

# IMPROVED SINGLE-CHANNEL NONSTATIONARY NOISE TRACKING BY AN OPTIMIZED MAP-BASED POSTPROCESSOR

Aleksej Chinaev<sup>1</sup>, Reinhold Haeb-Umbach<sup>1</sup>, Jalal Taghia<sup>2</sup>, Rainer Martin<sup>2</sup>

<sup>1</sup> University of Paderborn, Department of Communications Engineering, 33098 Paderborn, Germany, {chinaev,haeb}@nt.uni-paderborn.de  
<sup>2</sup> Ruhr-Universität Bochum, Institute of Communication Acoustics, 44780 Bochum, Germany, {jalal.taghia,rainer.martin}@rub.de

## Introduction

- An optimized Maximum A-Posteriori Based (MAP-B) noise power spectral density (PSD) estimator
- An extensive performance evaluation for nonstationary noise

## MAP-B Noise Tracker

- Idea
  - ▶ MAP-B estimate of noise variance in the presence of noise observations "corrupted" by speech [Chinaev, et al., 2012]

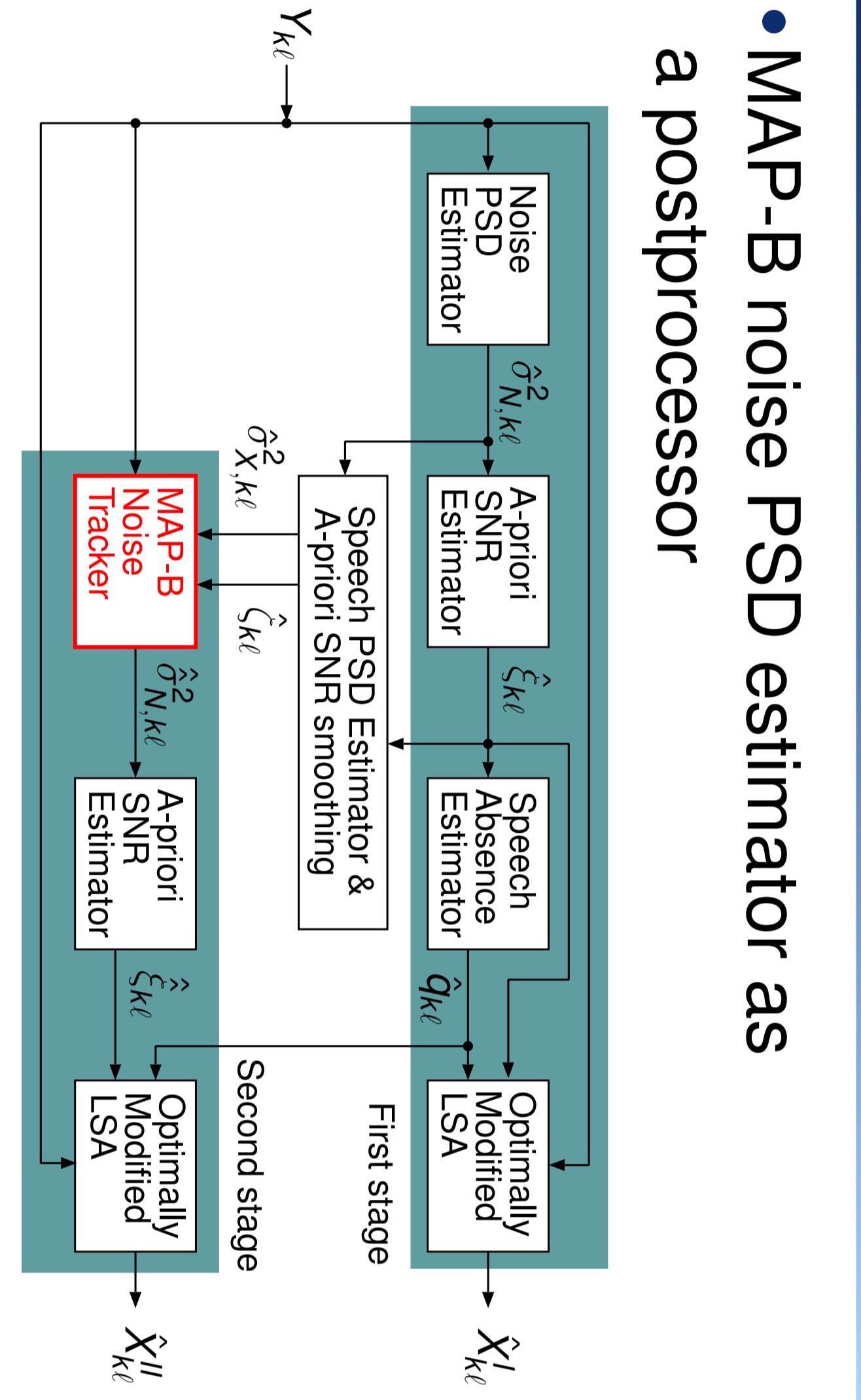


Fig. 1: Two-stage single-channel speech enhancement system

## Speech Enhancement System

- MAP-B noise PSD estimator as a postprocessor

- Input
  - ▶ Noisy speech  $Y_{k\ell} = X_{k\ell} + N_{k\ell}$
  - ▶ Speech power estimate of  $\sigma_{X_{k\ell}}^2 = E[|X_{k\ell}|^2]$
  - ▶ Smoothed SNR estimate
 
$$\hat{\zeta}_{k\ell} = \alpha_\zeta \cdot \hat{\zeta}_{k,\ell-1} + (1 - \alpha_\zeta) \cdot \hat{\xi}_{k\ell}$$
- Output
  - ▶ MAP noise PSD estimate  $\sigma_{N_{k\ell}}^2 = E[|N_{k\ell}|^2]$

## MAP-B Optimization

- Bias reduction
 
$$\hat{\sigma}_{N_{k\ell}}^2 = \left(1 - \beta(\hat{\zeta}_{k\ell})\right) \cdot \tilde{\sigma}_{N_{k\ell}}^2$$

with a bias compensation factor  $\beta(\hat{\zeta}_{k\ell})$  and a biased MAP-B estimate  $\tilde{\sigma}_{N_{k\ell}}^2$

- Tracking bandwidth reduction at high SNR
 
$$\nu(\hat{\zeta}_{k\ell}) = \nu_0 + \frac{\Delta\nu}{\pi} \cdot \arctan\left(\hat{\zeta}_{k\ell}\right)$$
- with a constant degrees of freedom  $\nu_0 = 40$  and an adjustment range  $\Delta\nu = 10$

## Experimental Setup

- 3 nonstationary noise types
- 8 state-of-the-art noise PSD estimators in the first stage
- Evaluation of MAP-B gain in

- ▶ noise PSD tracking performance
- ▶ quality of estimated speech

- Overall performance measures
- ▶ Estimation error reduction

$$\Delta \text{LogErr}_{\text{mean}} = \text{LogErr}_{\text{mean}}^{(I)} - \text{LogErr}_{\text{mean}}^{(II)}$$

## Experimental Results

## Conclusions

- Reduction of noise estimation error variance for almost all setups
- Slight improvement in PESQ scores
 
$$\Delta \text{PESQ} = \text{PESQ}^{(II)} - \text{PESQ}^{(I)},$$
- For all performance measures
  - ▶ the larger values show better performance
- Estimator's variance reduction
 
$$\Delta \text{LogErr}_{\text{var}} = \text{LogErr}_{\text{var}}^{(I)} - \text{LogErr}_{\text{var}}^{(II)}$$
- Speech quality improvement
 
$$\Delta \text{SNT} = \text{SNT}^{(II)} - \text{SNT}^{(I)}$$
- MS: Minimum Statistics, [Martin, 2001] [Hendriks, 2008]
- EMCRA: Enhanced MCRA, [Fan, 2007]
- MCRA: Minima Controlled Recursive Averaging, [Cohen, 2002]
- IMCRA: Improved MCRA, [Cohen, 2003]
- MCRA-MAP, [Kum, 2009]
- MMSE-Yu, [Yu, 2009]

