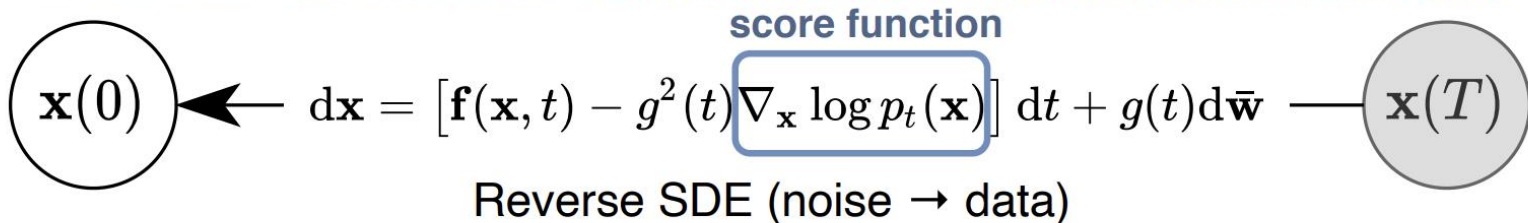
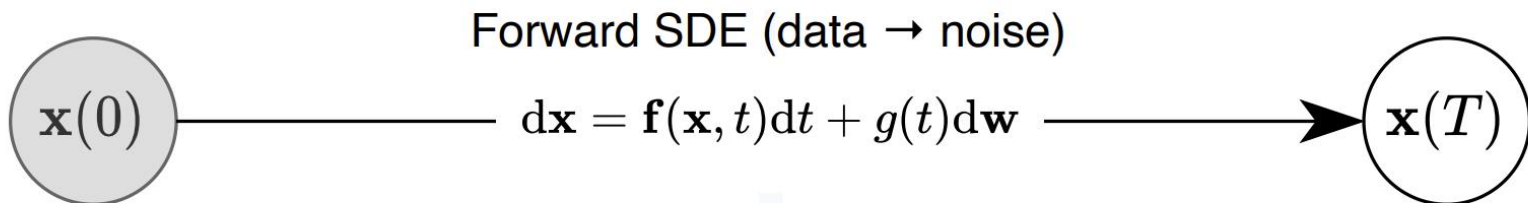


Smoothed Energy Guidance: Guiding Diffusion Models with Reduced Energy Curvature of Attention

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Introduction



Introduction

Classifier-Free Guidance [Ho and Salimans, 2022]

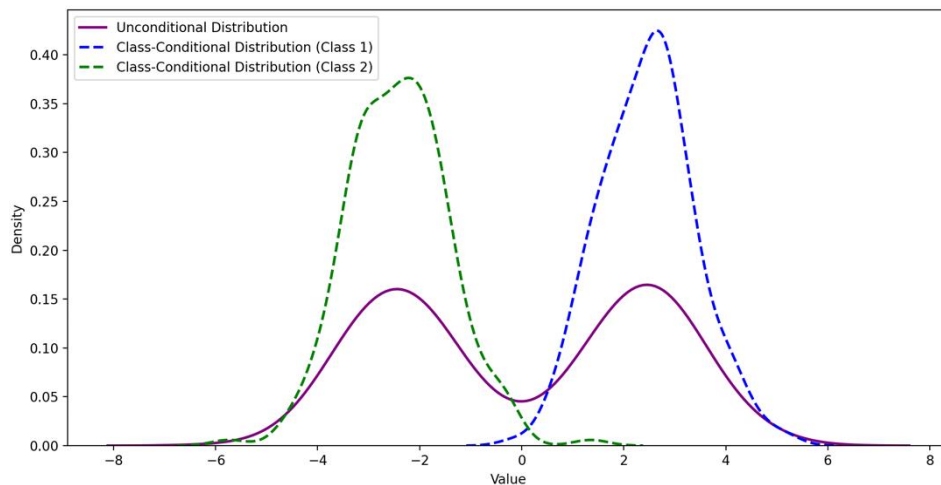
- Guidance without a classifier
- Reverse SDE

$$d\mathbf{x} = [\mathbf{f}(\mathbf{x}, t) - g(t)^2 (\gamma_{\text{cfg}} \mathbf{s}_\theta(\mathbf{x}, t, c) - (\gamma_{\text{cfg}} - 1) \mathbf{s}_\theta(\mathbf{x}, t))] dt + g(t) d\bar{\mathbf{w}}$$

Introduction

Classifier-Free Guidance [Ho and Salimans, 2022]

- Moves samples towards “**sharper**” class-conditional distributions



Main Idea

We propose using the energy of the attention weights, inspired by the modern (continuous) **Hopfield networks** (Definition 2.1)

$$E(\mathbf{A}) := \sum_{i=1}^H \sum_{j=1}^W E'(\mathbf{a}_{:(i,j)})$$

$$E'(\mathbf{a}) := -\text{lse}(\mathbf{a}) = -\log \left(\sum_{k=1}^H \sum_{l=1}^W e^{a_{(k,l)}} \right)$$

Main Idea

Blurring attention weights

- Preserves the mean and decreases the variance (Lemma 3.1)
- Increases the value of the energy function (Lemma 3.2)
- Results in decreasing the Gaussian curvature of the energy landscape (Theorem 3.1)

Main Idea

Blurring attention weights

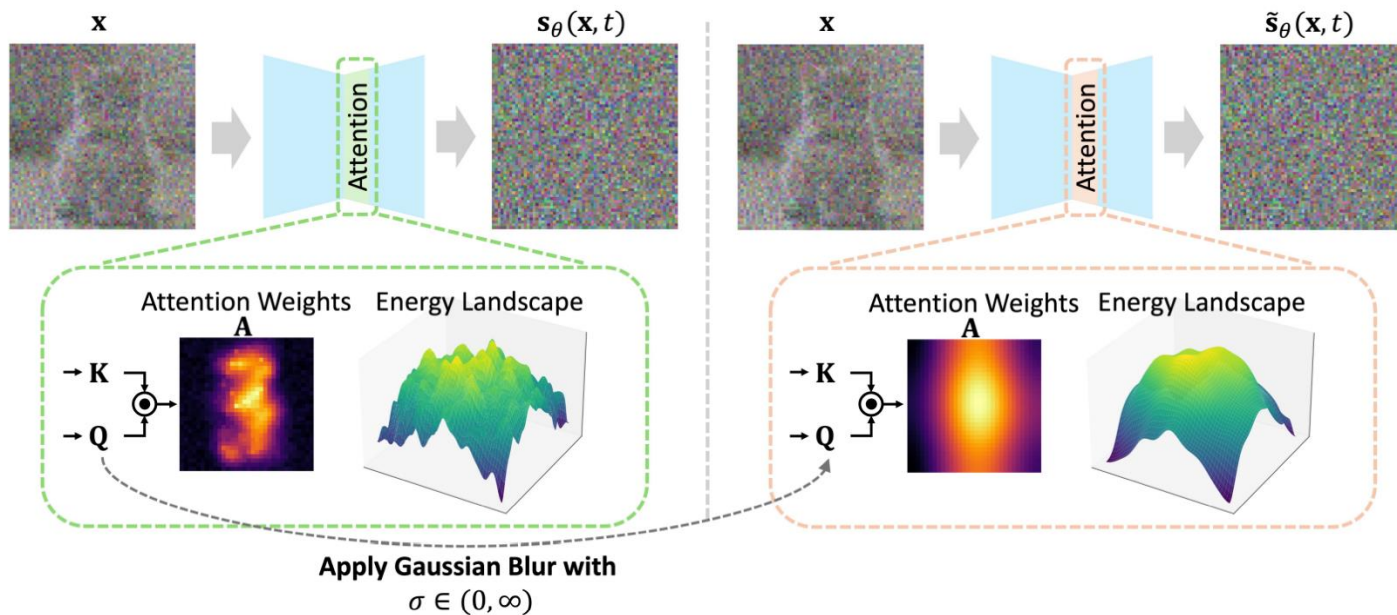
- Is equivalent to blurring queries (Proposition 3.1)

$$G * (\mathbf{Q}\mathbf{K}^\top) = \mathbf{B}(\mathbf{Q}\mathbf{K}^\top)$$

$$\mathbf{B}(\mathbf{Q}\mathbf{K}^\top) = (\mathbf{B}\mathbf{Q})\mathbf{K}^\top = (G * \mathbf{Q})\mathbf{K}^\top$$

Main Idea

Smoothed Energy Guidance



Main Idea

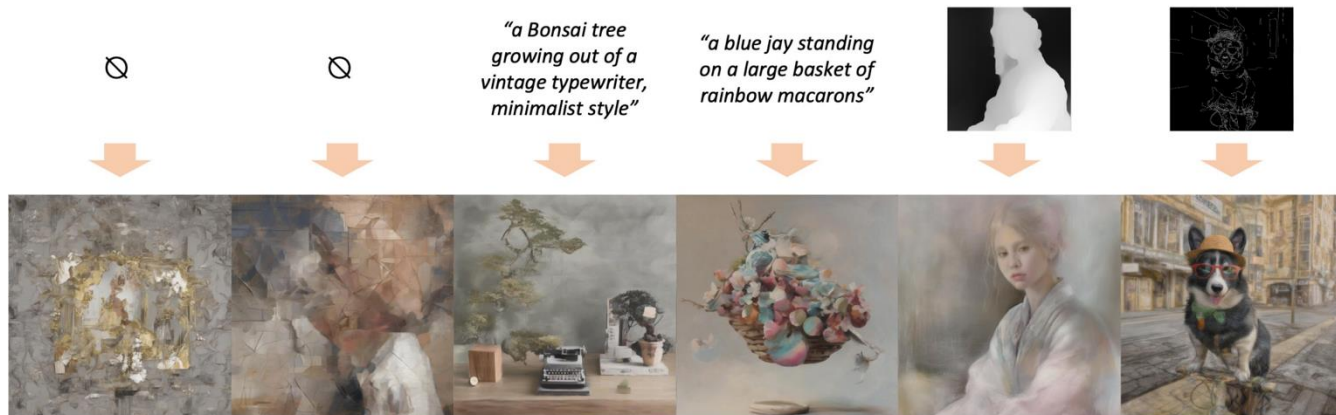
Smoothed Energy Guidance

- We propose training- and condition-free guidance

$$d\mathbf{x} = [\mathbf{f}(\mathbf{x}, t) - g(t)^2(\gamma_{\text{seg}}\mathbf{s}_\theta(\mathbf{x}, t) - (\gamma_{\text{seg}} - 1)\tilde{\mathbf{s}}_\theta(\mathbf{x}, t))]dt + g(t)d\bar{\mathbf{w}}$$

- We can control the effect with $\sigma \in (0, \infty)$ of the Gaussian blur

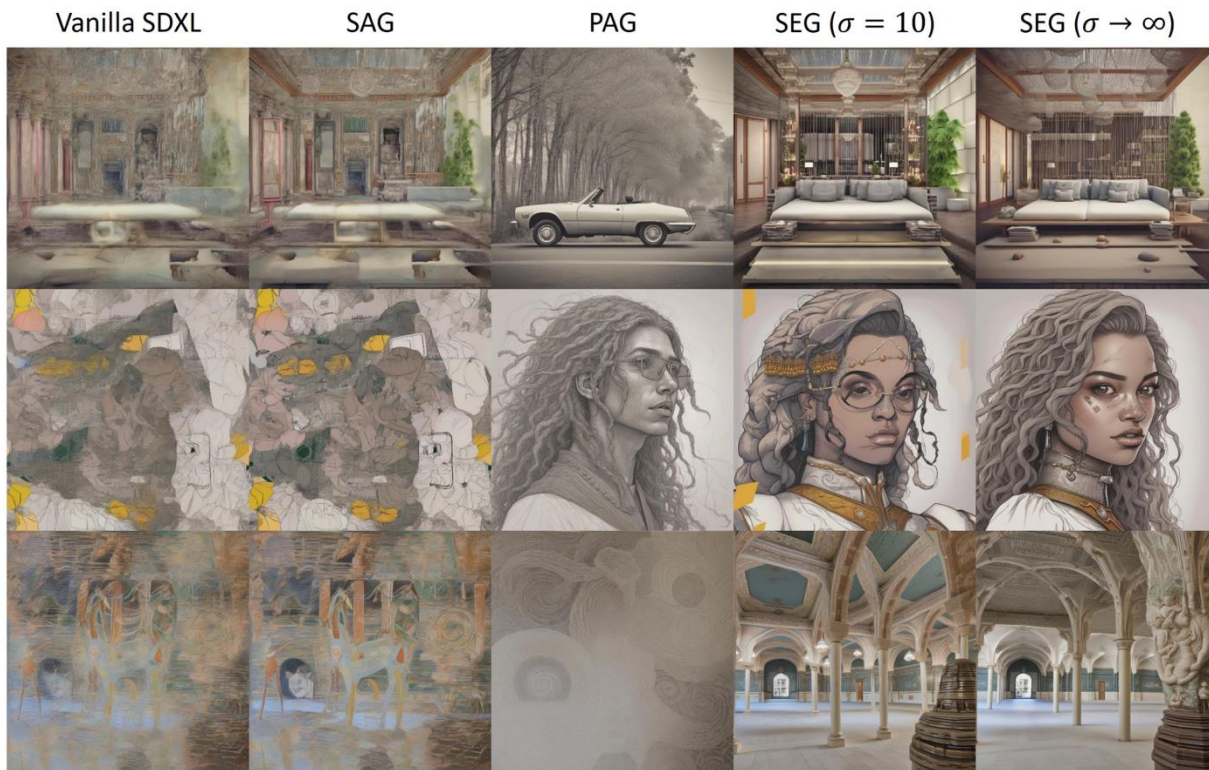
Experiments



Experiments



Experiments



Hong, Susung. "Smoothed Energy Guidance: Guiding Diffusion Models with Reduced Energy Curvature of Attention." NeurIPS 2024.

Ahn, Donghoon, et al. "Self-Rectifying Diffusion Sampling with Perturbed-Attention Guidance." ECCV 2024.

Hong, Susung, et al. "Improving sample quality of diffusion models using self-attention guidance." ICCV 2023.

Experiments



Experiments

| Metric | Vanilla SDXL [35] | SAG [17] | PAG [1] | SEG $\sigma = 10$ | SEG $\sigma \rightarrow \infty$ |
|-------------------------|-------------------|----------|---------|----------------------|------------------------------------|
| FID↓ | 129.496 | 106.683 | 105.271 | <u>95.316</u> | 88.215 |
| LPIPS _{vgg} ↓ | - | 0.706 | 0.542 | 0.522 | <u>0.536</u> |
| LPIPS _{alex} ↓ | - | 0.644 | 0.472 | 0.454 | <u>0.472</u> |

| Metric | Vanilla SDXL [35] | SEG | | | | |
|-------------------------|-------------------|--------------|--------------|--------|---------------|---------------|
| | | 1 | 2 | 5 | 10 | ∞ |
| FID↓ | 53.423 | 48.284 | 41.784 | 33.819 | <u>29.325</u> | 26.169 |
| CLIP Score↑ | 0.271 | 0.273 | 0.278 | 0.285 | <u>0.290</u> | 0.292 |
| LPIPS _{vgg} ↓ | - | 0.361 | <u>0.410</u> | 0.449 | 0.472 | 0.493 |
| LPIPS _{alex} ↓ | - | 0.295 | <u>0.347</u> | 0.390 | 0.416 | 0.440 |

