

## We propose a framework for **lifelong learning** that leverages **zero-shot composition** to solve new tasks that are expressible as **logical combinations** of learned ones.

### Goal Oriented RL

We define an **extended value function (EVF)** that decouples the values for each absorbing state:

$$Q(s, g, a) = \bar{r}(s, g, a) + \int_S V^{\pi g}(s') \rho_{(s,a)}(ds')$$

$$\bar{r}(s, g, a) = \begin{cases} N & \text{if } g \neq s \in G \\ r(s, a) & \text{otherwise} \end{cases}$$

Similar to UVFAs [3] but uses **extended rewards**.

### Compositionality

**Theorem 1:** Let  $\mathbf{M}$  be the set of tasks. Then  $\mathbf{M}$  forms a Boolean algebra when equipped with the **or**, **and**, and **not** operators given by:

$$\blacksquare \text{ or } \blacksquare = (S, A, p, r(\blacksquare \text{ or } \blacksquare)), \text{ where } r(\blacksquare \text{ or } \blacksquare) = \max\{r(\blacksquare), r(\blacksquare)\}$$

$$\blacksquare \text{ and } \blacksquare = (S, A, p, r(\blacksquare \text{ and } \blacksquare)), \text{ where } r(\blacksquare \text{ and } \blacksquare) = \min\{r(\blacksquare), r(\blacksquare)\}$$

$$\text{not } \blacksquare = (S, A, p, r(\text{not } \blacksquare)), \text{ where } r(\text{not } \blacksquare) = (r_{MAX} + r_{MIN}) - r(\blacksquare)$$

where,  $r_{MAX}$  and  $r_{MIN}$  are the reward functions for the maximum and minimum tasks.

**Theorem 2:** Let  $\mathbf{Q}$  be the set of extended value functions. Then  $\mathbf{Q}$  forms a Boolean algebra when equipped with the **or**, **and**, and **not** operators given by:

$$Q^*(\blacksquare) \text{ or } Q^*(\blacksquare) = \max\{Q^*(\blacksquare), Q^*(\blacksquare)\}$$

$$Q^*(\blacksquare) \text{ and } Q^*(\blacksquare) = \min\{Q^*(\blacksquare), Q^*(\blacksquare)\}$$

$$\text{not } Q^*(\blacksquare) = (Q^*_{MAX} + Q^*_{MIN}) - Q^*(\blacksquare)$$

where,  $Q^*_{MAX}$  and  $Q^*_{MIN}$  are the extended value functions for the maximum and minimum tasks.

**Theorem 3:** The task and extended value function spaces are homomorphic.

### Lifelong RL With Composition

**Algorithm 1:** Lifelong RL with Composition

```

Input: Learning method  $\mathcal{A}$ , task distribution  $\mathcal{D}$ 
 $\mathcal{T} \leftarrow \emptyset$ 
 $\bar{Q}^* \leftarrow \emptyset$ 
while True do
     $T, M \sim \mathcal{D}$ 
     $\mathcal{E}\mathcal{X}\mathcal{P} \leftarrow \text{SumOfProducts}(T, T)$ 
    if  $T = \text{Evaluate}(\mathcal{E}\mathcal{X}\mathcal{P}, T)$  then
        // Zero-shot transfer
         $\bar{Q}^*_T \leftarrow \text{Evaluate}(\mathcal{E}\mathcal{X}\mathcal{P}, \bar{Q}^*)$ 
    else
        // Learn and add to library
         $\bar{Q}^*_T \leftarrow \mathcal{A}(M)$ 
         $\mathcal{T} \leftarrow \mathcal{T} \cup T$ 
         $\bar{Q}^* \leftarrow \bar{Q}^* \cup \bar{Q}^*_T$ 
    end
     $\pi^*(s) \in \arg \max_{a \in \mathcal{A}} \max_{g \in \mathcal{G}} \bar{Q}^*_T(s, g, a)$ 
    Solve task  $M$  using  $\pi^*(s)$  for all  $s \in S$ 
end
    
```

Goals	$M_0$	$M_1$	$M$
$g_0$	0	0	0
$g_1$	0	1	1
$g_2$	1	0	1
$g_3$	1	1	0

#### Sum Of Products

- >  $M' := (\neg M_0 \wedge M_1) \vee (M_0 \wedge \neg M_1)$
- >  $M = M' ?$  **Yes!**
- >  $Q' := (\neg Q'_0 \wedge Q'_1) \vee (Q'_0 \wedge \neg Q'_1)$

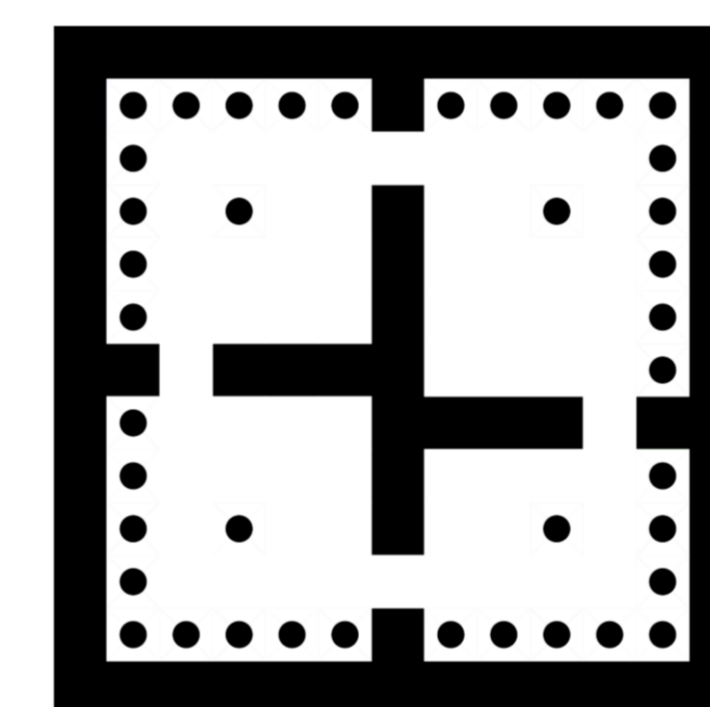
Goals	$M_0$	$M_1$	$M$
$g_0$	0	1	0
$g_1$	0	1	1
$g_2$	1	0	1
$g_3$	1	1	0

#### Sum Of Products

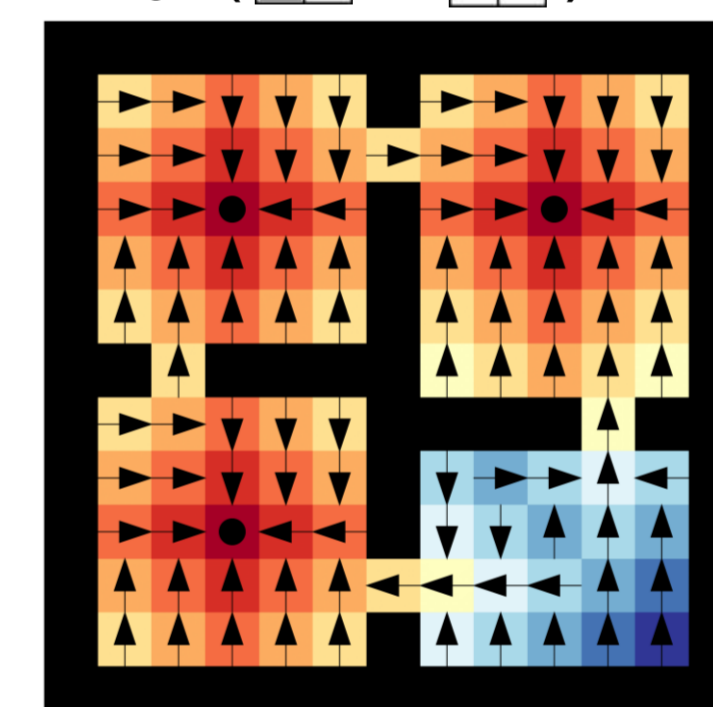
- >  $M' := (\neg M_0 \wedge M_1) \vee (M_0 \wedge \neg M_1)$
- >  $M = M' ?$  **No!**
- > Learn and add to library

### Experiment: Zero-shot Composition

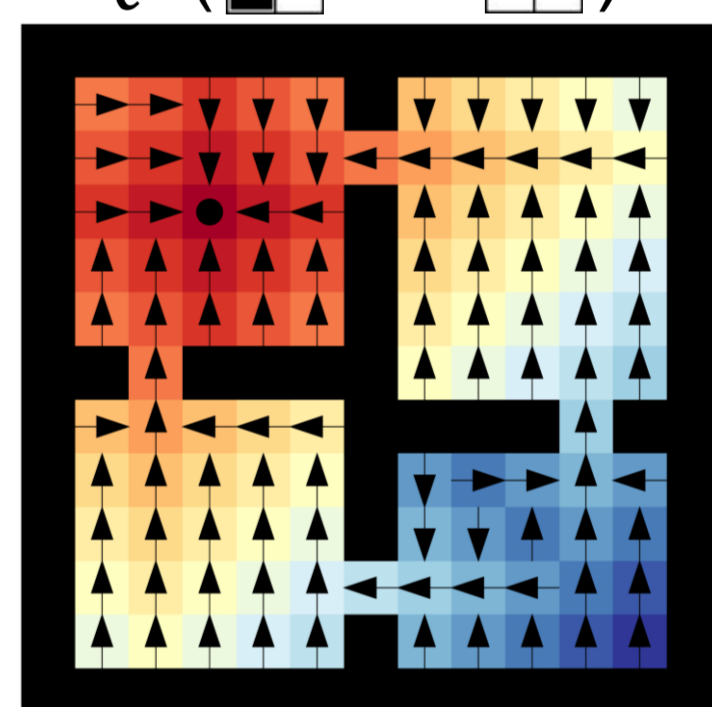
$\sim 10^{12}$  total tasks



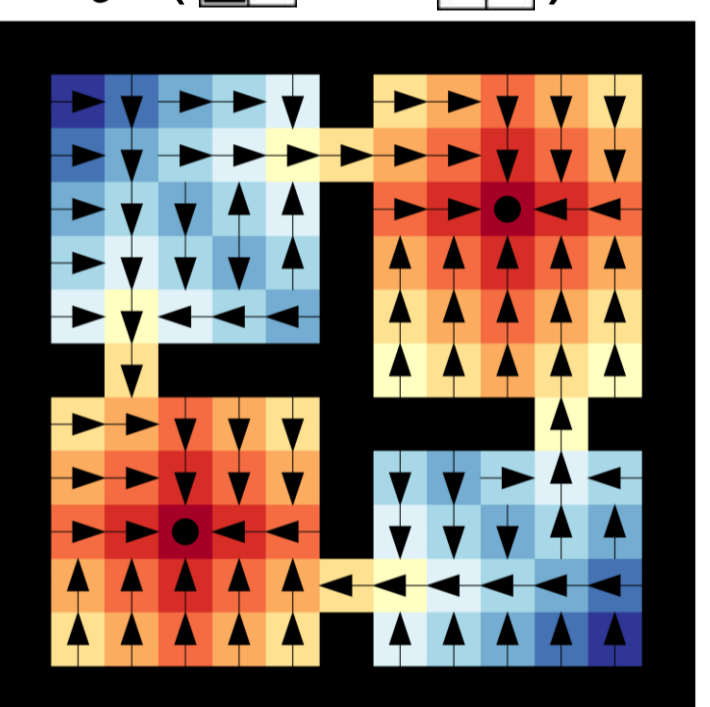
$Q^*(\blacksquare \text{ or } \blacksquare)$



$Q^*(\blacksquare \text{ and } \blacksquare)$



$Q^*(\blacksquare \text{ xor } \blacksquare)$



### Experiment: Lifelong RL With Composition

