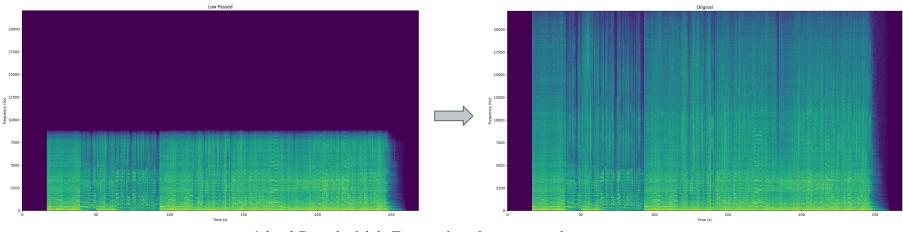
GAN-Based Bandwidth Extension for Music

**Cassius Close** 

# What is Bandwidth Extension (BWE)?

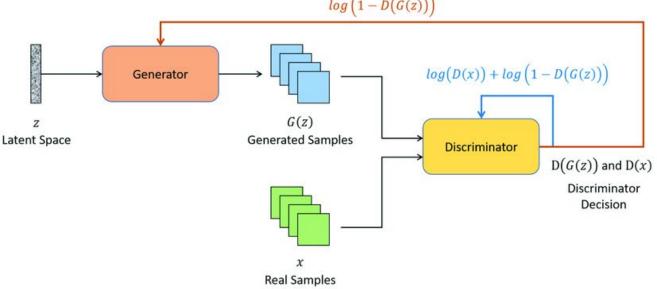
- Guess high frequency content from a low-resolution signal
- Applications
  - Telephone systems
  - Old recordings w/ missing high-freq content (music & speech)



Ideal Bandwidth Extension for a sample pop song

## **Generative Adversarial Networks**

- Generator tries to maximize loss, discriminator tries to minimize loss
- More detailed output

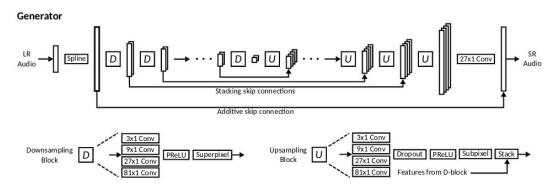


log(1-D(G(z)))

https://www.researchgate.net/figure/Typical-Generative-Adversarial-Networks-GAN-architecture\_fig2\_349182009

## **Existing Methods**

- Three recent methods work on music
  - All use U-Net architecture
- Recently, a HiFi-GAN based approach (BWE Is All You Need) has good results for speech
  - Uses WaveNet-based architecture for the generator



Kim, Sung, and Visvesh Sathe. "Bandwidth Extension on Raw Audio via Generative Adversarial Networks." arXiv, March 21, 2019. http://arxiv.org/abs/1903.09027.

## My Method

• Try to apply the WaveNet architecture (dilated convolution) to BWE

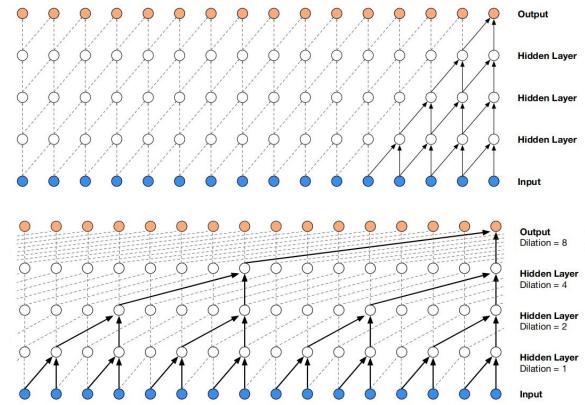
• Mono input signal

• Using the time domain (phase is implicit)

• Upsample from 16kHz to 44.1kHz

## **Methods - Generator**

- Dilated convolution layers
  - similar to WaveNet architecture
  - $\circ$  Non-causal

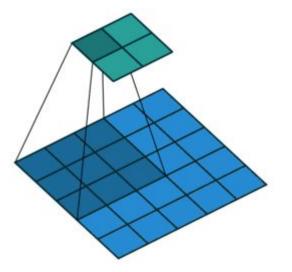


https://www.deepmind.com/blog/wavenet-a-generative-model-for-raw-audio

## **Methods - Discriminators**

- Strided convolutions
  - Reduce size of input

- Discriminator output is average value of last layer
  - Large number if real, small number if fake



## **Methods - Losses**

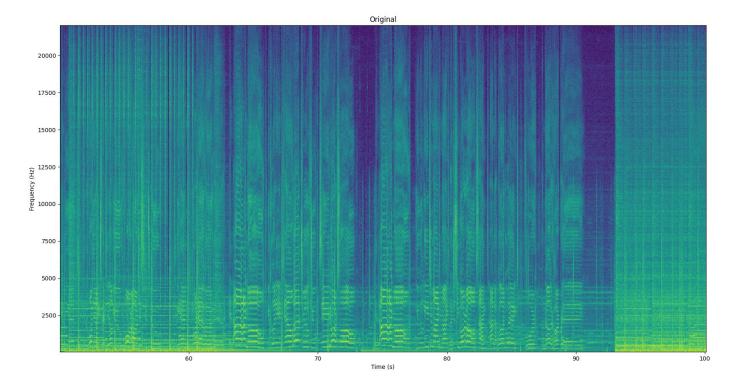
- Training Discriminators
  - Loss ~= Discriminator(fake) + (1 Discriminator(real))
- Training Generators
  - Adversarial loss
    - Loss ~= (1 Discriminator(fake))
    - Spectrogram discriminators
    - 3 waveform discriminators (1x, 2x, and 4x downsampled)
  - L1 waveform loss
  - L1 spectrogram loss

## **Objective Results**

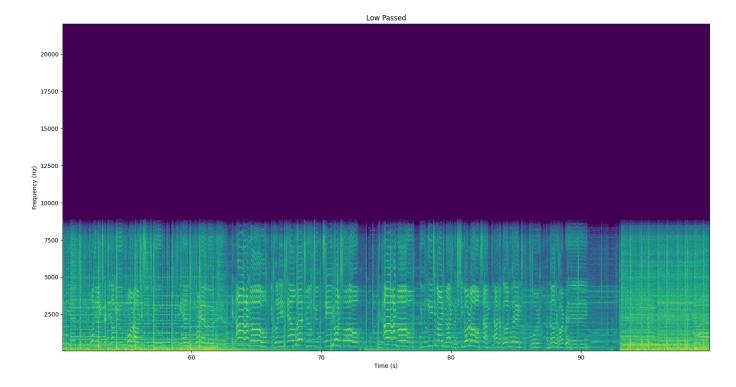
- Was looking at Frechet Audio Distance (FAD), but it's only trained with 16kHz data.
- Not accurate for perceptual quality

	Generator Only	Waveform Discriminators	Full Model
Signal-to-Distortion Ratio (SDR)	13.91 dB	15.35 dB	14.87 dB
Signal-to-Noise Ratio (SNR)	8.44 dB	7.45 dB	8.45 dB
Log-Spectral Distortion (LSD)	3.65 dB	2.64 dB	2.59 dB

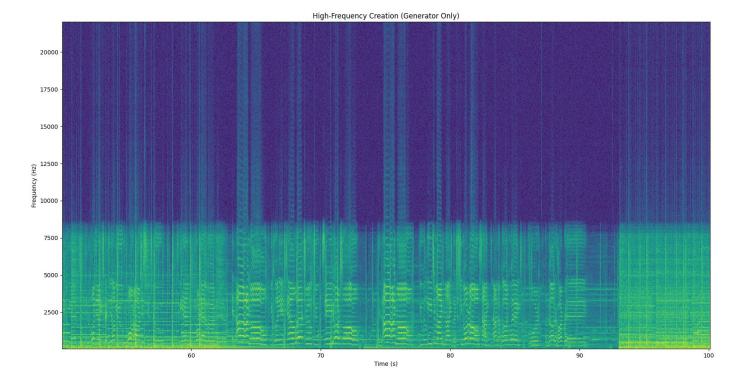
#### **Visual Results: Original Spectrogram**



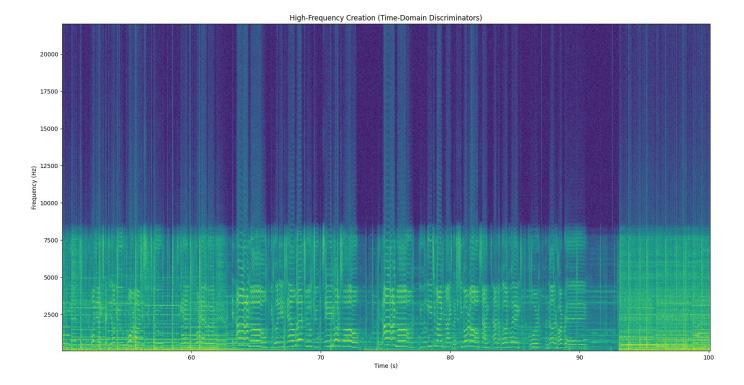
## **Visual Results: Input**



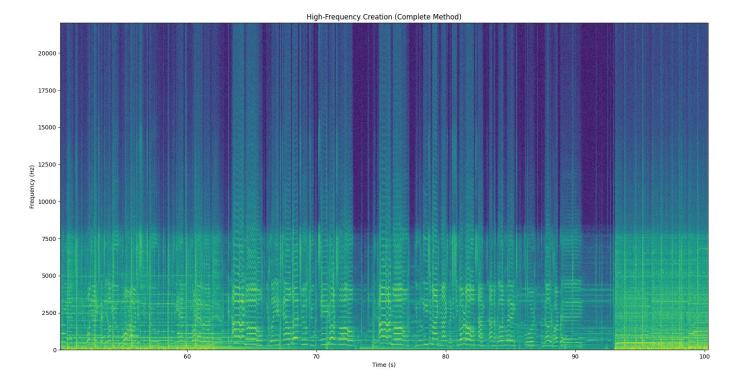
#### **Visual Results: Generator Only**



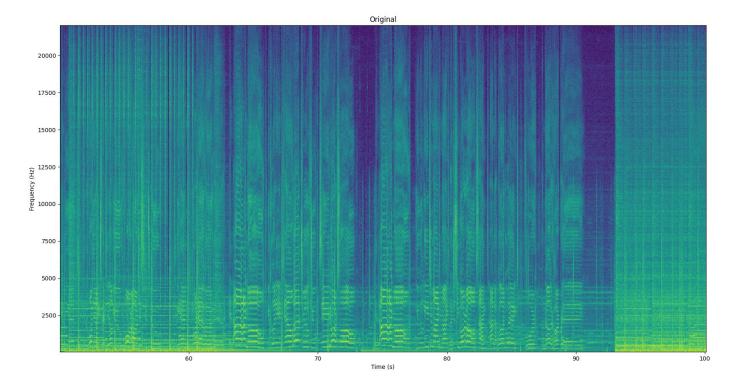
#### **Visual Results: Waveform Discriminators**



#### **Visual Results: Waveform + Spect. Discriminators**



#### **Visual Results: Original Spectrogram**



## **Aural Results**

Original:

Input:

Waveform + Spectrogram Discriminators:

Waveform Discriminators:

## **Conclusions/Limitations**

• Results are passable, but certainly not wonderful

- Difficult domain, could have used more data
- Memory limitations (greater batch sizes, more layers)

