Artificial Intelligence for 3D Asset Model Reconstruction from 2D Images

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Problem

3D asset models are becoming increasingly popular in various fields such as the gaming and animation industries. However, creating a 3D model from scratch is a complex process that requires significant time and patience, leading to a growing demand for skilled 3D artists.

Currently, Artificial Intelligence (AI) is widely used to assist in creating 2D images. However, the process of generating 3D asset models remains more complex and laborintensive. This project aims to explore how AI can be leveraged to generate 3D asset models from 2D images by utilizing 3D Point Cloud Reconstruction. This technique involves the use of a 3D Point Cloud, which is a set of points represented by their coordinates (x, y, z) in 3D space [1],[2].





Reprojection Loss Analysis

Image	Reprojection Loss
1	2.55
2	4.34
3	3.65
4	4.32
5	3.15
Average	3.60

2. Data Processing



Use Blender's Python API to capture 2D images of the model and convert the model's format from VRM to PLY.

Use Open3D to convert the mesh into a 3D Point Cloud.

Use ControlNet for pose Use Stable Video 3D to generate additional views of the image.

estimation.

3. Feature Extraction & Matching



4.Structure from Motion (SfM) and Multi-View Stereo (MVS)



The above table indicates that the results are based on a reprojection loss analysis, comparing the similarity between the 2D image and the 3D model from the same angle. The analysis shows that the average error rate is 3.6.



This project successfully demonstrates a method for generating 3D asset models from 2D images by utilizing techniques such as 3D Point Cloud Reconstruction, Multi-View Stereo, and Iterative Closest Point alignment. By combining tools like Blender's Python API, Open3D, Deep Point, and Deep Glue, we were able to reconstruct 3D models from 2D images. The reprojection error analysis, which compared the similarity between the 2D images and the reconstructed 3D models, revealed an average error rate of 3.6. This result highlights the potential of Al-driven approaches to simplify and accelerate 3D model creation, offering significant benefits to industries such as gaming, animation, and media. The proposed framework not only reduces the complexity of manual 3D modeling but also opens the door to further exploration of AI in automating 3D asset generation.

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5.Displaying The Model



Use reprojection loss analysis to compare the similarity between the 2D image and the 3D model from the same angle.

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References

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