

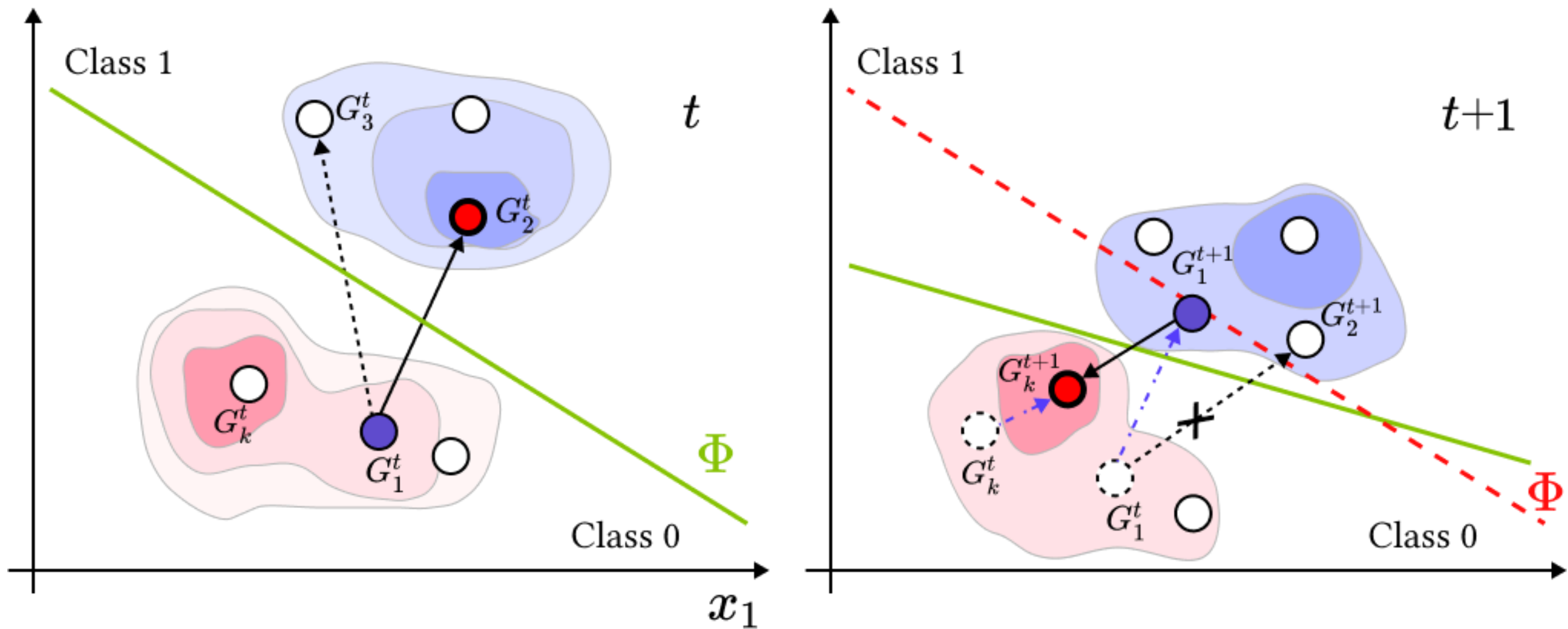
Unifying Evolution, Explanation, and Discernment: A Generative Approach for Dynamic Graph Counterfactuals

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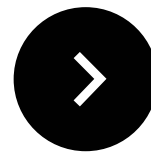




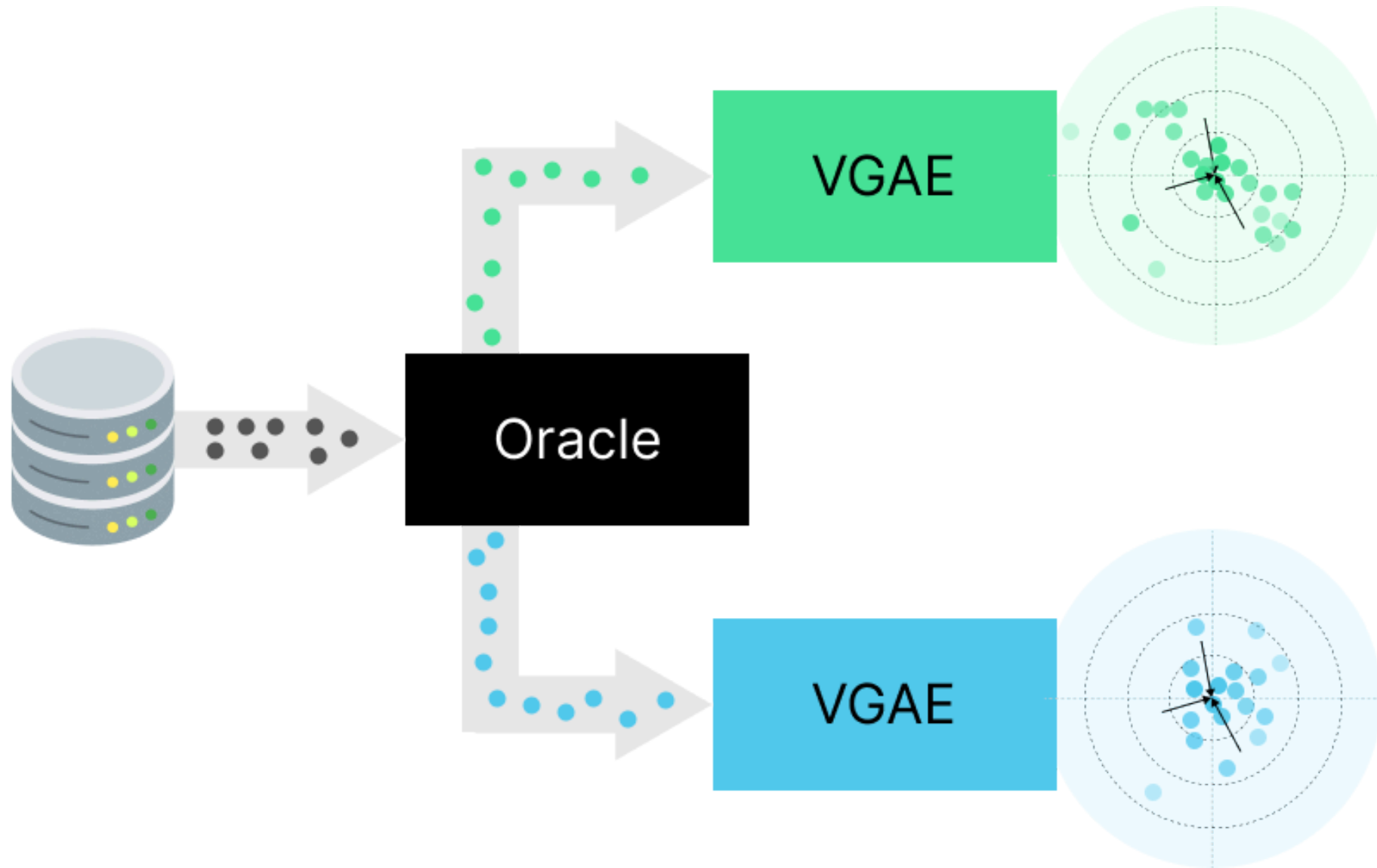
- Correct oracle decision boundary
- - - Old oracle decision boundary
- - - - Potential counterfactual association

- Correct counterfactual association
- - - → Old counterfactual association
- · - · → Graph movement from t to $t+1$

- Graph at current time
- ⊘ Graph at previous time



Class Representation Experts



Variational Autoencoders

We consider the following generative model where the graphs G are generated from factored latent representation and the true class label y

$$p(G|y) = \int_{\mathbf{z} \in \mathcal{Z}} p(G|\mathbf{z}, y) p(\mathbf{z}|y) dz$$



Variational Autoencoders (Decoder)

To represent $p(G|y)$, we use a single VGAE for each class $y \in Y$, which is dependent on the class where each node has a latent vector and then define

$$\begin{aligned} p_{\theta_y}(G|\mathbf{z}, y) &= p_{\theta_y}(\mathbf{A}, \mathbf{X}|\mathbf{z}, y) \\ &= p_{\theta_y}(\mathbf{X}|\mathbf{A}, \mathbf{z}, y) p_{\theta_y}(\mathbf{A}|\mathbf{z}, y) \end{aligned}$$



Variational Autoencoders (Encoder)

$$q_{\varphi_y}(\mathbf{z}|G, y) = \prod_{v_i} q_{\varphi_y}(z_{v_i}|G, y)$$

$$q(z_{v_i}|G, y) = \mathcal{N}\left(z_{v_i} | \mu_{v_i}, \gamma^2 \mathbf{I}\right),$$

> 0 and fixed
hyperparameter

$$\mu = \left[\mu_{v_1}, \dots, \mu_{v_n} \right] = \text{GCN}_{\varphi_y}(G)$$

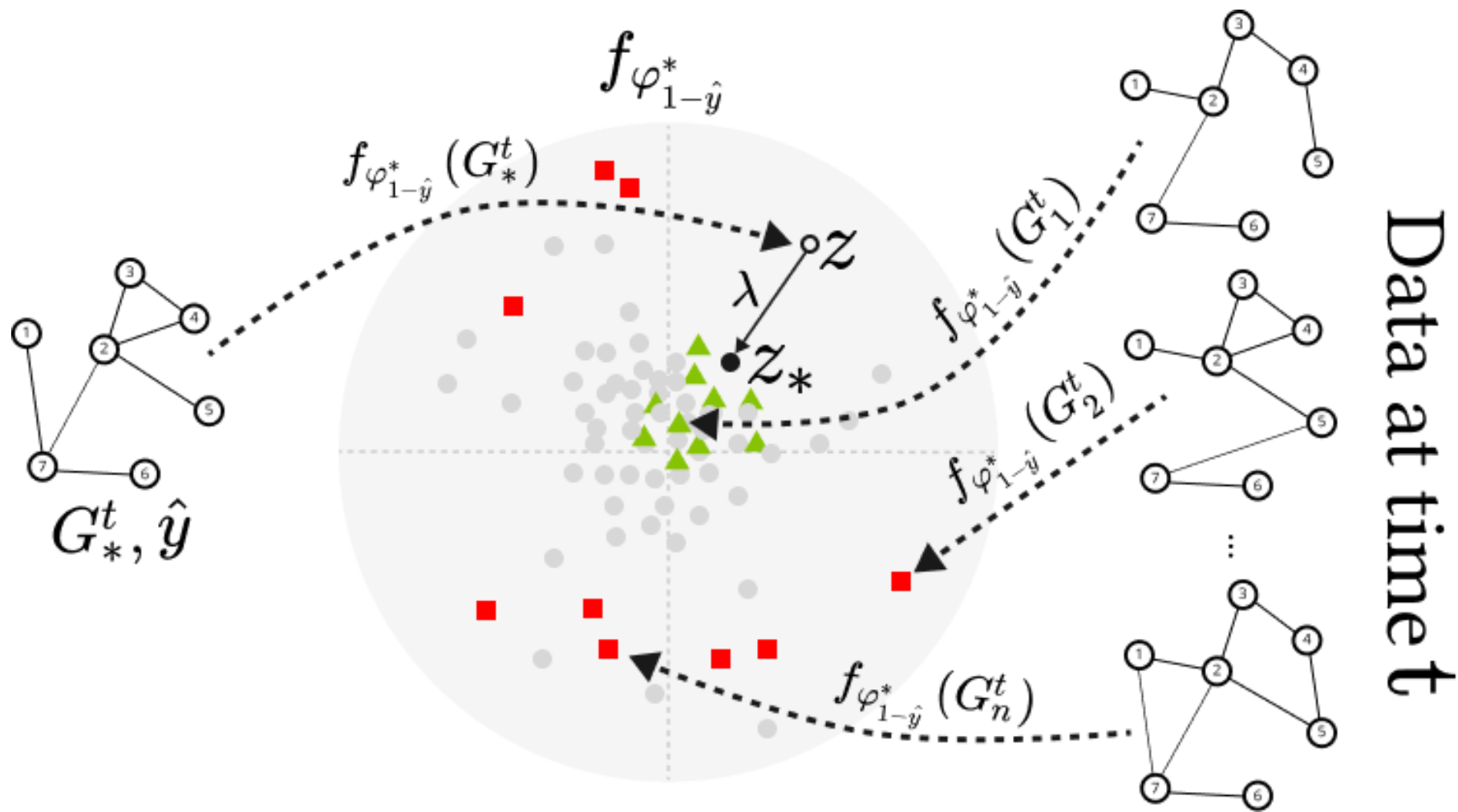


Training GRACIE

$$\begin{aligned} -\text{ELBO}_y(\theta_y, \varphi_y) &= \mathcal{L}_{rec} + \mathcal{L}_{dist} \\ &= \frac{1}{2} \left(\mathbb{E}_{q_{\varphi_y}(\mathbf{z}|G)} \left[\frac{\|g_{\theta_y}(\mathbf{z}) - G\|_2^2}{\sigma^2} \right] + \|f_{\varphi_y}(G)\|_2^2 \right) \end{aligned}$$



Inference & Update

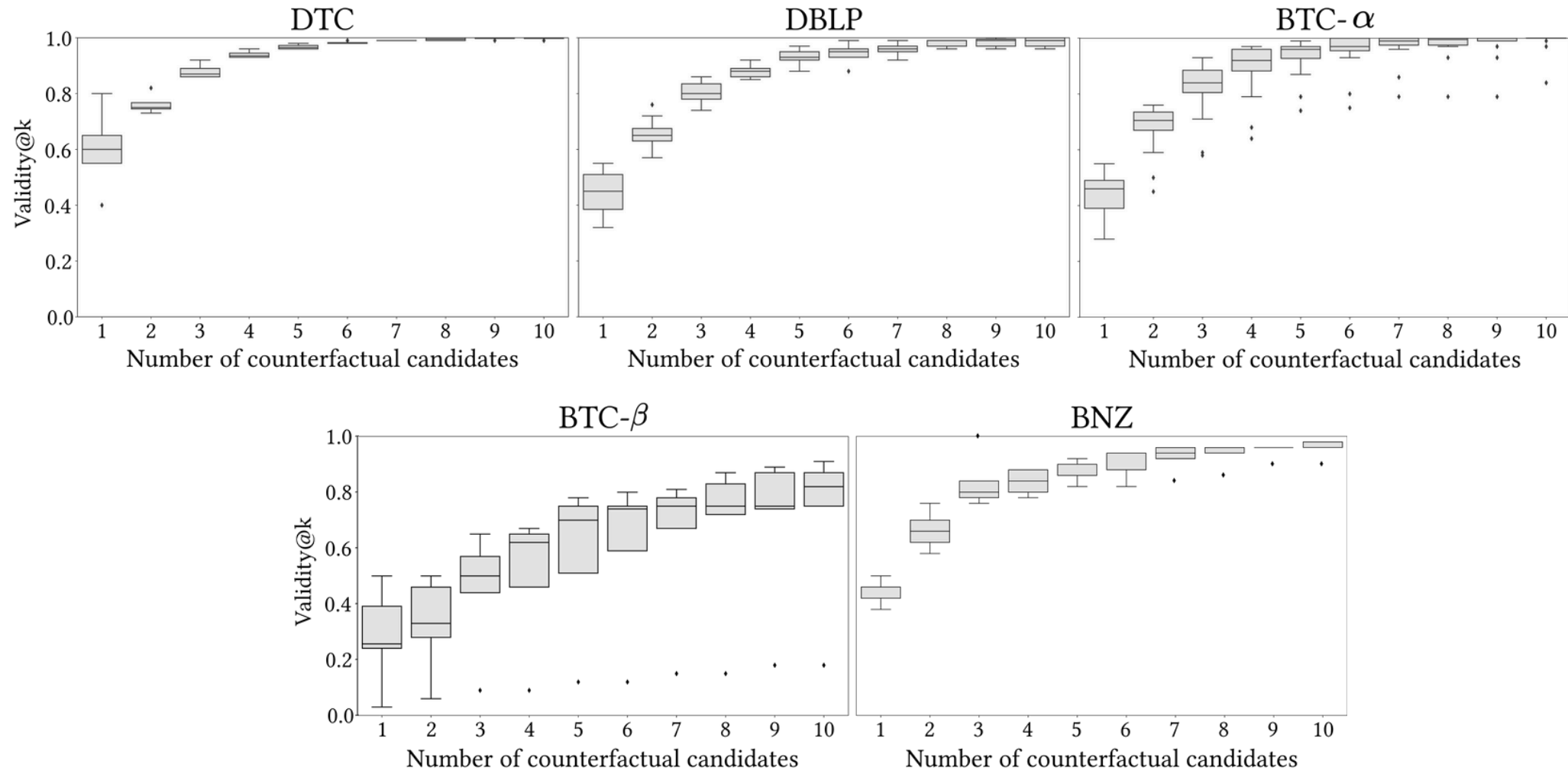


GRACIE is SoTA

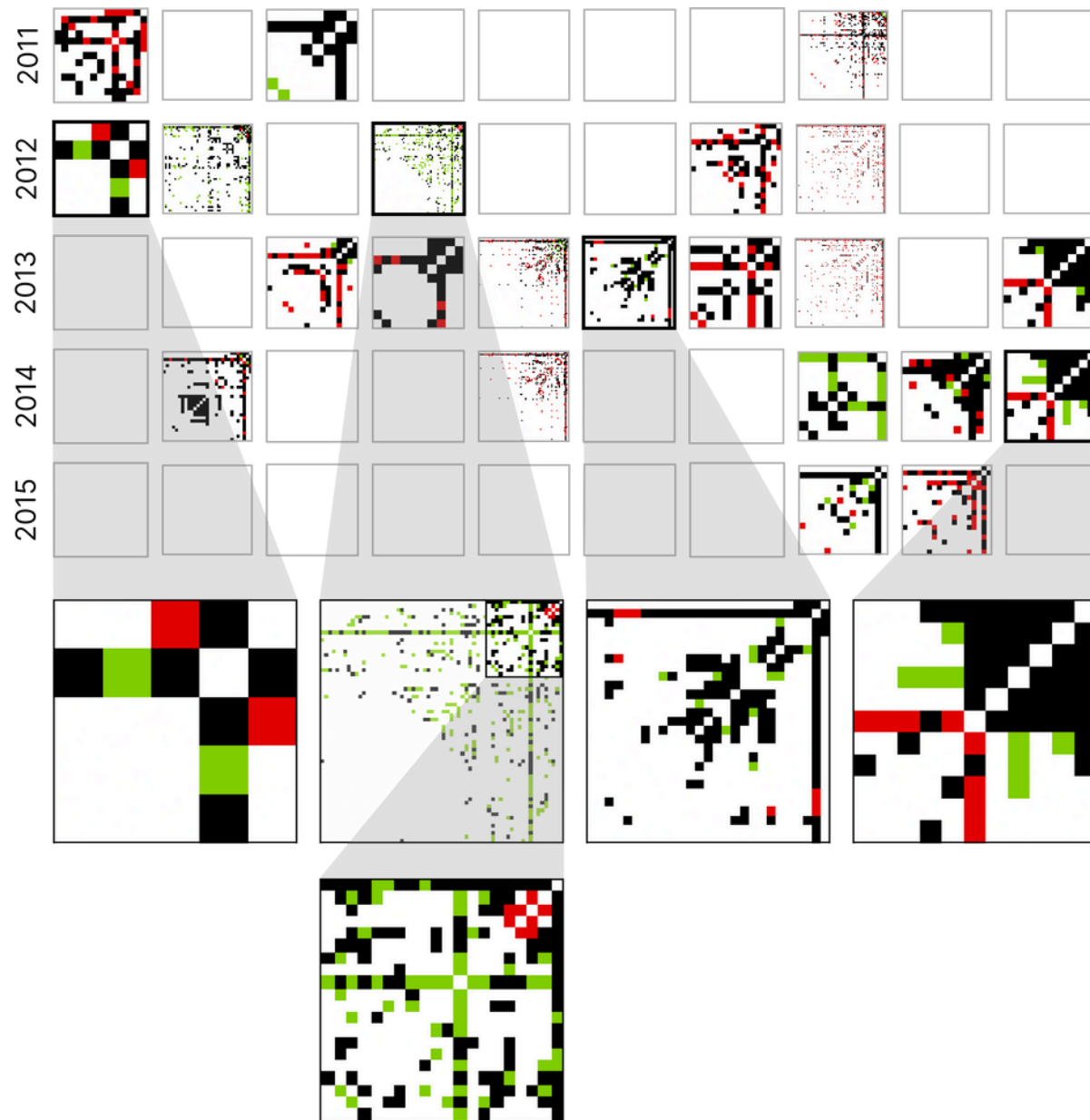
	DTC	DBLP	BTC- α	BTC- β	BNZ
BDDS	0.465	<u>0.381</u>	<u>0.360</u> [†]	0.235	0.136
MEG	0.250	0.209	×	<u>0.260</u>	0.120 [†]
CLEAR	0.458	0.024	0.214	0.125	0.000
G-CounteRGAN	0.507	0.256	0.236	×	<u>0.404</u>
DyGRACE	<u>0.525</u>	0.307	0.232	0.000 [†]	0.232
GRACIE	0.600	0.442	0.440	0.284	0.441

^aThe criterion of non-convergence is to fail to produce at least one counterfactual within 14 days of execution on an HPC SGE Cluster of 6 nodes with 360 cumulative cores, 1.2Tb of RAM, and two GPUs (i.e., one Nvidia A30 and one A100).

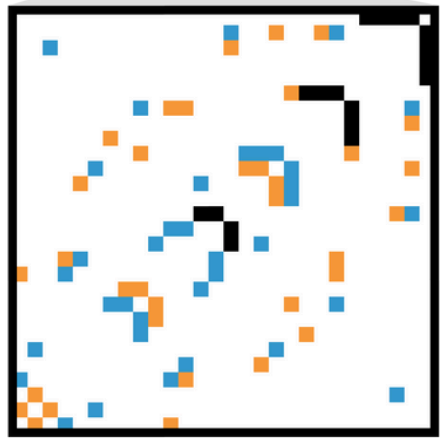
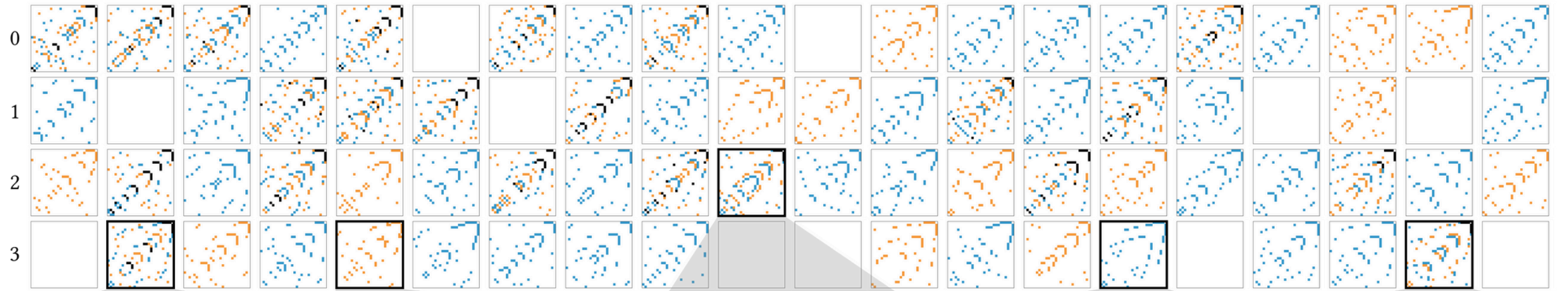
More sampling = more validity



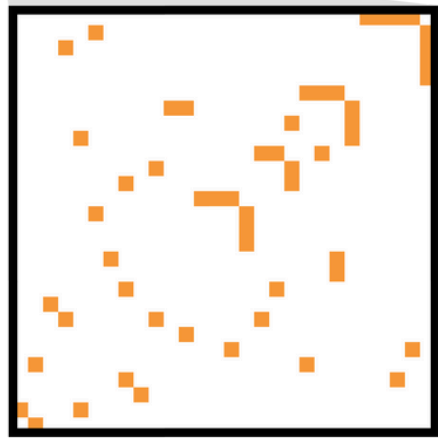
Qualitative on BTC- β



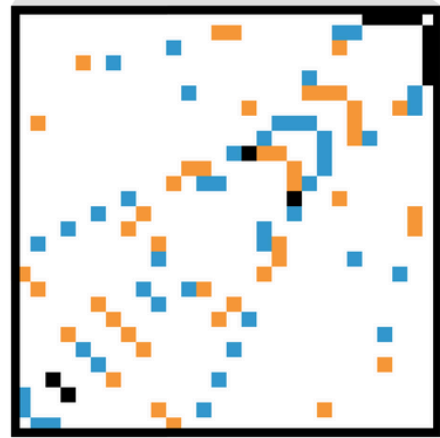
GRACIE vs BDDS



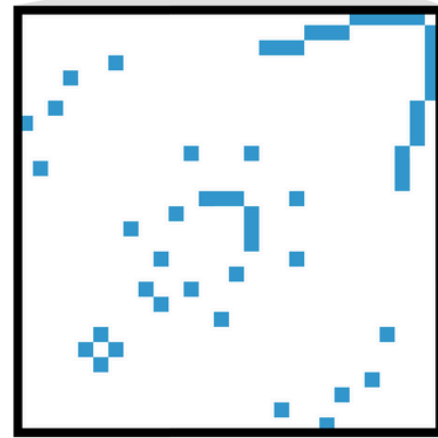
a)



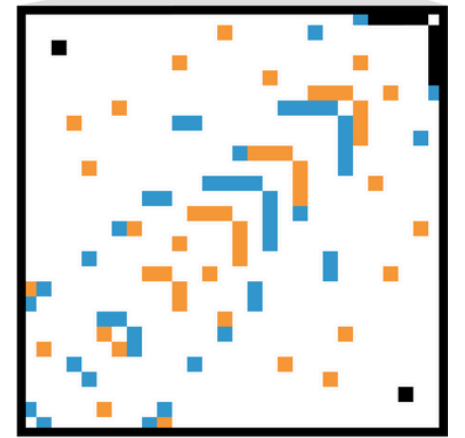
b)



c)



d)



e)

Thank you!

**Come to my poster
(#10 today at 13:00)**

