

# Deliberate model misspecification for weather radar signal Doppler spectrum moments estimation

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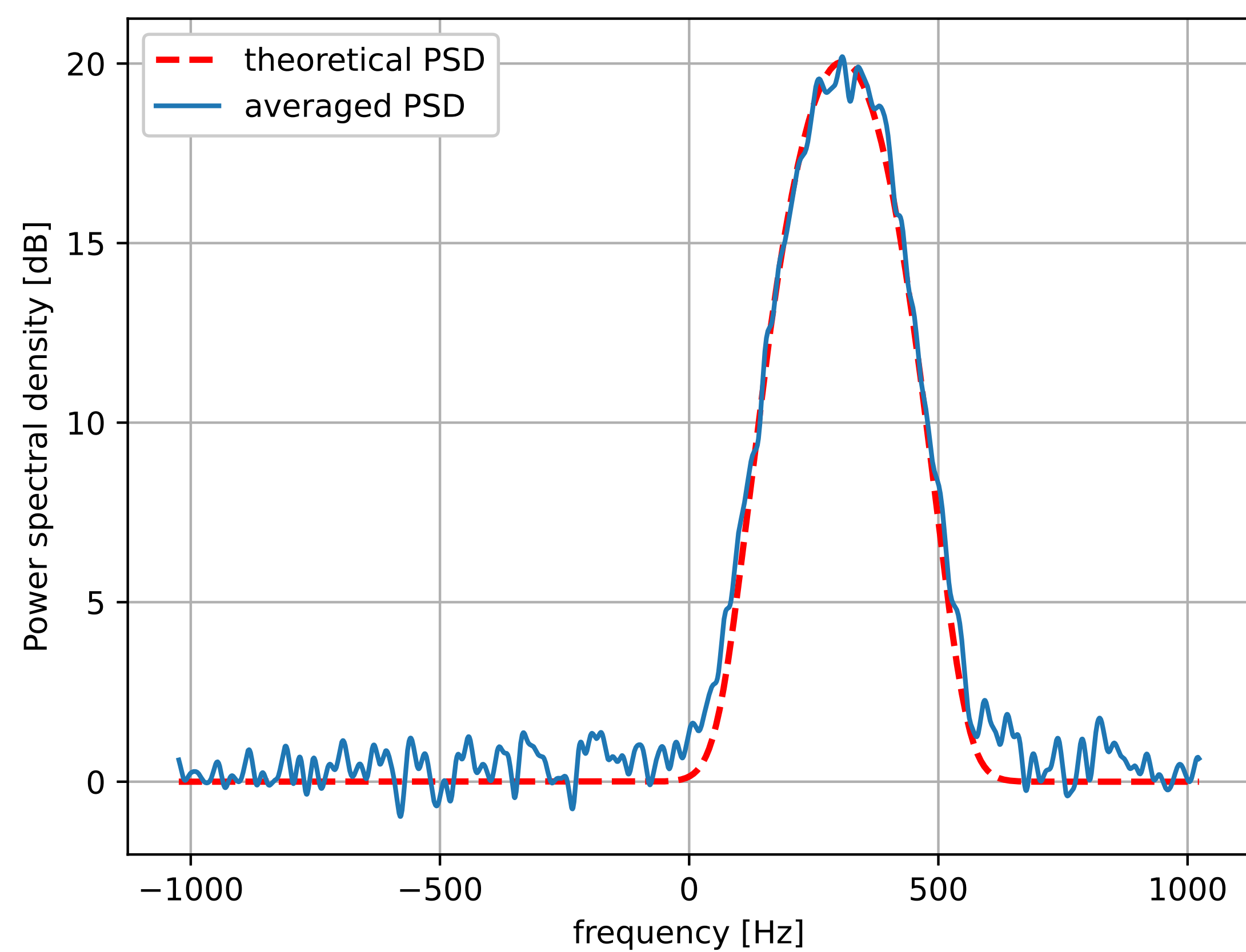
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## Context

In weather applications, ground based pulsed radars are used to estimate precipitation rates. The main features (moments of the Doppler spectrum) of the collected signals are the following:

- moment 0: mean power  $P_0 \rightarrow$  reflectivity,
- moment 1: mean Doppler frequency  $f_0 \rightarrow$  mean radial velocity of raindrops,
- moment 2: spectral width  $\sigma_0 \rightarrow$  dispersion of radial velocities (seldom used).



## Misspecification estimation

There is model misspecification when the assumed signal model (based on which an estimator is designed) differs from the true signal model.

- imperfect knowledge of the true data model,
- true data model known but the corresponding estimation problem is too involved.

## Signal models

A general signal model for the problem at hand is the following:

$$\mathbf{y} = \mathbf{x} + \mathbf{w},$$

with  $\mathbf{w} \sim \mathcal{CN}(\mathbf{0}, \sigma_n^2 \mathbf{I}_N)$ , an additive white complex Gaussian noise of power  $\sigma_n^2$  and  $\mathbf{x}$  the signal of interest.

## True signal model

The shape of the weather signals power spectrum is assumed Gaussian.

$$\mathbf{x} \sim \mathcal{CN}(\mathbf{0}, \mathbf{B}_t(P_0, f_0, \sigma_0^2)),$$

where,

$$\begin{aligned} \mathbf{B}_t(P_0, f_0, \sigma_0^2) &= P_0 \mathbf{D}(f_0) \mathbf{C}(\sigma_0^2) \mathbf{D}(f_0)^H, \\ \mathbf{D}(f_0) &= \text{diag}(\dots, \exp(j2\pi f_0(n-1)T_s), \dots), \\ [\mathbf{C}(\sigma_0^2)]_{k,l} &= \exp(-2\pi^2 \sigma_0^2 (k-l)^2 T_s^2). \end{aligned}$$

## Assumed signal model

One may ignore the presence of the spectral width  $\sigma_0$ . The resulting power spectrum would shrink to a dirac function centered on  $f_0$ .

$$\mathbf{x} \sim \mathcal{CN}(\mathbf{0}, \mathbf{B}_a(P_0, f_0)),$$

where

$$\begin{aligned} \mathbf{B}_a(P_0, f_0) &= P_0 \mathbf{d}(f_0) \mathbf{d}(f_0)^H, \\ \mathbf{d}(f_0) &= (\dots, \exp(j2\pi f_0(n-1)T_s), \dots). \end{aligned}$$

## Considered estimators

### Mismatched unconditional maximum likelihood estimator

$$L(\boldsymbol{\mu}) = \ln(|\mathbf{R}(\boldsymbol{\mu})|) + \text{Tr}\{\mathbf{R}(\boldsymbol{\mu})^{-1} \hat{\mathbf{R}}_y\},$$

with the sample covariance matrix  $\hat{\mathbf{R}}_y = \frac{1}{K} \sum_{k=1}^K \mathbf{y}_k \mathbf{y}_k^H$  and the asymptotic covariance matrix  $\mathbf{R}(\boldsymbol{\mu}) = \mathbf{B}_a(P_0, f_0) + \sigma_n^2 \mathbf{I}_N$ . Then,

$$\hat{f}_0 = \arg \max_f \sum_{k=1}^K |r_k(f)|^2, \quad \hat{P}_0 = \frac{1}{KN(N-1)} \sum_{k=1}^K (|r_k(\hat{f}_0)|^2 - \|\mathbf{y}_k\|^2),$$

where  $r_k(f) = \mathbf{y}_k^H \mathbf{d}(f)$ .

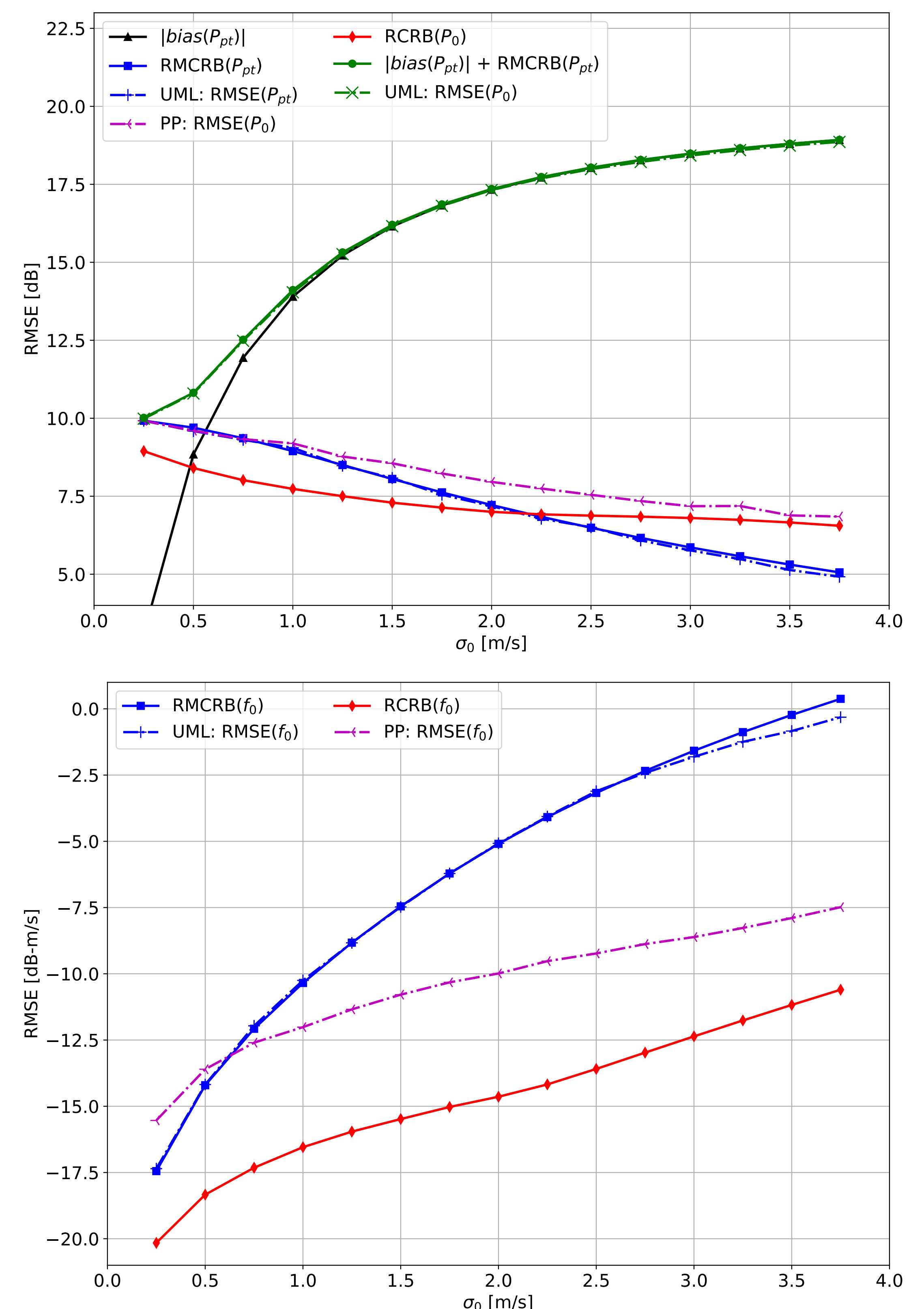
### Pulse-pair estimator

Widely used in operational systems, it requires an estimate of the auto-correlation.

$$\hat{A}_k(nT_s) = \frac{1}{N-|n|} \sum_{i=0}^{N-1-|n|} [\mathbf{y}_k]_i^* [\mathbf{y}_k]_{i+n}.$$

$$\hat{P}_{PP} = \frac{1}{K} \sum_{k=1}^K \hat{A}_k(0), \quad \hat{f}_{PP} = \frac{1}{2\pi T_s K} \sum_{k=1}^K \arg(\hat{A}_k(T_s)).$$

## Results



## Conclusions and perspectives

- ignoring  $\sigma_0$  may be interesting for a limited range of spectral width value,
- the approach may be applied to other misspecifications: spectrum shape, noise,
- operational constraints should be accounted for: pulse repetition time