## **SORA: A Review**

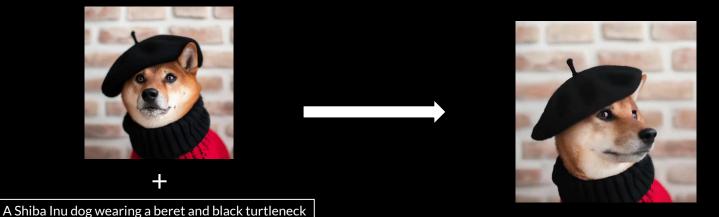
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## Sora and its capabilities

- A text to image and text to video model by OpenAl released in Feb 2024.
- Generates realistic videos/images with text prompt alone. Allows images/videos as well.
- Videos upto 1920x1080 and 1080x1920 and Images upto 2048x2048.
- It follows scaling laws of LLMs and has emergent properties.
- □ First model that can generate minute long videos.

# Sora using image and video prompts

- Uses image/video prompts as context for video generation.
- Video editing Connecting videos, extending videos forwards/backwards etc.
- Static image animation Uses image and text prompt to create video.



## **History of generative models**

- Image generation models like GANs, VAEs, DALLE etc.
- Video generation models like Imagen Video and VideoLDM.
- Text generation models like ChatGPT, Gemini etc.

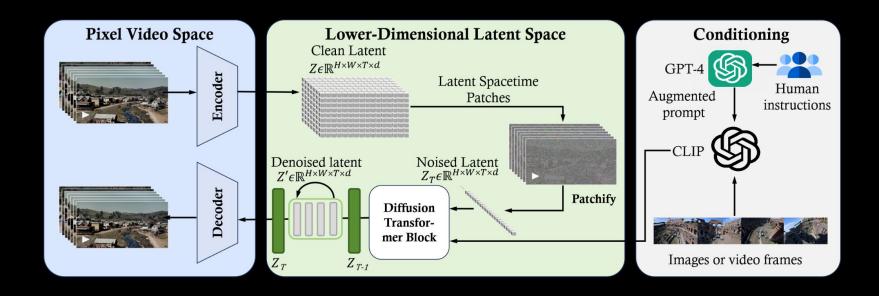
Generate a minute long video that would help me make this presentation



That's not even possible man.

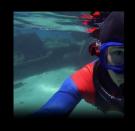
:(

## **Overview of Sora framework**



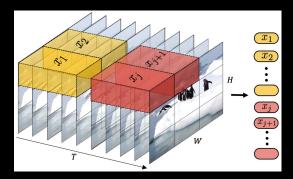
## **Data Pre-processing**

Training on data in their native sizes:
improves composition and framing.



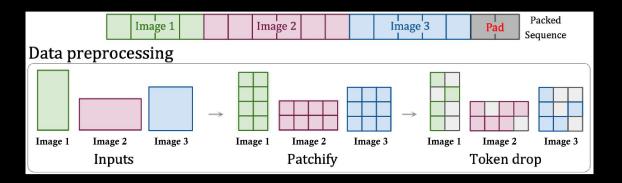


- □ Video compression using VAE or VQ-VAE.
- Decompose into unified spacetime latent patches.
- Spacetime patches :



## **Spacetime Latent Patches**

- Variability in number of patches.
- Patch and pack :
  - Arrange patches in fixed length sequences.
  - Padding and token drop.
- Large context window : can have multiple videos in a sequence.
- Noise addition.

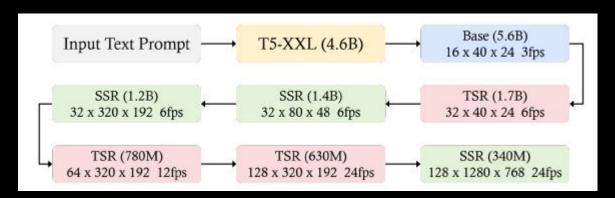


#### **Diffusion Transformers**

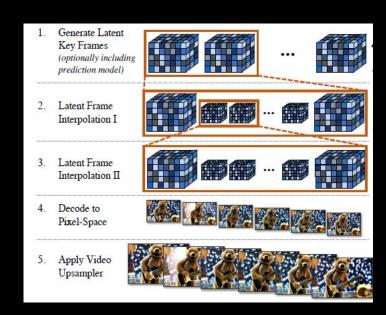
- Success of transformers over U-Nets in visual domain: ViT, U-ViT, MDT.
- Transformers in place of U-Nets for diffusion.
- Diffusion transformers for increasing spatial and temporal resolution.
  - Iteratively go from low resolution image to high resolution image.
  - Interpolate for intermediate frames to increase frame rate.
- Imagen Video and Video LDM.

## Imagen Video

- Contextual embeddings from input prompts.
- □ Base model (3D U-Net).
- Cascade of Spatial Super Resolution and Temporal Super Resolution diffusion models.



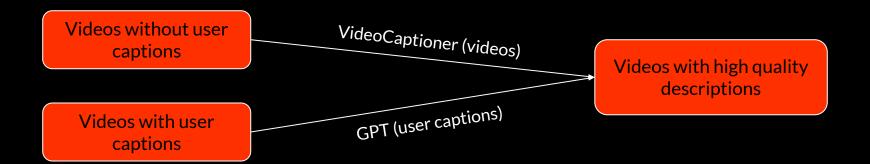
- Video Latent Diffusion Model:
  - Cascade of diffusion models.
  - Additional temporal layers that learn to align individual frames.
  - Fine-tuned for temporal consistency.
- Sora likely uses a cascade of diffusion models:
  - Base model and spacetime refiner models.
- Temporal consistency is more important: it likely trains with longer videos of low resolution.



**Video LDM Architecture** 

## Training data for the text-to-video model

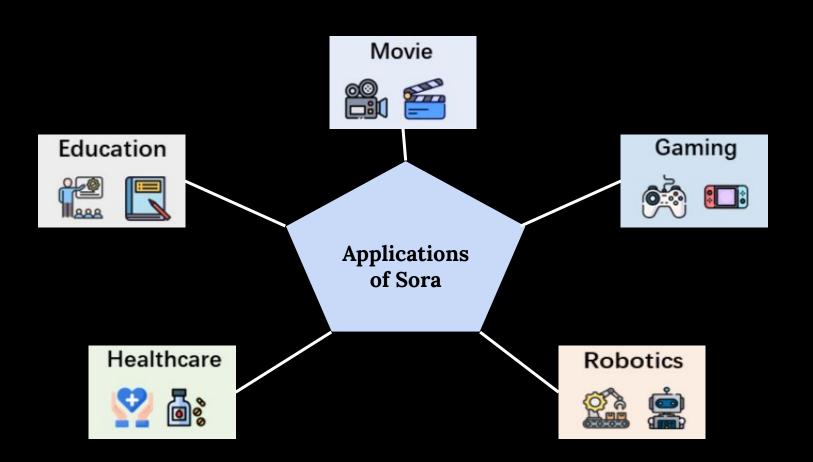
- □ First train a highly descriptive video captioner.
- Apply re-captioning on all videos like DALLE-3.
- Also utilizes GPT-enhanced user prompts.
- Gives high quality video caption pairs



# Video captioner training using VideoCoCa

- Sample multiple frames from a video.
- Image encoder gives frame token embeddings.
- □ Flatten and concatenate into a long sequence of video representation.
- Processing token information using attention pooler.
- Uses contrastive and captioning loss for training.

Sampling frames	Encoding	Concatenating	Pooling	Training
Sampling key frames from a video	Encoding frames independently	Concatenating the embeddings	Attention pooling to aggregate temporal information	Training using obtained video, caption pairs



### **Security and Protection**

- Misuse and jailbreak attacks
- Hallucination
- Bias
- Privacy

- □ RLHF
- Usage permissions
- Privacy protection

## **Limitations and Challenges**

- Physical realism
- Spatial and temporal complexities
- Human-computer interaction
- Length of videos



# Thank you