### A Training Framework for Stereo-aware Speech Enhancement using Deep Neural Networks





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Propose a stereo-aware speech enhancement training framework.

1. Preserve the stereo image while performing speech enhancement.

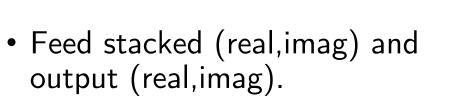
2. Evaluate perceptual enhancement through subjective tests.

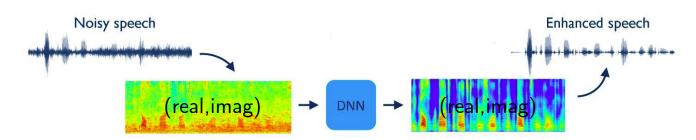
### Mono Speech Enhancement



Enhanced speech

• Enhance spectrogram and add mixture phase at the output.





Phase

DNN

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Image from https://acousticsresearchcentre.no/speech-enhancement-with-deep-learning/

spectrogram

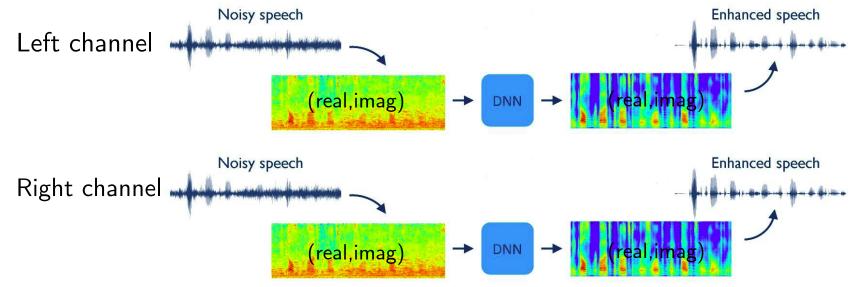
Noisy speech

spectrogram

## Stereo Speech Enhancement (LRindp)



Train one mono network and feed L/R independently.



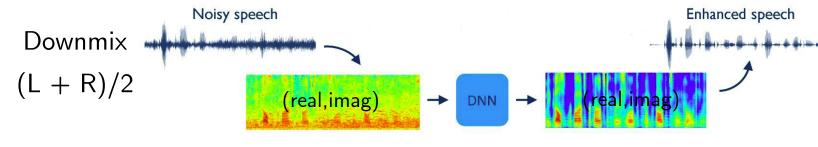
Drawbacks:

- Channel coherence info is completely ignored.
- Inference time is approximately doubled.

## Stereo Speech Enhancement (downmix)



Train using downmix.



Prediction:

- Enhance downmix.
- Add phase difference between mixture stereo and enhanced downmix.

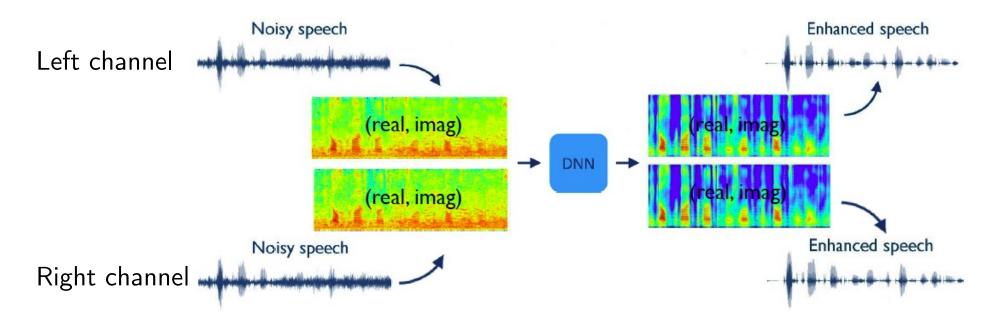
Drawbacks:

• Added noisy phase at prediction time is not optimal.

## Stereo Speech Enhancement (end-to-end)



End-to-end stereo input stereo output.



No guarantee to preserve the stereo image.

### Stereo-aware Training



Training loss

$$\mathcal{L}(\mathbf{s}, \hat{\mathbf{s}}) = \mathcal{L}_{speech-rec}(\mathbf{s}, \hat{\mathbf{s}}) + \mathcal{L}_{image-pres}(\mathbf{s}, \hat{\mathbf{s}})$$

Speech reconstruction

Stereo image preservation

# $\mathcal{L}_{\text{speech-rec}}(\mathbf{s}, \hat{\mathbf{s}}) = \text{LSD}(\mathbf{s}, \hat{\mathbf{s}}) + \alpha_{\text{TL}} \text{ TL}(\mathbf{s}, \hat{\mathbf{s}})$

Speech Reconstruction Loss

Log spectral distortion

$$LSD(\mathbf{s}, \hat{\mathbf{s}}) = \frac{1}{2T} \sum_{c=1}^{2} \sum_{t=1}^{T} \sqrt{\frac{1}{F} \sum_{f=1}^{F} \left( g(\mathbf{S}_{c}[t, f]) - g(\hat{\mathbf{S}}_{c}[t, f]) \right)^{2}}$$
TL(s,  $\hat{\mathbf{s}}) = \frac{1}{2} \sum_{c=1}^{2} \sqrt{\frac{1}{T} \sum_{t=1}^{T} (\mathbf{s}_{c}[t] - \hat{\mathbf{s}}_{c}[t])^{2}}$ 



### Stereo Image Preservation Loss



$$\mathcal{L}_{image-pres}(\mathbf{s}, \hat{\mathbf{s}}) = \sum_{M \in \{IID, IPD, IC, OPD\}} \alpha_M \mathcal{L}_M(\mathbf{S}, \hat{\mathbf{S}})$$

Intensity  
IID<sub>b</sub>(**S**) = 10 log<sub>10</sub> 
$$\frac{\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_1[f] \mathbf{S}_1^*[f]}{\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_2[f] \mathbf{S}_2^*[f]}$$

Phase  

$$PD_b(\mathbf{S}) = \angle \left( \sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_1[f] \mathbf{S}_2^*[f] \right)$$

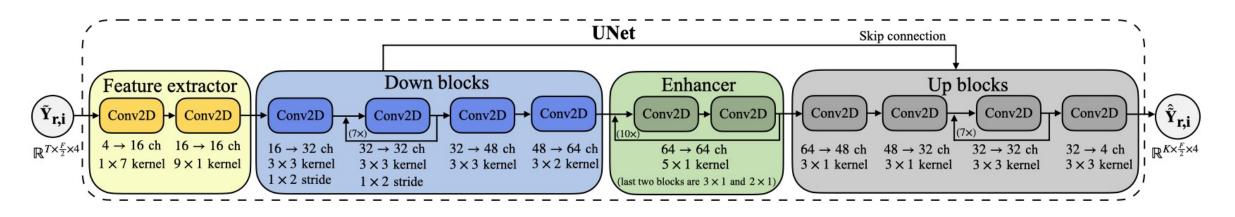
 $\begin{aligned} \text{Coherence} & |\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_1[f] \mathbf{S}_2^*[f]| \\ \text{IC}_b(\mathbf{S}) &= \frac{|\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_1[f] \mathbf{S}_2^*[f]]}{\sqrt{(\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_1[f] \mathbf{S}_1^*[f])(\sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}_2[f] \mathbf{S}_2^*[f])} \end{aligned}$ 

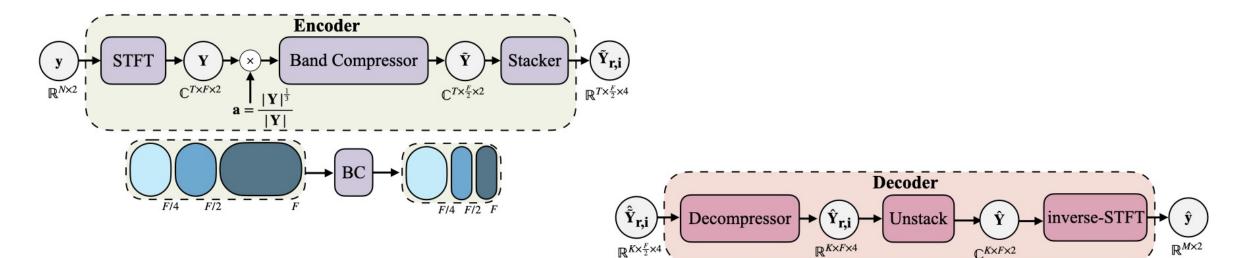
Overall phase  

$$OPD_b(\mathbf{S}, \hat{\mathbf{S}}) = \angle \left( \sum_{f=f_b}^{f_{b+1}-1} \mathbf{S}[f] \hat{\mathbf{S}}^*[f] \right)$$

### Network Architecture







#### Tolooshams and Koishida

#### A Training Framework for Stereo-aware Speech Enhancement using Deep Neural Networks

### Presence of Time Loss

• Higher SDR.

• Overall phase preservation.

0	$\operatorname{spec}$	spec-time	
0			-3
T frames			-2
T fr			-1
2003			
-1	B bands	B bands	32

STOLEN ST		Test set I						
Network	Method	Objective						
		SDR	POLQA	ĨD	IPD	IC	OPD	
	noisy	11.61	2.51	1.56	1.92	0.20	0.78	
1	downmix - spec	6.46	2.98	2.68	2.79	0.30	1.61	
	LRindp - spec	6.82	3.26	2.36	1.99	0.28	1.62	
	downmix - spec - time	10.10	2.95	2.39	2.78	0.29	1.40	
	LRindp - spec - time	12.89	3.31	2.42	1.92	0.27	1.27	
U-Net	stereo - spec - time	12.56	3.01	1.85	1.91	0.26	1.25	
U-INEL	stereo - spec - time - IID	14.17	3.33	1.55	1.76	0.35	1.42	
	stereo - spec - time - IPD	13.88	3.36	1.67	1.71	0.32	1.27	
	stereo - spec - time - IC	12.09	3.04	1.80	2.08	0.21	1.43	
	stereo - spec - time - OPD	14.05	3.33	1.86	2.10	0.23	0.99	
	stereo - spec - time - all	13.78	3.32	1.64	1.81	0.21	1.10	
U-NetCM	stereo - spec	6.28	3.34	2.24	2.14	0.25	2.48	
	stereo - spec - time - all	15.02	3.28	1.96	1.93	0.24	1.05	



### Mono to Stereo

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	Network	Network Method		Objective					
	12000000000		SDR	POLQA	ĬID IPI	) IC	OPD		
<ul> <li>Stereo preserves IID better than LRindp.</li> <li>LRindp has higher SDR and POLQA.</li> </ul>		noisy	11.61	2.51	1.56 1.9	2 0.20	0.78		
than LRindp.		downmix - spec	6.46	2.98	2.68 2.7	0.30	1.61		
		LRindp - spec	6.82	3.26	2.36 1.9	0.28	1.62		
		downmix - spec - time	10.10	2.95	2.39 2.7	3 0.29	1.40		
	U-Net	LRindp - spec - time	12.89	3.31	2.42 1.9	2 0.27	1.27		
		stereo - spec - time	12.56	3.01	1.85 1.9	0.26	1.25		
		stereo - spec - time - IID	14.17	3.33	1.55 1.7	5 0.35	1.42		
		stereo - spec - time - IPD	13.88	3.36	1.67 1.7	0.32	1.27		
• LRINDP has higher SDR and		stereo - spec - time - IC	12.09	3.04	1.80 2.0	3 0.21	1.43		
POLOA		stereo - spec - time - OPD	14.05	3.33	1.86 2.1	0.23	0.99		
		stereo - spec - time - all	13.78	3.32	1.64 1.8	0.21	1.10		
	U-NetCM	stereo - spec	6.28	3.34	2.24 2.14	1 0.25	2.48		
		stereo - spec - time - all	15.02	3.28	1.96 1.9	3 0.24	1.05		

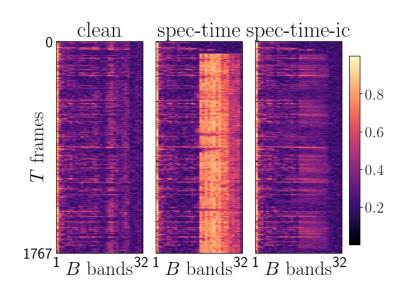


Test set I

### Image Preservation Loss



- Higher POLQA and SDR.
- IID to improve SDR.
- IPD to increase POLQA.



100000000000000000000000000000000000000		Test set I						
Network	Method	Objective						
		SDR	POLQA	ĪID	IPD	IC	OPD	
	noisy	11.61	2.51	1.56	1.92	0.20	0.78	
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### Subjective Evaluation



### MUSHRA test.

			Test set I								
	Network	work Method		Objective					Subjective		
			SDR	POLQA	IID	IPD	IC	OPD	OVRL	IMG	
		noisy	11.61	2.51	1.56	1.92	0.20	0.78	0	0	
Annual 2750 listonaus	88 - 95.	downmix - spec	6.46	2.98	2.68	2.79	0.30	1.61	х	х	
Approx. 2,750 listeners.		LRindp - spec	6.82	3.26	2.36	1.99	0.28	1.62	х	х	
		downmix - spec - time	10.10	2.95	2.39	2.78	0.29	1.40	0.34	0.30	
	U-Net	LRindp - spec - time	12.89	3.31	2.42	1.92	0.27	1.27	0.42	0.35	
		stereo - spec - time	12.56	3.01	1.85	1.91	0.26	1.25	0.38	0.37	
		stereo - spec - time - IID	14.17	3.33	1.55	1.76	0.35	1.42	0.45	0.41	
		stereo - spec - time - IPD	13.88	3.36	1.67	1.71	0.32	1.27	0.63	0.46	
Overall quality (OVRL).		stereo - spec - time - IC	12.09	3.04	1.80	2.08	0.21	1.43	0.31	0.37	
		stereo - spec - time - OPD	14.05	3.33	1.86	2.10	0.23	0.99	0.42	0.49	
		stereo - spec - time - all	13.78	3.32	1.64	1.81	0.21	1.10	0.45	0.43	
	U-NetCM	stereo - spec	6.28	3.34	2.24	2.14	0.25	2.48	Х	Х	
	U-INCICIVI	stereo - spec - time - all	15.02	3.28	1.96	1.93	0.24	1.05	х	Х	

# Stereophonic image quality (IMG).

### Model Independence



Proposed stereo-aware training improves SDR and preserves stereo image (e.g., IID, IPD, IC, and OPD) independent of the network architecture.

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	noisy	11.61	2.51	1.56	1.92	0.20	0.78	
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### Session: SPE-33: Speech Enhancement: Training Schemes and Losses.

### Tuesday, 10 May, 20:00 – 20:40 (Singapore Time, UTC +8)

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