

## APPENDIX

The following Table I shows the hyper-parameters of each competence-based algorithm used in this research.

TABLE I: Per algorithm sets of hyper-parameters.

<b>SMM hyper-parameters</b>	<b>Value</b>
Skill dim.	4
Skill discriminator lr	$10^{-3}$
VAE lr	$10^{-2}$
<b>DIAYN hyper-param.</b>	<b>Value</b>
Skill dim	16
Skill sampling frequency	50 steps
Discriminator net arch.	512→1024→1024 →16 ReLU MLP
<b>APS hyper-parameters</b>	<b>Value</b>
Representation dim.	512
Reward transformation	$\log(r + 1.0)$
Successor feat. dim.	10
Successor feat. net arch.	$ \mathcal{O}  \rightarrow 1024 \rightarrow 1024$ →10 ReLU MLP
k in NN	12
Avg top k in NN	True
Least square batch size	4096

The Homeostasis Algorithm 1, which is defined in [1], is used for the evaluation implementation of our model.

### Algorithm 1 Homeostasis

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1: Require:
   Target rate  $\rho$ 
2: Initialize:
    $\bar{x} \leftarrow 0$ ,  $\bar{x}^2 \leftarrow 1$ ,  $\bar{x}^+ \leftarrow 1$ 
3: for  $t \in \{1, \dots, T\}$  do
4:   obtain next scalar signal return  $x_t$ 
5:   set time-scale  $\tau \leftarrow \min(t, \frac{100}{\rho})$ 
6:   update moving average  $\bar{x} \leftarrow (1 - \frac{1}{\tau})\bar{x} + \frac{1}{\tau}x_t$ 
7:   update moving variance  $\bar{x}^2 \leftarrow (1 - \frac{1}{\tau})\bar{x}^2 + \frac{1}{\tau}(x_t - \bar{x})^2$ 
8:   standardise and exponentiate  $x^+ \leftarrow \exp\left(\frac{x_t - \bar{x}}{\sqrt{\bar{x}^2}}\right)$ 
9:   update transformed moving average
10:   $\bar{x}^+ \leftarrow (1 - \frac{1}{\tau})\bar{x}^+ + \frac{1}{\tau}x^+$ 
11:  sample  $y_t \sim \text{Bernoulli}\left(\min\left(1, \rho \frac{x^+}{\bar{x}^+}\right)\right)$ 
12: end for

```

## REFERENCES

- [1] M. Pislár, D. Szepesvári, G. Ostrovski, D. Borsa, and T. Schaul, “When should agents explore?” *arXiv preprint arXiv:2108.11811*, 2021.