

Generative Al with Diffusion Models

Part 4: Classifier-Free Diffusion Guidance



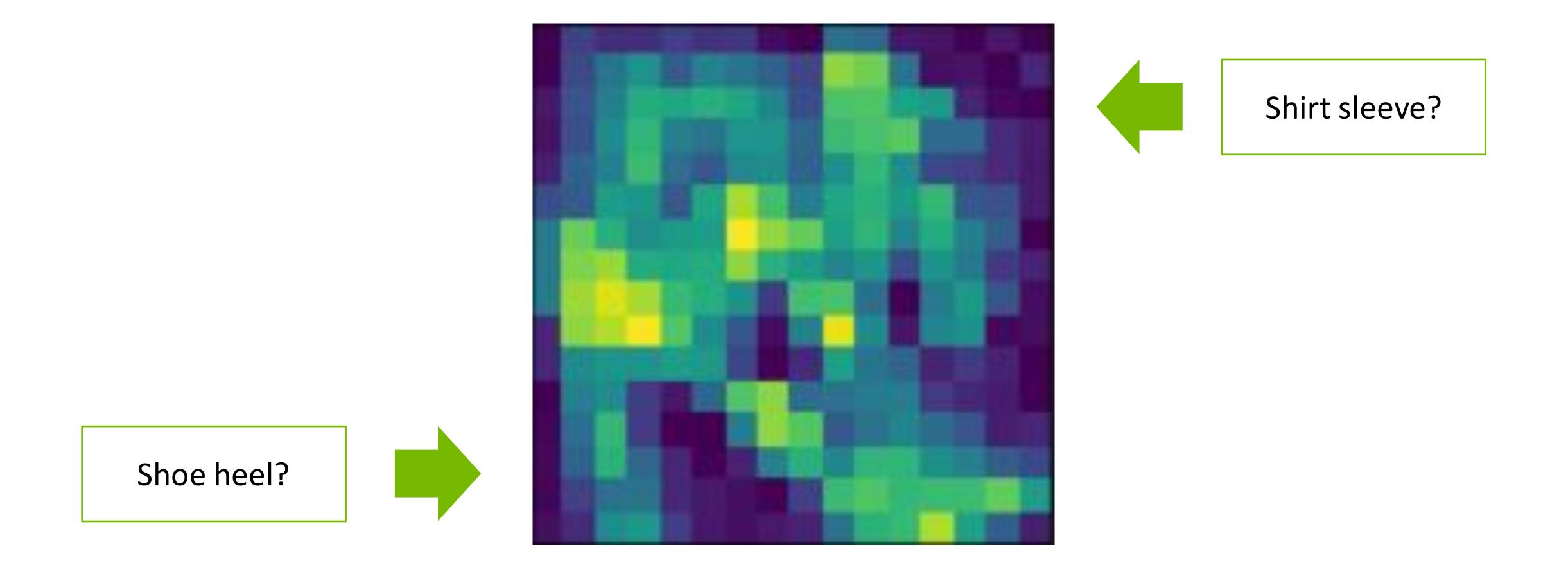
Agenda

- Part 1: From U-Nets to Diffusion
- Part 2: Denoising Diffusion Probabilistic Models
- Part 3: Optimizations
- Part 4: Classifier-Free Diffusion Guidance
- Part 5: CLIP
- Part 6: Wrap-up & Assessment



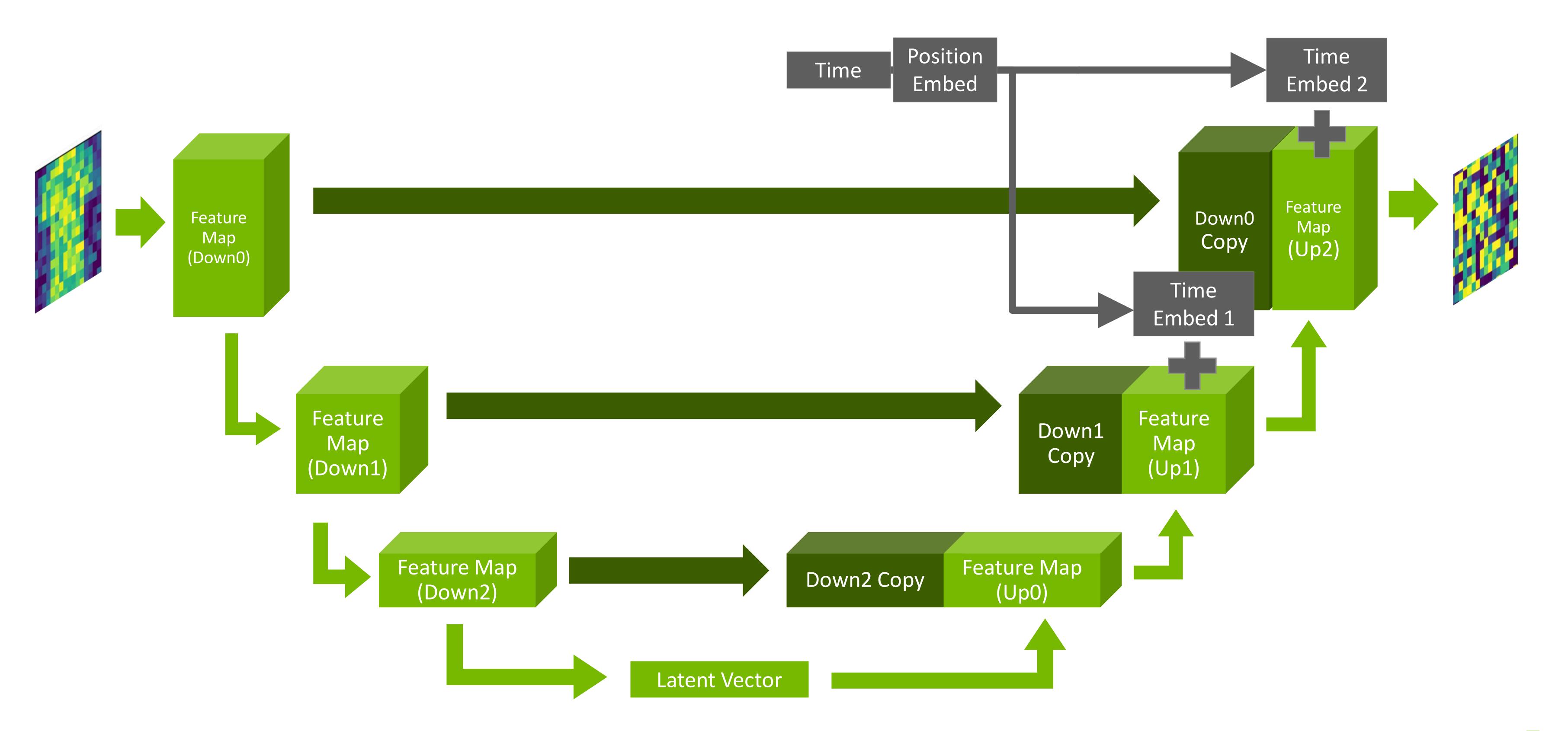


Adding Context



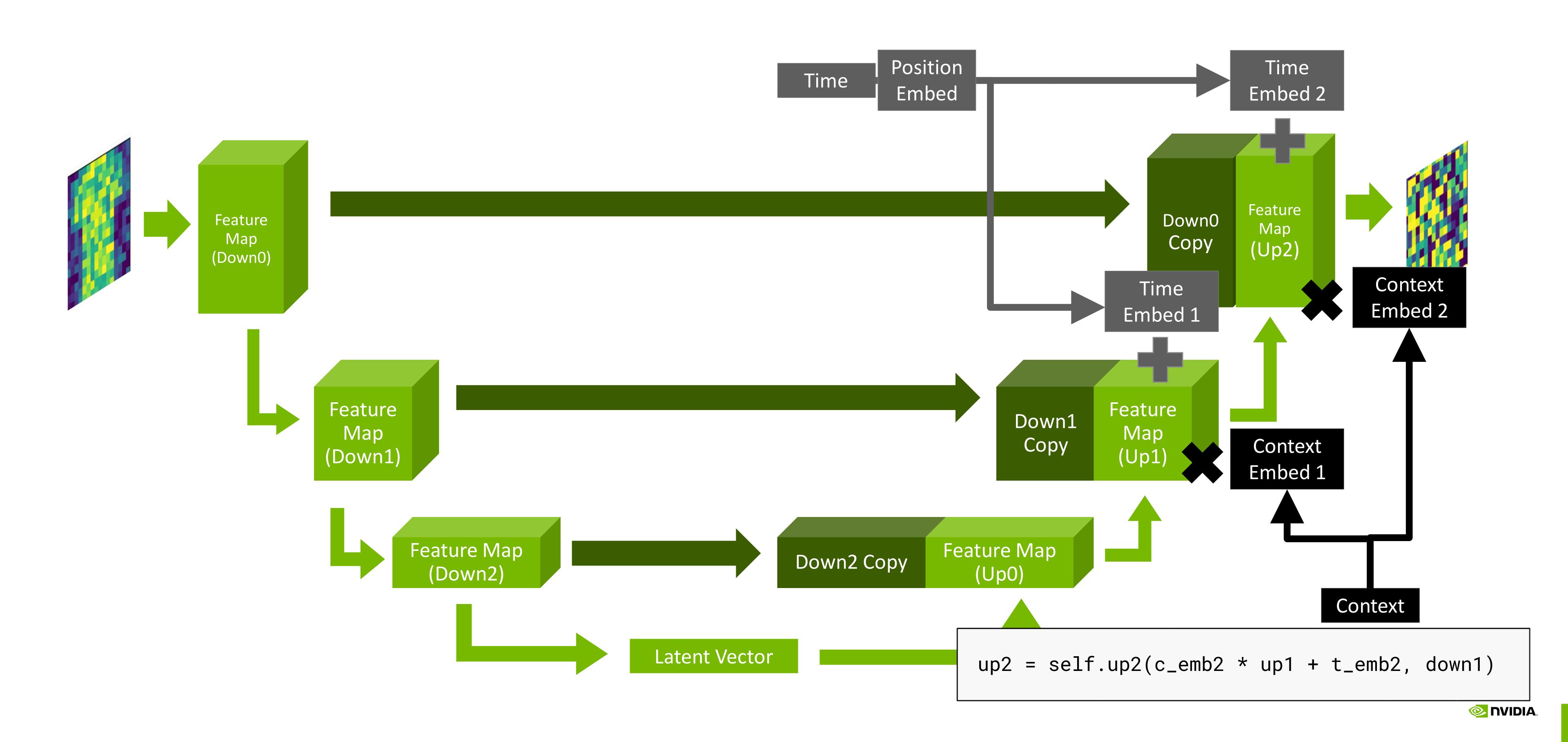


Adding Context



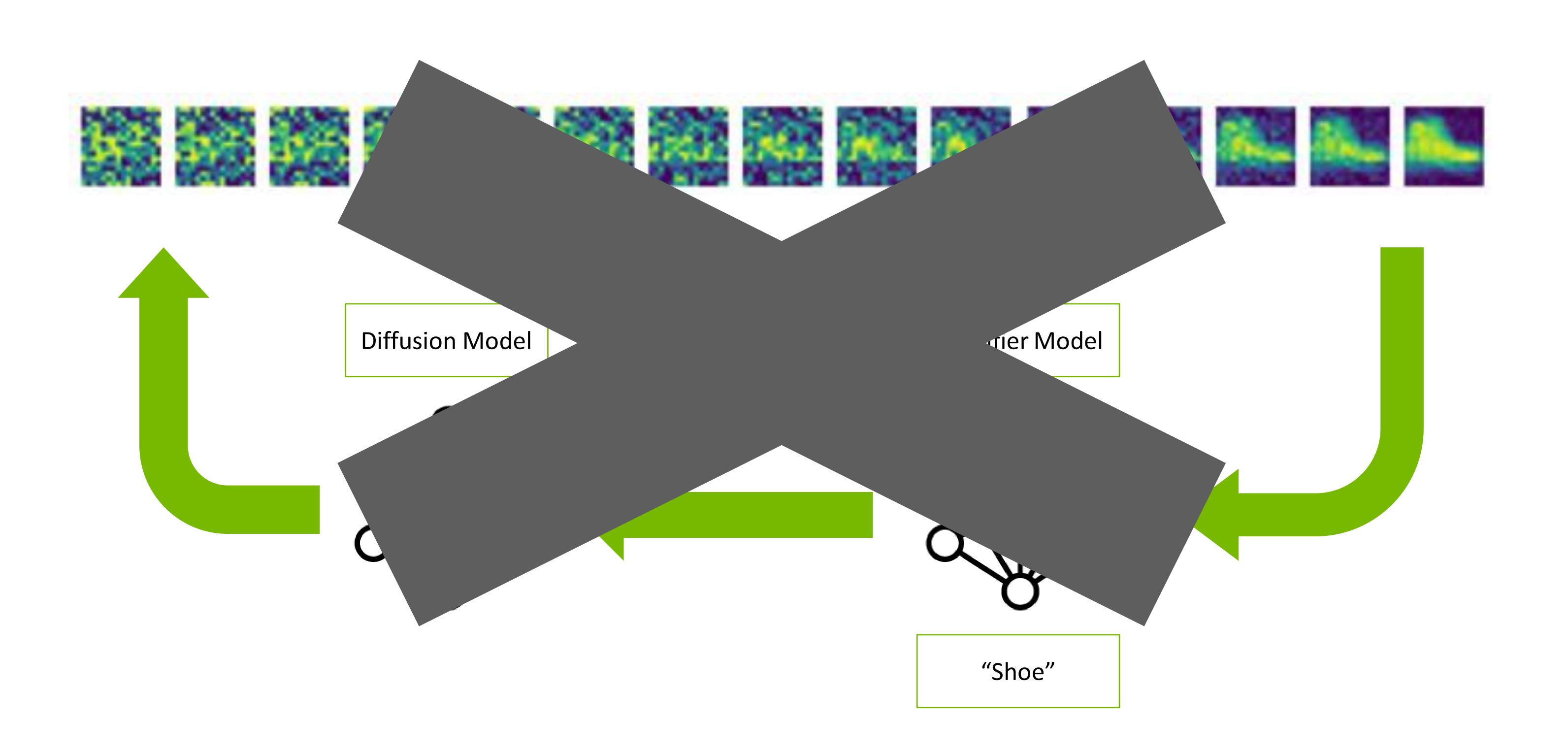


Adding Context





Make a Classifier Model?





CLASSIFIER-FREE DIFFUSION GUIDANCE

Jonathan Ho & Tim Salimans

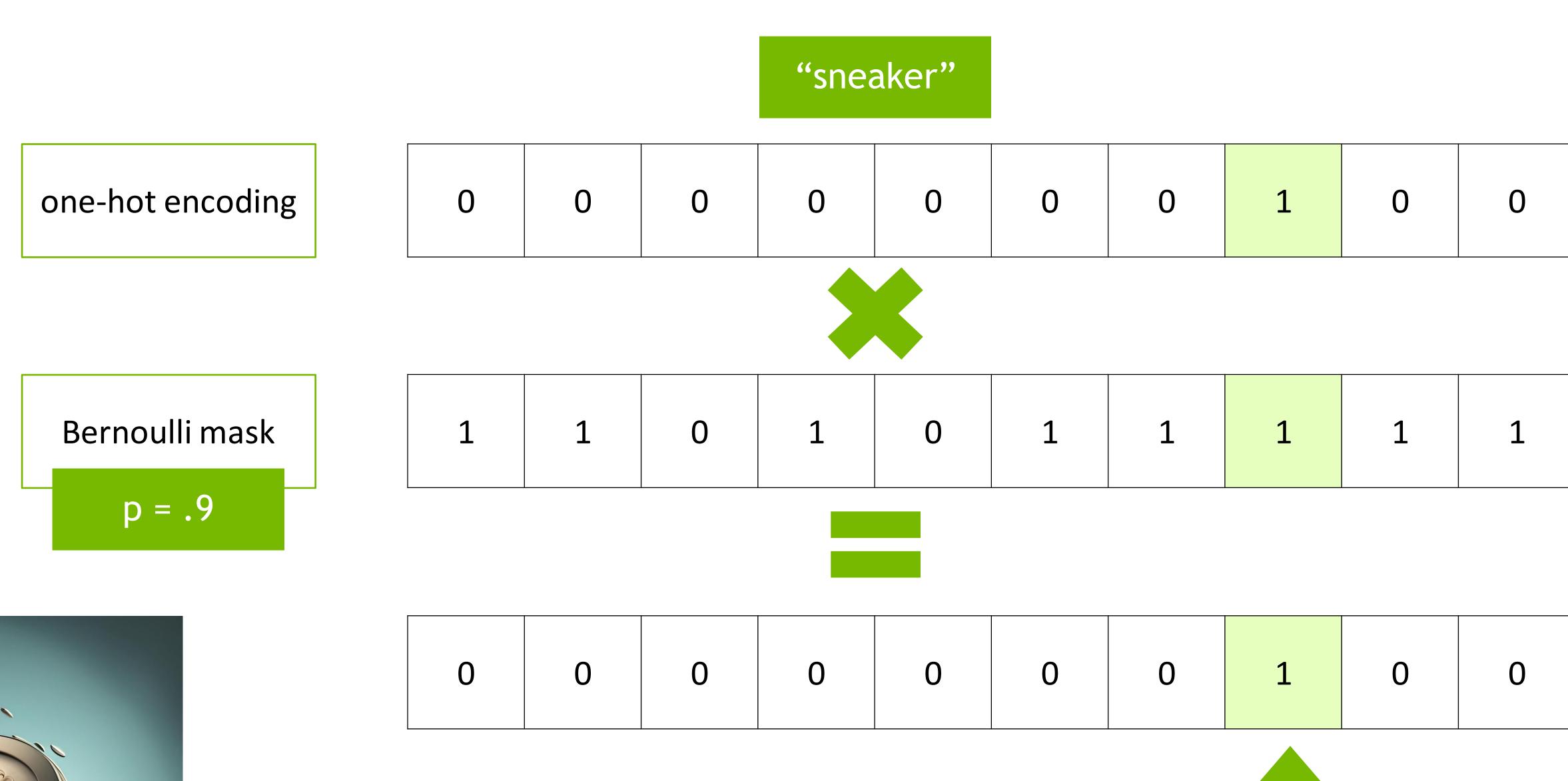
Google Research, Brain team {jonathanho, salimans}@google.com

ABSTRACT

Classifier guidance is a recently introduced method to trade off mode coverage and sample fidelity in conditional diffusion models post training, in the same spirit as low temperature sampling or truncation in other types of generative models. Classifier guidance combines the score estimate of a diffusion model with the gradient of an image classifier and thereby requires training an image classifier separate from the diffusion model. It also raises the question of whether guidance



Bernoulli Masks





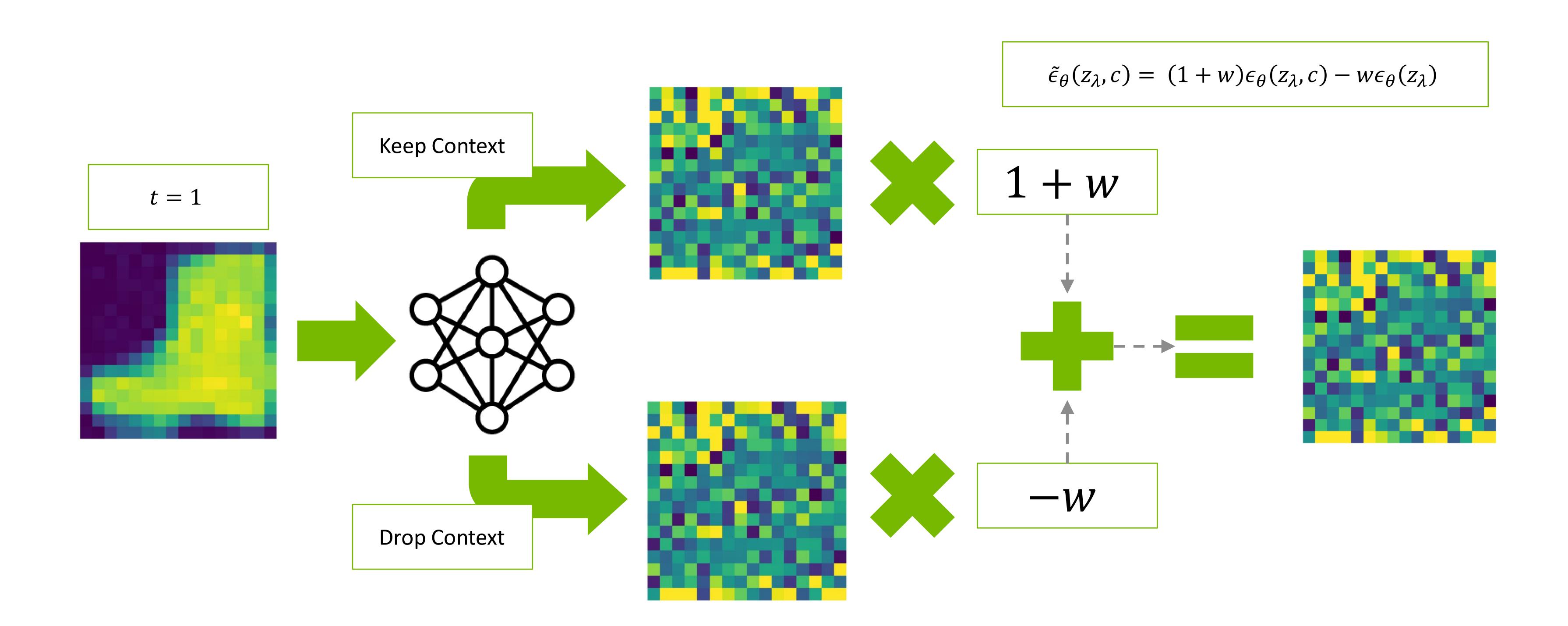
1 - p chance class will be dropped

A weighted coin flipping through the air like a cartoon

Adobe Firefly

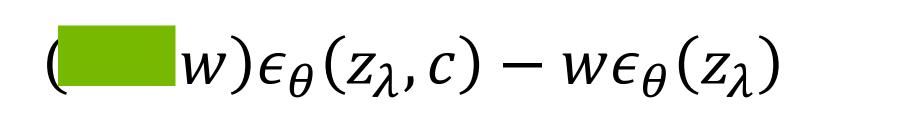


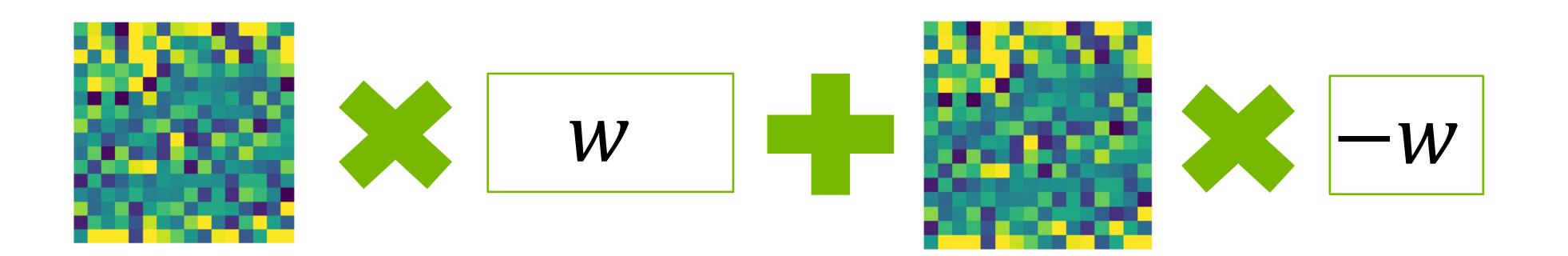
Weighted Reverse Diffusion





Weighted Reverse Diffusion



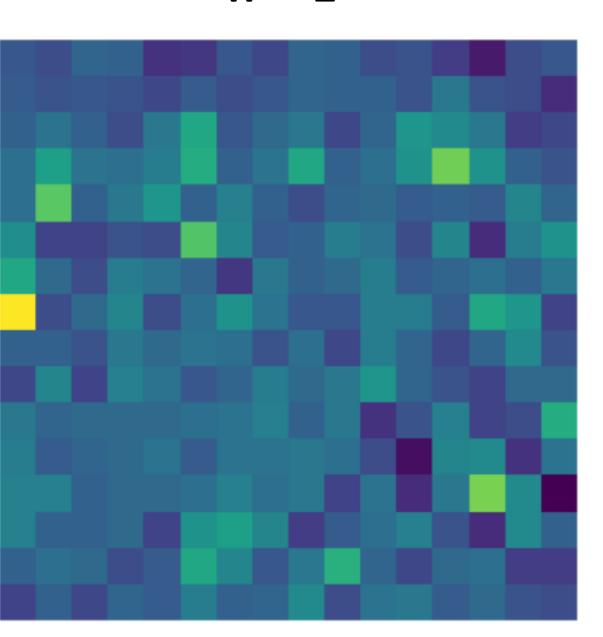


w = 0



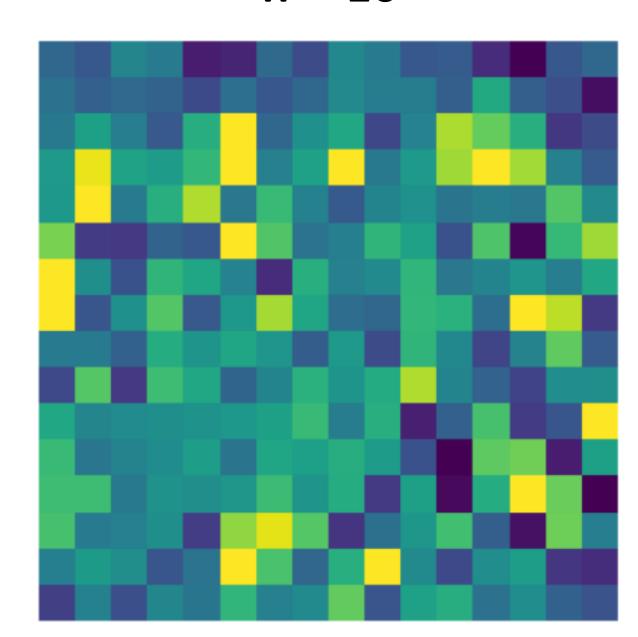
No context feature exaggeration

w = 1



Some context feature exaggeration

w = 10

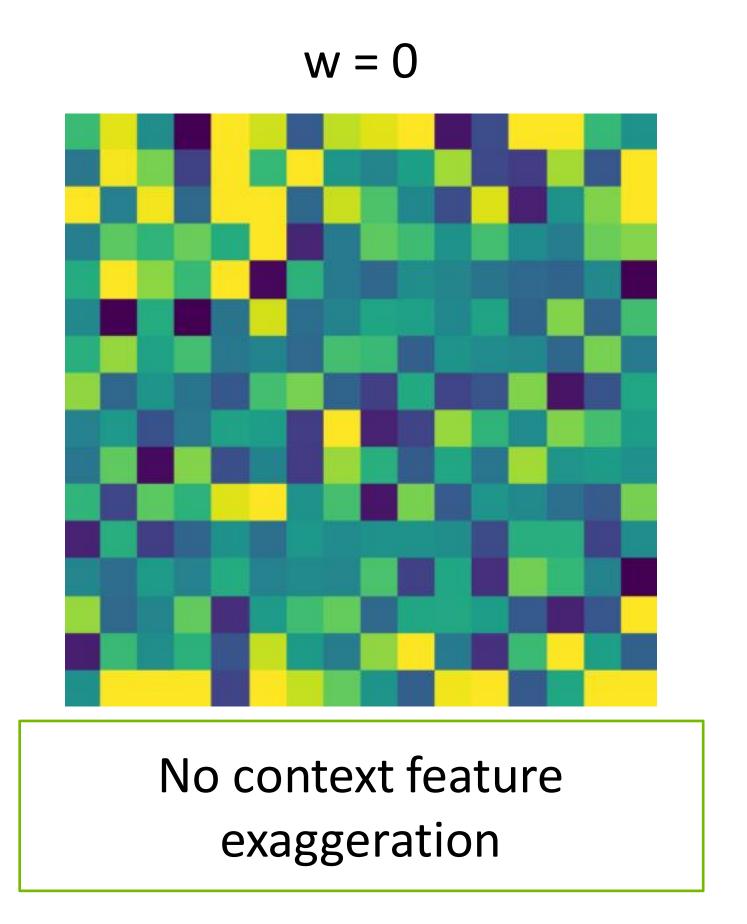


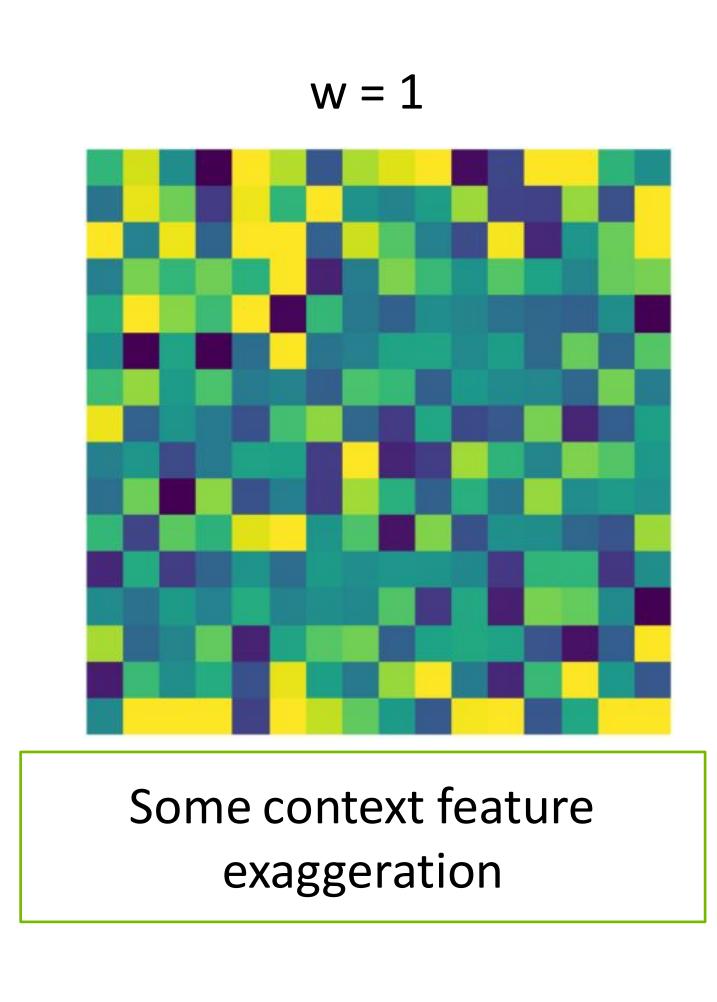
Huge context feature exaggeration

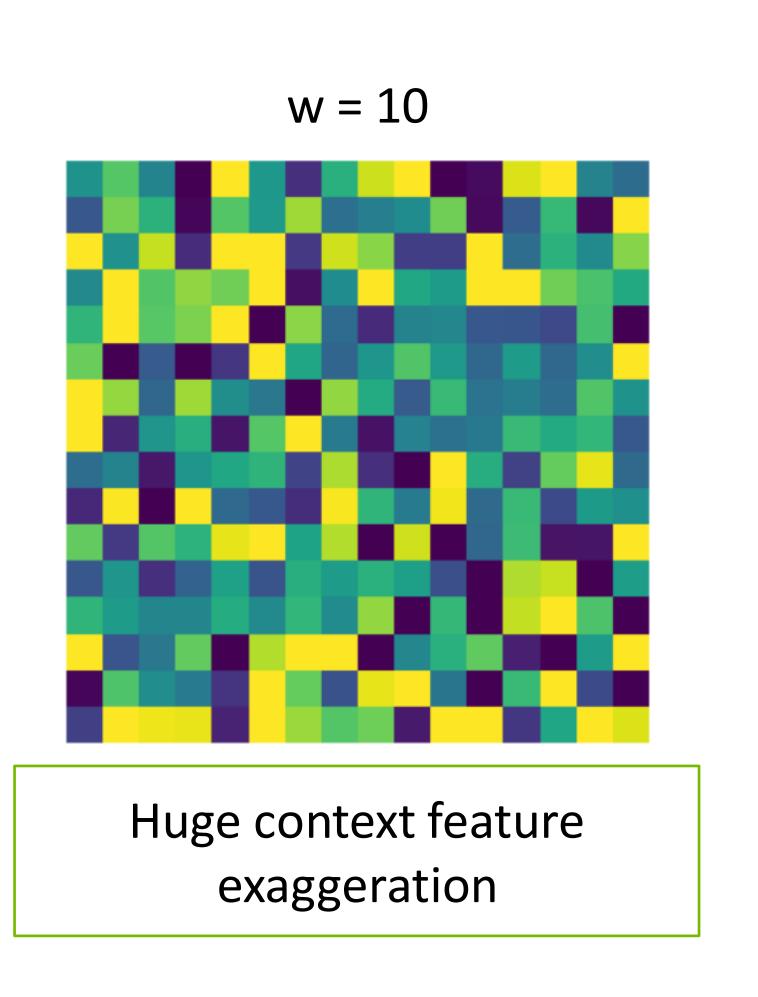
Weighted Reverse Diffusion

$$(1+w)\epsilon_{\theta}(z_{\lambda},c)-w\epsilon_{\theta}(z_{\lambda})$$









Weighted Reverse Diffusion

W	=	-2.	(
• •			•

$$w = -1.0$$

$$w = -0.5$$

$$w = 0.0$$

$$w = 0.5$$

$$w = 1.0$$

w = 2.0





Modified TF Flowers

Modified for Image Generation



Photo by _e.t





