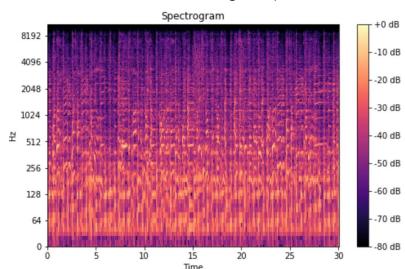
Images that Sound

Thanmaya Pattanashetty

Background: Spectrograms

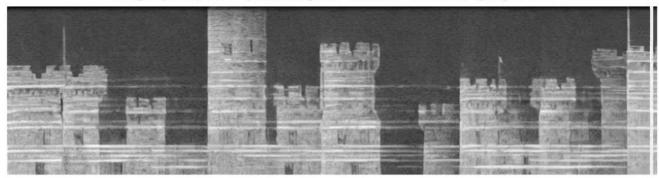
- Spectrograms: Visual representations of sound.
- Used in audio machine learning to depict sound features like frequency and amplitude.



Overview

- Goal: Create **images that sound**—spectrograms that are meaningful as both **images** and **audio**.
- Combines **text-to-image** and **text-to-spectrogram** diffusion models (Stable Diffusion & Auffusion).
- Opens new possibilities for multimodal art and audio-visual learning.

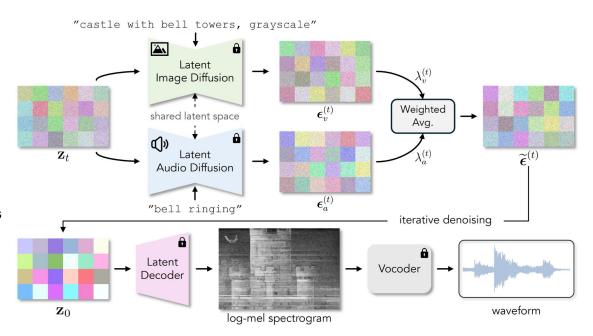
Image prompt: a painting of castle towers, grayscale



Audio prompt: bell ringing

Methods

- Diffusion Models: Iterative denoising to generate both images and sounds.
- Multimodal Denoising: Combines audio and image diffusion models using shared latent space.



Results

- Metrics: Evaluated using CLIP (image quality) and CLAP (audio quality).
- **Human Study**: Participants preferred the authors' method for both **visual** and **audio** quality.
- **Examples**: E.g., a **castle** that looks like bell towers and sounds like **bells ringing**.

Limitations

- Cannot achieve **high-fidelity audio** and **high-quality visuals** at the same time all the time.
- Depends on well-crafted **prompts** for optimal results.
- Some visual and audio prompts do not work well together.

Conclusion

- Introduces images that sound using diffusion models.
- Potential in art, cross-modal learning, and audio-visual applications.
- Future improvements in **audio model quality** and multimodal interactions.

References

- Chen, Ziyang, Daniel Geng, and Andrew Owens. "Images That Sound: Composing Images and Sounds on a Single Canvas." *arXiv*, version 1, 20 May 2024, https://arxiv.org/pdf/2405.12221.
- Das, Saptarshi. "Understanding the Mel Spectrogram." Medium, 9 July 2020, https://medium.com/analytics-vidhya/understanding-the-mel-spectrogram-fca2afa2ce53.