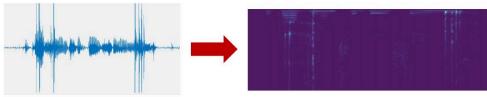


Evaluating Classical Image Denoising for Speech Enhancement

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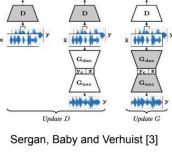
Motivation

- Speech enhancement (SE) is an important task required for the success of technologies like VOIP and speech recognition
- Many SE algorithms act on signals in the time-frequency domain (spectrogram) to take advantage of underlying structure
- With the problem transformed into the denoising of a 2D image, techniques like CNN denoisers can be used
- If classical, non-data-driven denoising methods can be applied to this domain of images, they could offer time and memory savings over deep neural networks



Related Work

- R-CED architecture is an early example of a CNN used for spectral SE [1]
- Alternate approach predicts "noise mask" from spectrum instead of mapping directly from noisy to clean spectrum [2]
- Time-domain methods like Srgan [3] address lack of phase information in magnitude spectrum

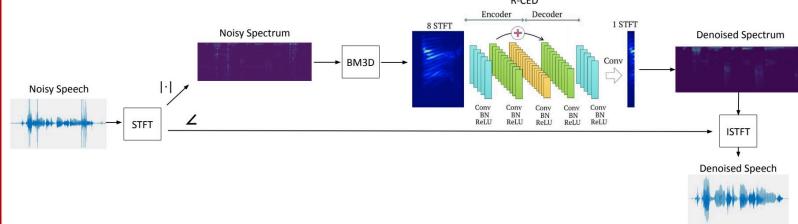


References

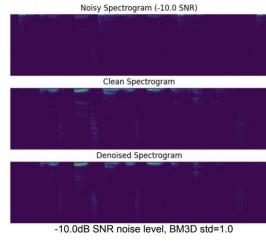
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Method

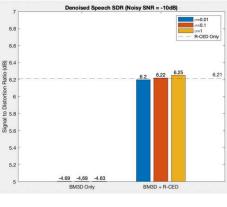
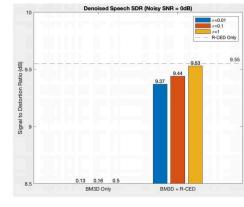
- Train R-CED [1] on speech spectrum preprocessed with classical denoising algorithm (BM3D [4]) to produce denoised spectrum
- MS-SNSD dataset [5] used for training on various noise types at SNR of 0dB and -10dB
- Compare performance against applying BM3D only or R-CED only to noisy spectra
- Experimental std estimates for BM3D chosen as $\sim(0.1x, 1x, 10x)$ average spectrum pixel value



Experimental Results



-10.0dB SNR noise level, BM3D std=1.0



- BM3D preprocessing enhances CNN performance at higher noise levels, potentially enabling the use of smaller networks
- However, BM3D incurs additional computational cost for minimal performance gains
- As an idea for future work, perhaps a simpler preprocessing method like local filtering could have similar benefits without the complexity of BM3D