

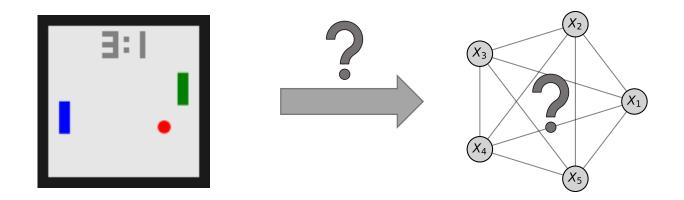
CITRIS: Causal Identifiability from Temporal Intervened Sequences



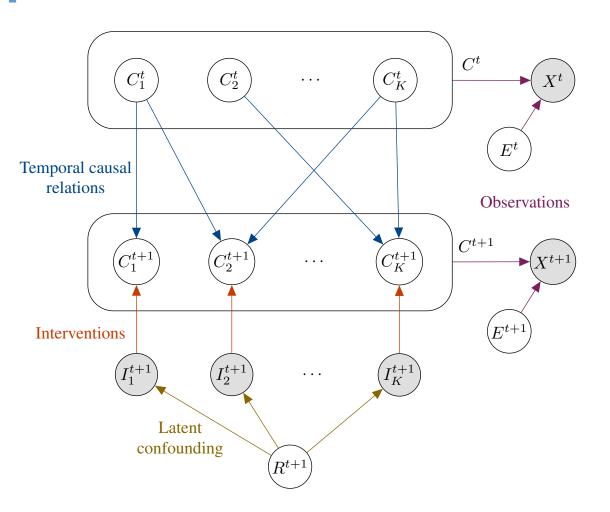


Causal Representation Learning

- Given high-dimensional observations of a dynamical system, what are the true varying factors?
- Crucial for reasoning, planning, generalization, and more
- Most works consider only scalar causal variables, but can we abstract further to multidimensional causal factors?
 - We may not be able to intervene on all scalar variables separately
 - Represent single causal variable by multiple latent dimensions for better optimization



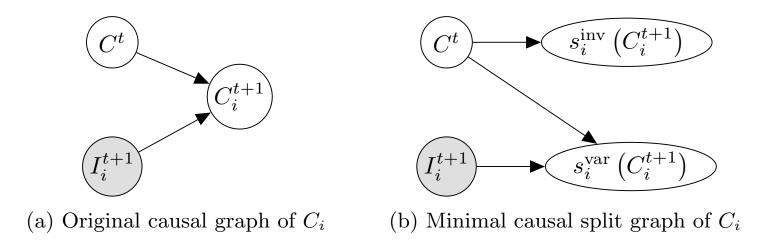
Causal Identifiability from Temporal Intervened Sequences Setup



- Multidimensional latent causal variables C_1^t, \ldots, C_K^t
- Goal: identify causal variables from observation pairs X^t , X^{t+1} and observed intervention targets I^{t+1}

Causal Identifiability from Temporal Intervened Sequences Minimal Causal Variables

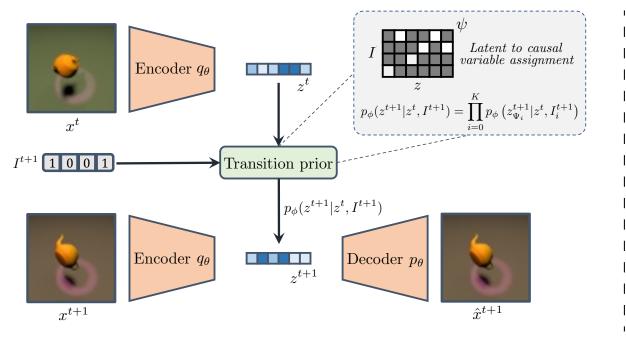
• Main theoretical result: we can identify the *minimal causal variables*, i.e. the information/mechanism of a causal variable which strictly depends on the interventions



CITRIS Architecture

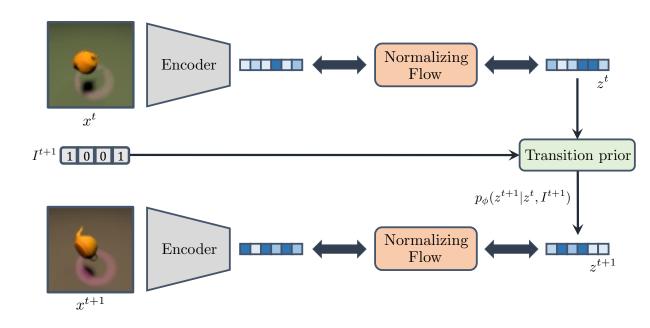
CITRIS-VAE

VAE with learned latent-to-causal variable assignment



CITRIS-NF

Normalizing Flow on pretrained autoencoder



Experiments Temporal Causal3DIdent dataset

Novel combinations of causal factors



Image 1



Image 2

Image 1



Image 2



Ground Truth

Ground Truth

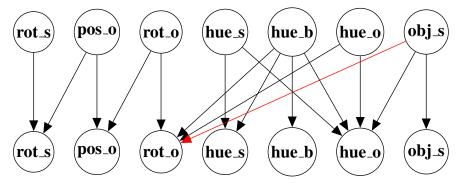


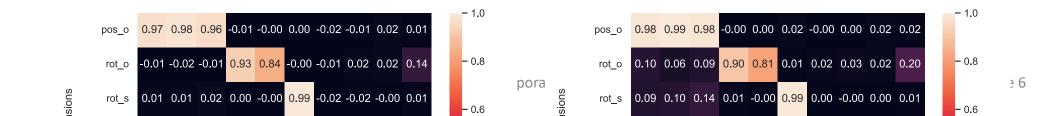


Prediction

Prediction

Causal Graph





Conclusion

- CITRIS: finding multidimensional causal variables from temporal sequences with interventions
- Main characteristics of approach:
 - Identifiable minimal causal variables depend on provided interventions
 - Supports modeling causal variables in arbitrary number of latent dimensions
 - Disentangling latent representation of pretrained autoencoder with normalizing flow
 - Scales to visually complex 3d scenes

Hope to see you at our poster later!

