

# Echo-aware signal processing for audio scene analysis

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Diego DI CARLO  
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**PhD supervisors:** Antoine DELEFORGE  
Nancy BERTIN

**Jury members:** Laurent GIRIN (reviewer - president)  
Simon DOCLO (reviewer)  
Fabio ANTONACCI (EXAMINER)  
Renaud SEGUIER (EXAMINER)

Université de Rennes 1, IRISA/INRIA, Panama research group



## Echo-aware Application

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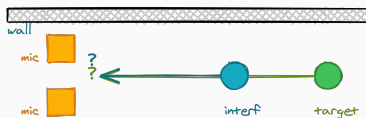


Echoes = same content, different time/direction

Image Source Model



Image Microphone Model



Recent literature on echo-aware processing:

## What?

Echoes = repetitions

- Sound Source Separation  
[Leglaive et al., 2016]
- Speech Enhancement  
[Flanagan et al., 1993,  
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## Where?

Echoes  $\leftarrow$  image

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[Ribeiro et al., 2010,  
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- Room Geometry Estimation

## How?

Echoes  $\in$  sound propagation

- Blind Channel Estimation  
[Lin et al., 2007,  
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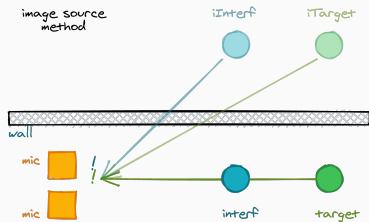


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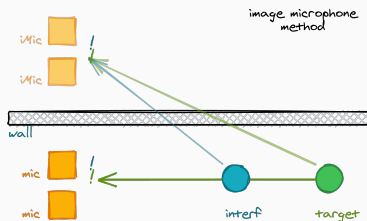


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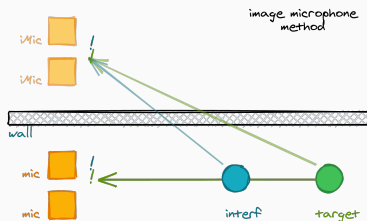


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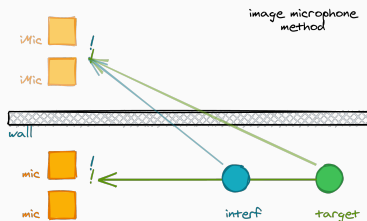


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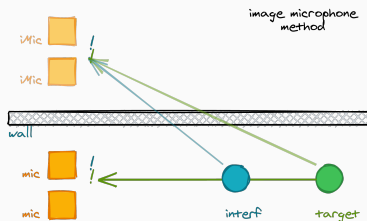


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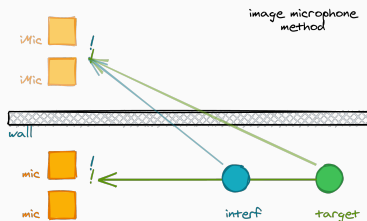


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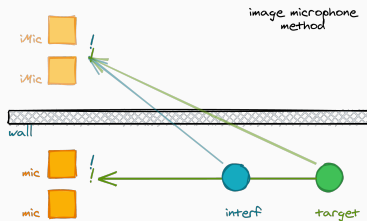


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# Sound Source Localization (SSL)

(common knowledge) 

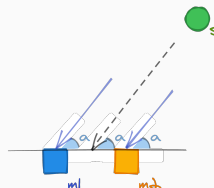
SSL → 3D position of sound source

## SSL with 2 microphones


- Only angle of arrival (AOA) ↗
- can be approximated from TDOA using e.g. GCC PHAT

[Knapp and Carter, 1976]

(known limitation, but good in practice)



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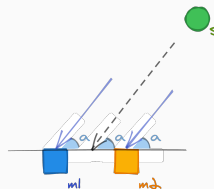
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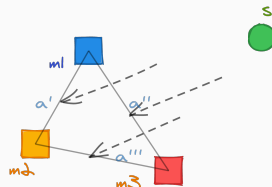
[Knapp and Carter, 1976]

(known limitation, but good in practice)



## SSL with more microphones

- Only Direction of Arrival (DoA): azimuth ( $\leftrightarrow$ ) and elevation ( $\updownarrow$ )
- AOA for each pair can be “fuse” together (e.g. angular spectra in SRP-PHAT [DiBiase et al., 2001])  
(known limitation, but good in practice)

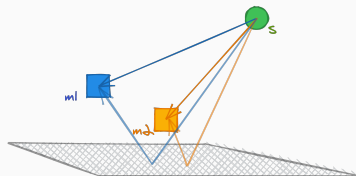


# Sound Source Localization with Echoes



## The Picnic Scenario:

- One source
- Two microphones
  - passive scenario
  - generalizable to any array geometry

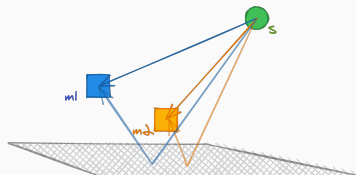


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- Close to a very reflective surface
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  - $\alpha_{\text{picnic}} \text{ const. } \forall f$
  - table-top device

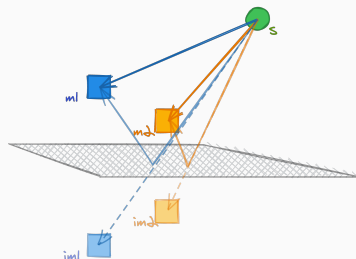


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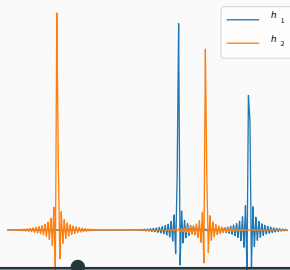


## Each pair is augmented with echoes

### Mirage Array

(Microphone Array Augmentation with Echoes)

How to access the *image* microphones?

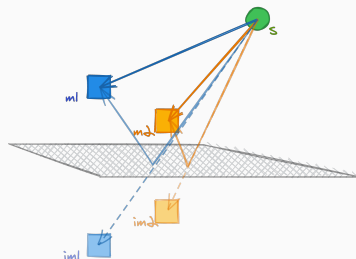


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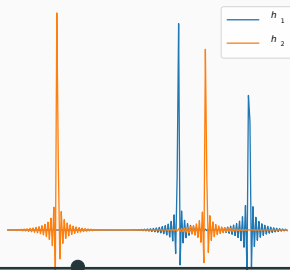


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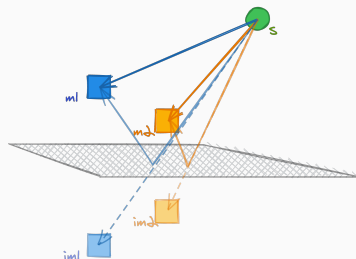


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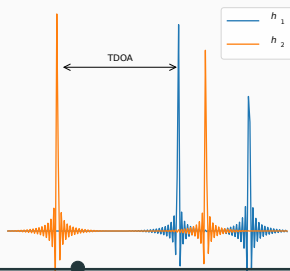


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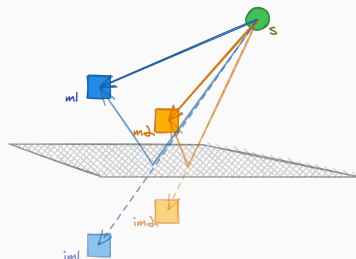


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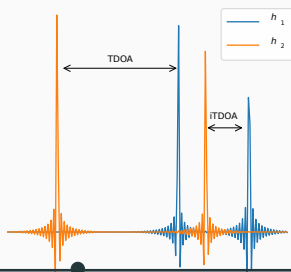


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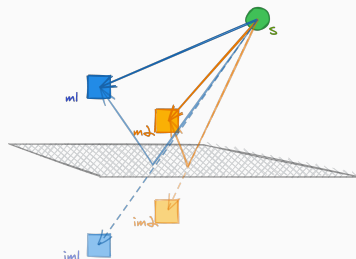


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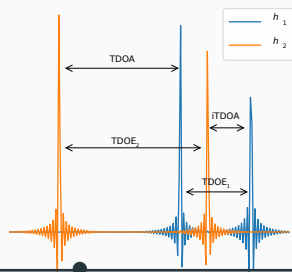


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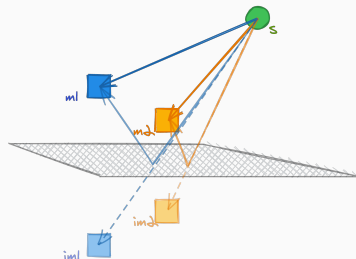


# Sound Source Localization with Echoes



**Idea:** DoA estimate on the MIRAGE array.

**Recall:** these TDOAs are the same of the DNN-based method



## Proposed Approach:

1. use proposed MLP model for TDOAs estimation
2. fuse together the estimation ...
  - of the Mirage array (similar to SRP-PHAT<sup>1</sup>)
  - knowing the position of the microphones;
  - use the error on a validation set as measure of uncertainty.

## Baseline: GCC PHAT on true microphones<sup>2</sup>

<sup>2</sup> [DiBiase et al., 2001]

<sup>1</sup> [Knapp and Carter, 1976]

# Experimental results



**Proposed:** MLP with **Mirage**

**Baseline:** GCC PHAT<sup>1</sup>

**Data:** 200 synthetic stereophonic recordings for close-surface scenario

**Metric:** accuracy in % ( $<10^\circ$ ,  $<20^\circ$ ) (↩ also error in the manuscript)

AOA ↗	Input	ACCURACY	
		$\alpha < 10^\circ$	$\alpha < 20^\circ$
<b>Mirage</b>	wn	77	97
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## Observation

✓ comparable to baseline when white noise source in noiseless case

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<b>Mirage</b>	wn+n	18	26	35	66
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- ✗ not generalize to noisy and speech data
- ✓ Solved “impossible” localization
- ⚠ Performance depending on echo estimation methods



Crocco, M., Trucco, A., and Del Bue, A. (2017).  
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