

# Learning at the Edge: Tailed-Uniform Sampling for Robust Simulation-Based Inference

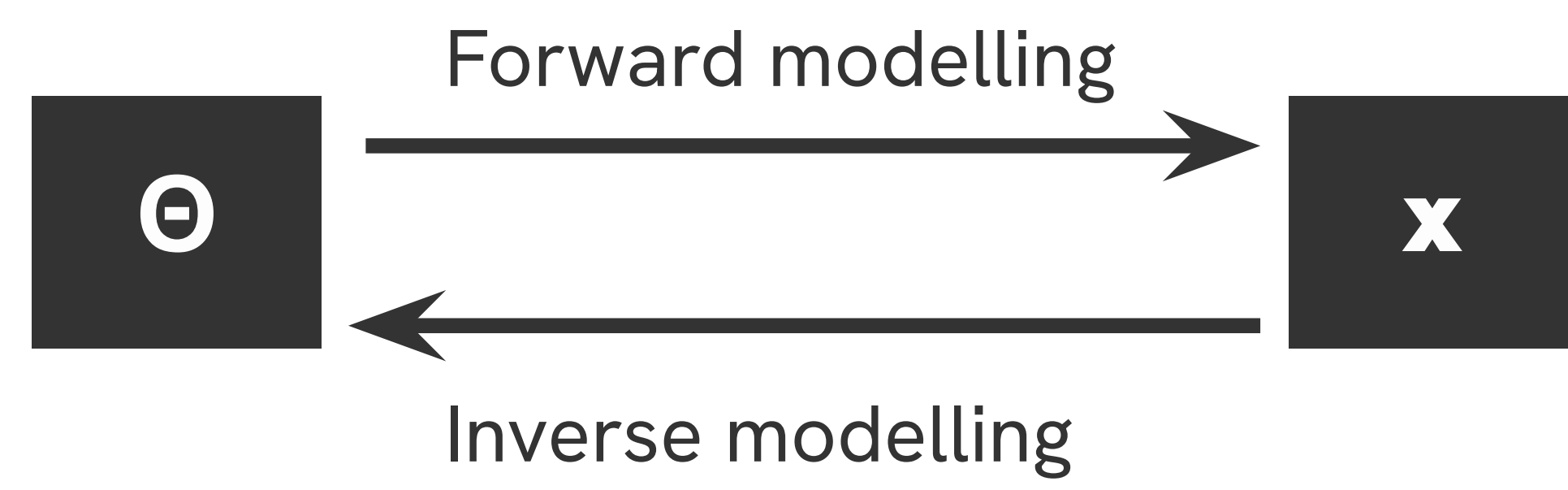
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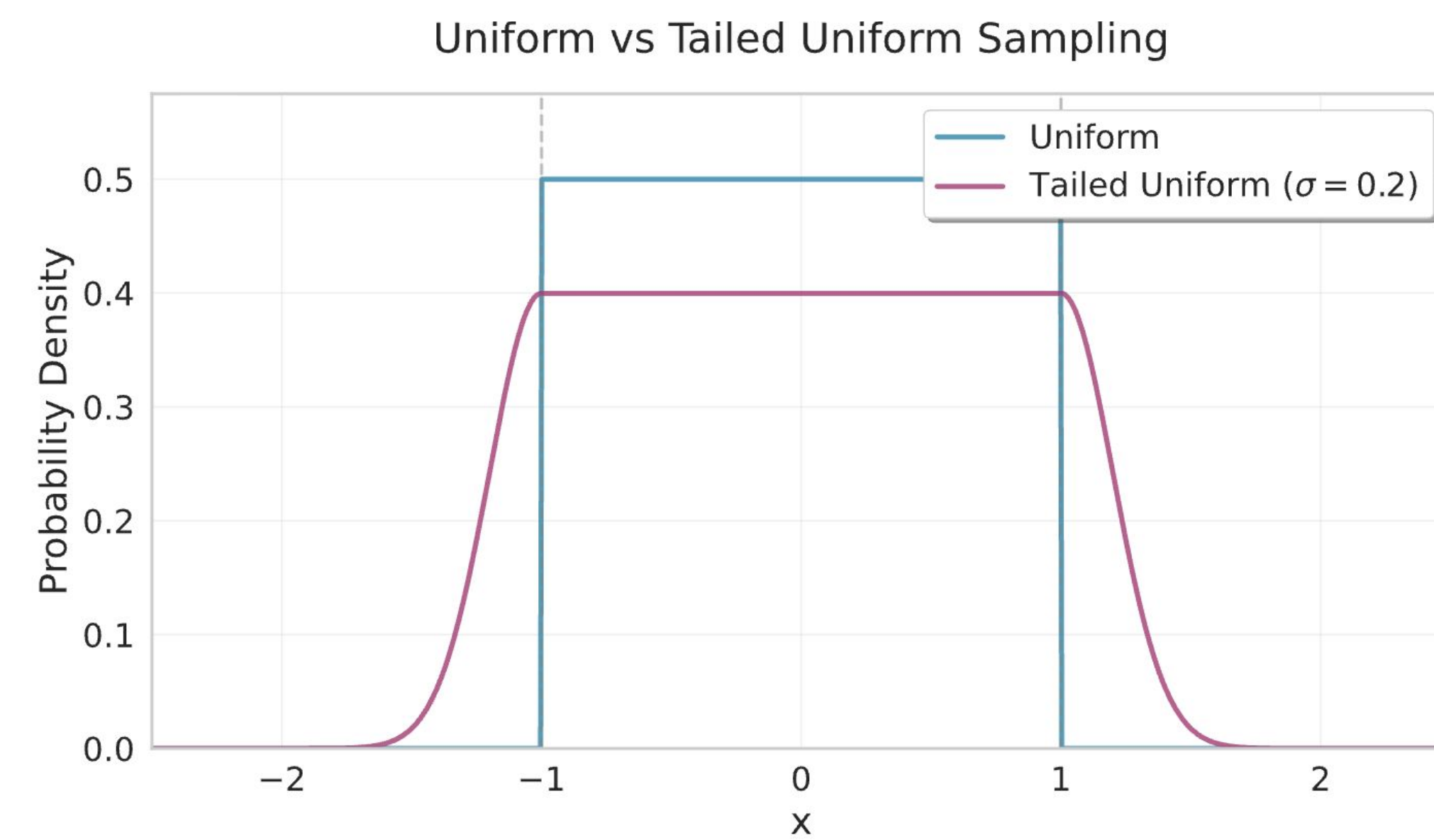
## Introduction



**Bayesian inference:**  $p(\Theta | \mathbf{x}_1, \dots, \mathbf{x}_n) \sim \prod_i p(\mathbf{x}_i | \Theta) p(\Theta)$

- Simulation-based inference (SBI) leverages neural networks to learn posterior distributions directly from simulation data.

- Simulators for large-scale public campaigns such as CAMELS choose to simulate via a Latin hypercube bounded within a specific region of interest.



**Problem:** Without gradient information at the boundaries, our flow-based density estimator cannot learn the correct density transitions near the cutoff.

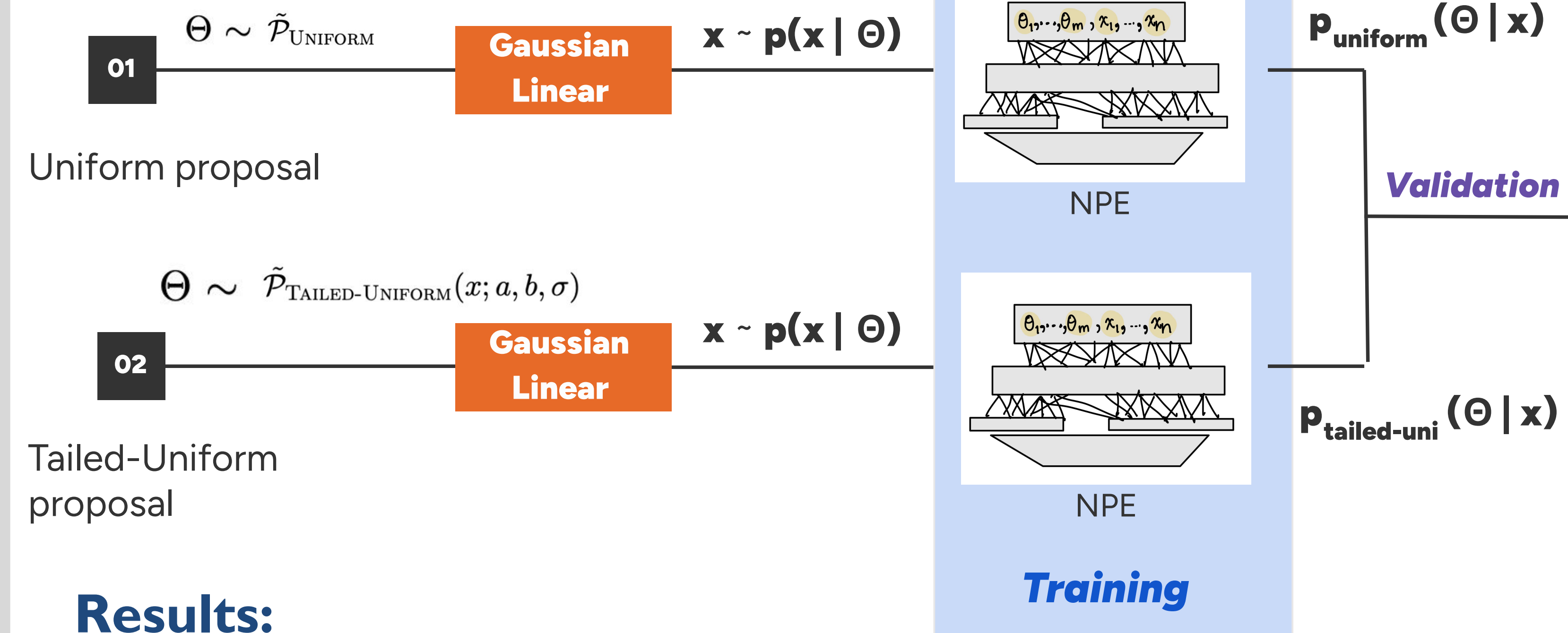
$$\tilde{P}_{\text{TAILED-UNIFORM}}(x; a, b, \sigma) = \begin{cases} A \cdot \mathcal{N}(a, \sigma^2), & x \leq a \\ B \cdot \mathcal{U}(a, b), & x \in [a, b] \\ A \cdot \mathcal{N}(b, \sigma^2), & x \geq b \end{cases}$$

By feeding the network training samples from outside the uniform region (the extended Gaussian support beyond the original grid)

- Tailed-Uniform captures the posterior density outside the assumed prior bounds, and
- Tailed-Uniform learns a smooth, well-calibrated posterior that extends into boundary-adjacent test points.

## Toy Problem

2D Gaussian simulator with an assumed Gaussian prior



### Results:

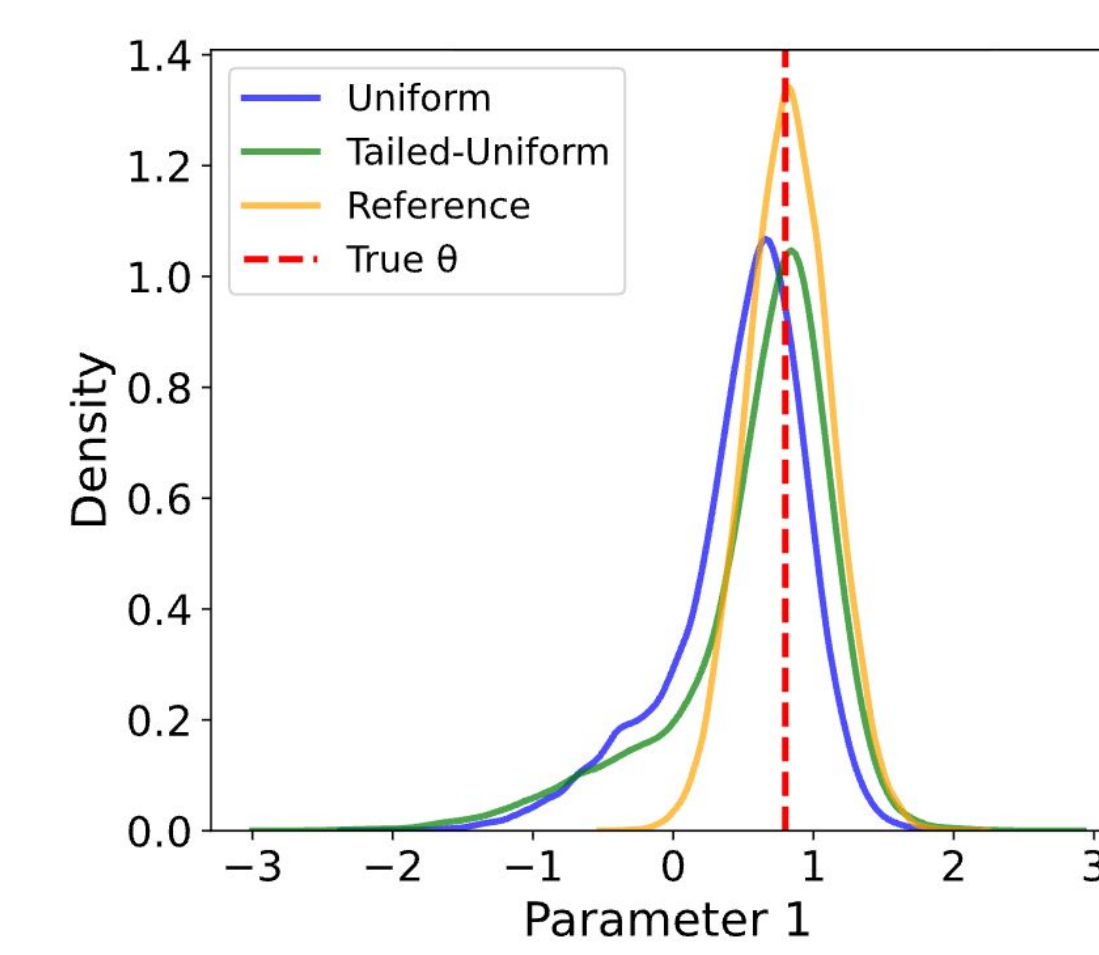
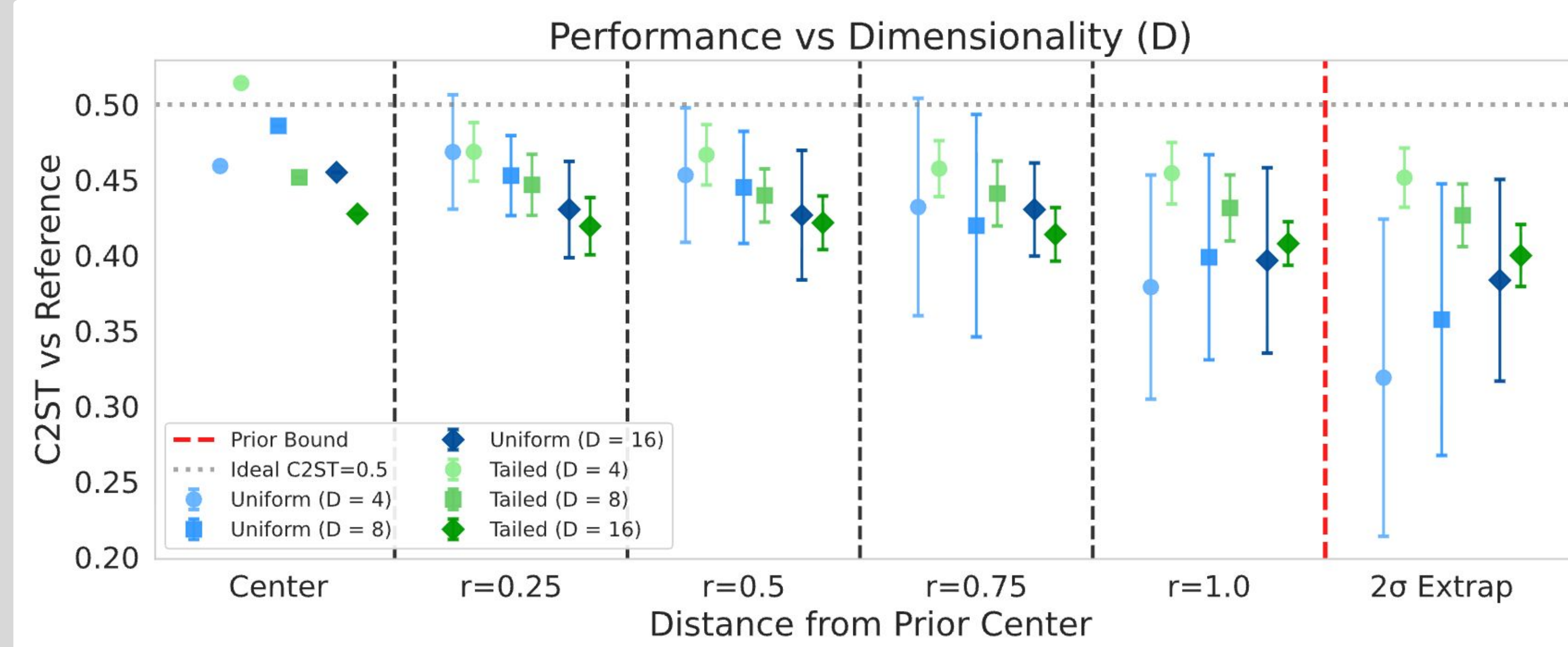
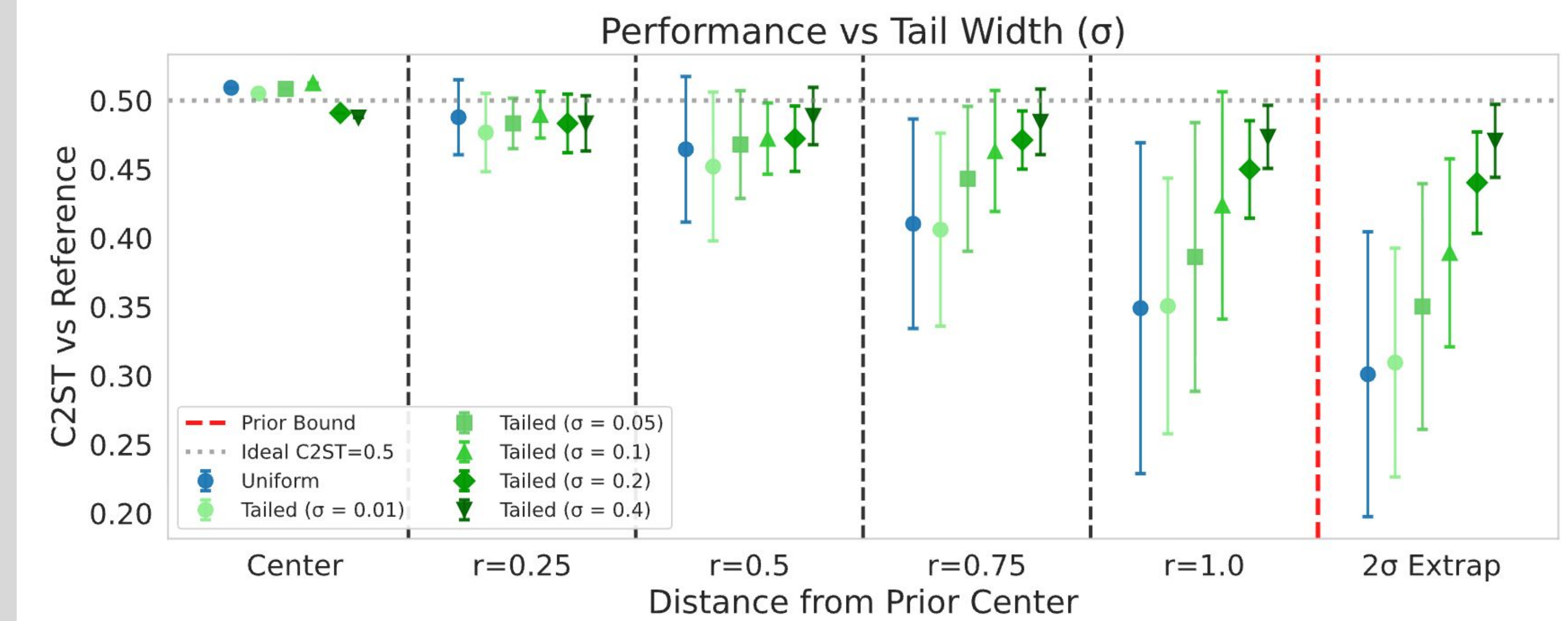
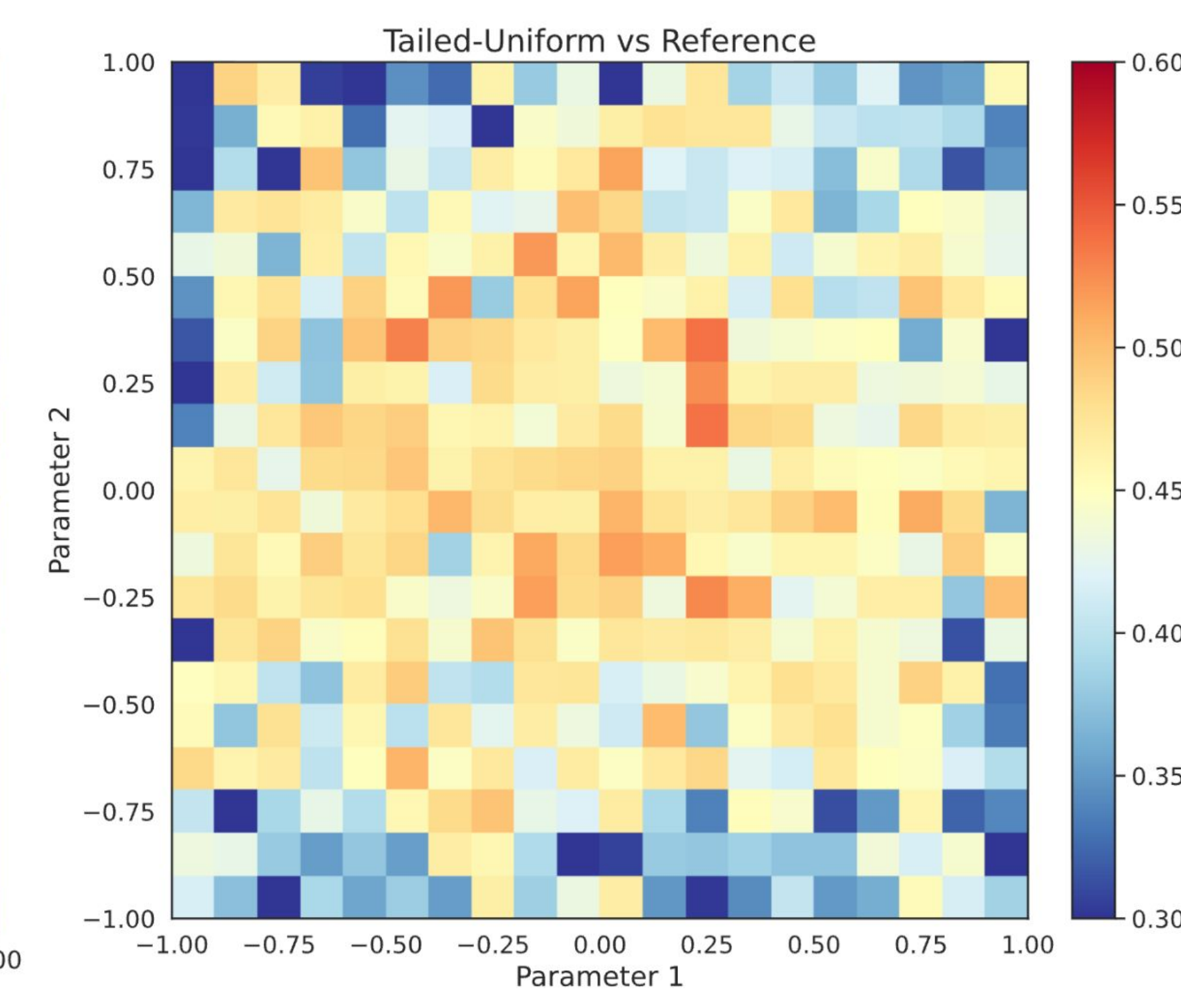
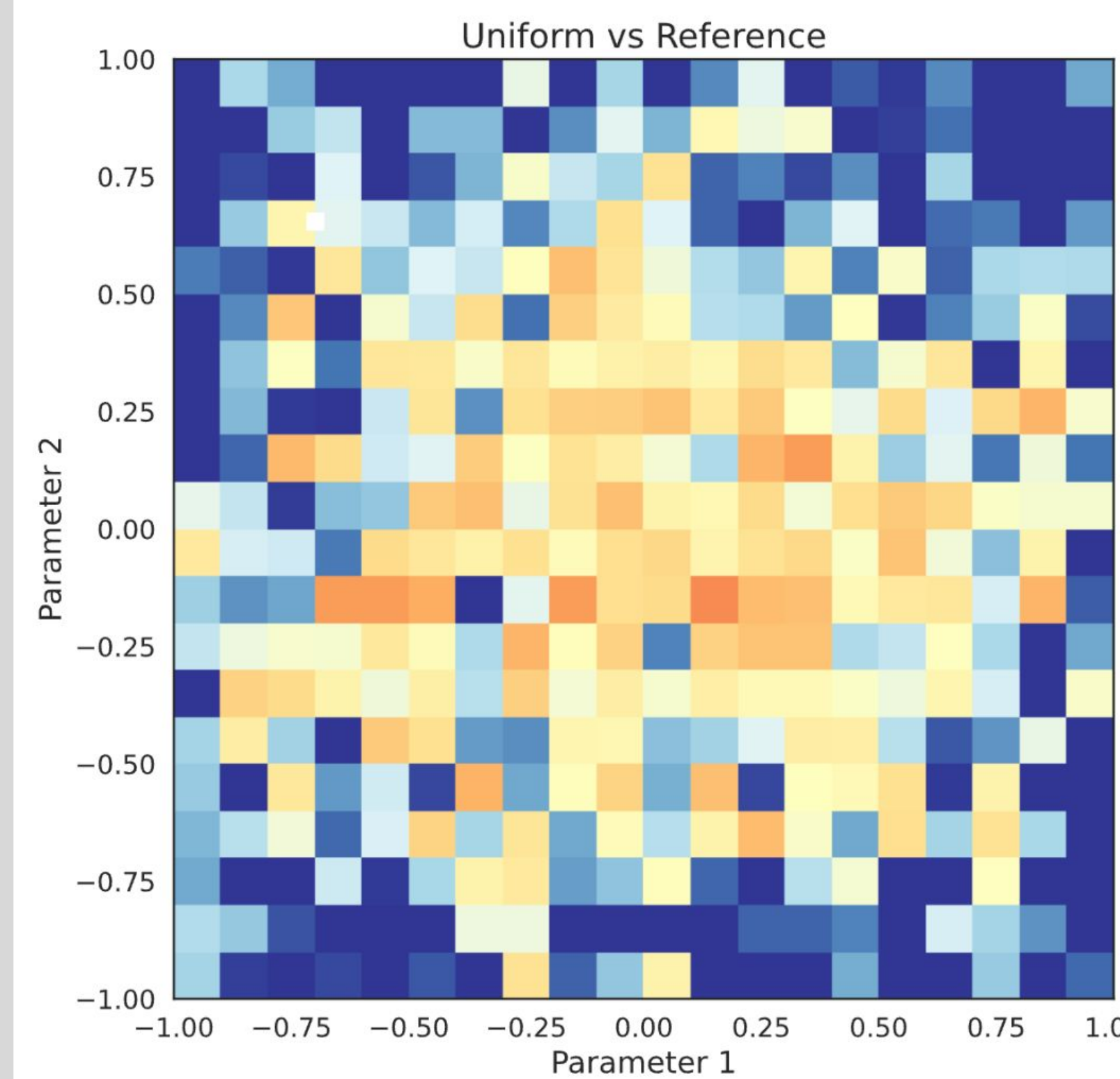
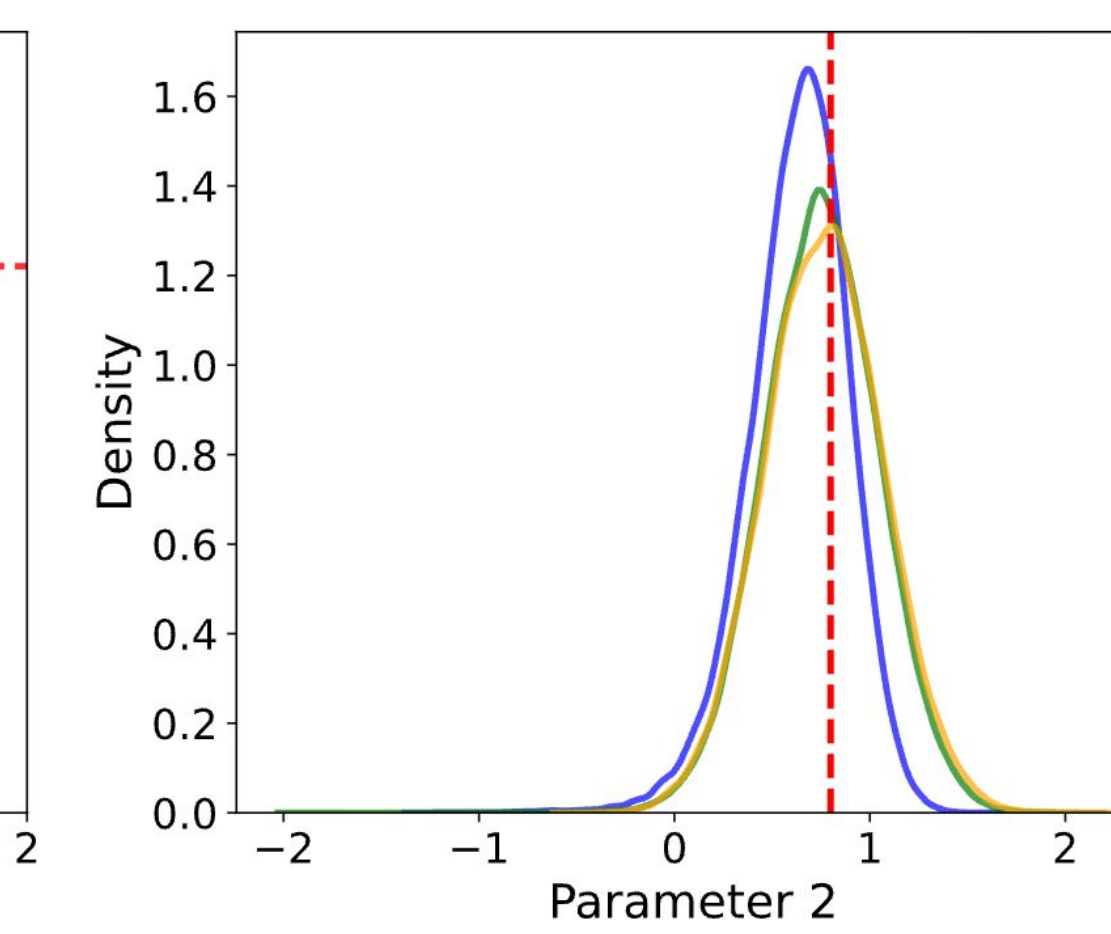
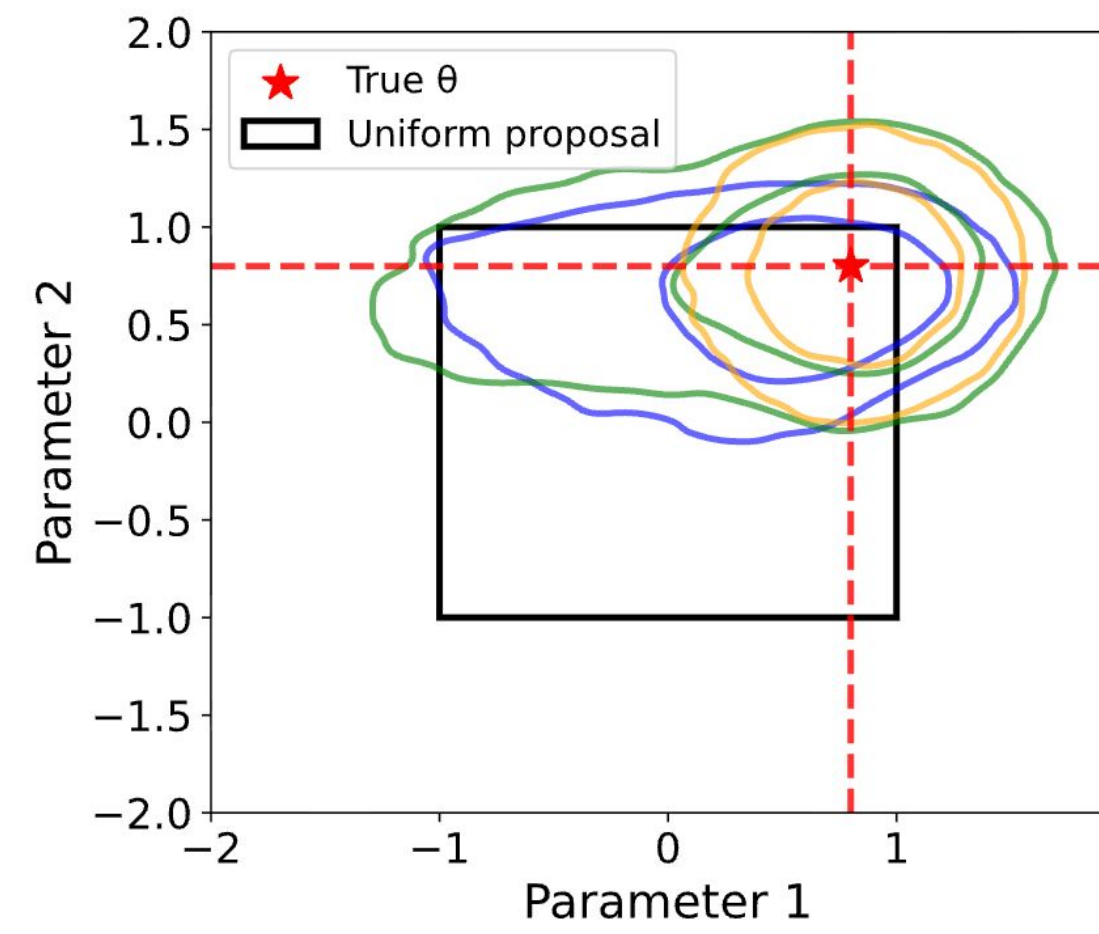


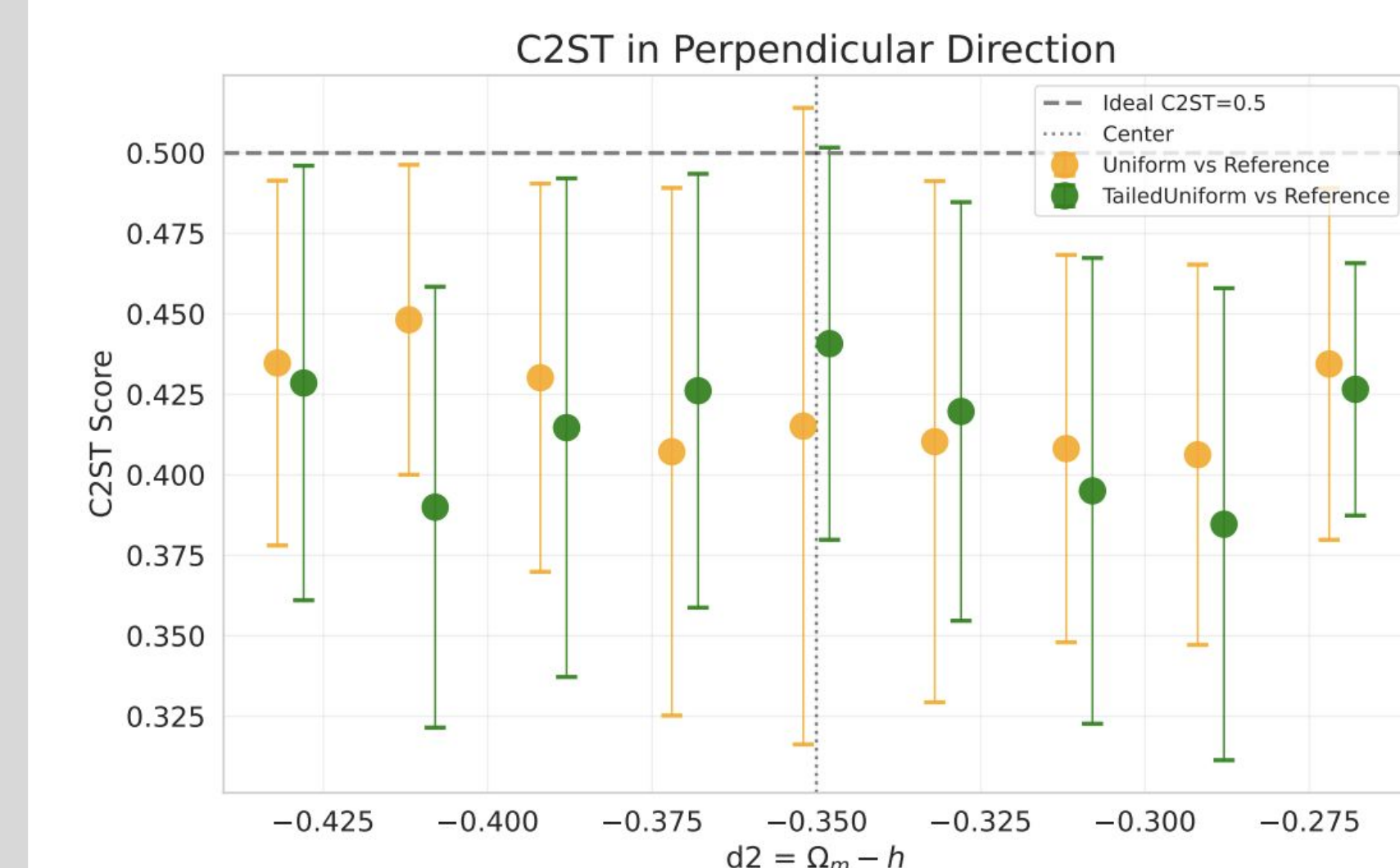
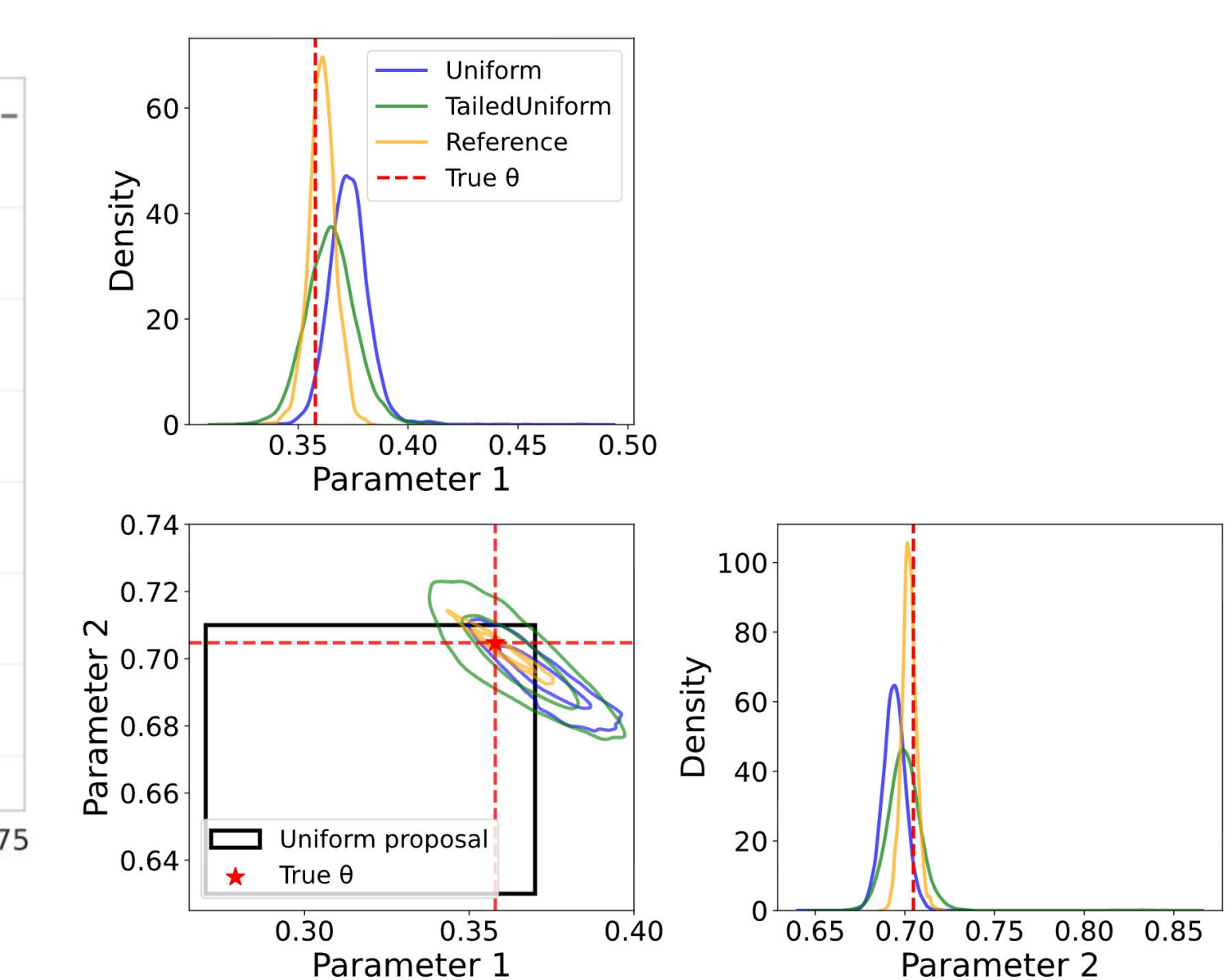
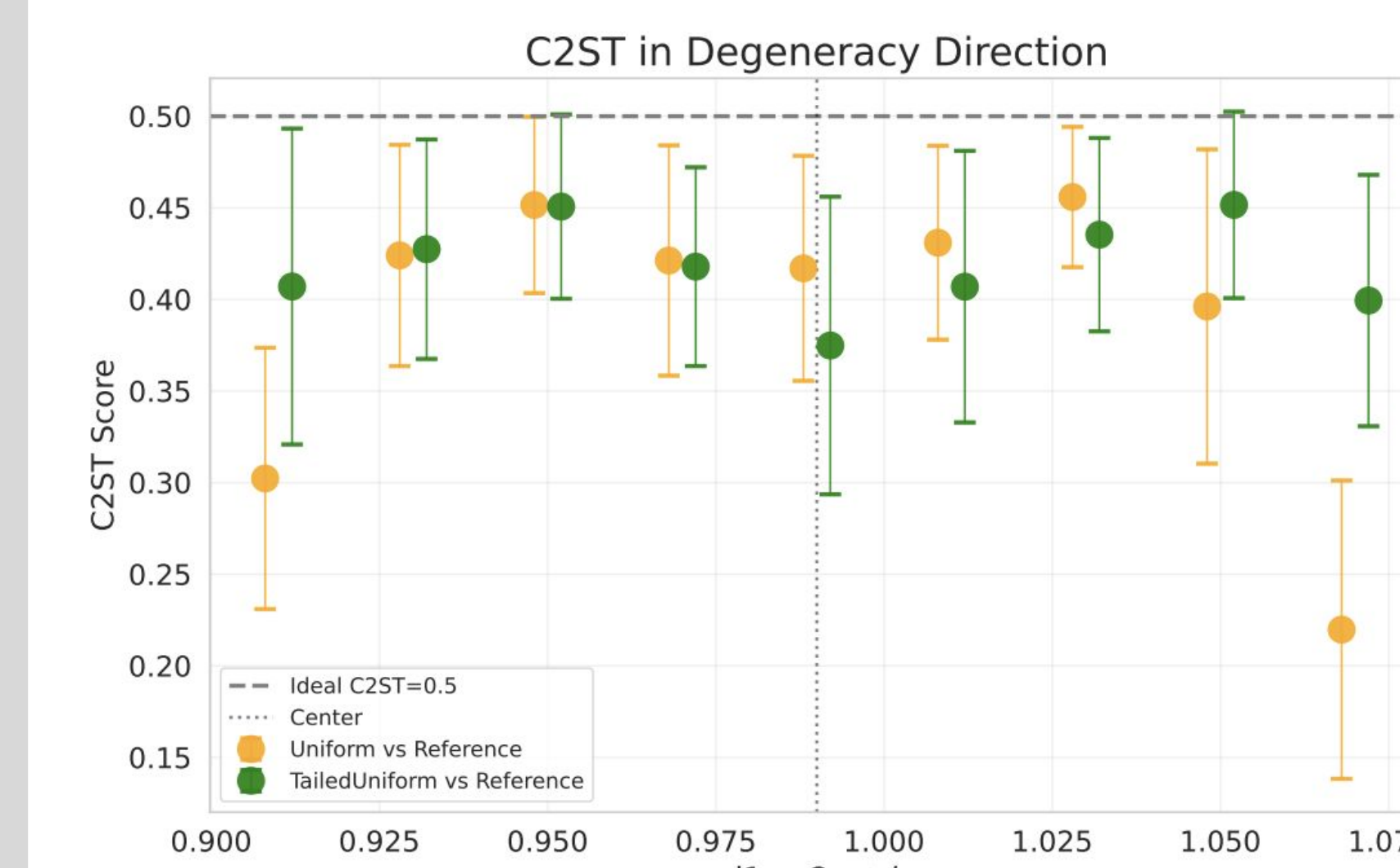
TABLE I. Posterior Performance for  $\theta_{\text{true}} = (0.8, 0.8)$ .

Method	Parameter Estimates		C2ST
	$\theta_1$	$\theta_2$	
Analytical Reference	$0.825 \pm 0.301$	$0.763 \pm 0.299$	0.500
TAILED-UNIFORM	$0.568 \pm 0.590$	$0.777 \pm 0.279$	<b>0.417</b>
UNIFORM	$0.469 \pm 0.519$	$0.644 \pm 0.286$	0.329
<b>C2ST Improvement</b>			<b>+27%</b>



## Science Experiments

**Goal:** Want to infer  $(\Omega_m, h)$  from observations generated using the syren-new emulator with heteroskedastic cosmic noise.



- Degeneracy direction ( $d1 = \Omega_m + h$ ): Uniform degrades at both ends; Tailed-Uniform is stable.
- Perpendicular direction ( $d2 = \Omega_m - h$ ): both methods perform about the same