

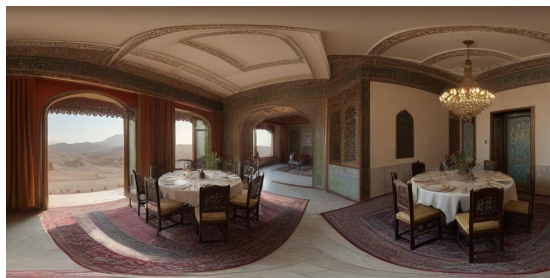
# Depth-Controlled-Spherical-Panorama- Generation

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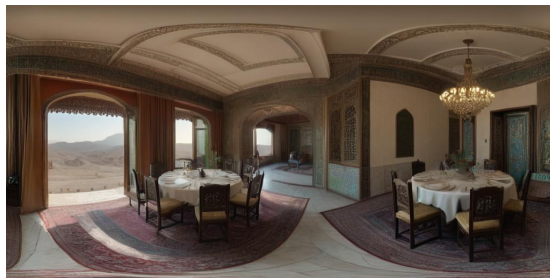
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## 1 Introduction

This project demonstrates the generation of text-controlled 360-degree panoramas using Stable Diffusion. I have implemented seam-tiling to achieve coherent panoramas and also experimented with depth conditioning using a ControlNet. In this study, I compare the results of panoramas with and without depth conditioning and seam-tiling.



(a) A non coherent panorama with inconsistent seams



(b) A coherent panorama: The right and left ends match seamlessly.

Figure 1: Comparison of panoramas with and without seam-tiling

## 2 Requirements

To set up this project, run the following commands:

```
$ pip install diffusers transformers accelerate safetensors huggingface_hub  
$ git clone https://github.com/replicate/cog-sdxl cog_sdxl
```

### 2.1 Approach

In this section, I describe the approach that I used to solve both tasks. I describe the method to obtain a 360 panorama, followed by performing seam correction and depth conditioning.

### 2.2 Generating a Prompt Conditioned Panorama

For generating 360-degree panoramas, I utilized the LoRA weights from [this model](#). This is an XL Stable diffusion model that is tuned to generate 360 panorama images when referred to the T0k token.

### 2.3 Seam Correction

Initially, the seams of these panoramas were not coherent. To address this, I rolled the panorama by half its width, placing the incoherent transition in the centre. Then, by masking and inpainting, the center using XL Stable Diffusion, a coherent panorama was achieved.



Figure 2: Fig(a): An image generated by rolling over the pixels by half the width. The centre has an inconsistent transition.

Fig(b): The coherent panorama generated by masking and inpainting the central region.

### 2.4 Generating a Prompt and Depth Conditioned Panorama

For depth-conditioned panoramas, I used the [ControlNet pipeline](#). Similar to the previous method, I corrected the non-coherent seams through rolling and inpainting.

### 3 Results

I have tested this methodology with the following prompts:

1. 'A dining room in Persia'
2. 'Inside Bag End'
3. 'Inside the cantina in Star Wars'
4. 'Inside a bedroom in France'
5. 'Inside a medieval prison'

Below are the results for each prompt:



Figure 3: Dining Room in Persia

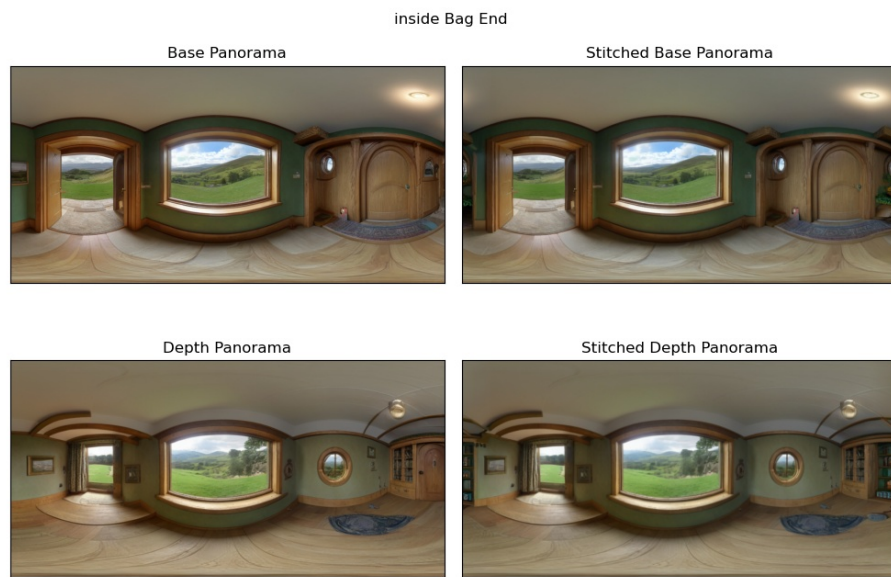


Figure 4: Inside Bag End



Figure 5: Inside the Cantina in Star Wars

inside a bedroom in France



Figure 6: Inside a bedroom in France

Inside a medieval prison

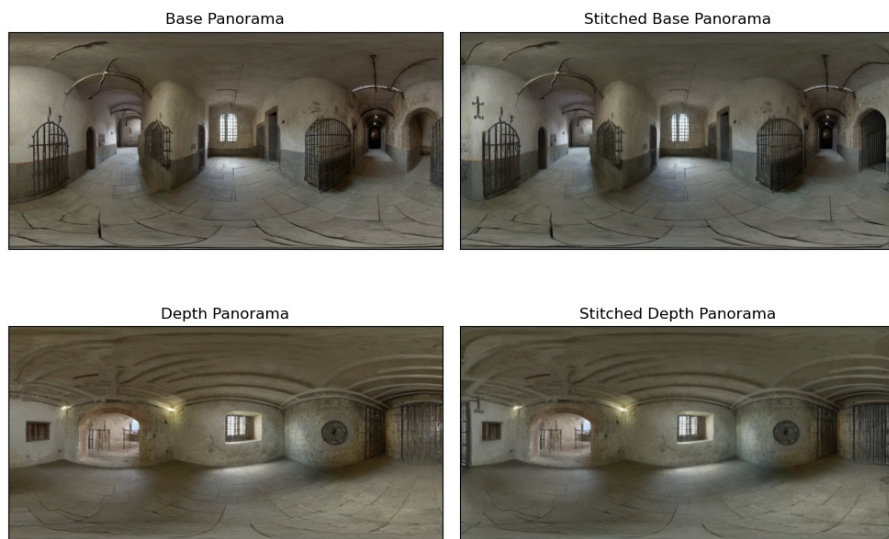


Figure 7: Inside a medieval prison

## 4 Usage of Code

To run the code, execute the following command:

```
python3 main.py
```

`main.py` contains the prompts, which can be edited as needed.

- `part1.py`: Contains `create_pano` function to generate panoramas using a prompt.
- `part2.py`: Contains `create_pano_depth` function for generating panoramas using prompt and depth information.
- `cleaner.py`: Contains `clean_pano` function for seam-stitching.

## 5 Rendering Panoramas

The generated images are stored in the `images/` directory. Use images from `images/raw/` to render panoramas at [360 Panorama Web Viewer](#).

The following are the (**prompt only**) coherent panoramas that can be rendered in the Web viewer:

1. [Dining Room in Persia](#)
2. [Inside Bag End](#)
3. [Inside the Cantina Star Wars](#)
4. [Inside a bedroom in France](#)
5. [Inside a medieval prison](#)

The following are the **depth-conditioned panoramas**:

1. [Dining Room in Persia](#)
2. [Inside Bag End](#)
3. [Inside the Cantina Star Wars](#)
4. [Inside a bedroom in France](#)
5. [Inside a medieval prison](#)