

A Simple and Effective Framework for A Priori SNR Estimation



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Abstract

- The **a priori SNR** is key-parameter in DFT-based speech enhancement schemes
- Decision-directed (DD)** *a priori* SNR estimation: linear combination of estimates along fixed DFT bin k .
- Can speech enhancement performance be improved by combining estimates along **harmonic trajectories** instead of fixed DFT bins?

Speech Enhancement

- DFT based speech enhancement: multiplicative gain function $G(\cdot)$
- Speech Estimate is obtained by
$$\hat{X}(k, \ell) = G(k, \ell, \xi(k, \ell), \zeta(k, \ell)) \cdot Y(k, \ell)$$
- $\zeta(k, \ell) = \frac{|Y(k, \ell)|^2}{\sigma_d^2(k, \ell)}$... a posteriori SNR
- $\xi(k, \ell) = \frac{\sigma_d^2(k, \ell)}{\sigma_d^2(k, \ell)}$... a priori SNR

The Decision-Directed A Priori SNR Estimator

DD a priori SNR estimate:

$$\hat{\xi}_{\text{DD}}(k, \ell) = (1 - \alpha_{\text{DD}}) \max[\hat{\xi}_{\text{ML}}(k, \ell), 0] + \alpha_{\text{DD}} \hat{\xi}_{\ell-1}(k, \ell)$$

with

$$\hat{\xi}_{\text{ML}}(k, \ell) = \hat{\zeta}(k, \ell) - 1$$

$$\hat{\xi}_{\ell-1}(k, \ell) = \frac{|\hat{X}(k, \ell-1)|^2}{\hat{\sigma}_d^2(k, \ell-1)}$$

- Is there a better choice for $\hat{\xi}_{\ell-1}(k, \ell)$?

PADDI - The Proposed Method

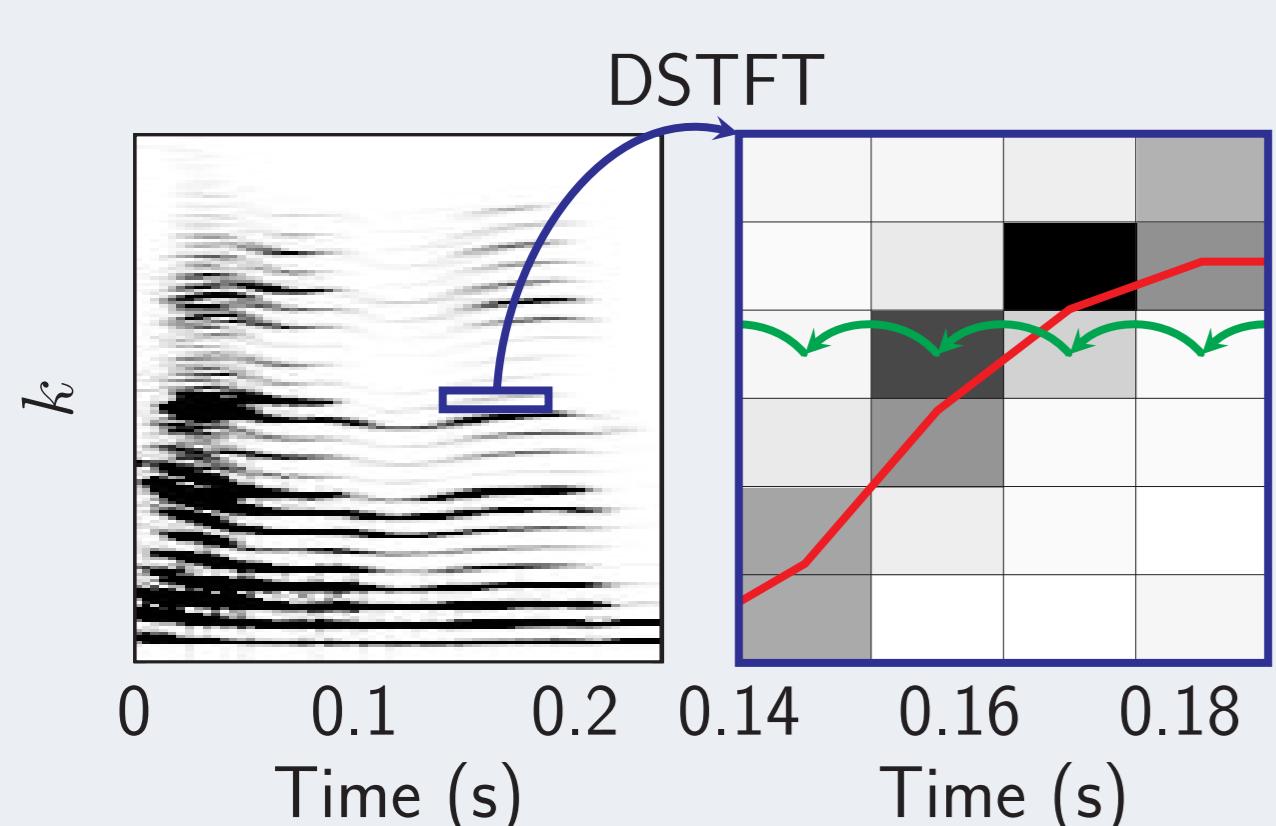
- Speech exhibits **harmonic structure**
- Fundamental frequency is **time-varying**
- Main idea of this work: ensure that k is dominated similarly by the same harmonic at frames ℓ' and $\ell' - 1$
- Pitch-adaptive discrete STFT (**PADSTFT**):

$$N_{\text{DFT}}(\ell) = \text{round} \left[K \frac{f_s}{f_0(\ell)} \right]$$

$$k_h(\ell) = \underset{k}{\operatorname{argmin}} \left| k - N_{\text{DFT}}(\ell) \frac{h f_0(\ell)}{f_s} \right| = K h$$

independent of ℓ !

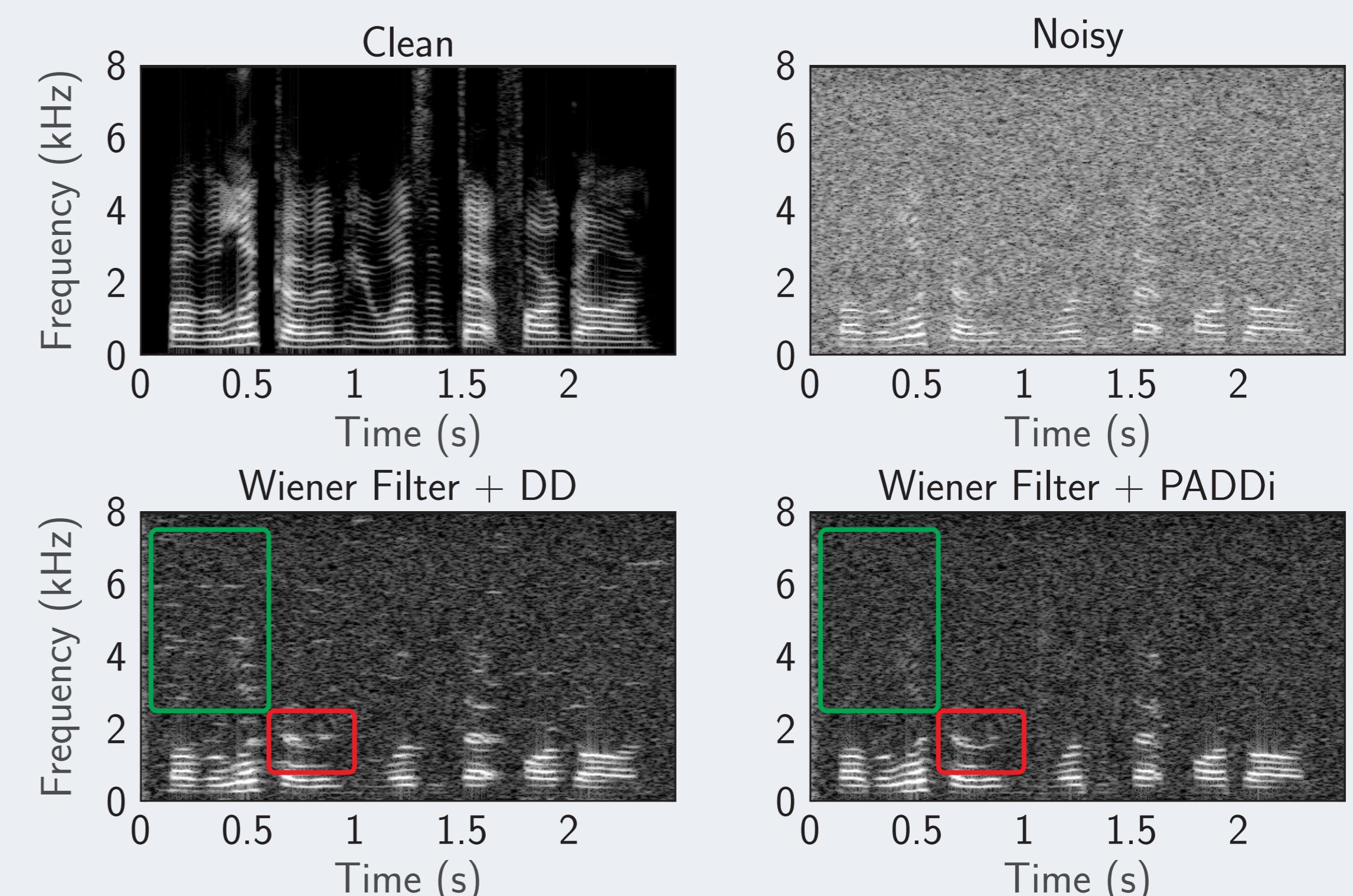
- Red: harmonic trajectory
- Green: smoothing path of *a priori* SNR estimator



- PADSTFT:**
- Fixed mapping from h to k
- Harmonic trajectory and smoothing path coincide

Proof-of-Concept

Noisy signal: Speech and white noise mixed at 0 dB SNR.



DD:

- Spurious spectral peaks → musical noise
- Harmonics are smeared along time

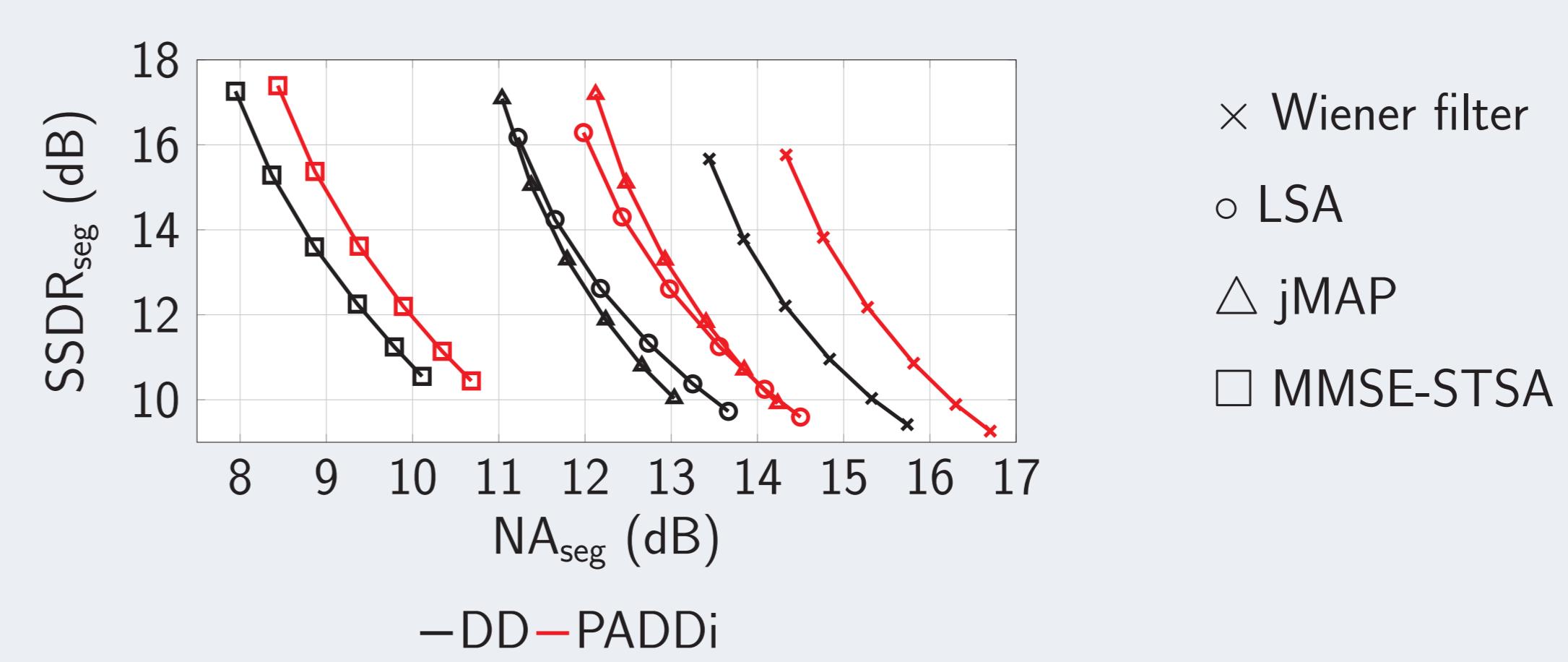
PADDI:

- Less isolated spectral peaks
- Harmonic fine structure is preserved

Results (1)

Characteristics of speech estimator **strongly depend on** $G(\cdot)$

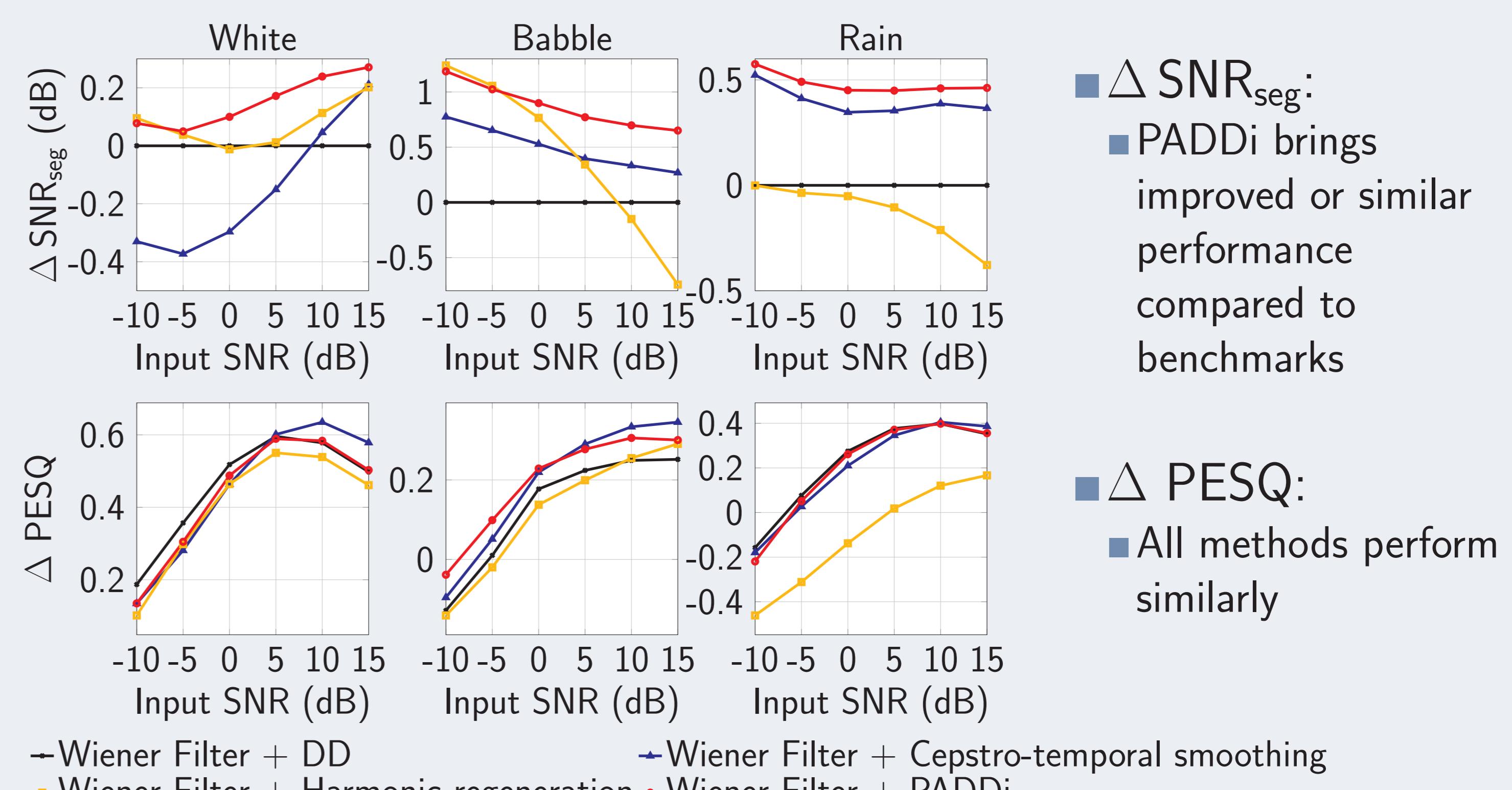
- We compared **DD** and **PADDI** for various $G(\cdot)$ s
- Evaluation: **Segmental Speech to Speech Distortion Ratio (SSDR_{seg})** vs. **Segmental Noise Attenuation (NA_{seg})**



PADDI increases NA_{seg} while preserving SSDR_{seg} compared to DD

Results (2)

Δ -improvement in terms of **PESQ** and **SNR_{seg}** over noisy speech



- Δ SNR_{seg}:
- PADDI brings improved or similar performance compared to benchmarks

- Δ PESQ:
- All methods perform similarly

Compared to the classical DD approach, PADDI enables

- more noise suppression** while
- preserving the level of speech distortions.**