

Misspecified Cramér-Rao Bound of RIS-aided Localization under Geometry Mismatch

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1 Motivation

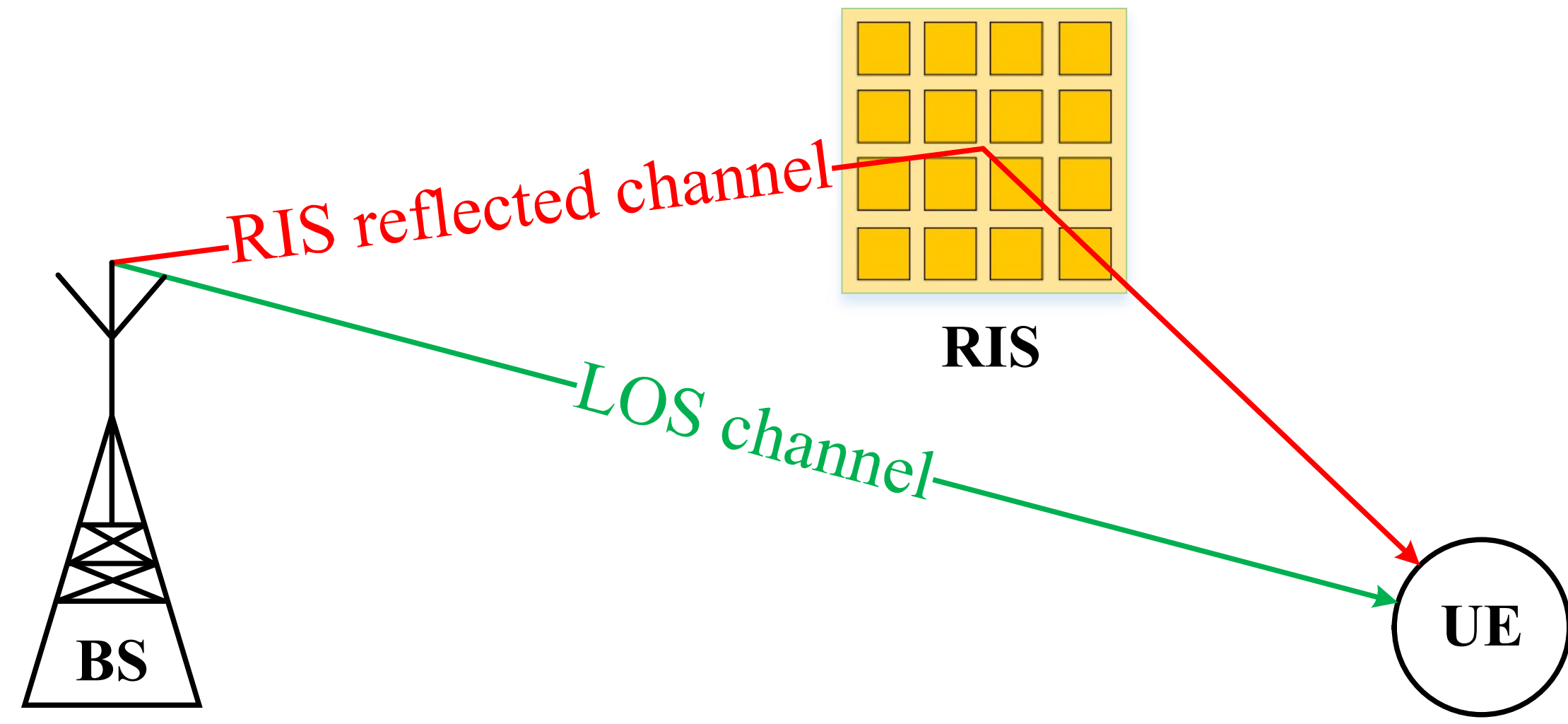


Figure 1: Illustration of a RIS-aided localization system.

- In 5G/6G wireless systems, reconfigurable intelligent surface (RIS) can play a role as a passive anchor to enable and enhance localization in various scenarios.
- However, most existing RIS-aided localization works assume that the geometry of the RIS is perfectly known, which is not realistic in practice.
- This work aims to characterize the effect of the model mismatch caused by RIS geometry error on the localization performance.
- Specifically, we derive the **misspecified Cramér-Rao bound (MCRB)** for a single-input-single-output (SISO) RIS-aided localization system with RIS geometry mismatch.
- The main contribution is that we derive a **closed-form** solution to the pseudo-true parameter determination problem (which is a core step to derive the MCRB).

2 Channel Model

- A downlink SISO wireless system with a BS, a UE, and a RIS.
- Transmission of L OFDM pilot symbols with K subcarriers.

The received baseband signal in the OFDM block with index ℓ is given by

$$\mathbf{y}_\ell = \underbrace{g_b \mathbf{d}(\tau_b)}_{\mu_\ell} \odot \mathbf{x} + \underbrace{g_r \mathbf{b}(\phi)^\top \gamma_\ell(\mathbf{d}(\tau_r) \odot \mathbf{x})}_{\mu_\ell} + \mathbf{n}, \quad (1)$$

3 Bound Derivation

The accessible geometry information of the RIS can be modeled as $\tilde{\mathbf{p}}_r = \mathbf{p}_r + \mathbf{u}$, $\tilde{\mathbf{R}}_r = \mathbf{R}_p(\mathbf{v})\mathbf{R}_r$, where \mathbf{u} and \mathbf{v} are the calibration errors. Define channel parameters and localization parameters:

$$\boldsymbol{\eta}_{\text{ch}} \triangleq [\phi_{\text{az}}, \phi_{\text{el}}, \tau_b, \tau_r, \Re(g_b), \Im(g_b), \Re(g_r), \Im(g_r)]^\top, \quad (2)$$

$$\boldsymbol{\eta} \triangleq [\phi_{\text{az}}, \phi_{\text{el}}, \tau_b, \tau_r]^\top, \quad (3)$$

$$\mathbf{r} \triangleq [\mathbf{p}^\top, \Delta]^\top \in \mathbb{R}^4. \quad (4)$$

Two-stage localization: $\mathbf{y} \longrightarrow \boldsymbol{\eta} \longrightarrow \mathbf{r}$

The **true** likelihood function (f_T) and **mismatched** likelihood function (f_M):

$$\ln f_T = -\frac{1}{2}(\hat{\boldsymbol{\eta}} - \mathbf{g}(\mathbf{r}|\mathbf{p}_r, \mathbf{R}_r))^\top \boldsymbol{\Sigma}^{-1}(\hat{\boldsymbol{\eta}} - \mathbf{g}(\mathbf{r}|\mathbf{p}_r, \mathbf{R}_r)),$$

$$\ln f_M = -\frac{1}{2}(\hat{\boldsymbol{\eta}} - \mathbf{g}(\mathbf{r}|\tilde{\mathbf{p}}_r, \tilde{\mathbf{R}}_r))^\top \boldsymbol{\Sigma}^{-1}(\hat{\boldsymbol{\eta}} - \mathbf{g}(\mathbf{r}|\tilde{\mathbf{p}}_r, \tilde{\mathbf{R}}_r)).$$

The lower bound matrix of the estimation MSE based on f_M :

$$\text{LBM}(\hat{\mathbf{r}}, \bar{\mathbf{r}}) = \underbrace{\mathbf{A}_{\mathbf{r}_0}^{-1} \mathbf{B}_{\mathbf{r}_0} \mathbf{A}_{\mathbf{r}_0}^{-1}}_{\text{MCRB}(\mathbf{r}_0)} + \underbrace{(\bar{\mathbf{r}} - \mathbf{r}_0)(\bar{\mathbf{r}} - \mathbf{r}_0)^\top}_{\text{Bias}(\mathbf{r}_0)}, \quad (5)$$

User PEB: $\sqrt{\mathbb{E}\{\|\mathbf{p} - \hat{\mathbf{p}}\|^2\}} \geq \sqrt{\text{tr}([\text{LBM}(\hat{\mathbf{r}}, \bar{\mathbf{r}})]_{1:3,1:3})} \triangleq \text{LB}$. Here, \mathbf{r}_0 is the pseudo-true parameter vector that minimizes the KLD between f_T and f_M :

$$\mathbf{r}_0 = \arg \min_{\mathbf{r}} D(f_T(\hat{\boldsymbol{\eta}}|\bar{\mathbf{r}}) \| f_M(\hat{\boldsymbol{\eta}}|\mathbf{r})), \quad (6)$$

$$\begin{aligned} D(f_T(\hat{\boldsymbol{\eta}}|\bar{\mathbf{r}}) \| f_M(\hat{\boldsymbol{\eta}}|\mathbf{r})) &= \mathbb{E}_{f_T} \{ \ln f_T(\hat{\boldsymbol{\eta}}|\bar{\mathbf{r}}) - \ln f_M(\hat{\boldsymbol{\eta}}|\mathbf{r}) \}, \\ &= \dots = \frac{1}{2} \mathbf{h}(\mathbf{r})^\top \boldsymbol{\Sigma}^{-1} \mathbf{h}(\mathbf{r}). \end{aligned} \quad (7)$$

Let $\mathbf{r}_0 = [\mathbf{p}_0^\top, \Delta_0]^\top$ be the pseudo-true parameters and $\bar{\mathbf{r}} = [\bar{\mathbf{p}}^\top, \bar{\Delta}]^\top$ be the true parameters, then $\mathbf{h}(\mathbf{r}_0) = \mathbf{0}$ gives

$$(\alpha + \|\mathbf{p}_0 - \tilde{\mathbf{p}}_r\| - \|\mathbf{p}_0 - \mathbf{p}_b\|) \frac{\mathbf{p}_0 - \tilde{\mathbf{p}}_r}{\|\mathbf{p}_0 - \tilde{\mathbf{p}}_r\|} = \tilde{\mathbf{R}}_r \mathbf{R}_r^\top (\bar{\mathbf{p}} - \mathbf{p}_r), \quad (8)$$

It can be inferred that a pseudo-true UE position \mathbf{p}_0 can be obtained as the intersection of a line s_l and a hyperboloid s_h given by

$$s_l: \quad \mathbf{p} = x \tilde{\mathbf{R}}_r \mathbf{R}_r^\top (\bar{\mathbf{p}} - \mathbf{p}_r) + \tilde{\mathbf{p}}_r, \quad (9)$$

$$s_h: \quad \|\mathbf{p} - \tilde{\mathbf{p}}_r\| - \|\mathbf{p} - \mathbf{p}_b\| = \beta. \quad (10)$$

4 Results Validation

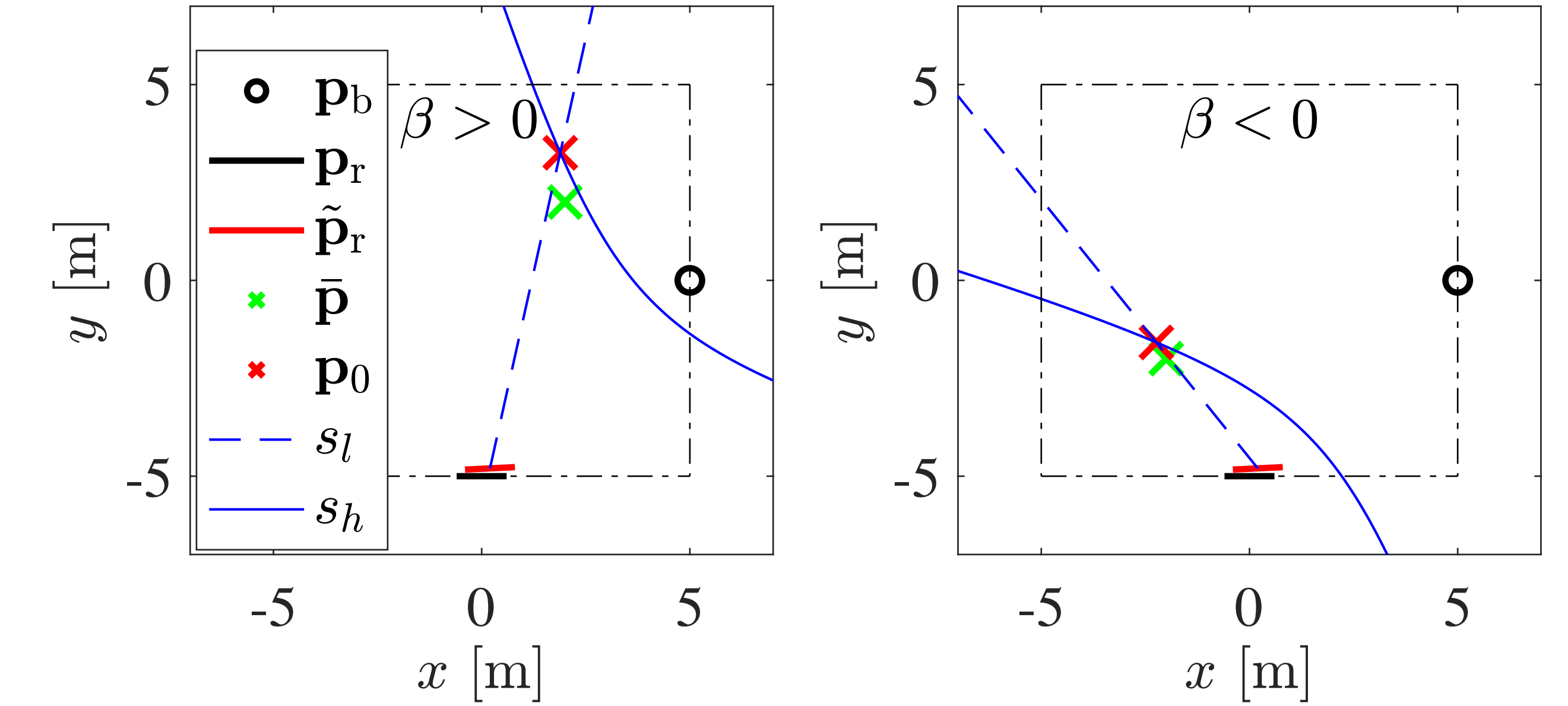


Figure 2: The geometric relationship between the pseudo-true UE position \mathbf{p}_0 and true UE position $\bar{\mathbf{p}}$.

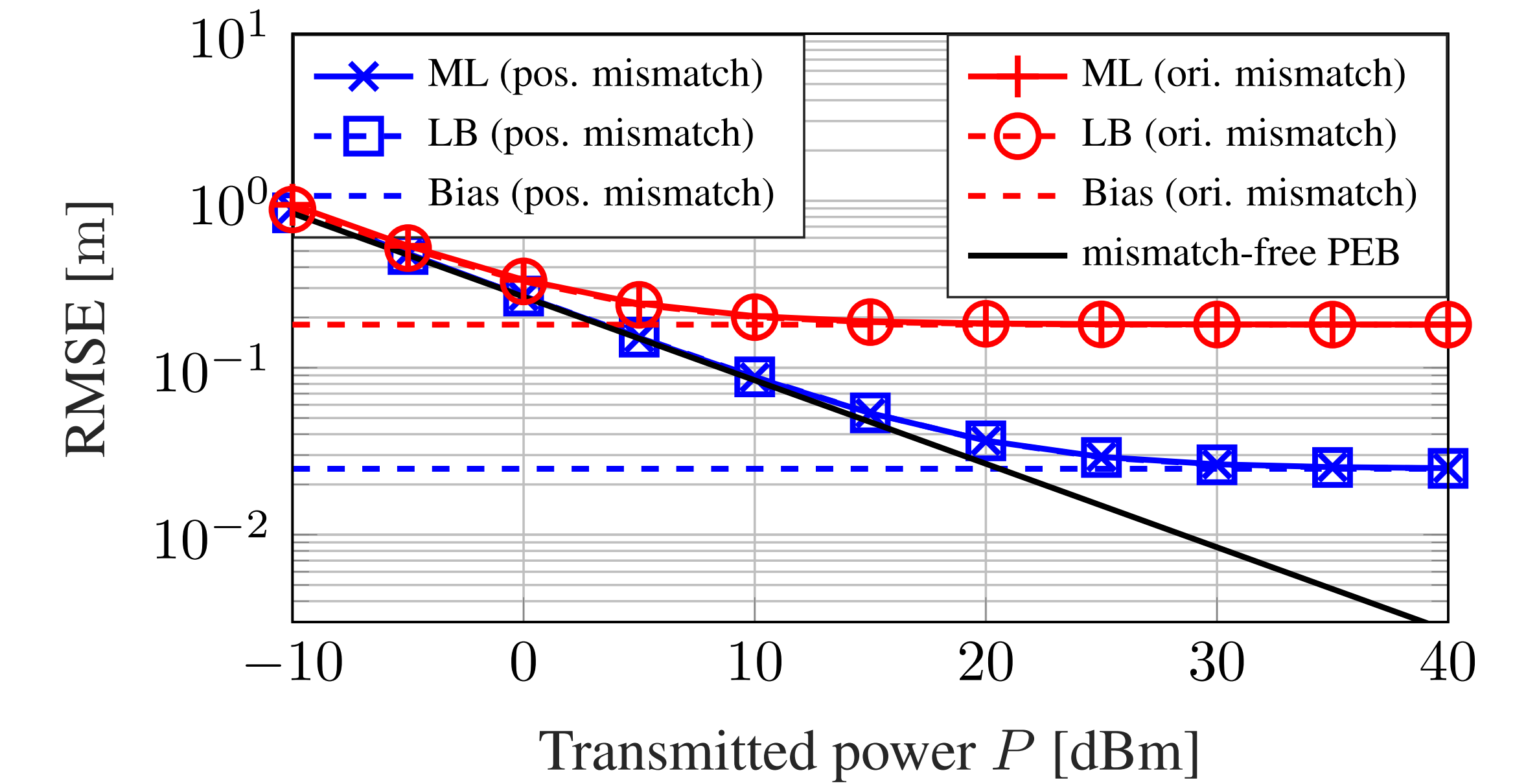


Figure 3: ML-RMSE, LB, and bias term versus transmitted power (SNR) for RIS position mismatch ($\mathbf{u} = 0.01 \times \mathbf{1}$ m) and orientation mismatch ($\mathbf{v} = 0.5 \times \mathbf{1}$ deg).

- 1 The RMSE of the ML estimator closely follows the LB, which demonstrates the validity of our derivation.
- 2 At low SNR levels, LB and mismatch-free PEB coincide \rightarrow The RIS geometry mismatch is not the main source of error.
- 3 At higher SNR, LB deviates from the mismatch-free PEB and saturates \rightarrow The positioning performance is more severely affected by RIS geometry mismatch.