

# AI-Driven Diffusion and LoRA Models for customizable 3D Room Visualization and Design Enhancement

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## Introduction

Existing 3D design tools lack real-time customization and are not user-friendly. This project uses **Diffusion Models** for room generation and **LoRA** for interactive refinement, enabling intuitive, multimodal 3D room design from text, sketches, or mood boards. Previous tools focused on static AR/VR visualizations but lacked real-time customization and support for multimodal user input—key gaps this project addresses.

## Motivation

To make 3D room design faster, more intuitive, and accessible by replacing complex tools with an AI-driven system that supports real-time, personalized customization.

## Scope of the Project

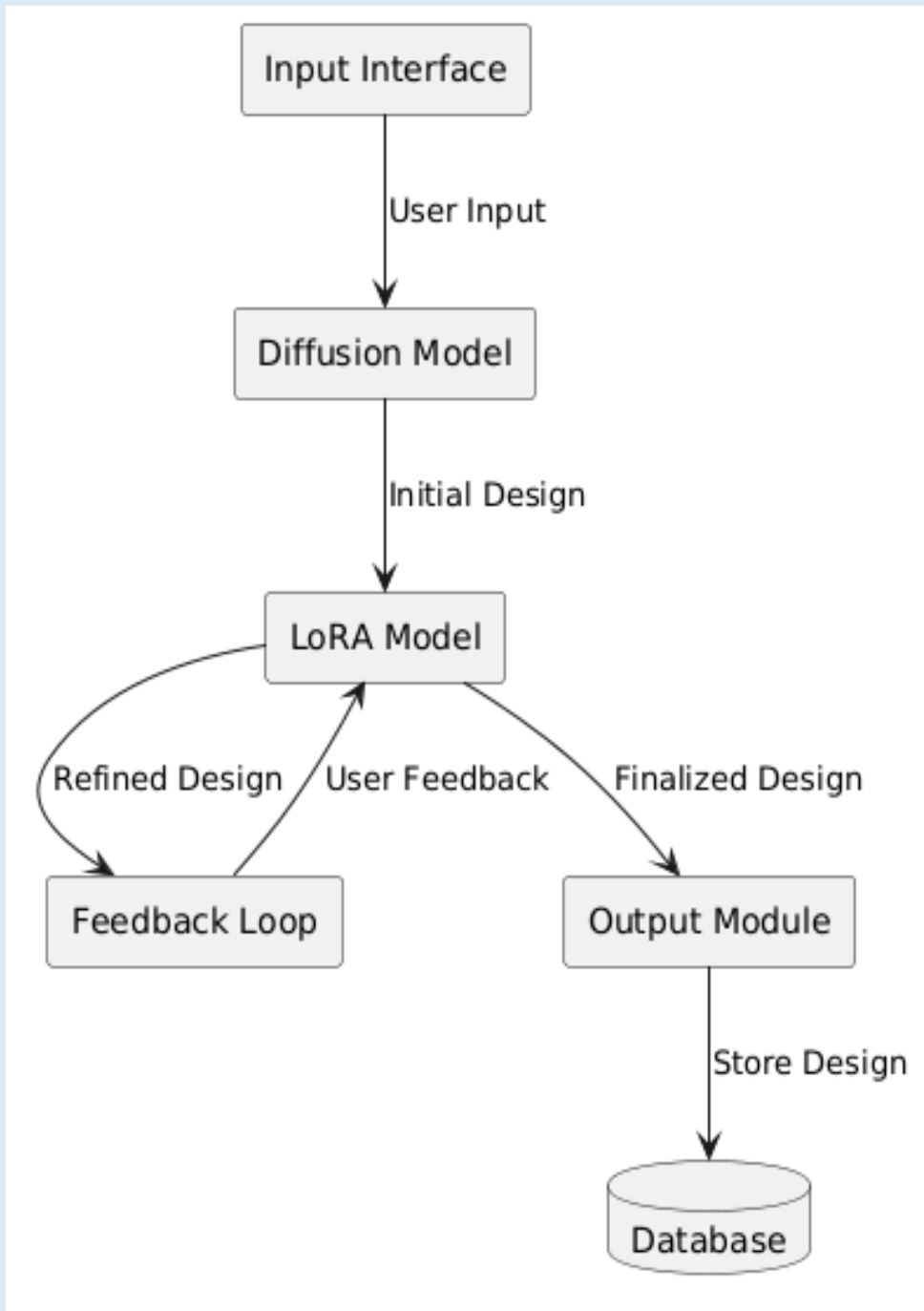
An AI-powered 3D modeling tool designed for ease of use and accessibility.  
**Key Features:**

- Text, sketch & mood board input
- Diffusion & LoRA models for design & customization
- Photorealistic 3D output with iterative feedback

**Tech:** PyTorch-based, optimized for performance  
**Deliverables:** Prototype, docs, performance report, future roadmap  
**Not Included:** CAD integration, full-scale architecture, real-time VR

Aiming to make 3D interior design intuitive and accessible for all.

## Methodology



System Architecture of the AI-Driven 3D Modeling Tool

This project implements a modular, AI-driven pipeline for generating and customizing 3D room visualizations based on multimodal user inputs.

### 1. Input Acquisition

The system accepts:

- Textual descriptions** processed via NLP to extract design semantics.
- Sketches**, analyzed using OpenCV for spatial structure detection.
- Mood boards**, evaluated using CNN-based feature extraction to infer stylistic themes.

### 2. Diffusion-Based 3D Generation

A pretrained **diffusion model** converts structured input into a preliminary 3D room layout. This includes basic architectural components, textures, and furniture placement, synthesized through iterative denoising.

### 3. LoRA-Based Customization

The **Low-Rank Adaptation (LoRA) module** allows real-time, user-driven modifications such as furniture replacement, material changes, or lighting adjustments. This process is non-destructive and preserves design context.

### 4. Iterative Refinement

User feedback is integrated through an iterative loop, enabling granular control and progressive enhancement of the 3D output. This supports adaptive co-creation and design personalization.

### 5. Rendering & Output

Finalized scenes are rendered photorealistically using high-resolution visualization techniques. Outputs are exportable in common formats (e.g., PNG, FBX) for integration with external tools or presentations.

## Results

### 1. Design Quality (DQ)

Users evaluated the generated 3D room layouts based on realism, spatial coherence, and aesthetic alignment with input prompts.

- Average User Rating: 9.2 / 10
- Users highlighted strengths such as:
  - Lighting realism
  - Accurate material rendering
  - Consistency of spatial layout

The architectural coherence, including logical furniture placement, proportional room dimensions, and natural traffic flow, contributed to the realism and usability of the outputs.

### 2. Customization Accuracy (CA)

Customization accuracy was assessed by comparing user-specified modifications with the system's updates.

- Accuracy Score: 96%
- Customization types tested:
  - Furniture repositioning
  - Color and material updates
  - Lighting adjustments

The system successfully applied changes without regenerating the entire scene and preserved unrelated design elements.

### 3. Performance

Performance was measured through generation and refinement times across various hardware configurations.

- Initial Layout Generation: ~10 seconds
- Iterative Refinement: < 5 seconds

These results validated the system's capability for real-time interaction while maintaining design quality.

### 4. User Experience and Satisfaction

Usability testing showed high satisfaction across multiple parameters:

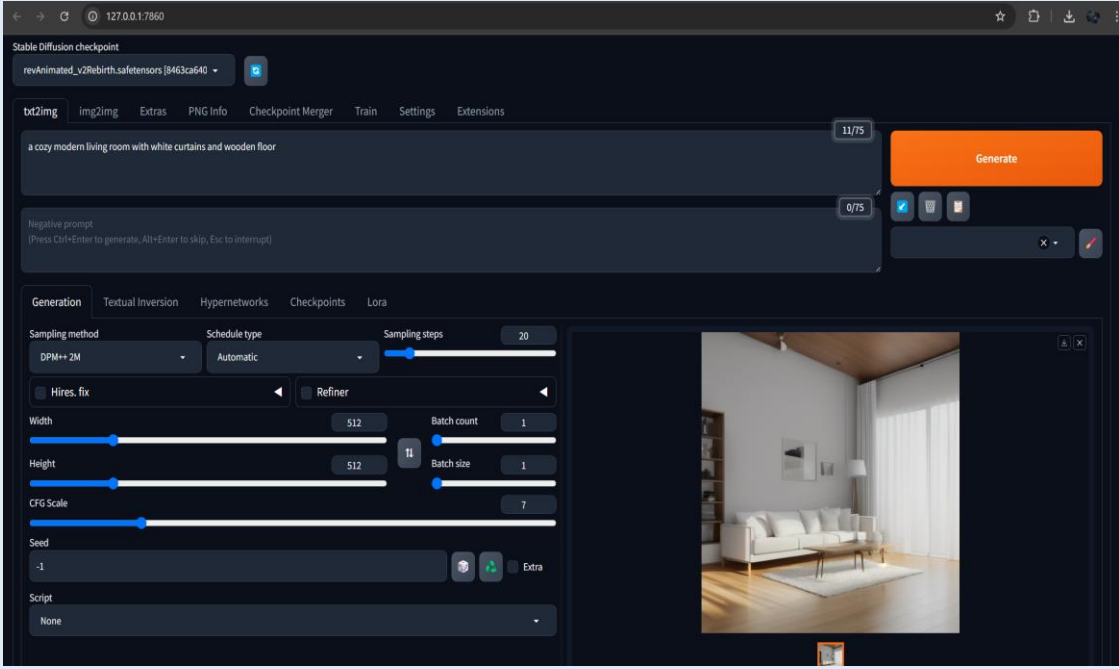
- Ease of input through text, sketch, and mood boards
- Real-time feedback loop supported by LoRA
- Positive reception among both professionals and non-professionals

Qualitative feedback affirmed the system's intuitive design and ability to translate user vision into photorealistic outputs.

## Sample Outputs from Project Demonstration

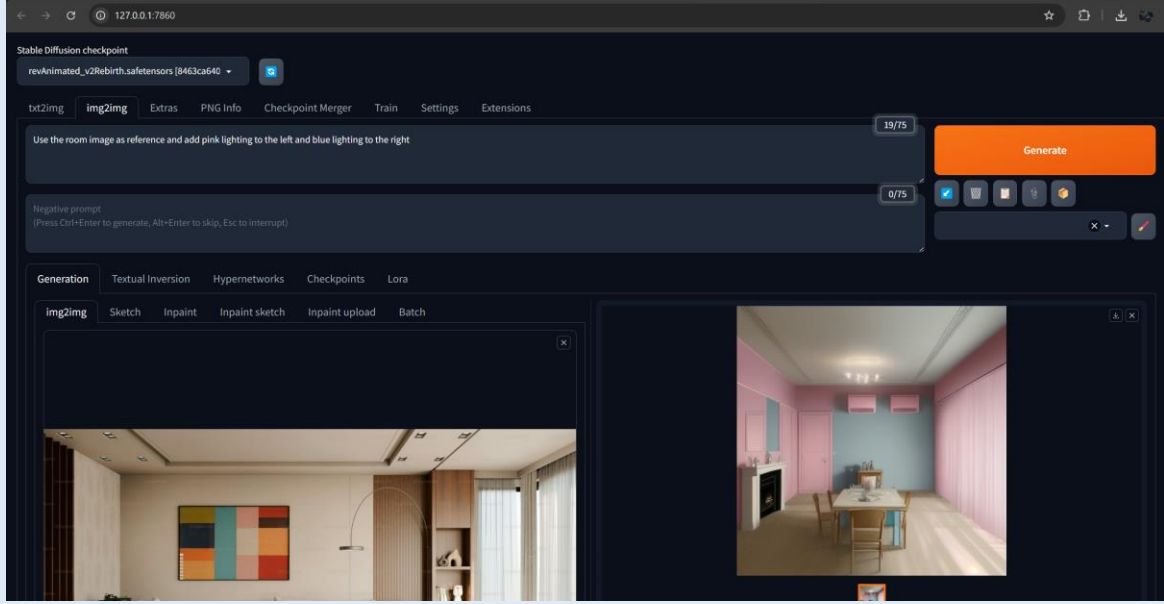
### 1. Text-to-Image (txt2img)

**Prompt:** "A cozy modern living room with white curtains and wooden floor"



Screenshot of the Web UI with text prompt and resulting image shown in txt2img tab.

### 2. Image-to-Image (img2img)



Screenshot of the img2img tab with input image and generated output

## Conclusion

Our AI-powered system successfully bridges creativity and technology—delivering fast, realistic, and user-customizable 3D room designs from simple text or sketches. With **96% customization accuracy** and **9.2/10 design satisfaction**, the results strongly validate our approach.

**Future scope** includes immersive VR integration, voice/gesture input, and real-time collaboration—paving the way for smarter, more inclusive spatial design experiences.

## References

1. Patel, R., & Mandekar, A. (2024). *Augmented Reality in Interior Design: Enhancing Client Engagement Through Real-Time Visualization*. International Journal of AR Applications.

2. Dunston, P. S., Arns, L. L., McGlothlin, J. D., Lasker, G. C., & Kushner, A. G. (2011). *An Immersive Virtual Reality Mock-Up for Design Review of Hospital Patient Rooms*. Journal of Building Performance Simulation, 4(3).