

Final Project Proposal

Paper Implementation:
[CVPR 2023] MoFusion: A Framework for
Denoising-Diffusion-based Motion Synthesis

Group 32

110612117 張仲瑜

110550022 賴柏允

Goal

To implement the a diffusion-based motion synthesis model that can generate a sequence of motion based on the given text prompt or music.



Related work

[ICLR2023]

MDM: Human Motion Diffusion Model

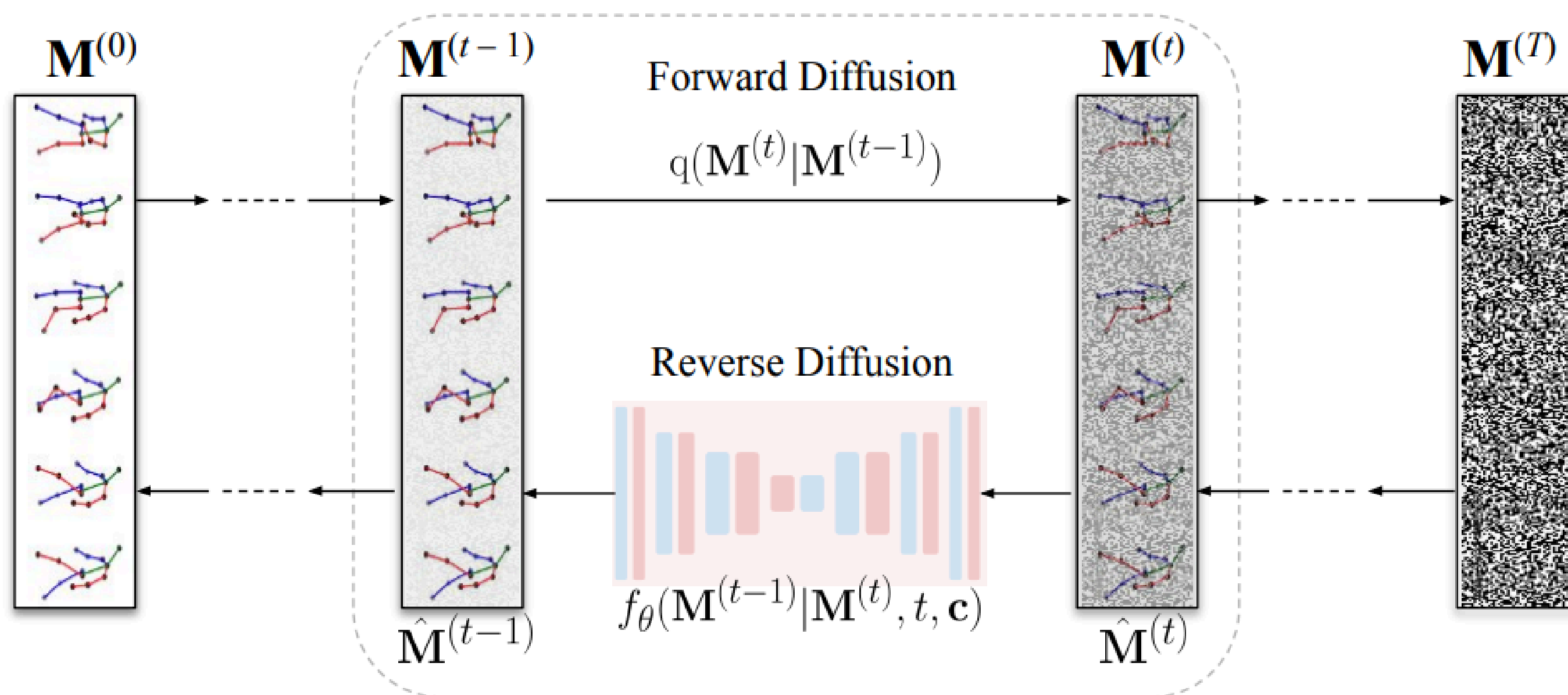
[ICCV 2023]

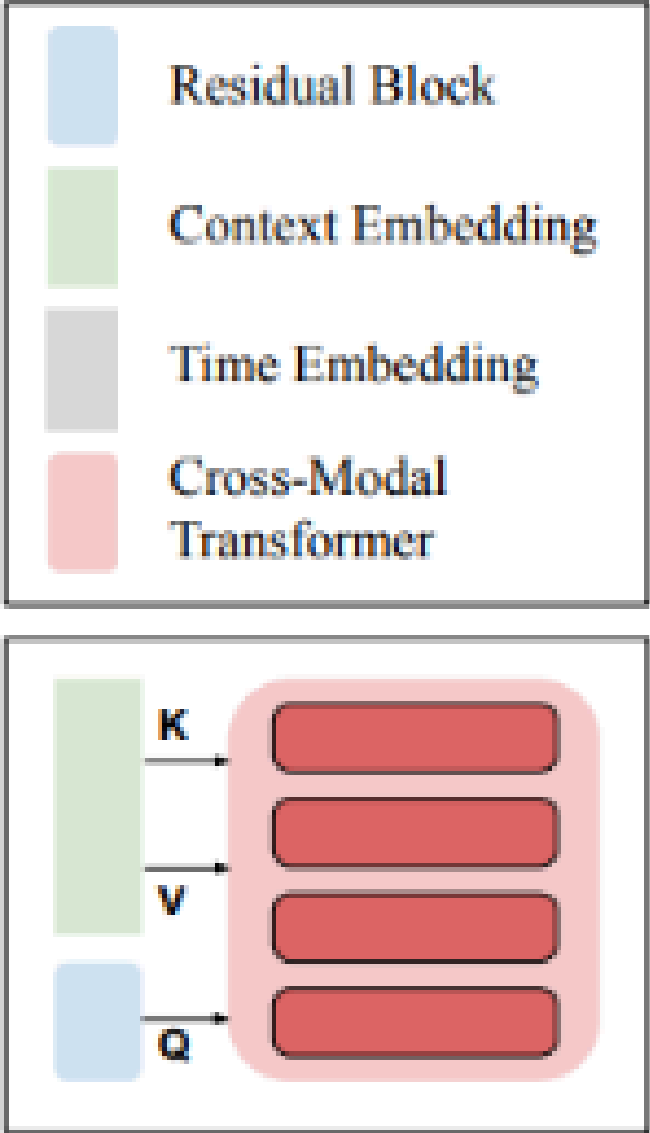
PhysDiff: Physics-Guided Human Motion Diffusion Model

[SIGGRAPH 2023]

Listen, denoise, action! Audio-driven motion synthesis with diffusion models

Core idea





Pipeline

```
graph LR; FE[Feature Extraction] --> FD[Forward Diffusion]; FD --> BD[Backward Diffusion];
```

Backward Diffusion

- Train a model to denoise
- Loss
 - L2 distance between the predicted noise and that added in the forward process
 - Constraints
 - Symmetric Bone length
 - Consistent Bone length
 - Ground Truth supervision

Forward Diffusion

- Adding Gaussian noise to a motion sequence
- Timestamp and context embeddings are included during the training

Feature Extraction

- Transformer to get the context embeddings

Expected Result

Things that-

- Are guaranteed (We'll try our best!) to work:
 - Code of building the Unconditional generation
 - Code of text to motion synthesis
 - Plan B: We will make a blender animation if things above don't work out the way we hope.
- May (might not) work:
 - Audio to motion synthesis
 - Same performance as the thesis (due to poor hardware, RTX 3060 Ti vs RTX A40)

Example

Text-to-Motion Synthesis

