

# Playing Minecraft Game with Human Feedback

## SDSC8006 Presentation

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# Content

- ▶ Intro to the game and competition
- ▶ Methodology
- ▶ Experiment Results
- ▶ Conclusion

# Minecraft

- ▶ The Game: A famous open-world game with high freedom



Figure: Minecraft

# Competition

The Competition: NeurIPS 2022 MineRL BASALT [1]

- ▶ "Towards Solving Fuzzy Tasks with Human Feedback"

Challenge

- ▶ Complex Environments
- ▶ Hundreds Actions
- ▶ Sparse Reward
- ▶ Human Feedback

# Tasks

- ▶ Data:  
Gameplay recordings for each task
- ▶ Four Tasks:  
Find-Cave/Make-Waterfall/Create-AnimalPen/Build-Village  
House

# Methodology

- ▶ Random (bottom)
- ▶ Behavior Cloning (baseline—)
- ▶ KABasalt (RL)
- ▶ Human Expert (by Zichuan)

# Behavior Cloning

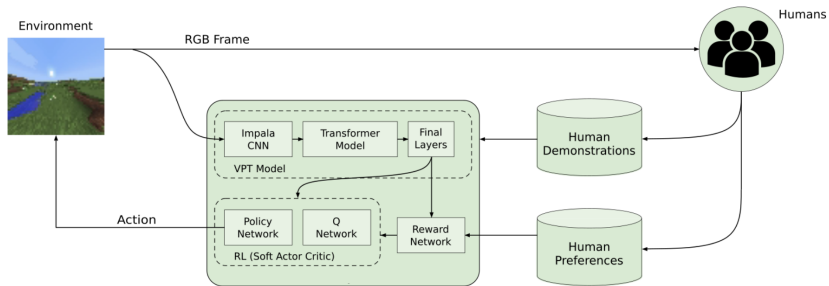
Behavior Cloning(BC): An imitation learning technique[2] that learns a policy to mimic the expert.[3]

- ▶ Expert Demonstrations:  $D = \{(s_1, a_1), (s_2, a_2), \dots, (s_N, a_N)\}$
- ▶ Learn the policy from the images
- ▶ Compare the two minimize the loss by update the policy
- ▶ Move to next step and loop the above

- ▶ Value Function: output of the OpenAI VPT model [4]
- ▶ State:  $128 \times 128 \times 3$  images
- ▶ Reward: learned from the Human preference
- ▶ Action: restricted to 16, eg. "Forward", "Left", "Jump"



# KAB Framework



## Video Results

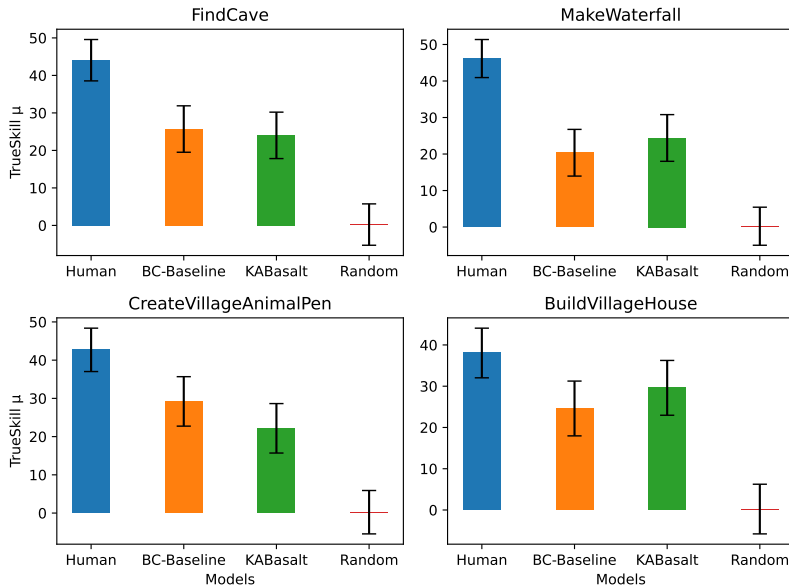
Find-Cave video:  
video link

# Evaluation

"TrueSkill" is a probabilistic rating system developed by Microsoft, primarily used to rate and match players in competitive gaming environments.[5]

- ▶ View each method as a kind of player
- ▶ The skill of each player follows a Normal Distribution with unknown parameters
- ▶ Rank players by pairs for each task in each round
- ▶ Update our belief about players by Bayesian approach, and get the posterior skill estimation

# Performance Comparison



# Performance Comparison

Model	FindCave	MakeWaterfall	CreateVillageAnimalPen	BuildVillageHouse	Average
Human	1.479	1.431	1.267	1.116	<b>1.323</b>
KABasalt	-0.108	0.021	-0.105	0.412	<b>0.055</b>
BC-Baseline	0.061	-0.283	0.446	-0.062	<b>0.041</b>
Random	-1.432	-1.169	-1.608	-1.466	-1.419

**Table:** Normalized TrueSkill scores for each model across the four tasks

# Discussion

- ▶ The game is tough. Human's Feedback is helpful.
- ▶ Behavior Cloning simplifies the learning by supervised learning. Efficiency in the early stage. Challenges with distribution shift, Mix of expert demonstrations.
- ▶ RL method learns the environment. Partially observable, Hard to define the reward.
- ▶ Further: Learn the reasoning, Better reward design,...

# References

- [1] Stephanie Milani et al. “Towards solving fuzzy tasks with human feedback: A retrospective of the miner1 basalt 2022 competition”. In: *arXiv preprint arXiv:2303.13512* (2023).
- [2] Adam Gleave et al. *imitation: Clean Imitation Learning Implementations*. 2022. arXiv: 2211.11972 [cs.LG].
- [3] Anssi Kanervisto, Janne Karttunen, and Ville Hautamäki. “Playing minecraft with behavioural cloning”. In: *NeurIPS 2019 Competition and Demonstration Track*. PMLR. 2020, pp. 56–66.
- [4] Antonin Raffin et al. “Stable-Baselines3: Reliable Reinforcement Learning Implementations”. In: *Journal of Machine Learning Research* 22:268 (2021), pp. 1–8. URL: <http://jmlr.org/papers/v22/20-1364.html>.
- [5] B Schölkopf, J Platt, and T TrueSkill Hofmann. “A Bayesian Skill Rating System”. In: *Advances in Neural Information Processing Systems* 20 (2006), pp. 569–576.