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# EXPLORING “WORLD MODELS”

(HA ET AL., 2018)

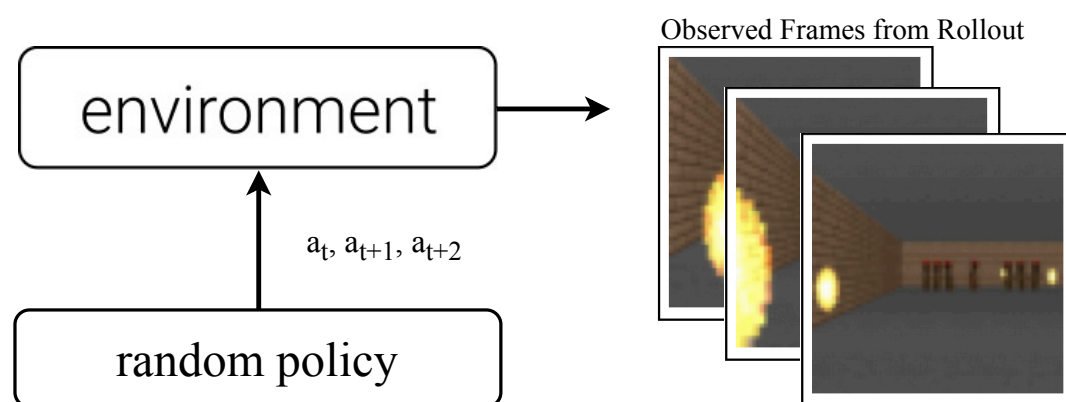
## WHY WORLD MODELS?

- ▶ Focus on the representation
- ▶ Decoupling of the model components
- ▶ Action-conditioned prediction in latent space
- ▶ Many avenues of exploration and extension

# APPROACH

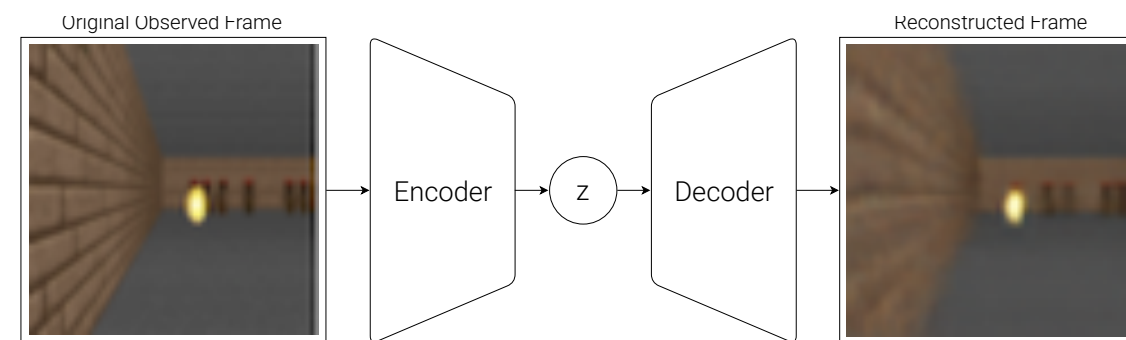
## THEIR APPROACH

1



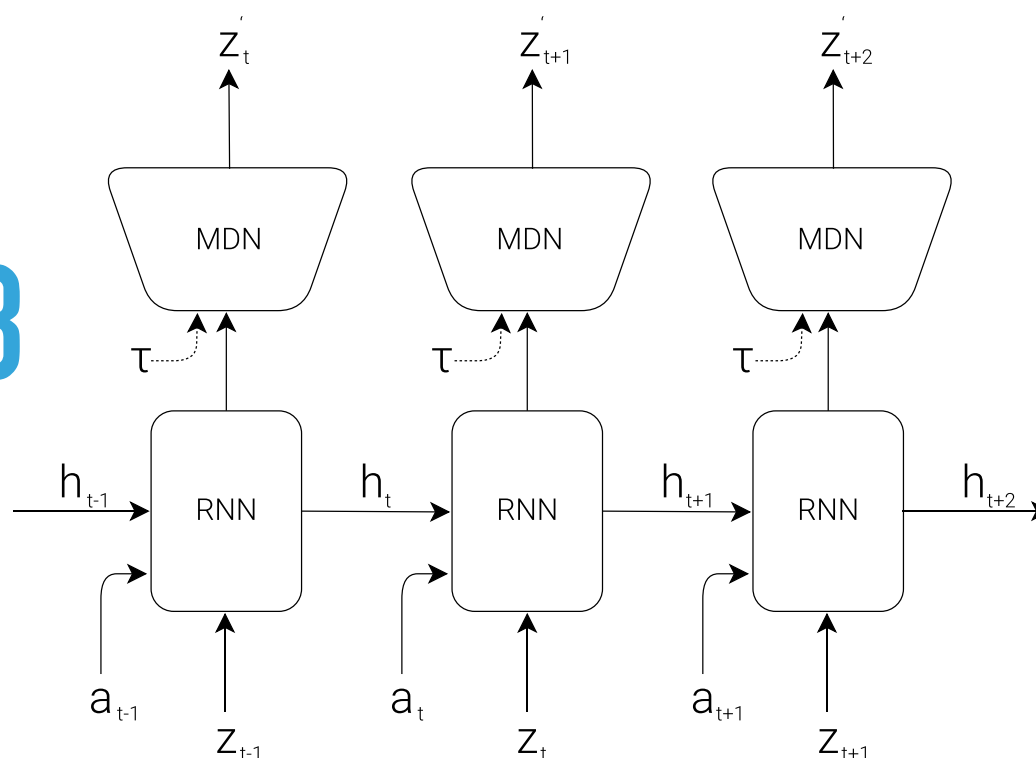
Generate actions, frames from random policy (10,000 rollouts)

2



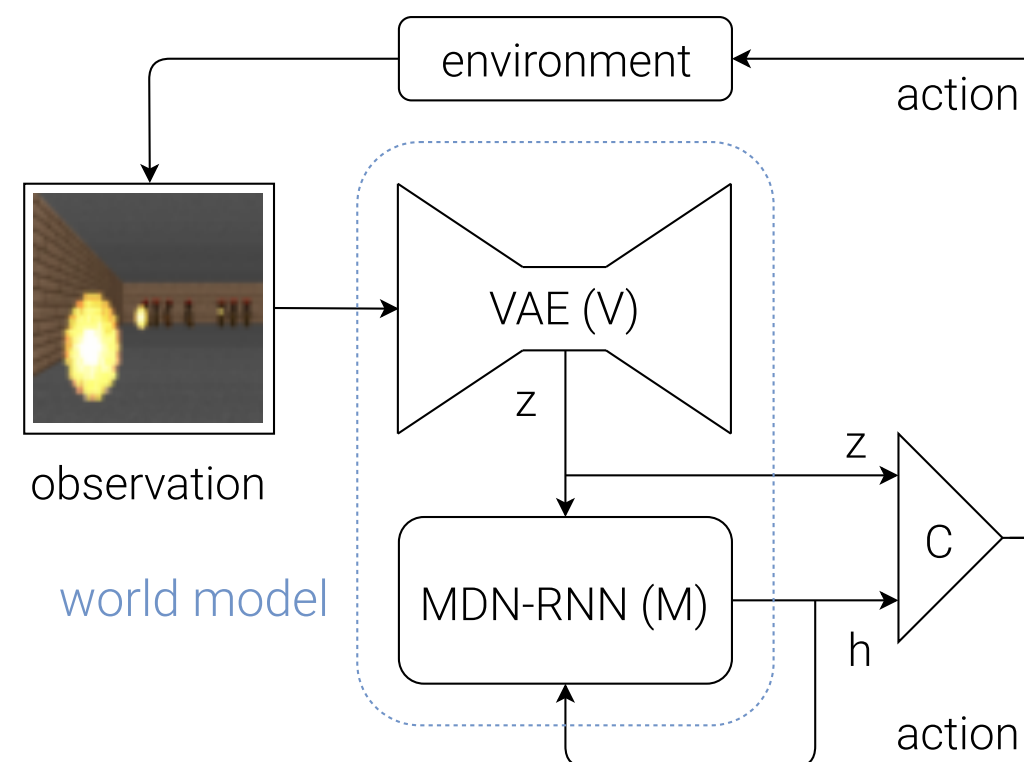
Train VAE to encode these frames into  $z$

3



Learn to predict future  $z$ , given previous  $z$  and the action that caused it from random policy

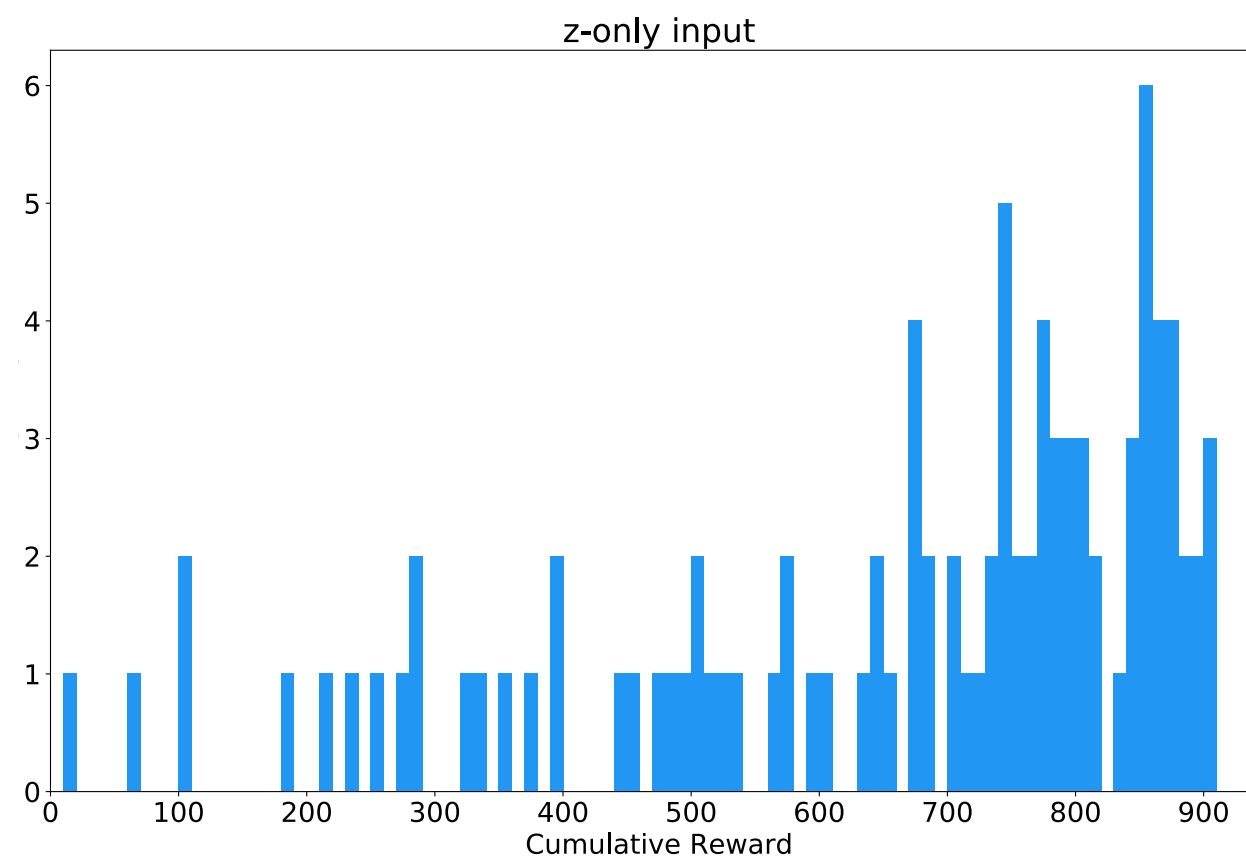
4



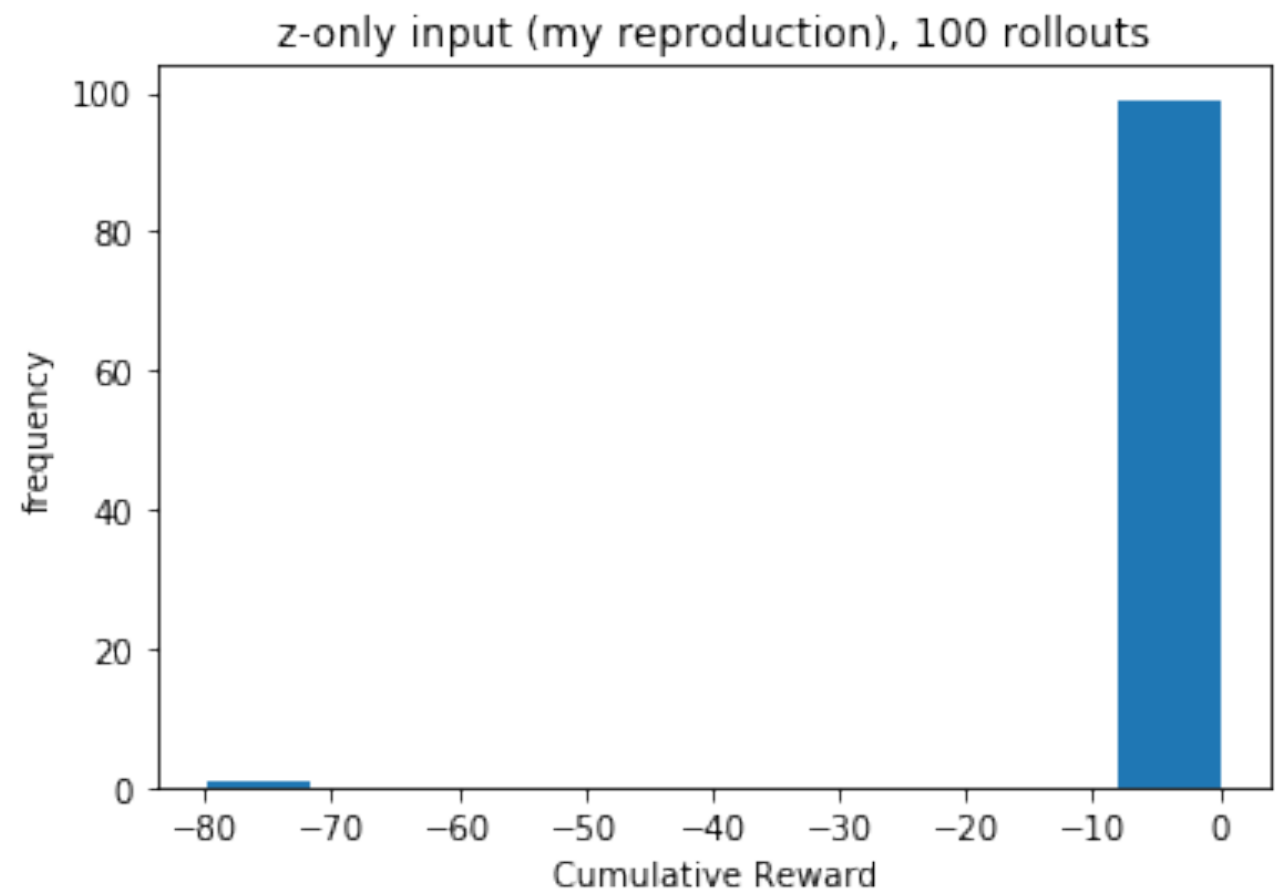
Learn best action using  $z$  from VAE and  $h$  from RNN as input using evolutionary algorithms

# RESULTS

## THEIR RESULTS VS. MINE



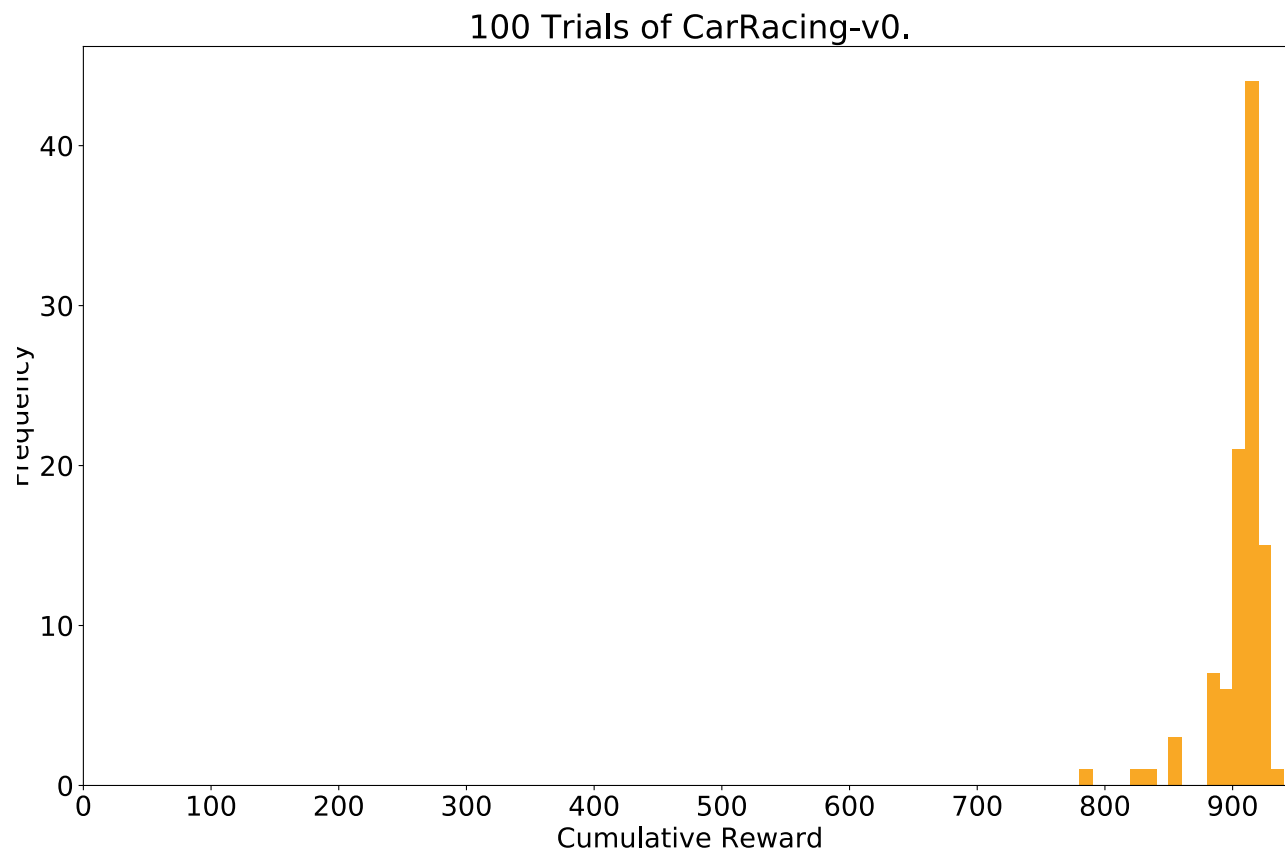
Their reward distribution for 100 rollouts with only input from VAE to controller



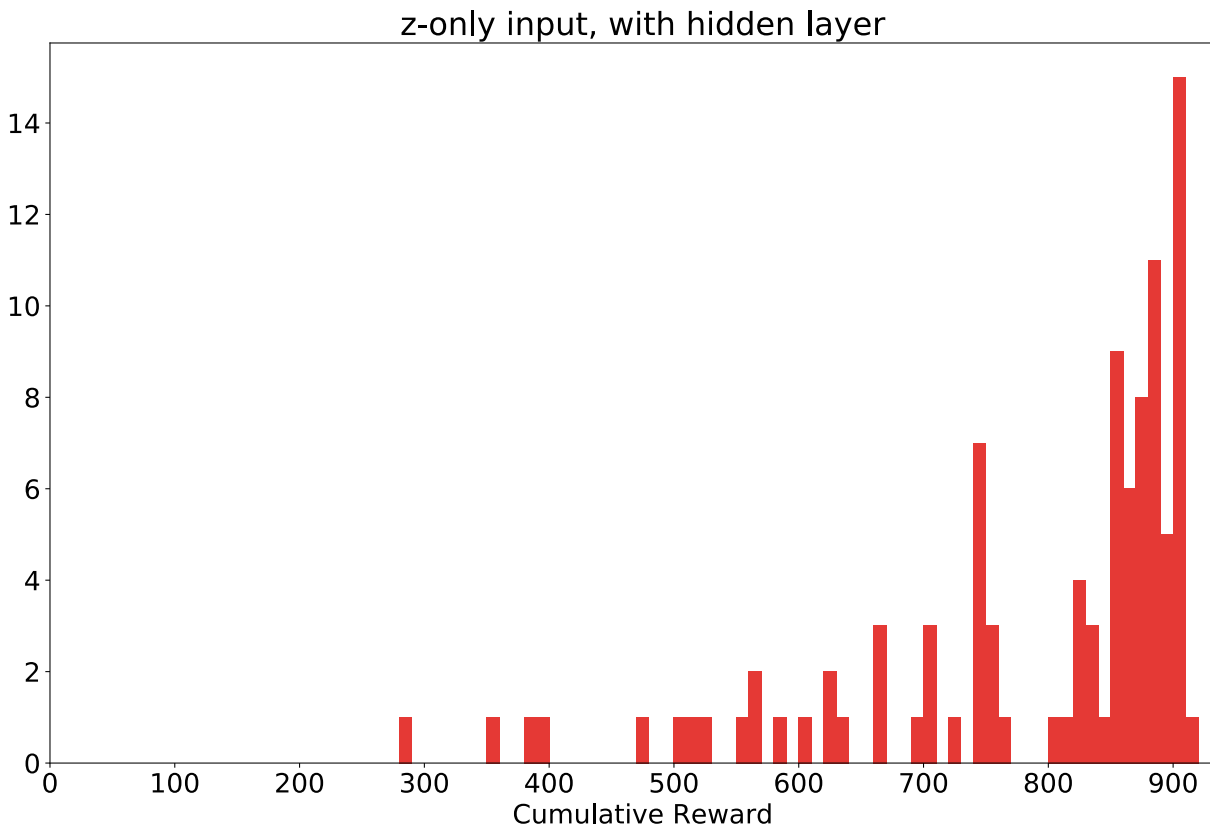
My reproduction of this with the time I had

# RESULTS/DISCUSSION

## MORE OF THEIR RESULTS



100 rollouts with z from VAE and h from LSTM



Just z from VAE, but added complexity to controller

# HOW DID IT GO?

- ▶ What more needs to be done to reproduce
  - ▶ Add in features from LSTM
  - ▶ Run evolutionary search longer to match their performance
- ▶ Future Questions to Address
  - ▶ Policy gradient (just for controller and end-to-end)
  - ▶ Why can't VAE see actions too?
  - ▶ Add some disentangling inductive bias to VAE?
  - ▶ Does it work for Atari?