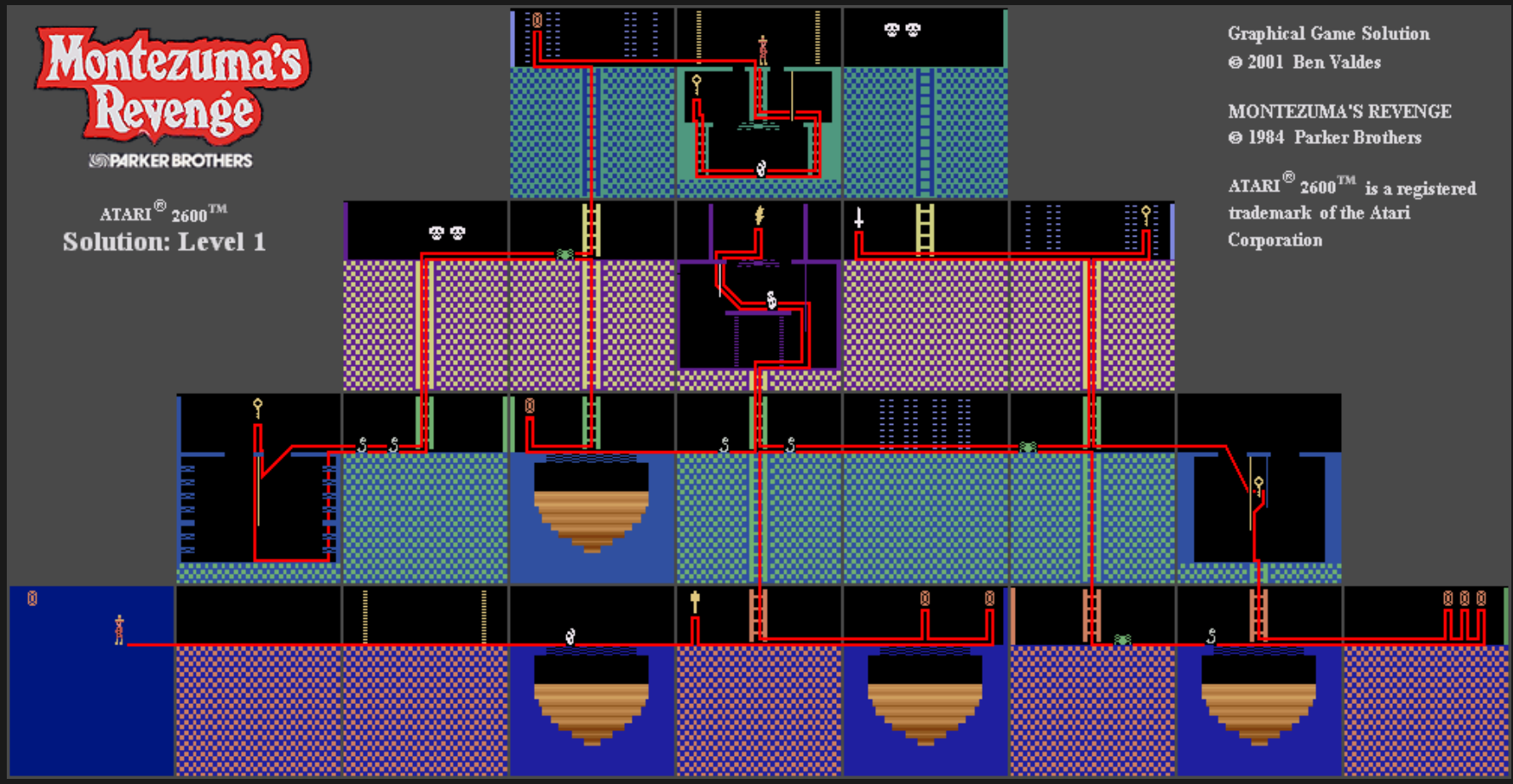
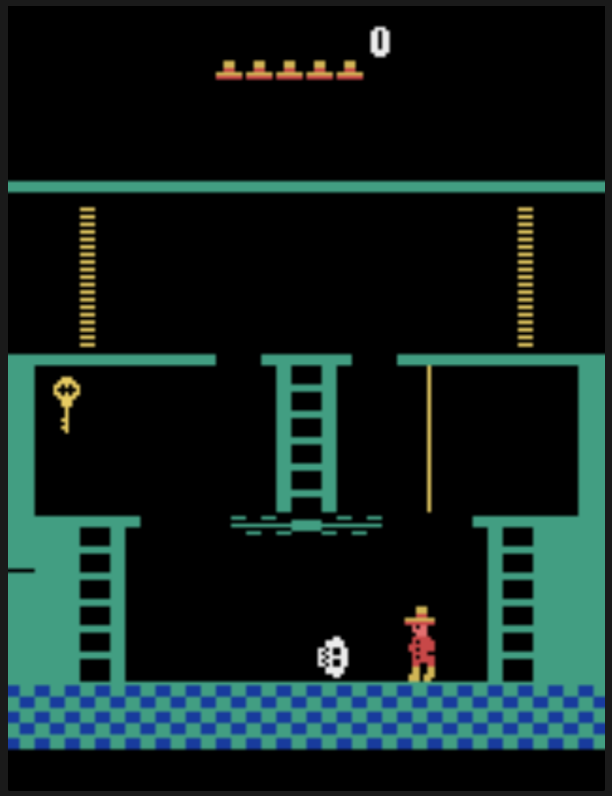
challenging Montezuma's Revenge intrinsic motivation in RL

Michal CHOVANEC

Montezuma's Revenge



Montezuma's Revenge



- **very sparse rewards** - hundrets of steps
- **huge state space**
- **hard exploration**
- **needs returns back**

state of the art score

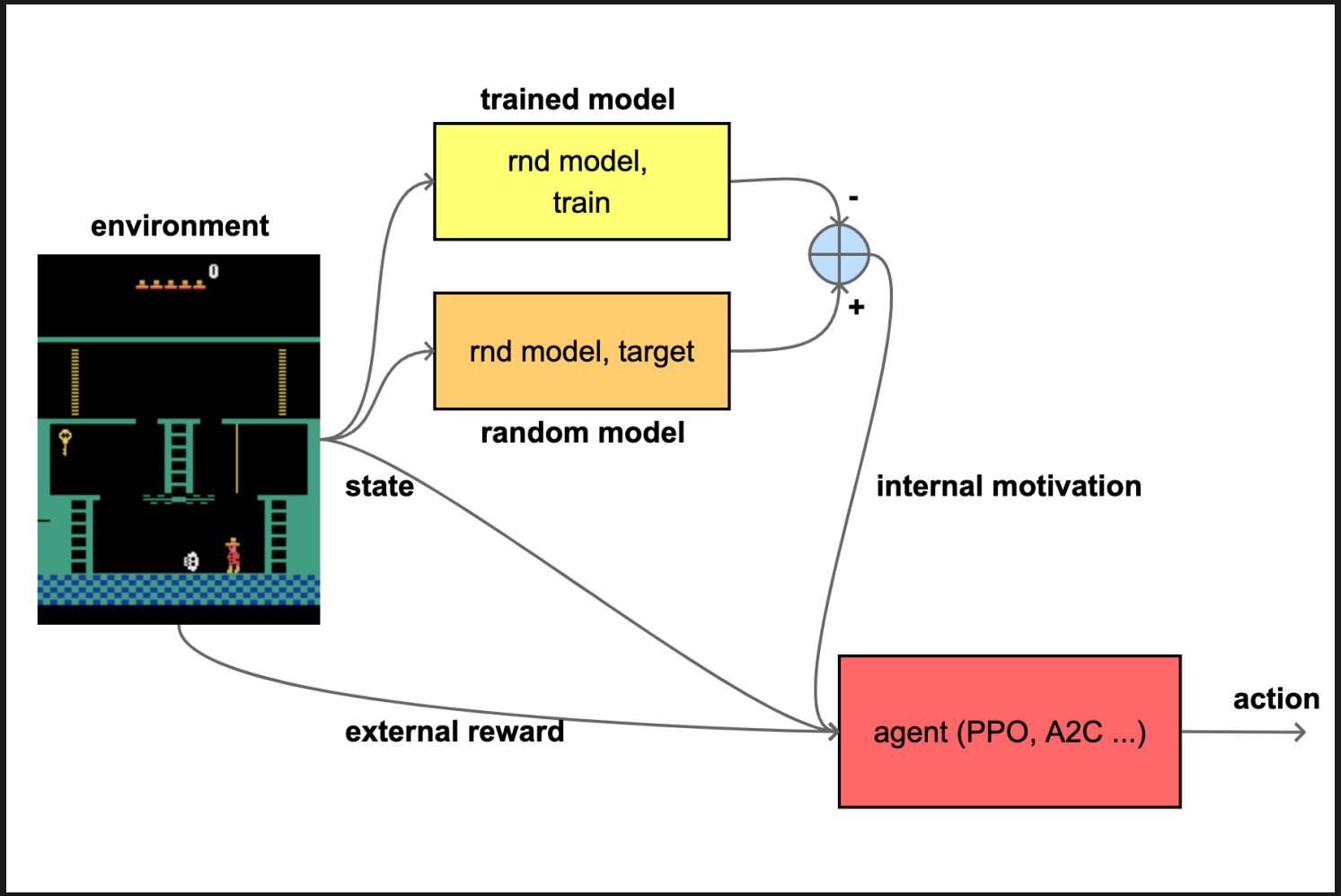
source : https://paperswithcode.com/sota/atari-games-on-atari-2600-montezumas-revenge

year	name	score
2015	Deep Reinforcement Learning with Double Q-learning	0
2017	Curiosity-driven Exploration by Self-supervised Prediction [a]	0
2021	MuZero	2500
2018	Count-Based Exploration with Neural Density Models [b]	3705
2019	**Exploration by Random Network Distillation** [c]	**8152**
2021	GoExplore* [d]	43 000

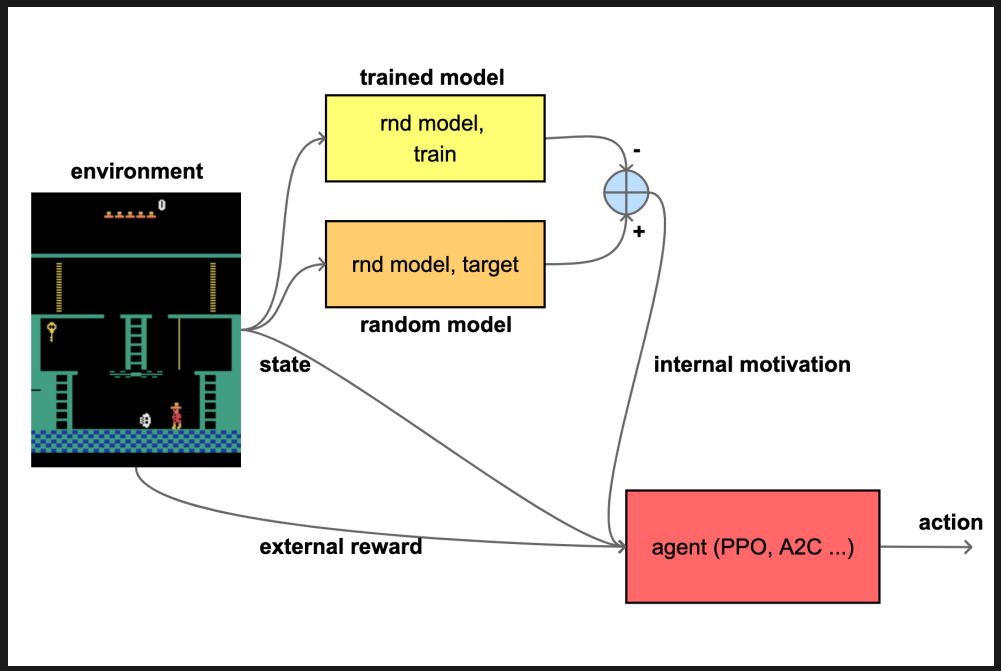
*** requires environment state saving/loading**

[a] https://arxiv.org/abs/1705.05363
[b] https://arxiv.org/abs/1703.01310
[c] https://arxiv.org/abs/1810.12894
[d] https://arxiv.org/abs/2004.12919

random network distillation

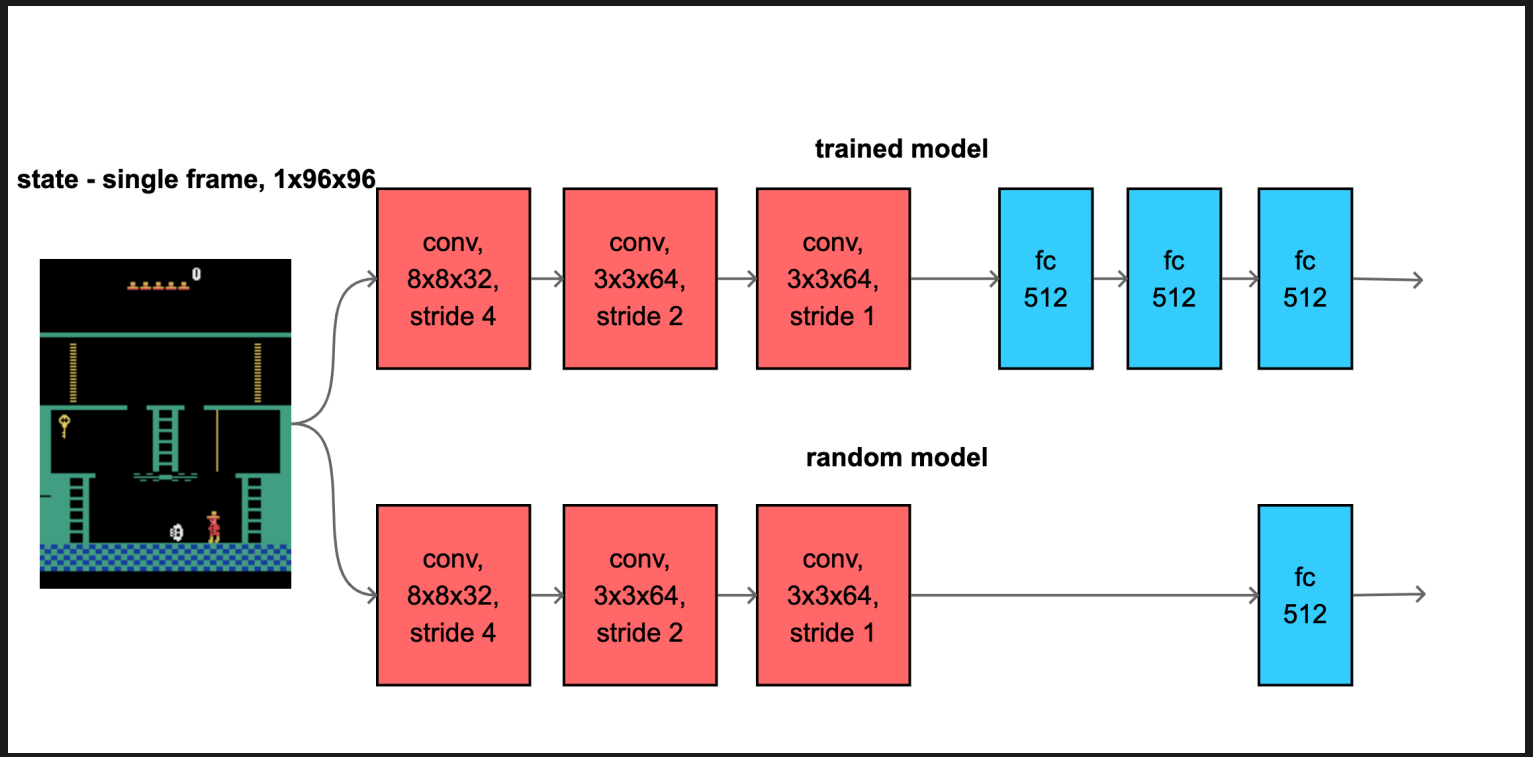


random network distillation

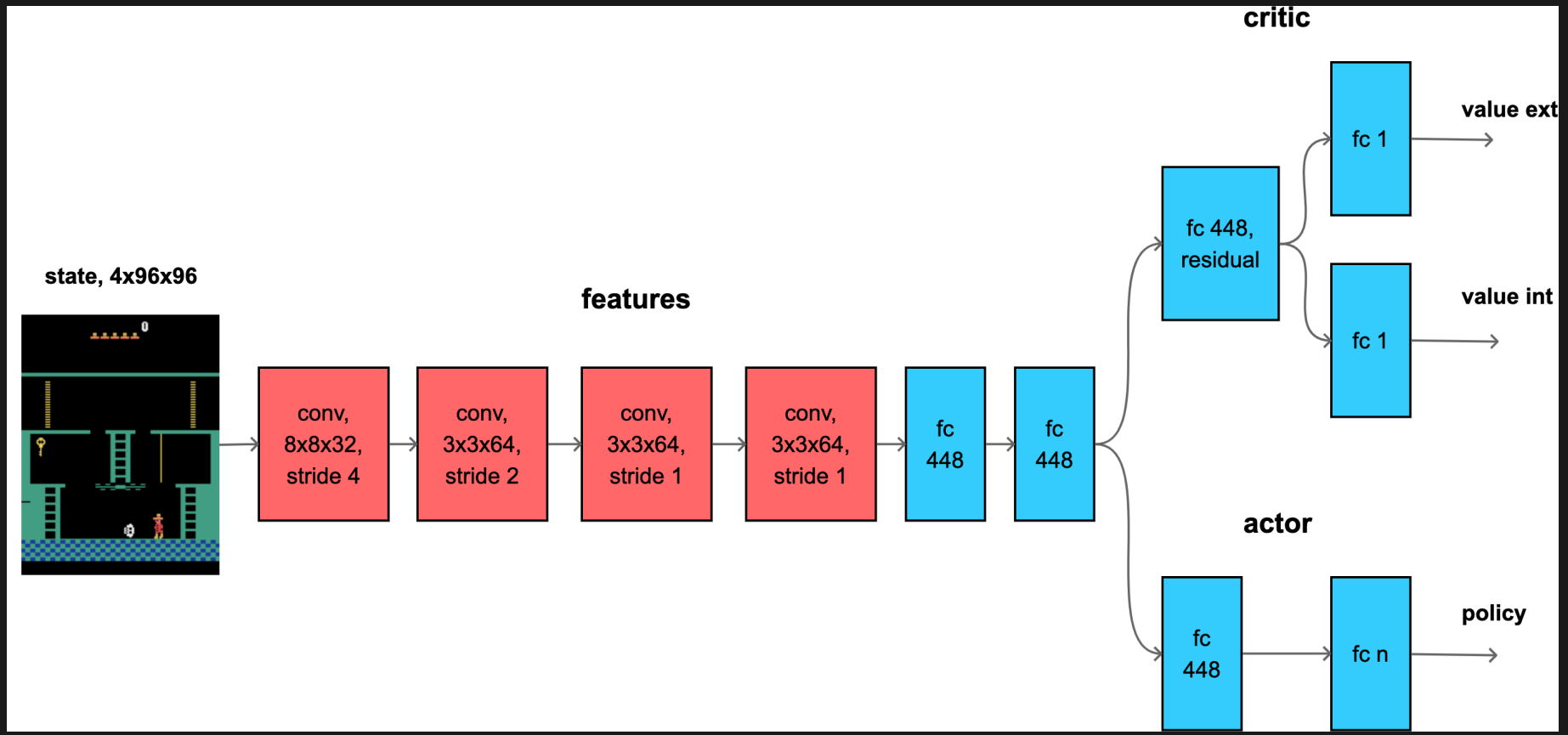


- neural network works as **novelty detector**
- model learns to imitate random (target) model
- **less visited states produce bigger motivation signal**
- orthogonal weights initialisation ($g = 2^{0.5}$) for strong signal
- lot of fully connected layers **to avoid generalisation**

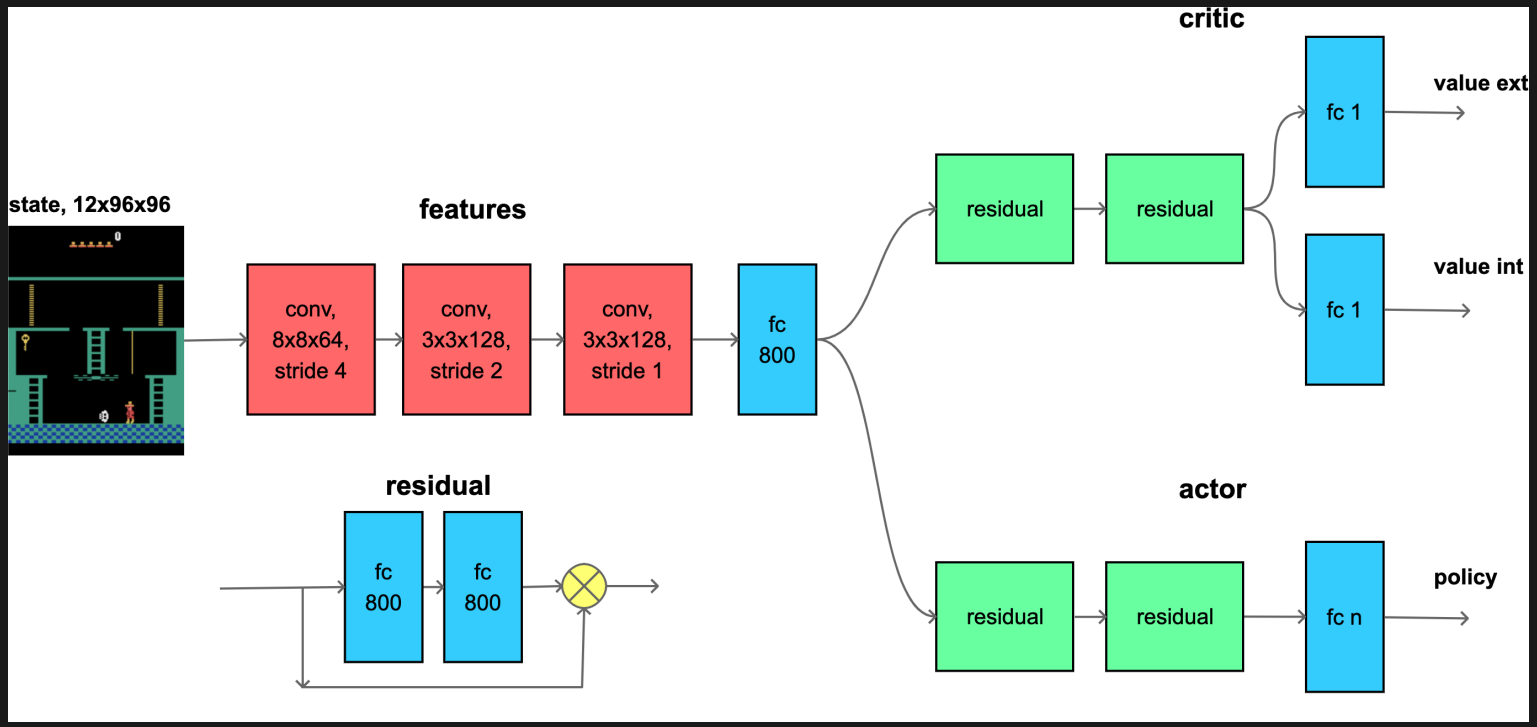
random network distillation architecture



ppo model architecture



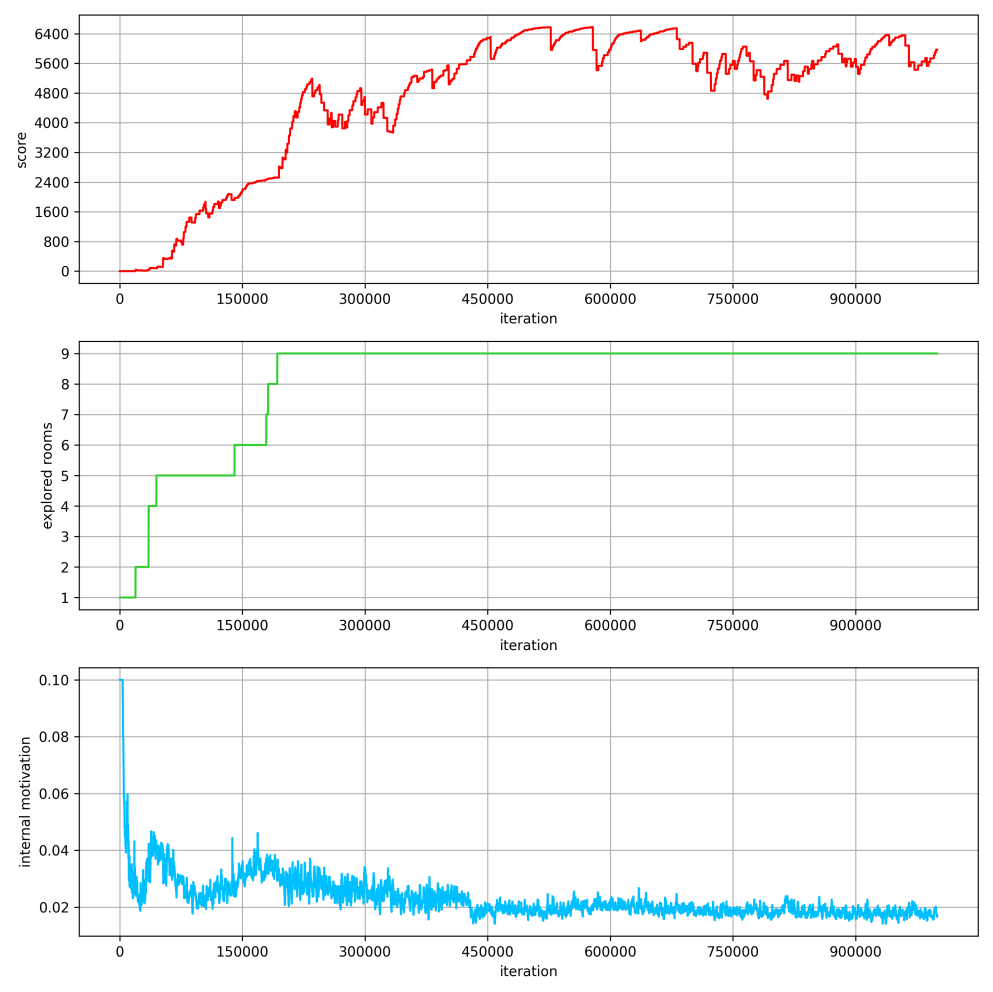
ppo model architecture



loss

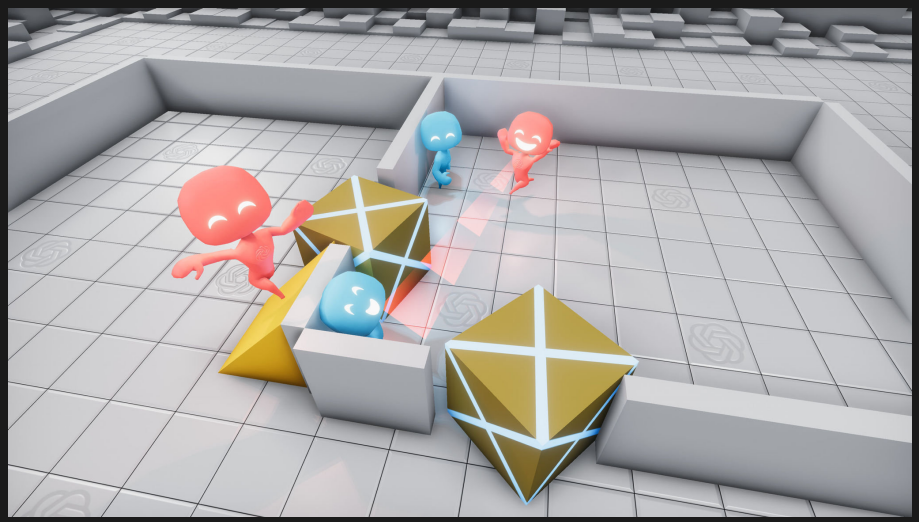
TODO

results



- 1M steps - **20% of original paper**
- 128 parallel envs = total 128M steps
- **score 6400**
- **9 rooms explored**

Emergent Tool Use From Multi-Agent Autocurricula



- multi-agent robotic environment
- hide and seek
- https://openai.com/blog/emergent-tool-use/
- https://arxiv.org/abs/1909.07528

Q&A