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# Versatile Diffusion: Text, Images and Variations

## All in One Diffusion Model

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# Outline

- History of Generative Models
- Versatile Diffusion
- Network Architecture
- Disentanglement of Style and Semantic
- Dual Context Blender
- Questions

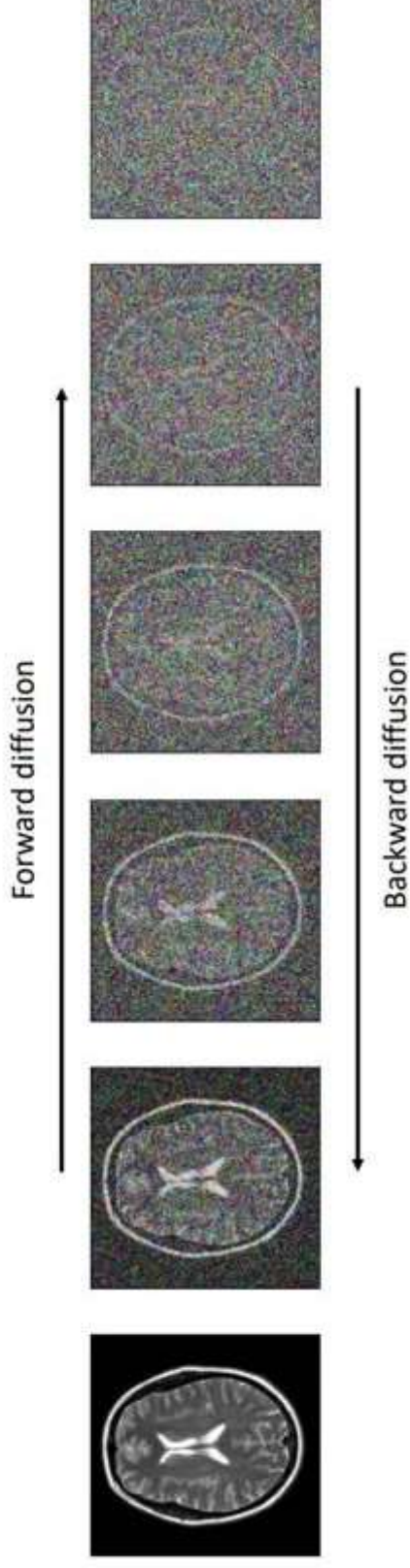
# Generative Adversarial Networks

GAN focus on specific domains and specific tasks i.e. faces



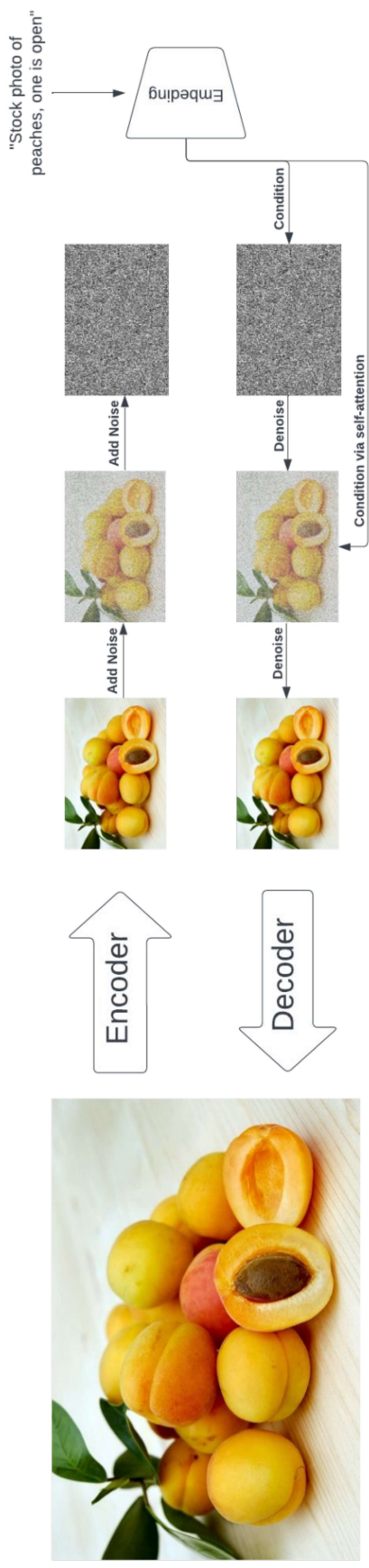
# New Horizons: Diffusion Model

Likelihood based models that gradually restore image contents from gaussian corruptions.



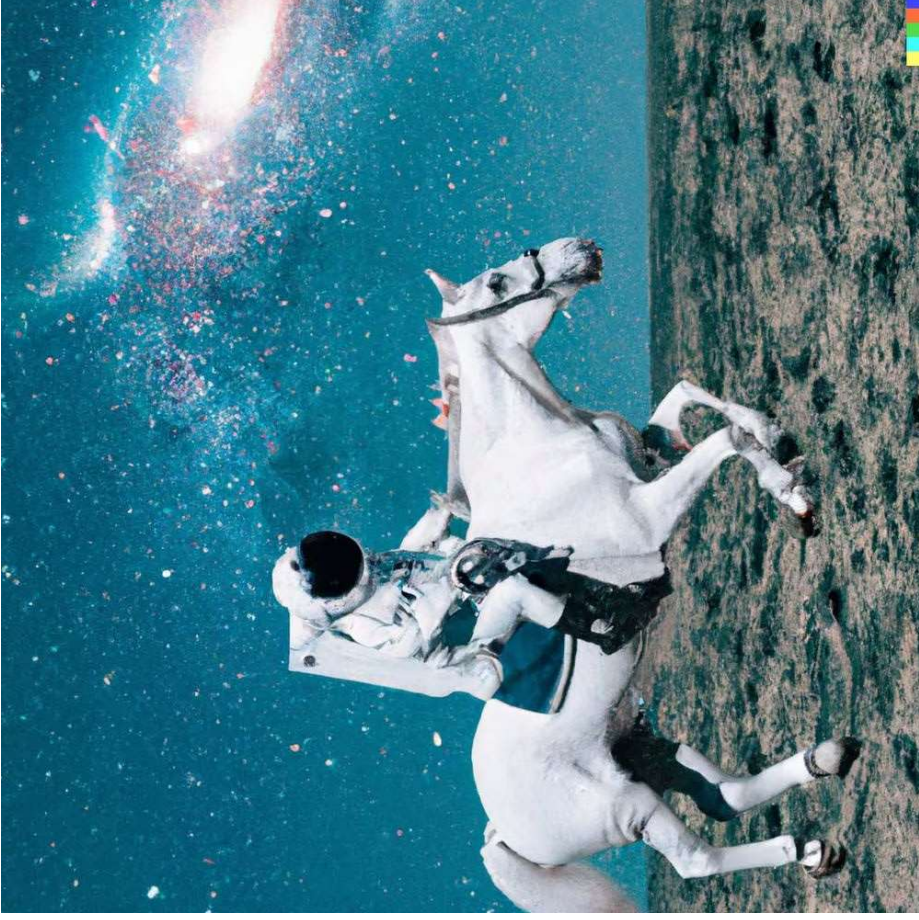
# New Horizons: Diffusion Model

It has proven to be effective in bridge modalities and tasks.



# DALL-E2 (2021)

Input: An astronaut riding a horse in photorealistic style.





# DALL-E3 (2023)

The sidewalks bustling with **pedestrians enjoying the nightlife.**

A bustling city street under the shine of a **full moon.**



At the corner stall, a **young woman** with fiery red hair, dressed in a signature velvet cloak, is **haggling with the grumpy old vendor.**

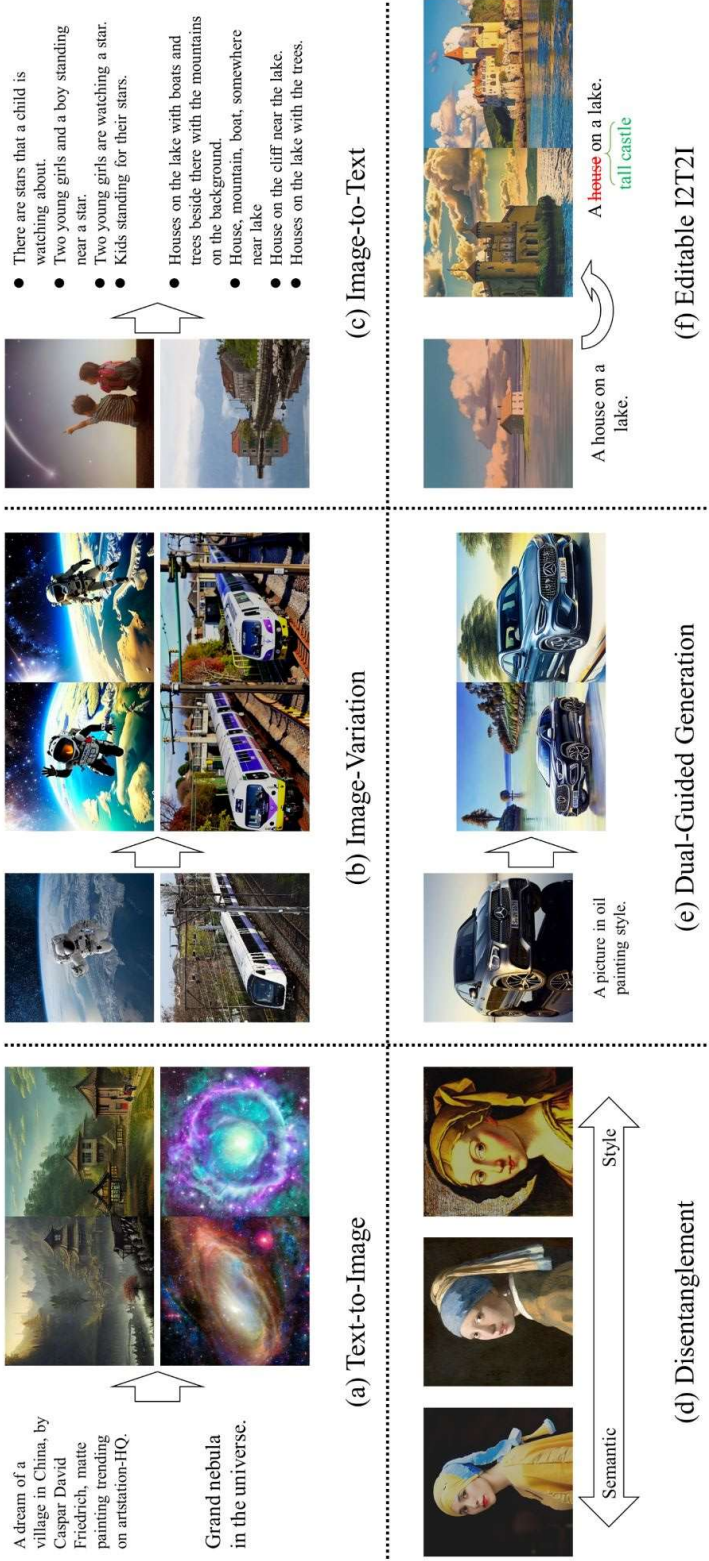
The grumpy vendor, a **tall, sophisticated man**, is wearing a sharp suit, sports a **noteworthy moustache** and is animatedly conversing on his **steampunk telephone.**

**What's Next?**

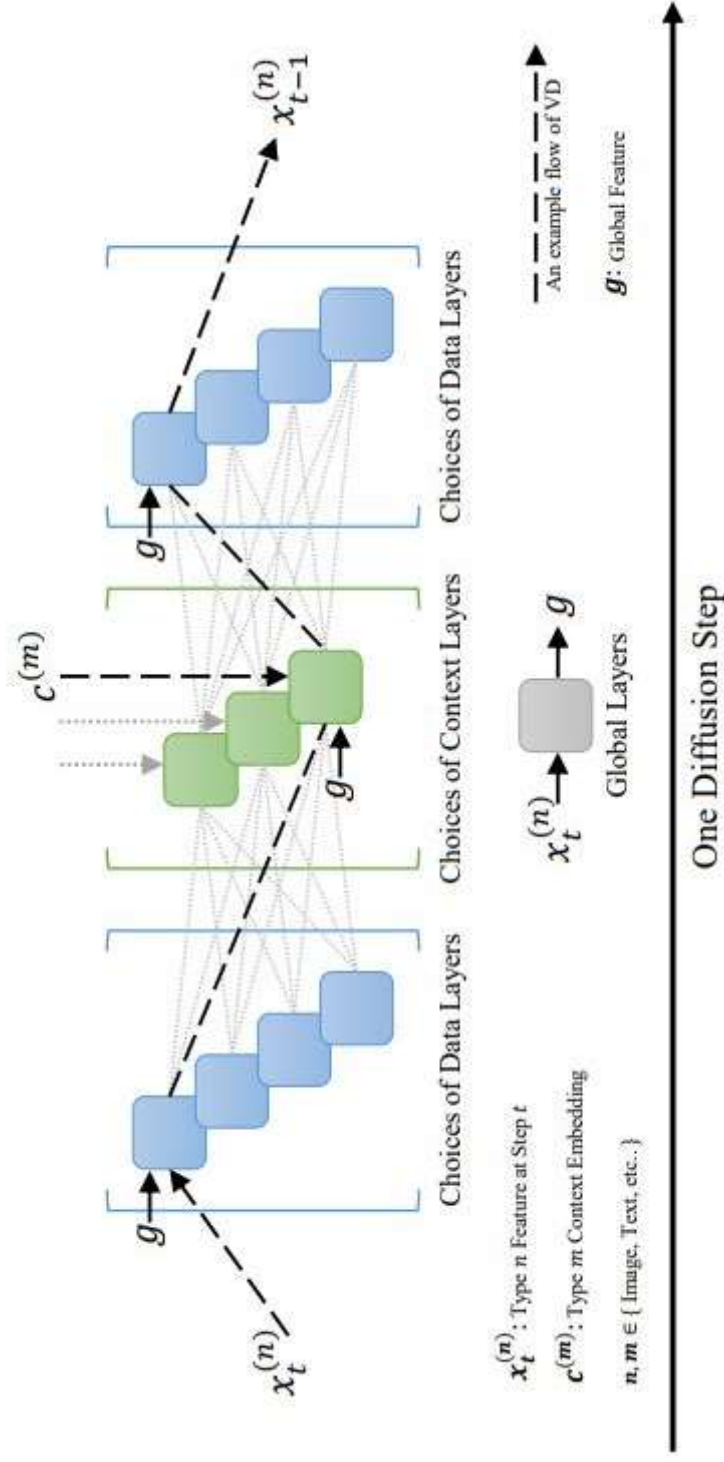




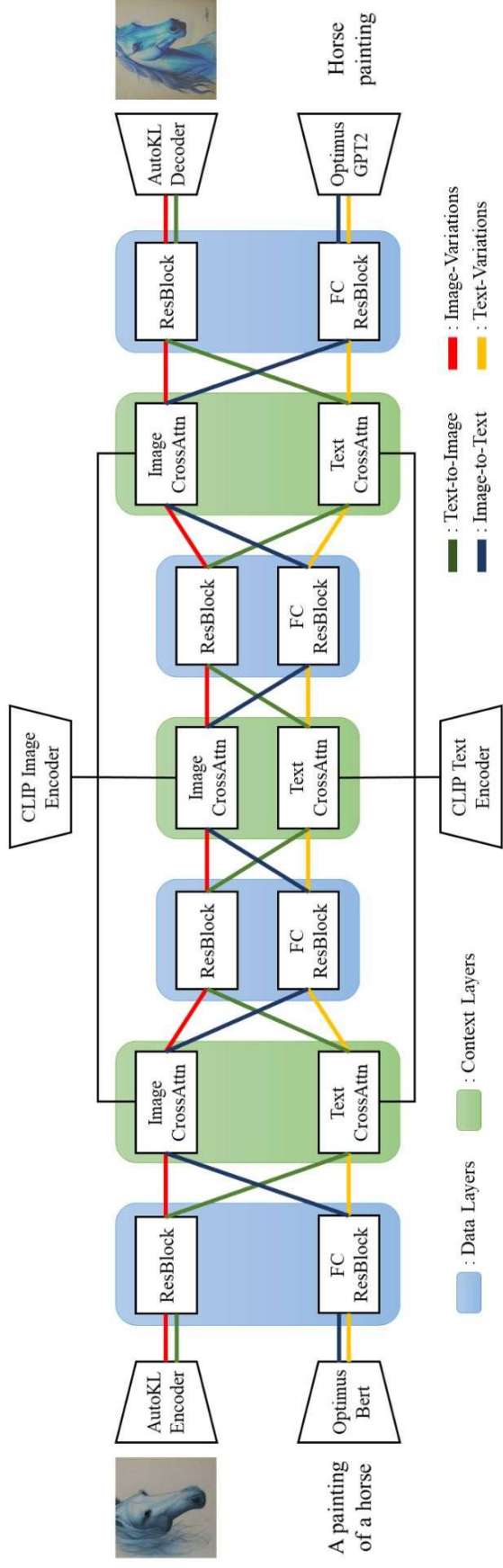
# Versatile Diffusion (2023)



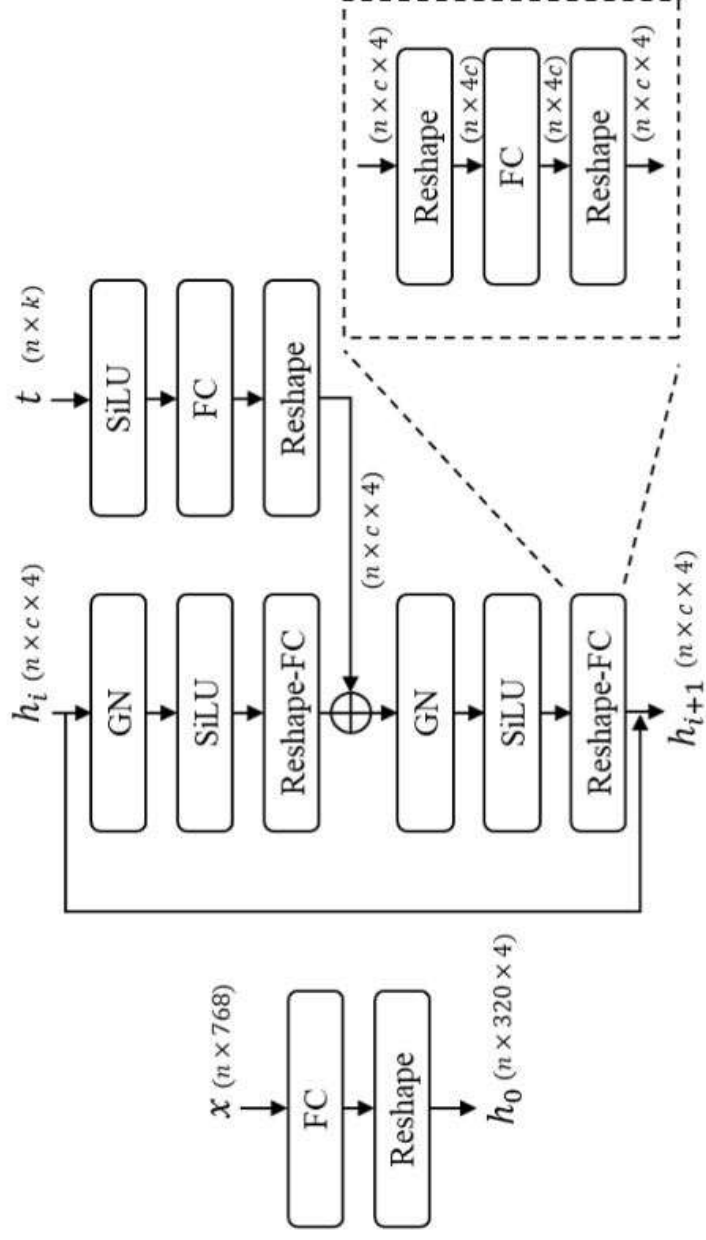
# One Diffusion Step



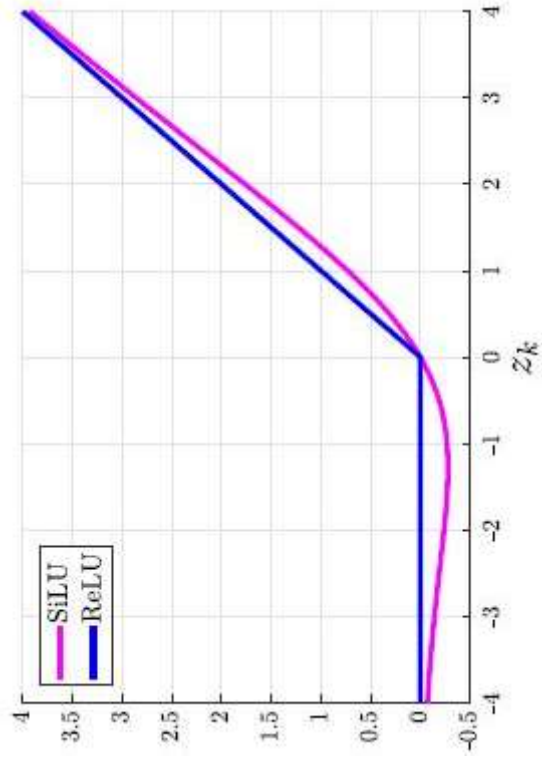
# Network Structure



# Text Data Layers



# SiLu



# Training

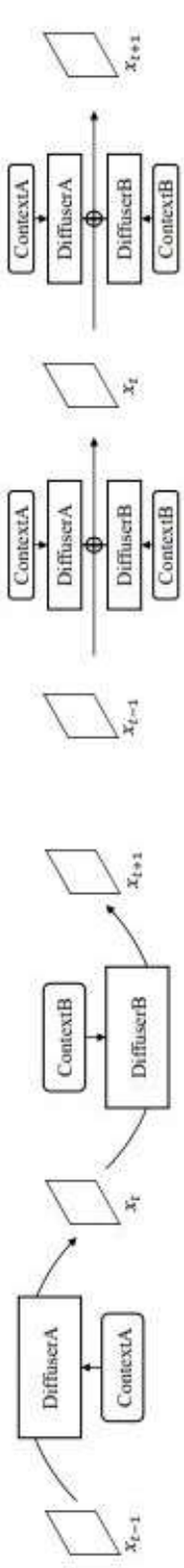
- For each of the flows compute the variational weighted losses and do regular backpropagation.
- Update model weights when the gradient in all flows are accumulated.
- VD is trained progressively in three settings:
  - Single Flow: Image Variation
  - Dual Flow: Text-to-Image and Image-variation
  - Four Flow: All the tasks together



# Disentanglement of Style and Semantic

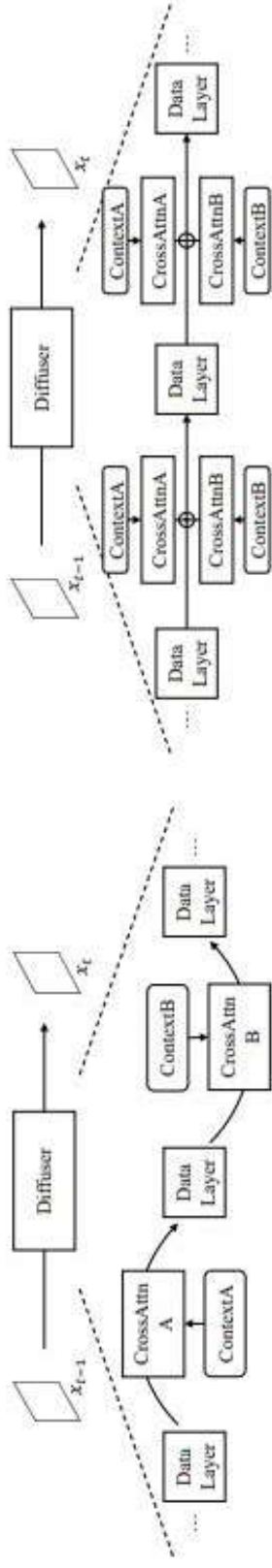


# Dual-context Blender



(a) Model-level Mixing A

(b) Model-level Mixing B



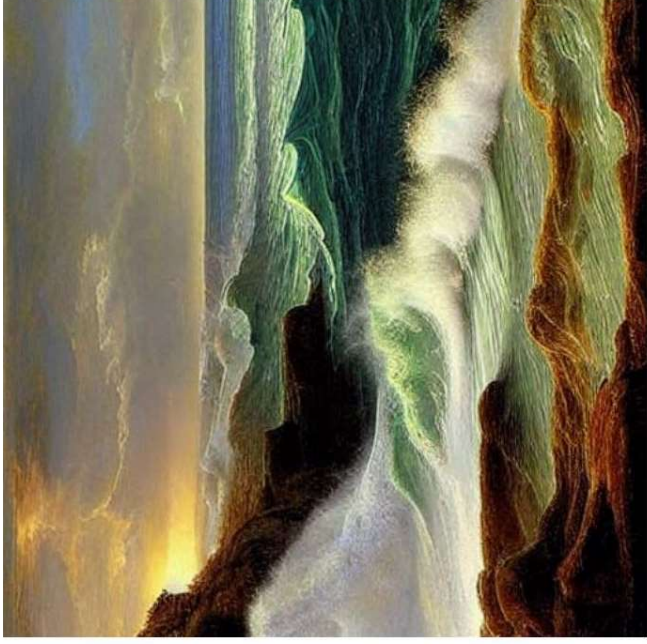
(c) Layer-level Mixing

(d) Attention-level Mixing

# Results: Text to Image



“A wonderful evening in New York City with a great view of Brooklyn Bridge and a magnificent city view of Manhattan, HD 8K”



“A beautiful painting of waves crashing on a cliff by Thomas Cole”



# Results: Image Variation



Input



Variation #1



Variation #2

# Results: Dual-Context Blender



Input



“100 mph”



“Traveling among the stars”

# Limitations

- **Limited Latent Space:** Optimus VAE's latent vector are 768 single dimension generated using Bert which might be inadequate for long text sentences. It is weak in understanding word locations and orders
- **Imperfect Text Data:** There is domain shift in Optimus VAE's training data compare to VD's training data making it difficult to reconstruct certain images.



# Question 1

How does the authors disentangle and semantic using the VD?



## Question 2

Give three example of basic tasks VD can achieve?

