

Table II Explanation

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The values in Table II are calculated similarly to Dilution of Precision (DOP), a common uncertainty metric in GPS [1]. Specifically, we are computing Position DOP (PDOP), the norm of the x, y, z covariances (i.e., $\sigma_{xx}, \sigma_{yy}, \sigma_{zz}$), after being propagated through the nonlinear geometry linearized with respect to position and then evaluated at some relative pose \mathbf{T}_B^A .

This document breaks down this computation and shows how these values are calculated in the accompanying Jupyter Notebook.

1 Problem Formulation:

Consider two agents, A and B , in world frame W with true state vectors:

$$\mathbf{x}_A^W = [x_A^W \ y_A^W \ z_A^W \ \alpha_A^W \ \beta_A^W \ \gamma_A^W]^\top = \mathbf{T}_A^W$$

$$\mathbf{x}_B^W = [x_B^W \ y_B^W \ z_B^W \ \alpha_B^W \ \beta_B^W \ \gamma_B^W]^\top = \mathbf{T}_B^W$$

where $\langle x, y, z \rangle$ refers to position and $\langle \alpha, \beta, \gamma \rangle$ refers to roll/pitch/yaw, superscript represents the frame, and subscript represents the target. While in general $\mathbf{x} \leftrightarrow \mathbf{T}$, \mathbf{x} refers to a column state vector while \mathbf{T} refers to $\mathbb{R}^{4 \times 4}$ transformation matrix. Additionally, let $\mathbf{T}(\cdot)$ refer to the function that constructs a transformation matrix. Specifically, the following notations and shorthands are equivalent:

$$\mathbf{T}(\mathbf{x}) = \mathbf{T}(x, y, z, \alpha, \beta, \gamma)$$

$$\mathbf{T}(x, y, z) = \mathbf{T}(x, y, z, 0, 0, 0)$$

$$\mathbf{T}(x, y, z, \gamma) = \mathbf{T}(x, y, z, 0, 0, \gamma)$$

Let the relative state from agent A to agent B be defined as:

$$\mathbf{x} = \mathbf{x}_B^A = [x_B^A \ y_B^A \ z_B^A \ \alpha_B^A \ \beta_B^A \ \gamma_B^A]^\top = (\mathbf{T}_A^W)^{-1} \mathbf{T}_B^W = \mathbf{T}_B^A$$

Relative coordinate frames are assumed to be oriented with respect to downward gravity.

Here, we have:

- $\hat{\mathbf{x}}$ are the estimated state parameters
- \mathbf{z} are the current measurements

where:

- $\hat{\mathbf{x}} \in \mathbb{R}^3 \implies \hat{\mathbf{x}} = [\hat{x}_B^A \quad \hat{y}_B^A \quad \hat{z}_B^A]$
- $\mathbf{z} \in \mathbb{R}^{37}$ has measurement model:

$$\mathbf{h}(\mathbf{x}_B^A) = [\mathbf{d} \quad z_B^A]^\top$$

where $z_B^A \in \mathbb{R}$ is the relative altitude and \mathbf{d} is the vector of true UWB range. In this case, there are $\mathbf{d} \in \mathbb{R}^{36}$ since agent A and B have $N_A = 6$ and $N_B = 6$ antennas respectively. In general, \mathbf{d} has the form:

$$\mathbf{d} = [d_{11} \quad d_{12} \quad \dots \quad d_{N_A N_B}]^\top$$

with each element defined as:

$$d_{ij} = d_{ij}(\mathbf{x}_B^A) = \|\mathbf{T}_B^A \mathbf{p}_j^B - \mathbf{p}_i^A\|_2$$

where \mathbf{p}_i^A and \mathbf{p}_j^B is the homogeneous coordinate of agent A 's i th and B 's j th antenna in A 's and B 's frame respectively.

With this measurement model, we can convince ourselves the system is fully observable. Specifically, the direct inclusion of z_B^A into the measurement model resolves the usual $\pm z$ ambiguity stemming from antennas only spanning their relative xy -plane.

We note that while z_B^A is not directly measurable, since both local z -axes are aligned with gravity down, z_B^A can be calculated as the difference of two realizable measurements:

$$z_B^A = z_B^W - z_A^W$$

In practice, while we can always locally measure z_A^W , we do not want to transmit z_B^W . This will need to be considered in our error analysis.

2 Computing PDOP:

Consider following formulas from [1]:

$$\Sigma_x = (A^\top \Sigma_z^{-1} A)^{-1}$$

in general, which can be simplified to:

$$\Sigma_x = (A^\top A)^{-1} \sigma^2$$

when covariances are consistent across all measurements and independent. Here, we have:

- A is the linearized measurement model about some point \mathbf{x}_0
- Σ_z is the covariance matrix of the measurements \mathbf{z}
- Σ_x is the covariance matrix of the state variables \mathbf{x} after being propagated through the linearized measurement model A

When calculating DOP, the diagonal values of $(A^\top A)^{-1}$ are XDOP, YDOP, ZDOP values (i.e., the variance amplification resulting from the linearized geometry in various state dimensions).

In general since d_{ij} is a UWB measurement and z_B^A is some calculated combination of a low variance LiDAR measurement and/or high variance assumption, we have:

$$\text{var}[d_{ij}] \neq \text{var}[z_B^A]$$

forcing us to use the former equation to calculate Σ_x .

In the paper (Figure 2a), we experimentally measured our UWB measurement variance to be:

$$\text{var}[d_{ij}] = 0.24\text{m}$$

For the purpose of this error analysis, we will assume the measurement error is modeled by a zero mean Gaussian.

Depending on our experimental setup, z_A^W and z_B^W are either measured by downward facing LiDAR (with a reported operating tolerance of $\pm 4\text{cm}$ in our operating range) or assumed by some pre-communicated (but locally monitored) global altitude tolerance (i.e., $\bar{z}_A^W \pm \tilde{z}_A^W$ and $\bar{z}_B^W \pm \tilde{z}_B^W$ respectively where \bar{z} is the commanded altitude and \tilde{z} is the altitude tolerance).

3 Jupyter Notebook:

Using the $\hat{\mathbf{x}} \in \mathbb{R}^3$ and $\mathbf{z} \in \mathbb{R}^{37}$ as defined above, we perform the above error analysis with different values for $\text{var}[z_B^A]$ (selected based on current communication/sensor assumptions and encoded in Σ_z) and different measurement model linearization points (select based on current position and encoded in A) and calculate the corresponding PDOPs. This implies A is of the form $\mathbb{R}^{37 \times 3}$, making this a cumbersome computation to do by hand. Thus, we make the Jupyter Notebook used to compute Table II (and other relevant facts) available here:

- <https://github.com/mit-acl/murp-datasets>

References

- [1] Richard B Langley et al. Dilution of precision. *GPS world*, 10(5):52–59, 1999.