## 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 ( $\sim 200$ /episode), typically after 50-200k learner steps.

		Input resolution		(160, 192, 3)	
State update $f_{\theta}$	Image encoder $e^i_{\theta}$	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 $e_{\theta}^{s}$ Memory	Tokenizer	tokenizer name vocabulary size max token length	subword 8000 19 (right-padded)	
		Linear embedding	embeddings per token	16	
		LSTM	hidden units	256	
		Input structure		$[e^i_{\theta}(i_t), e^s_{\theta}(s_t), a_{t-1}, r_{t-1}]$	
	core	LSTM	hidden units	512	
Policy head $h_{\theta}$		Policy MLP   hidden units action space		$200 \in [-1,1]^4$	
		Value MLP	hidden units	200	
$\begin{array}{c} \text{CST head} \\ g_{\theta} \end{array}$		MLP	hidden units	32 (one-hot) 512 (language)	

Table 1: Agent architecture.		Table	1:	Agent	architecture.
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V-Trace Loss	baseline cost entropy cost $\gamma$ max reward	$\begin{array}{c} 1.0 \\ 0.001 \\ 0.95 \\ 1.0 \end{array}$
Adam Optimizer	$\begin{vmatrix} \text{learning rate} \\ \beta_1 \\ \beta_2 \\ \text{clip grad norm} \\ \text{above} \end{vmatrix}$	
Schedule	batch size termination steps	$\begin{array}{c} 192 \\ 6e^7 \end{array}$

Table 2: Training hyperparameters.