

SICRN: Advancing Speech Enhancement through State Space Model and Inplace Convolution Techniques

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Introduction

- CRN typically employs consecutive downsampling and upsampling convolution for frequency modeling, which destroys the inherent structure of the signal over frequency.
- Hao et al. [1] propose FullSubNet, due to the introduction of full-band and sub-band model, these methods lead to a large number of parameters, increasing the complexity of the model.
- We propose an innovative module combing a State space model and Inplace Convolution [2] (SIC), and to replace the conventional convolution in CRN, called SICRN.
- The experimental results demonstrate notably high evaluation scores. Notably, this achievement is attained with less than 1/2 the parameters and 1/7 the computational complexity of FullSubNet.



Fig. 1: Overview of the proposed SICRN system

Experimental Results

Table 1: The performance in terms of WB-PESQ [MOS], NB-PESQ [MOS], STOI [%], and SI-SDR [dB] on the DNS challenge test dataset.

Method	#Para	Look Ahead		With Reve	erb		Without Reverb					
	(M)	(ms)	WB-PESQ	NB-PESQ	STOI	SI-SDR	WB-PESQ	NB-PESQ	STOI	SI-SDR		
Noisy	-	-	1.822	2.753	86.62	9.033	1.582	2.454	91.52	9.071		
NSNet [20]	5.1	0	2.365	3.076	90.43	14.721	2.145	2.873	94.47	15.613		
DTLN [21]	1.0	-		2.700	84.68	10.530	-	3.040	94.76	16.340		
Conv-TasNet [22]	5.08	33	2.750	-	-	-	2.730	-	-	-		
DCCRN-E [23]	3.7	37.5	-	3.077	-	-	-	3.266	-	-		
PoCoNet [24]	50	-	2.832	-	-	-	2.748	-	-	-		
FullSubNet [11]	5.64	32	2.969	3.473	92.62	15.750	2.777	3.305	96.11	17.290		
SICRN	2.16	0	2.891	3.433	92.59	15.137	2.624	3.233	95.83	15.998		

 Table 2: Comprehensive comparison with FullSubNet.

Table 3: Ablation experiment(With Reverb).

Table 4: Ablation experiment(Without Reverb).

				Method	WB-PESQ	NB-PESQ	STOI	SI-SDR	Method	WB-PESQ	NB-PESQ	STOI	SI-SDR
Method	#Para(M)	MACs(G/s)	Look Ahead(ms)	mixture	1.822	2.753	86.62	9.033	Noisy	1.582	2.454	91.52	9.071
FullSubNet	5.64	30.84	32	IICRN	2.797	3.378	91.71	14.929	IICRN	2.596	3.218	95.56	15.795
SICKI	2.10	4.24	U	SICKN	2.891	3.433	92.59	15.13/	SICRN	2.624	3.233	95.83	15.998

Replace the S4ND block in the SIC block with inplace convolution and name it IICRN.

Conclusion

- We propose an innovative model combing a state space model and inplace convolution, called SICRN.
- This network avoids downsampling operations throughout its architecture.
- The experimental results demonstrate superior performance achieved with fewer parameters and computational resources.

References

1. Hao X, Su X, Horaud R, et al. Fullsubnet: A full-band and sub-band fusion model for real-time single-channel speech enhancement[C]// ICASSP 2021-2021 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP). IEEE, 2021: 6633-6637. 1. Liu J, Zhang X. Iccrn: Inplace cepstral convolutional recurrent neural network for monaural speech enhancement[C]//ICASSP 2023-2023 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP). IEEE, 2023: 1-5.