# Structured World Belief

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# **Object-Centric Temporal Models**





SCALOR (Jiang et al., 2020)

SILOT (Crawford et al., 2020)





*GSWM* (*Lin et al., 2020*)



Not Visible

# Partial Observability and Belief



# Can we integrate **object-centric** and **belief** representations into one model?

# Belief

#### How can we represent the object-level uncertainty?



Previous Object-Centric Models



#### How can we represent the object-level uncertainty?



Previous Object-Centric Models



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Previous Object-Centric Models



# Learning Objective

We maximize the log probability of the AESMC ELBO objective.

where,

 $w_t^k = \tilde{w}_{t-1}^{a_{t-1}^k} \cdot p_{\theta}(x_t)$ 

#### We maximize the log probability of the image sequence i.e. $\log p(x_{1:T})$ via

$$\mathscr{L}_{\theta,\phi} = \frac{1}{T} \sum_{t=1}^{T} \log \sum_{k=1}^{K} w_t^k$$

$$s_{t}^{k} \frac{p_{\theta}(s_{t}^{k} | s_{t-1}^{a_{t-1}^{k}})}{q_{\theta,\phi}(s_{t}^{k} | s_{t-1}^{a_{t-1}^{k}}, x_{t})}.$$

### Belief tracking in 2D Branching Sprites data set.

Observation



Particles



Object Files





True Paths

#### Belief tracking and gameplay in 3D Food Chase.





R V

Particles

Object Files



#### Benefits of structured belief for A2C agents.



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# Conclusion

- I. Propose scene representation learning model combining object-centric representation and belief.
- 2. Show benefits in world modeling, RL, planning and supervised reasoning.
- 3. Solve the object matching problem without inductive biases such as spatial locality.