

BAND FILTER ASSISTED U-NET DESIGNS FOR DENOISING TRAFFIC SOUNDS IN MUSICAL RECORDINGS

Introduction

Modern deep learning models such as U-Net architectures have shown strong performance in audio restoration tasks but often suffer from high computational cost, limiting their practicality for users with constrained hardware. A combination of band filter and U-Net model is proposed to reduce the computational requirements of the U-Net while preserving denoising quality. Additionally, a composite loss function is also introduced that allows for the model to train more effectively on traffic noise.

U-Net Loss Function Design

- MAE treats all errors linearly
- MSE disproportionately penalizes larger errors
- Composite loss as a weighted sum of MAE and MSE losses
- · Weights are adjustable, add to 1.

Training Result

NMSE provides an objective measure of how much the desired signal remains relative to residual noise after denois-ing, with a lower NMSE indicating better preservation of the clei

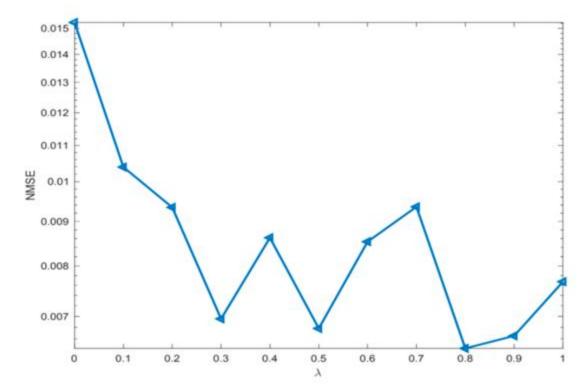


Fig. 2. Normalized Mean Squared Error (NMSE) versus λ in the composite loss function.

Hybrid U-Net Based Music Denoising Filter

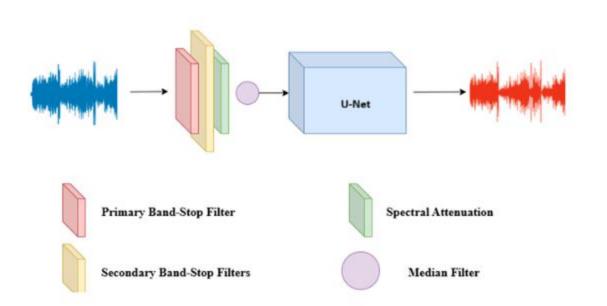


Fig. 1. Hybrid U-Net Based Music Denoising Filter

Filtering pipeline design

- · Broad band-stop filtering.
- narrowband notch filtering.
- STFT-based spectral suppression.
- · Weighted blending strategy.

Bio:

- Brady Zhang is a senior at Chattahoochee High School
- 5 time AIME Qualifier, Perfect Score on 2021 AMC 8, 2023 All-State Concert Band Bassoonist, 6-time GMTA Piano Solo Finalist
- President of the CHS Math Team and Band, Vice President of Math Honor Society and Physics Olympiad Team, and participates in community service activities including tutoring and musical performances.





