

# Generative Al with Diffusion Models

Part 4: Classifier-Free Diffusion Guidance



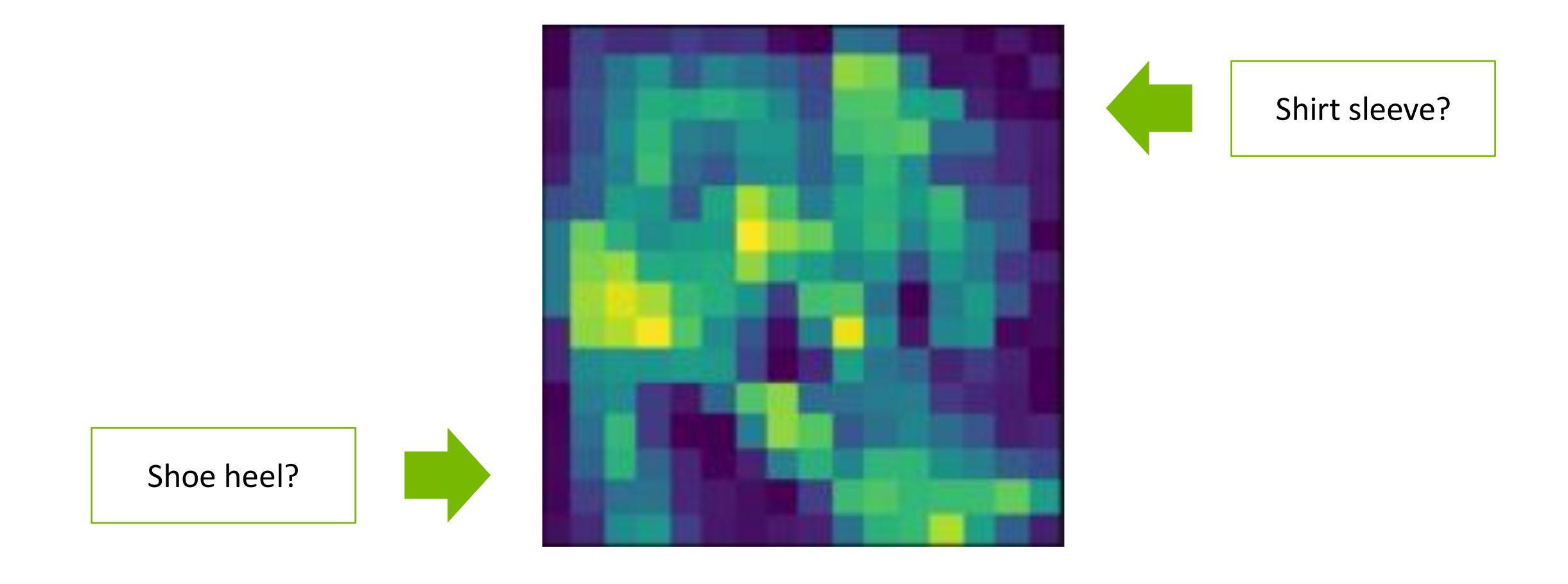
# Agenda

- 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



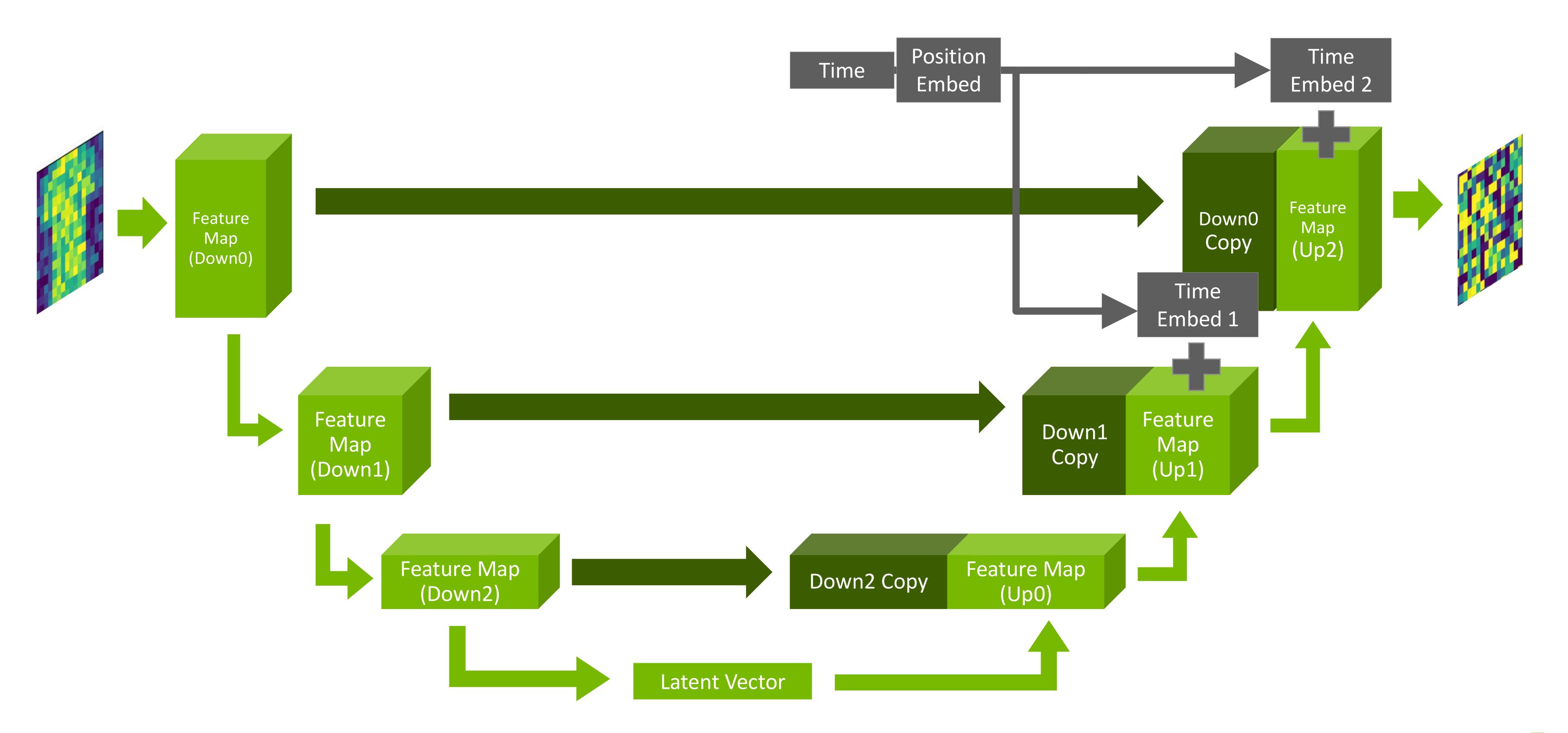


# Adding Context



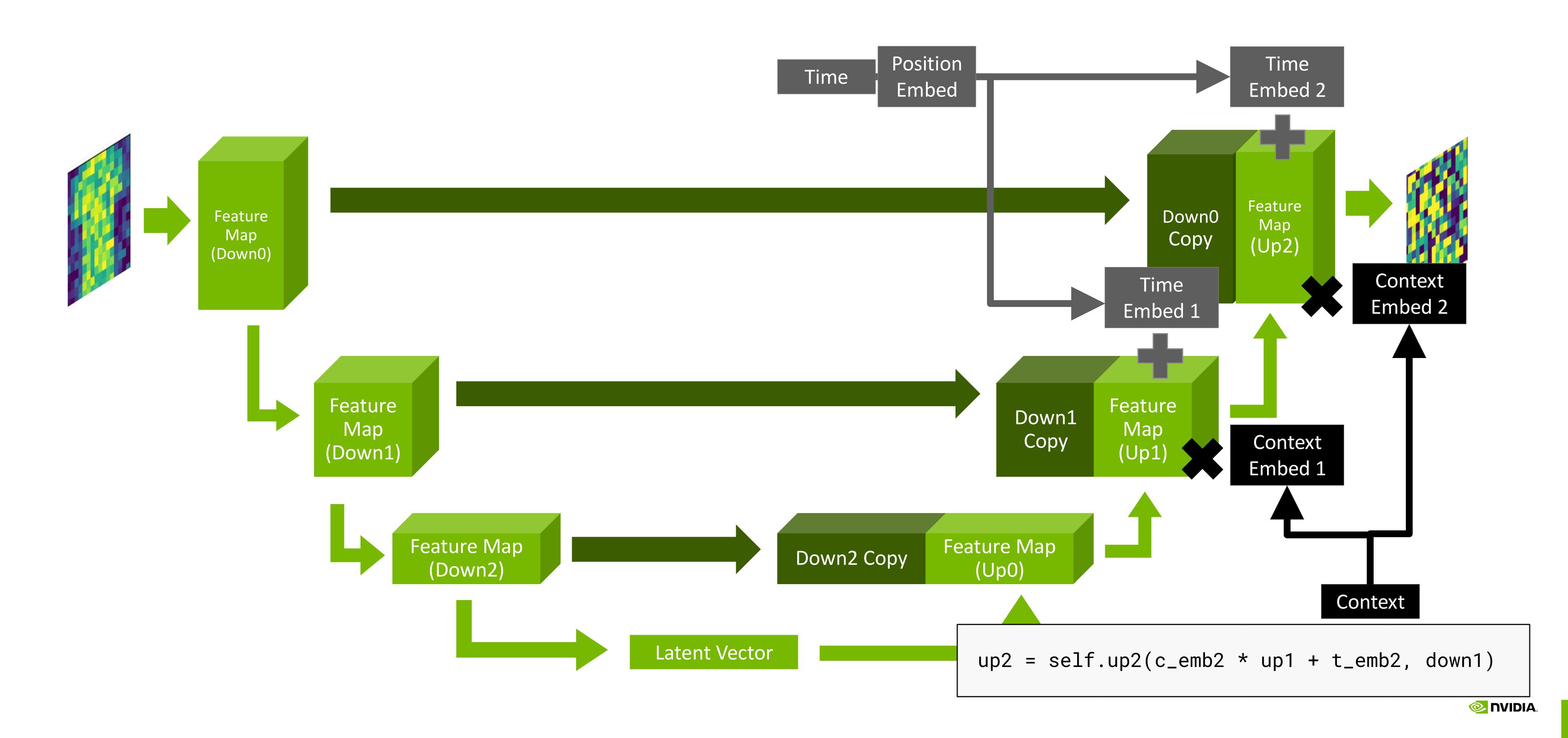


# **Adding Context**



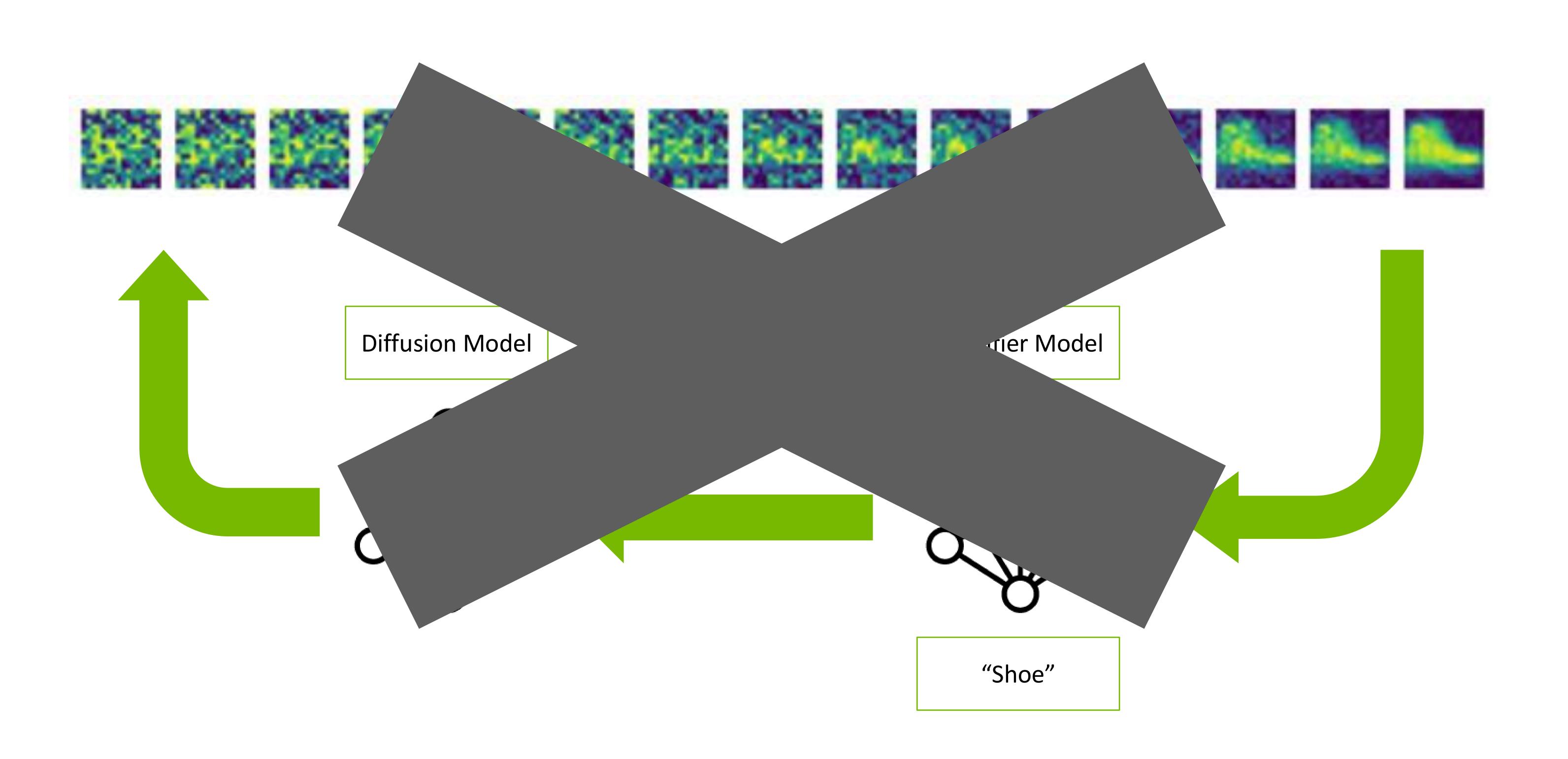


# **Adding Context**





## Make a Classifier Model?





## CLASSIFIER-FREE DIFFUSION GUIDANCE

#### Jonathan Ho & Tim Salimans

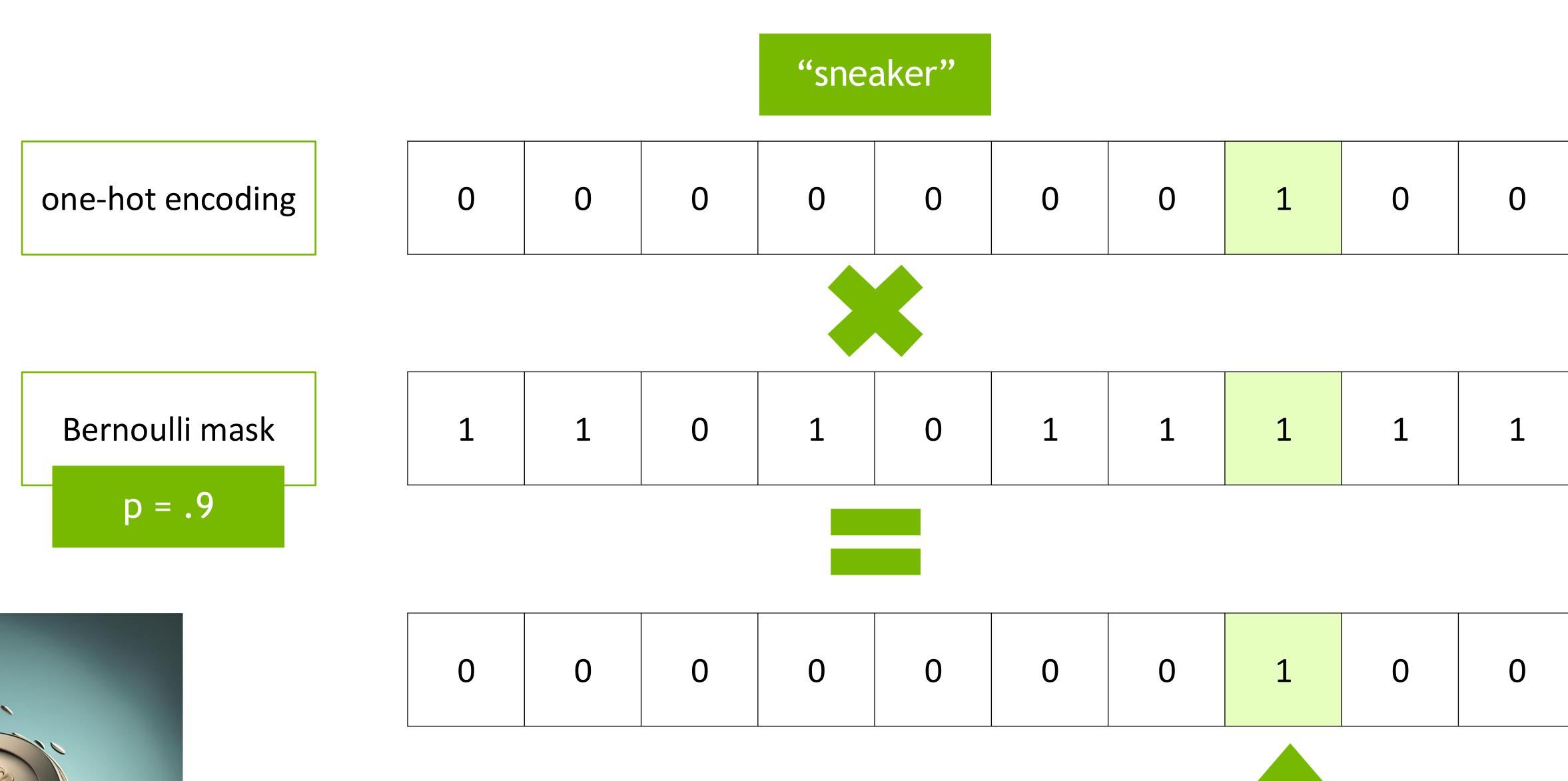
Google Research, Brain team {jonathanho, salimans}@google.com

#### 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



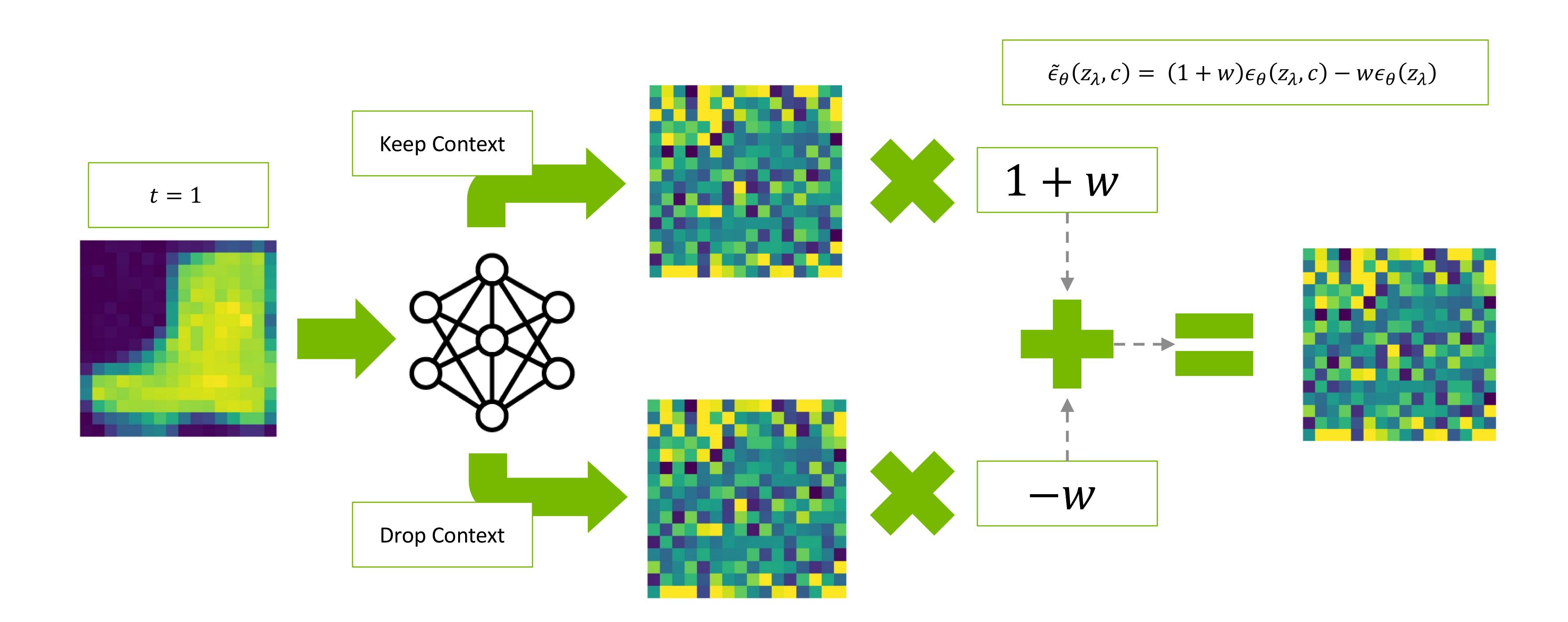


A weighted coin flipping through the air like a cartoon

1-p chance class will be

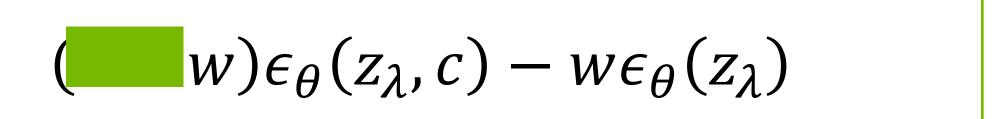
dropped

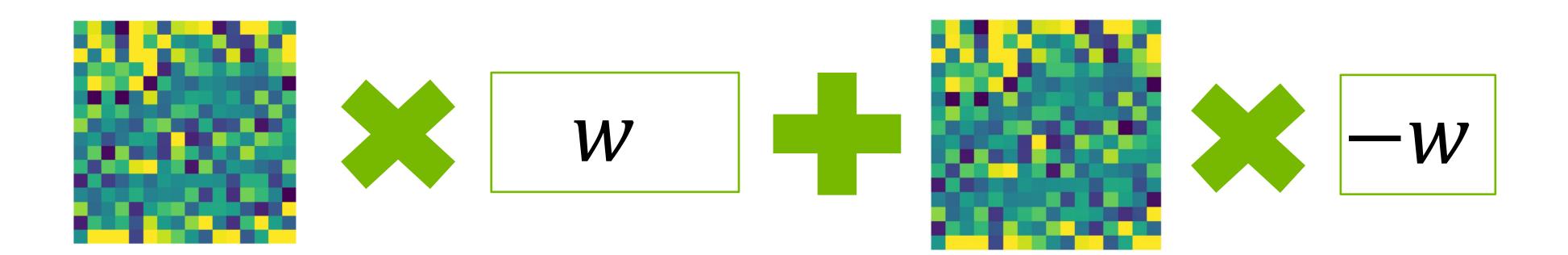
Weighted Reverse Diffusion





Weighted Reverse Diffusion

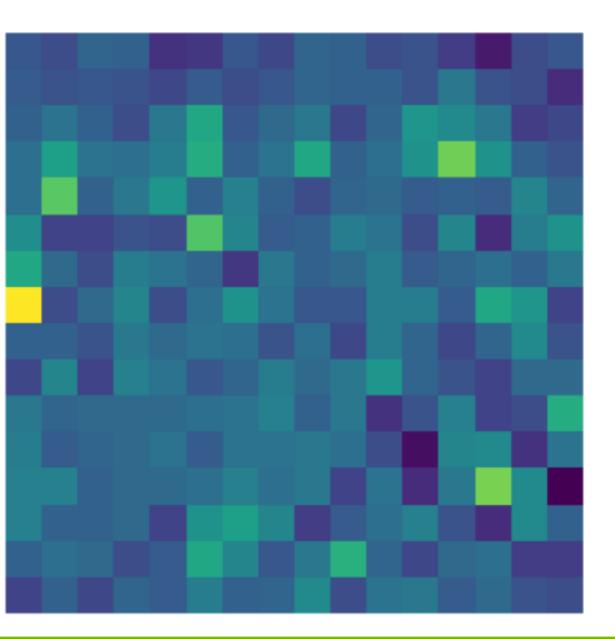




w = 0

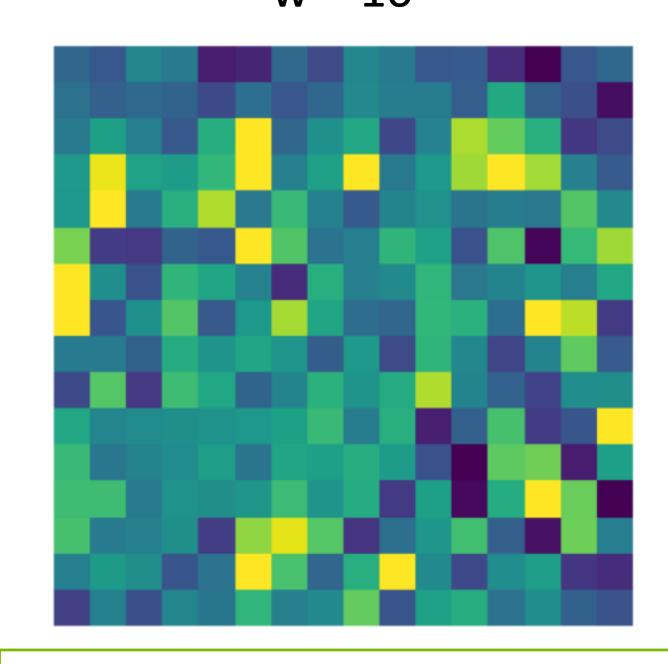
No context feature exaggeration

w = 1



Some context feature exaggeration

w = 10



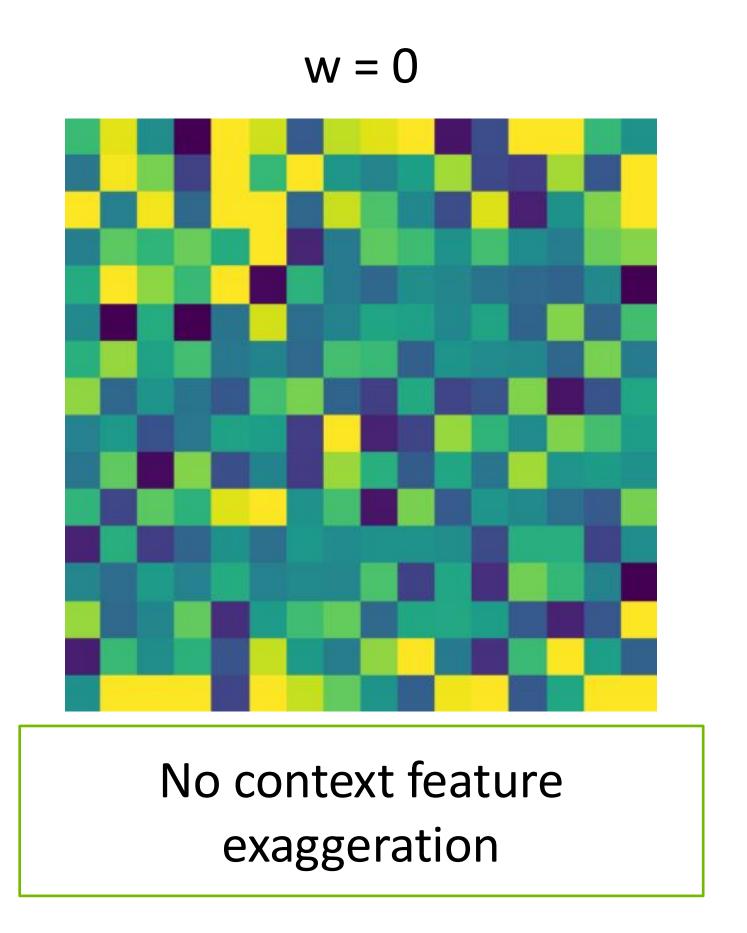
Huge context feature exaggeration

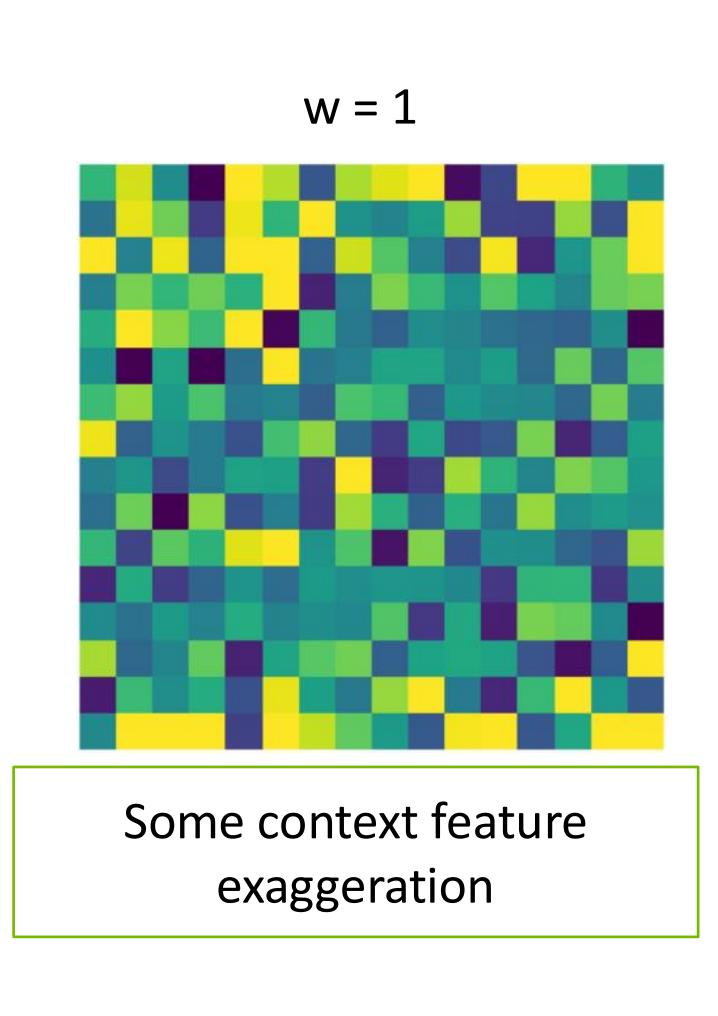


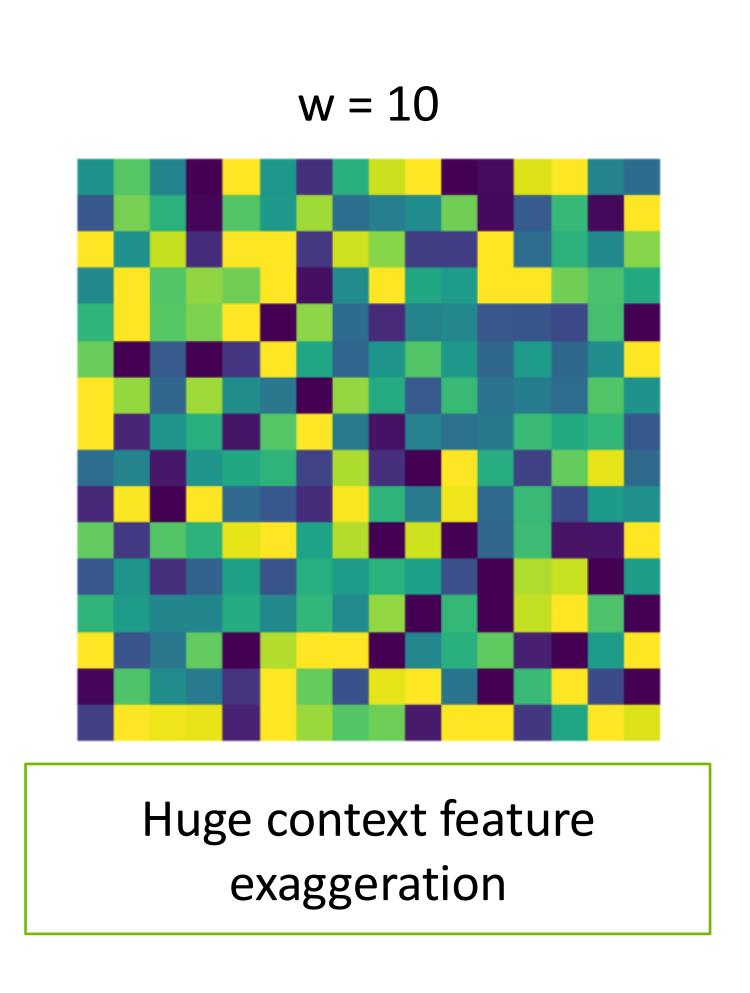
Weighted Reverse Diffusion

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









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





