

SynCamMaster: Synchronizing Multi-Camera Video Generation from Diverse Viewpoints

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Overview

TL; DR:

SynCamMaster generates multiple synchronized videos of the same dynamic scene.

Input and Output:

- > 1 text prompt + N camera parameters
- > N synchronized videos.

Main Features:

- ✓ Multi-camera synchronized video generation.
- ✓ Enable synthesis from diverse viewpoints.
- ✓ A simple and efficient module on top of pretrained text-to-video models.

SynCamVideo Dataset:

Release a multi-camera synchronized video dataset rendered with Unreal Engine 5.



Close-Up and Wide Shot



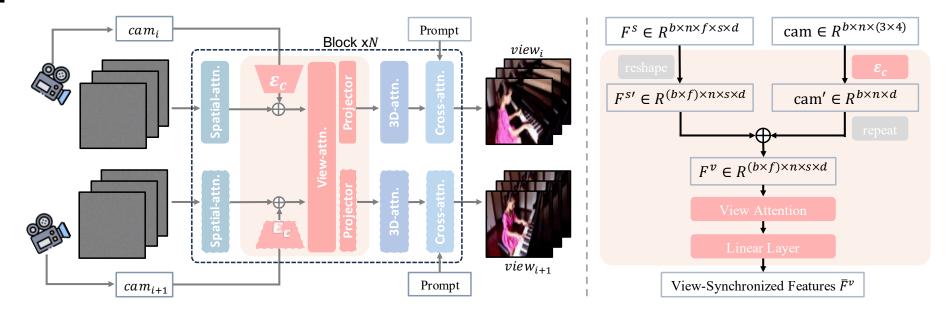
60° in Azimuth+30° in Elevation

Background & Motivation

- Multi-View Video Generation.
 - Existing works primarily focus on 4D object generation or generation on a specific domain (e.g., autonomous driving).
 - This paper explores how to achieve open-domain multicamera video generation.
- Why Multi-View Open-Domain Generation?
 - In filmmaking, switching back and forth between multiple cameras is commonly used to create a storytelling atmosphere.
 - It can be used as a data generator for various downstream tasks (e.g., robotic manipulation, 3D human pose estimation).



Method



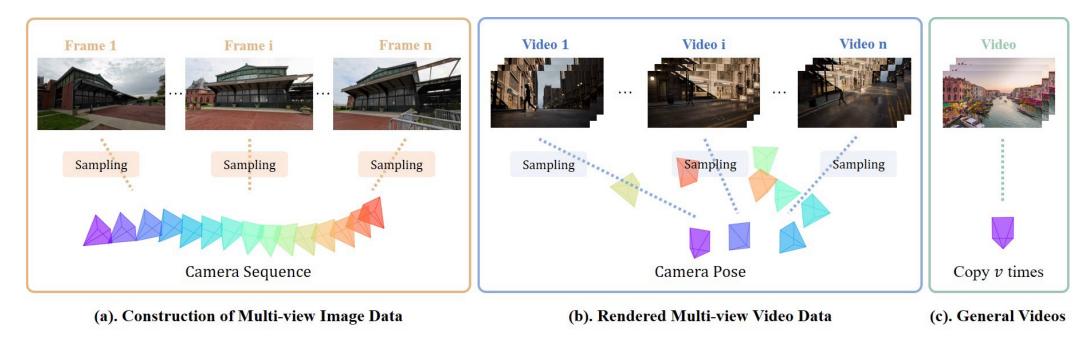
(a) Overview of SynCamMaster

- (b) Multi-View Synchronization Module
- Based on the pre-trained text-to-video model, two components are newly introduced:
 - The camera encoder projects the camera extrinsics into the embedding space.
 - <u>The multi-view synchronization module</u>, as plugged in each TransformerBlock, modulates multi-view features under the guidance of camera parameters.

$$\mathbf{F}_i^v = \mathbf{F}_i^s + \mathcal{E}_c(\mathsf{cam}^i), \tag{5}$$

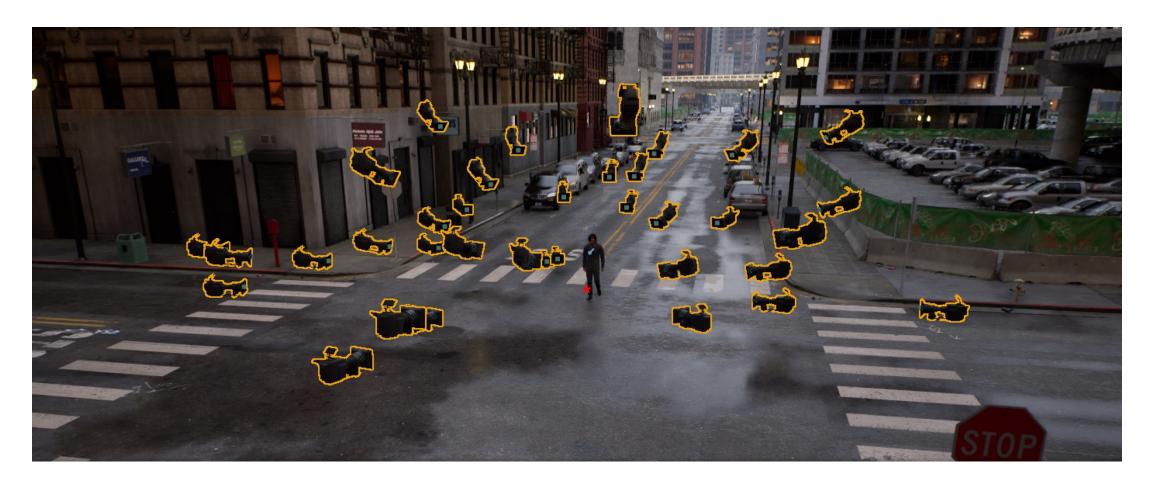
$$\overline{\mathbf{F}}_{i}^{v} = \mathbf{F}_{i}^{v} + \text{projector}(\text{attn_view}(\mathbf{F}_{1}^{v}, \dots, \mathbf{F}_{n}^{v})[i]), \tag{6}$$

Training Data



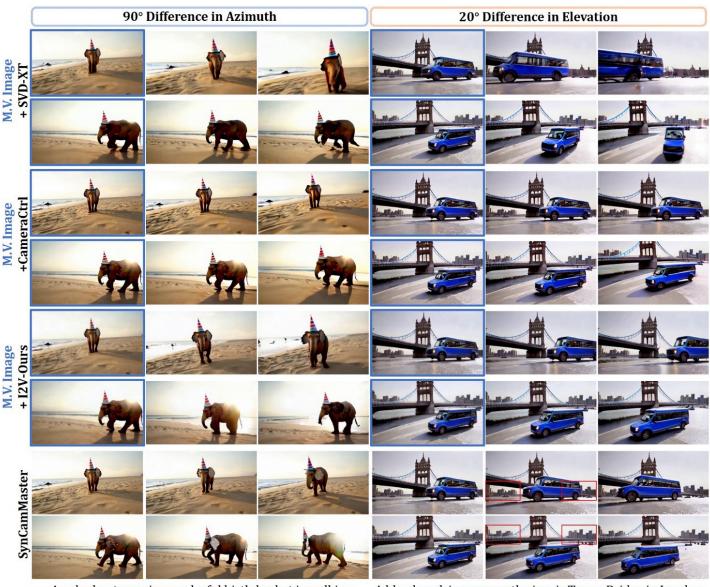
- Due to the scarcity of available multi-view videos, we used a hybrid training set to enhance the model's robustness and improve the visual quality of the generated videos, the training set is composed of:
 - Multi-view image data from videos with camera movements.
 - Multi-view video data from the rendered SynCamVideo Dataset.
 - General video data from the internet.

SynCamVideo Dataset



- Multi-Camera Synchronized Videos + Corresponding Camera Parameters
- Rendered with UnrealEngine 5

Results



An elephant wearing a colorful birthday hat is walking along the sandy beach.

A blue bus drives across the iconic Tower Bridge in London.

Results

Table 1: Quantitative comparison with state-of-the-art methods.

| | Visual Quality | | | View S | View Synchronization | | |
|-------------------------|----------------|------|---------|---------|----------------------|-----------|----------|
| Method | FID ↓ | FVD↓ | CLIP-T↑ | CLIP-F↑ | Mat. Pix.(K) | ↑ FVD-V ↓ | CLIP-V ↑ |
| M.V. Image + SVD-XT | 137.3 | 1755 | - | 97.56 | 150.4 | 1742 | 89.14 |
| M.V. Image + CameraCtrl | 152.8 | 2203 | - | 98.32 | 172.9 | 1661 | 89.33 |
| M.V. Image + I2V-Ours | 113.1 | 1376 | 33.48 | 99.27 | 116.8 | 1930 | 90.01 |
| SynCamMaster | 116.7 | 1401 | 33.40 | 99.36 | 527.1 | 1470 | 93.71 |

Table 2: Quantitative ablation on the joint training strategy.

| | Visual Quality | | | View Synchronization | | | ion |
|---|--------------------------------|-----------------------------|--------------------------------|--------------------------------|--------------------------|-----------------------------|--------------------------------|
| Method | FID ↓ | FVD↓ | CLIP-T↑ | CLIP-F↑ | Mat. Pix.(K) ↑ | FVD-V↓ | CLIP-V ↑ |
| Multi-View Video | 149.9 | 1971 | 30.97 | 99.37 | 460.5 | 1668 | 89.68 |
| + Multi-View Image + General Video + Both | 121.5 122.4 116.7 | 1655 1608 1401 | 33.02 32.54 33.40 | 99.36 99.38 99.36 | 533.0 471.9 527.1 | 1482 1514 1470 | 93.15 90.12 93.71 |

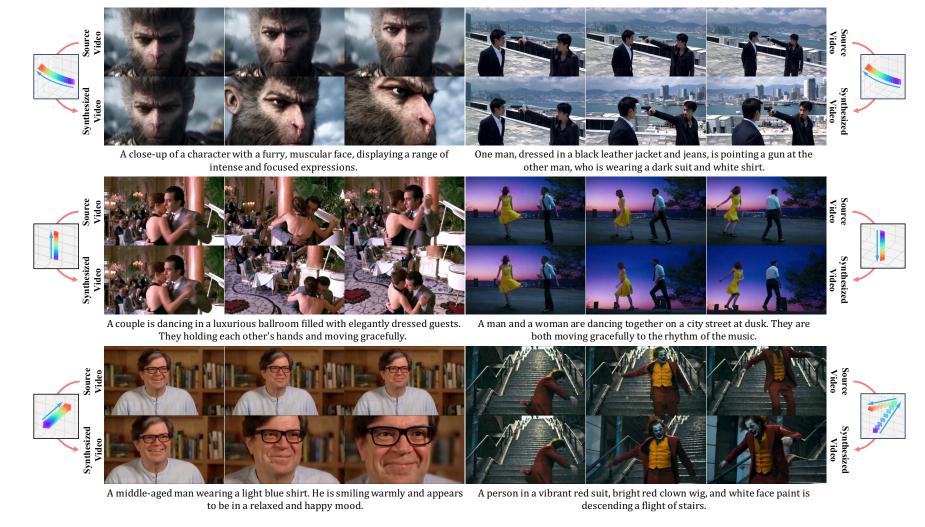
Table 3: Results of novel view video synthesis.

| Setting | LPIPS ↓ | PSNR ↑ | SSIM ↑ |
|------------------------|---------|--------------|--------|
| $s_V = 1.2, s_T = 5.0$ | 0.4899 | 16.29 | 0.4754 |
| $s_V = 1.2, s_T = 7.5$ | 0.4901 | 16.60 | 0.4783 |
| $s_V = 1.8, s_T = 7.5$ | 0.4761 | 16.47 | 0.4935 |
| $s_V = 2.5, s_T = 7.5$ | 0.5022 | 14.55 | 0.4667 |

Table 4: Accuracy of camera control.

| Method | RotErr ↓ | TransErr↓ |
|-------------------------|----------|-----------|
| M.V. Image + SVD-XT | 0.25 | 0.72 |
| M.V. Image + CameraCtrl | 0.16 | 0.67 |
| M.V. Image + I2V-Ours | 0.26 | 0.80 |
| SynCamMaster | 0.12 | 0.58 |

Subsequent Work: ReCamMaster



- Input: source video + target camera trajectory.
- Output: Video with the novel camera trajectory.

Take Home Messages

- We propose SynCamMaster to synthesize multi-camera videos from the text prompt and camera extrinsic.
- We release a multi-camera synchronized video dataset rendered with Unreal Engine 5.
- Our subsequent work, ReCamMaster, can recapture an input video using novel camera trajectories.
- For more information:



SynCamMaster Project Page



ReCamMaster Project Page

Thanks for your attention!