

Supplemental Materials for Paper 340
 Neurally-Guided Procedural Models:
 Amortized Inference for Procedural Graphics
 Programs using Neural Networks

1 Derivation of Equation 1

$$\begin{aligned}
 & \min_{\theta} D_{\text{KL}}(P_{\text{CM}} || P_{\text{GM}}) \\
 &= \min_{\theta} \mathbb{E}_{P(\mathbf{c})} \left[\mathbb{E}_{P_{\text{CM}}(\mathbf{x}|\mathbf{c})} \left[\log \frac{P_{\text{CM}}(\mathbf{x}|\mathbf{c})}{P_{\text{GM}}(\mathbf{x}|\mathbf{c}; \theta)} \right] \right] \\
 &= \min_{\theta} \mathbb{E}_{P(\mathbf{c})} \left[\mathbb{E}_{P_{\text{CM}}(\mathbf{x}|\mathbf{c})} \left[\log P_{\text{CM}}(\mathbf{x}|\mathbf{c}) - \log P_{\text{GM}}(\mathbf{x}|\mathbf{c}; \theta) \right] \right] \\
 &= \max_{\theta} \mathbb{E}_{P(\mathbf{c})} \left[\mathbb{E}_{P_{\text{CM}}(\mathbf{x}|\mathbf{c})} \left[\log P_{\text{GM}}(\mathbf{x}|\mathbf{c}; \theta) - \log P_{\text{CM}}(\mathbf{x}|\mathbf{c}) \right] \right] \\
 &= \max_{\theta} \mathbb{E}_{P(\mathbf{c})} \left[\mathbb{E}_{P_{\text{CM}}(\mathbf{x}|\mathbf{c})} \left[\log P_{\text{GM}}(\mathbf{x}|\mathbf{c}; \theta) \right] \right] \\
 &\approx \max_{\theta} \frac{1}{N} \sum_{s=1}^N \log P_{\text{GM}}(\mathbf{x}_s | \mathbf{c}_s; \theta) \quad \mathbf{x}_s \sim P_{\text{CM}}(\mathbf{x}|\mathbf{c}), \mathbf{c}_s \sim P(\mathbf{c})
 \end{aligned}$$

In the second-to-last step, the $\log P_{\text{CM}}(\mathbf{x}|\mathbf{c})$ term is dropped because it does not depend on θ . In the last step, we approximate the expectations with an average over a finite set of samples.

2 Additional Results

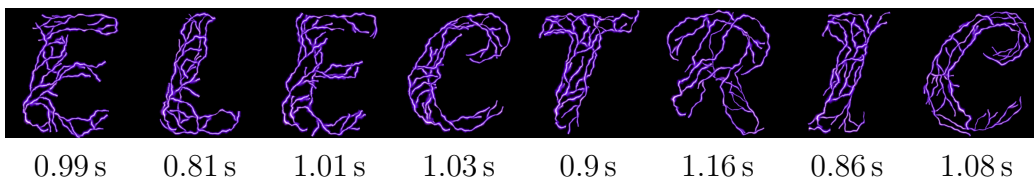


Figure 1: Targeting letter shapes with a neurally-guided procedural lightning program. Generated using SMC with 10 particles; compute time required is shown below each letter. Best viewed on a high-resolution display.

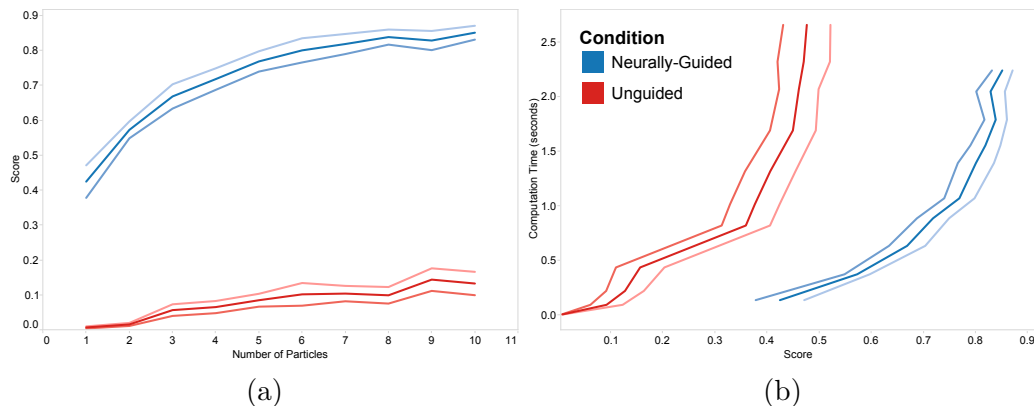


Figure 2: Performance comparison for the circuit design problem (section 4.3 in the main paper). “Score” is median normalized score (i.e. argument one to the Gaussian in Equation 4 of the main paper), averaged over 50 runs. The neurally-guided version achieves significantly higher average scores than the unguided version given the same number of particles or the same amount of compute time.




















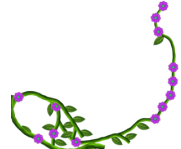





Target	Reference	Guided	Unguided (Equal N)	Unguided (Equal Time)
	 $N = 600$, 38.68 s	 $N = 5$, 0.86 s	 $N = 5$, 0.09 s	 $N = 30$, 0.83 s
	 $N = 600$, 33.5 s	 $N = 10$, 1.23 s	 $N = 10$, 0.14 s	 $N = 40$, 1.28 s
	 $N = 600$, 25.55 s	 $N = 15$, 1.75 s	 $N = 15$, 0.23 s	 $N = 50$, 1.73 s
	 $N = 600$, 20.76 s	 $N = 10$, 0.81 s	 $N = 10$, 0.15 s	 $N = 40$, 0.85 s
	 $N = 600$, 25.5 s	 $N = 10$, 1.04 s	 $N = 10$, 0.14 s	 $N = 40$, 1.05 s

Figure 3: Additional shape matching results (section 4.2 in the main paper).