A Agent architecture and hyperparameters

We note that our emphasis in this work was not on finding the overall best performing networks, so we did not extensively tune network and learning hyperparameters.

We trained agents using a distributed RL setup, with 4096 parallel actors. We trained the one-hot form using a 4x4 TPUv2, and the language form using a 4x4 TPUv3. Training runs took approximately 12-48 hours to reach maximum episode return (~ 200 /episode), typically after 50-200k learner steps.

		Input resolution		(160, 192, 3)
State update f_{θ}	Image encoder e^i_{θ}	ResNet	number of blocks channels per block conv layers per block conv filter size nonlinearity max-pool filter size max-pool strides	3 (16, 32, 32) (2, 2, 2) 3 ReLU 3 2
	String encoder e_{θ}^{s}	Tokenizer	tokenizer name vocabulary size max token length	subword 8000 19 (right-padded)
		Linear embedding	embeddings per token	16
		LSTM	hidden units	256
Memor		Input structure		$[e^{i}_{\theta}(i_{t}), e^{s}_{\theta}(s_{t}), a_{t-1}, r_{t-1}]$
cor	core	LSTM	hidden units	512
Policy head h_{θ}		Policy MLP	hidden units action space	$\begin{array}{c} 200 \\ \in [-1,1]^4 \end{array}$
		Value MLP	hidden units	200
CST head $g_{ heta}$		MLP	hidden units	32 (one-hot) 512 (language)

Table 1. Agent architecture

V-Trace Loss	baseline cost entropy cost γ max reward	$ \begin{array}{r} 1.0 \\ 0.001 \\ 0.95 \\ 1.0 \\ \end{array} $
Adam Optimizer	learning rate β_1 β_2 clip grad norm above	
Schedule	Schedule batch size termination steps	

Table 2: Training hyperparameters.