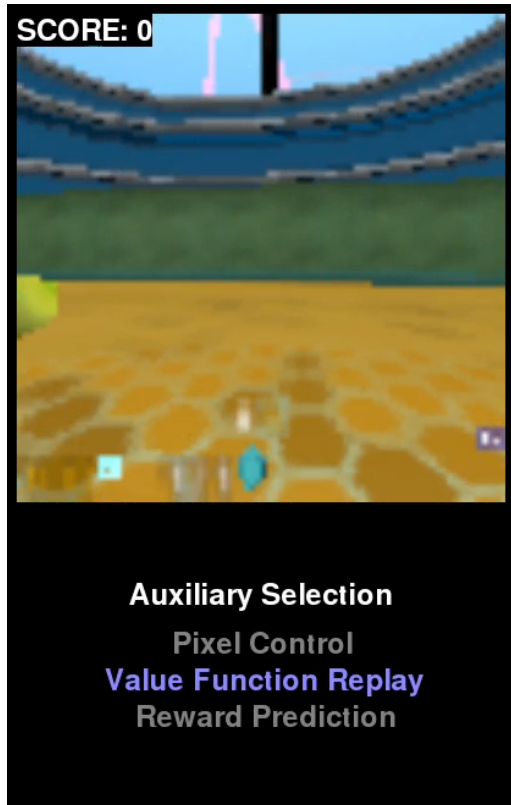
Selection percentage of each AT in one episode [%]

	Pixel Control	Value Function Replay	Reward Prediction
maze	48.3	54.1	41.0
seekavoid	0.1	100.0	0.0
horseshoe	94.9	0.1	99.9

※ 50 episodes average

→ All auxiliary tasks are equivalently selected

Analysis of the selected auxiliary tasks (seekavoid_arena_01)



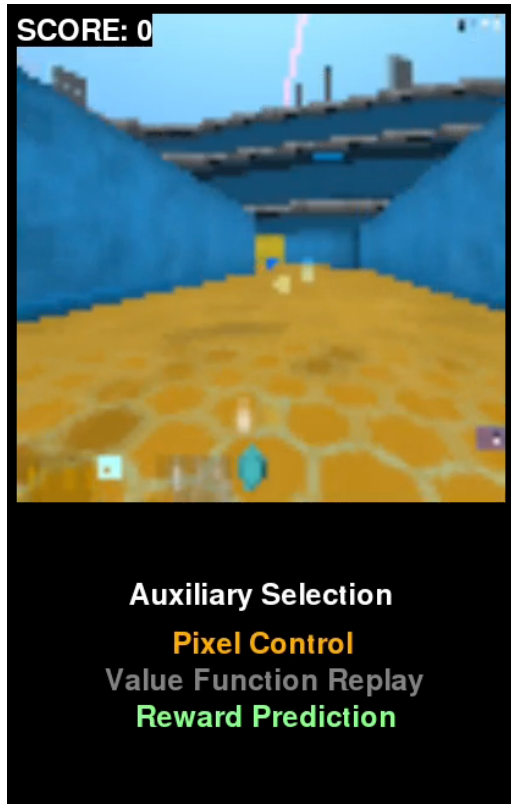
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horseshoe	94.9	0.1	99.9

※ 50 episodes average

→ VR is stably selected

Analysis of the selected auxiliary tasks (It_horseshoe_color)



Selection percentage of each AT in one episode [%]

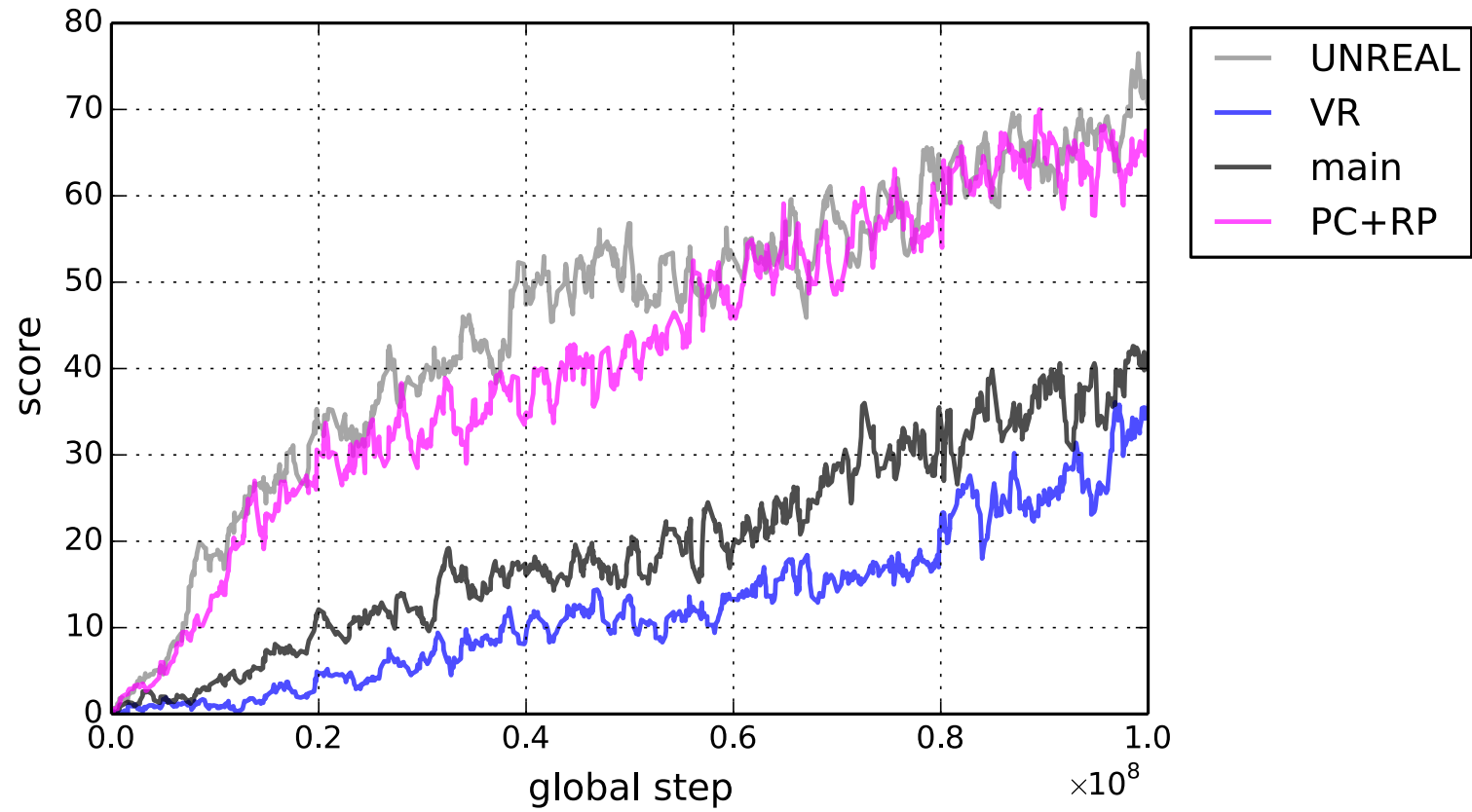
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※ 50 episodes average

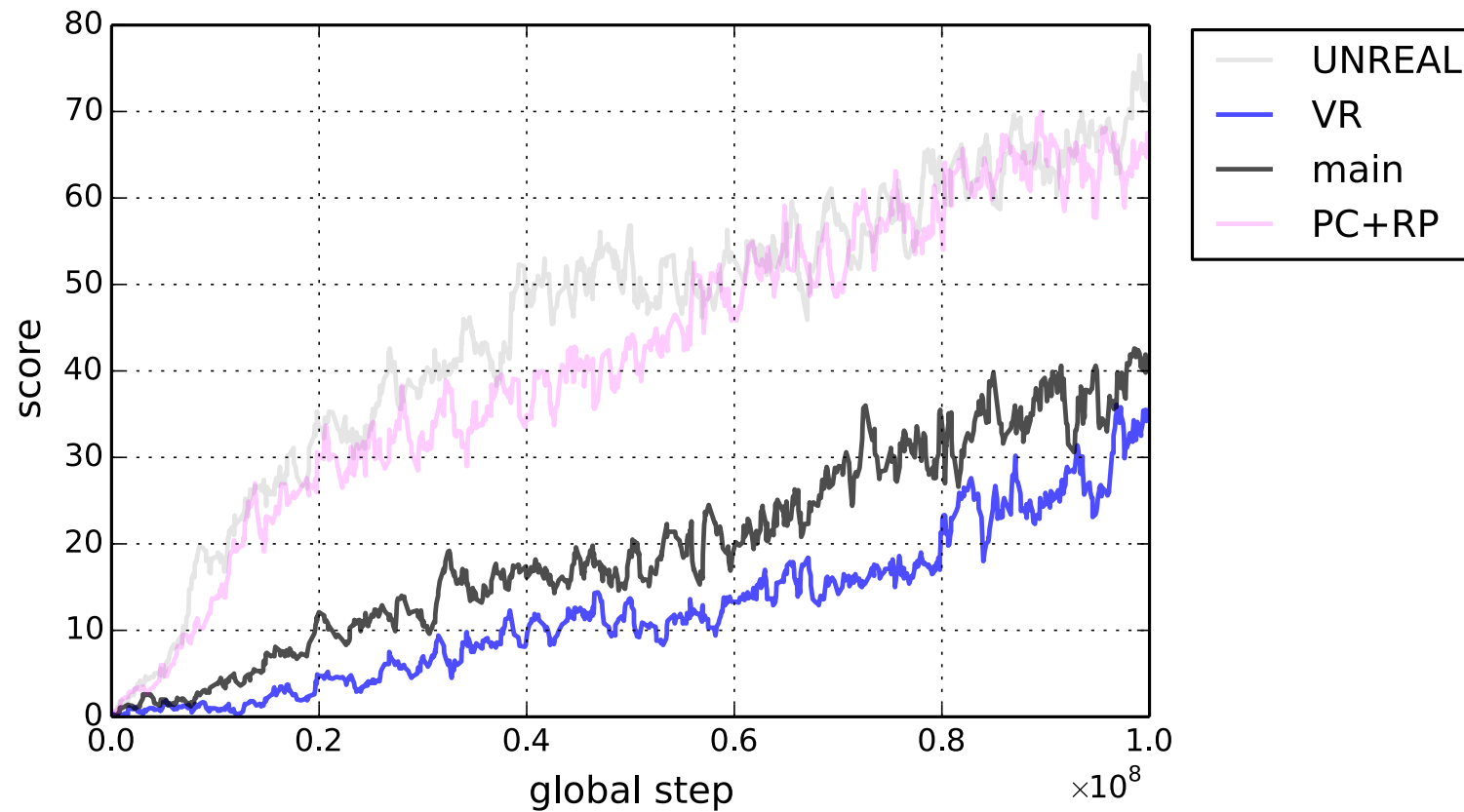
→ PC and RP are stably selected

- Investigation another combinations of auxiliary tasks
- Environment : It_horseshoe_color (DeepMind Lab)
- Comparison : Compare scores in horseshoe
 - Three auxiliary tasks (UNREAL)
 - Value Function Replay (VR)
 - Pixel Control and Reward Prediction (PC+RP)
 - Only main task (main)

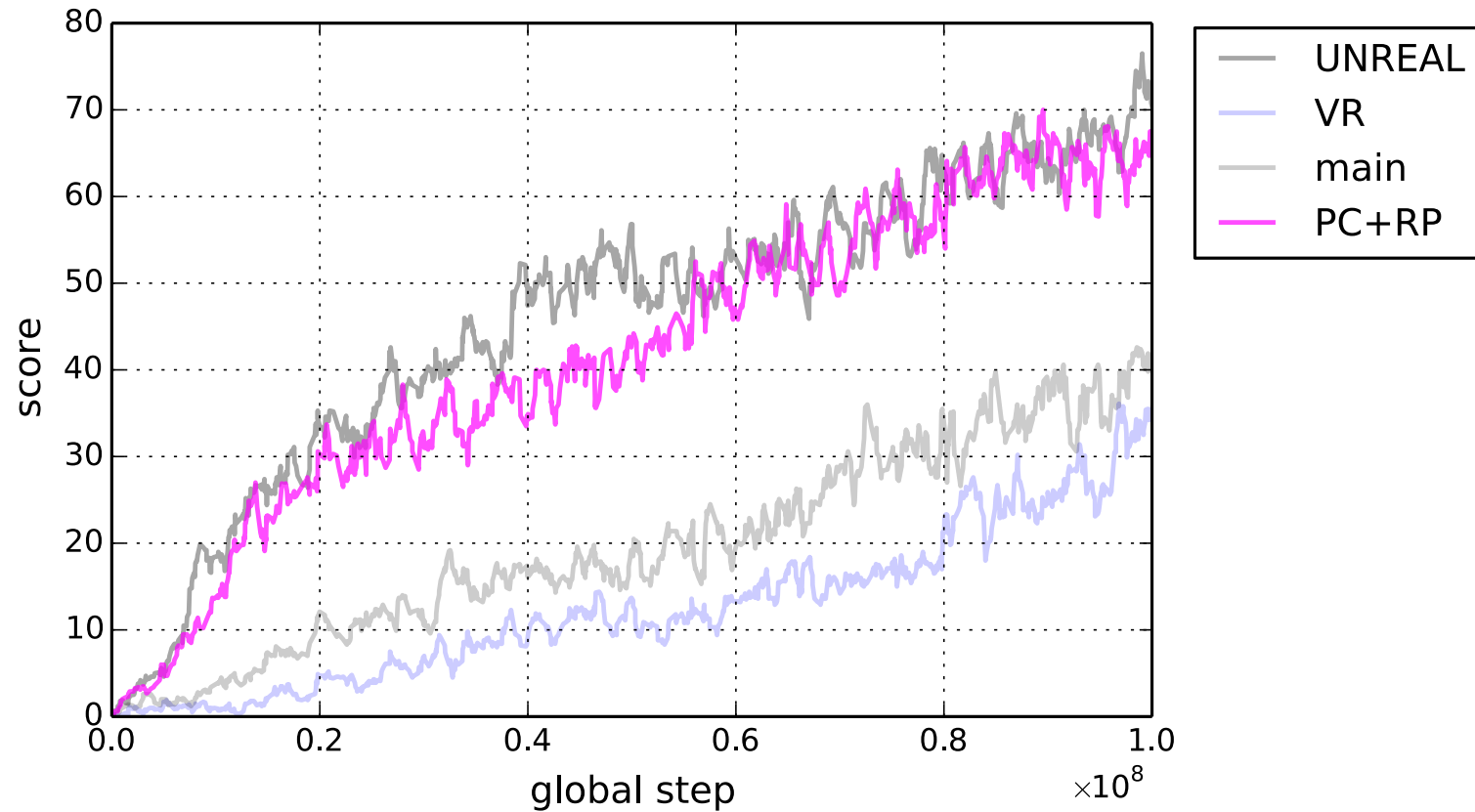
Result



Result



→ VR is lower than only main task



→ VR is lower than only main task

→ PC+RP achieve high score as same as UNREAL

- Auxiliary Selection
 - Achieves the score as same as the optimal auxiliary task
 - Can select appropriate auxiliary tasks for each games
 - nav_maze_static_01 : UNREAL, Pixel Control
 - seekavoid_arena_01 : Value Function Replay
 - lt_horseshoe_color : Pixel Control + Reward Prediction
- Future work
 - Evaluating the proposed method in various environments with other auxiliary tasks