







Weidentify multidimensional causal variables from videos with interventions.



Paper and code

CITRIS: Causal Identifiability from Temporal Intervened Sequences

Phillip Lippe¹ (p.lippe@uva.nl), Sara Magliacane^{2,3}, Sindy Löwe⁴, Yuki M. Asano¹, Taco Cohen⁵, Efstratios Gavves¹ ¹University of Amsterdam, QUVA lab; ²University of Amsterdam, INDE lab; ³MIT-IBM Watson AI Lab; ⁴University of Amsterdam, Bosch-Delta lab; ⁴Qualcomm AI Research, The Netherlands









Architectures

CITRIS-NF: Combining Pretrained Autoencoders with Normalizing Flows I^{t+1} 1 0 0 1 Transition prior $p_{\phi}(z^{t+1}|z^t, I^{t+1})$ Flow

Experiments

Temporal Causal3DIdent

3D rendered images with 7 causal factors





Image]



Image 2

Ground Truth



Prediction

Prediction

	$\mathbf{Triplet} \mathbf{dist} {\downarrow} $	Correlation metrics				
	Mean	$R^2 \operatorname{diag} \uparrow$	$R^2 \operatorname{sep} \downarrow$	Spearman diag \uparrow	Spearman sep \downarrow	
SlowVAE	0.40	0.61	0.23	0.59	0.27	
\mathbf{iVAE}^*	0.27	0.80	0.29	0.77	0.28	
CITRIS-VAE	0.21	0.89	0.10	0.88	0.12	
CITRIS-NF	0.07	0.98	0.04	0.97	0.08	
- 5 shapes	0.09	0.98	0.05	0.97	0.10	
- 2 unseen shapes	0.23	0.94	0.15	0.93	0.19	



CITRIS disentangles all seven causal variables well, while finding the causal graph and generalizing beyond seen instances!

Interventional Pong

Pong game dynamics with 5 intervened causal factors

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	$\mathbf{Triplets}\downarrow$	R^2 diag \uparrow	$R^2 \operatorname{\mathbf{sep}} \downarrow$	Spearman diag	Spearman sep
SlowVAE	0.34	0.61	0.17	0.66	0.23
\mathbf{iVAE}^*	0.09	0.91	0.04	0.92	0.06
CITRIS-VAE	0.03	0.99	0.01	0.99	0.05
CITRIS-NF	0.02	1.00	0.04	1.00	0.10

CITRIS identifies the five causal variables while disentangling them from the unintervened causal variables!