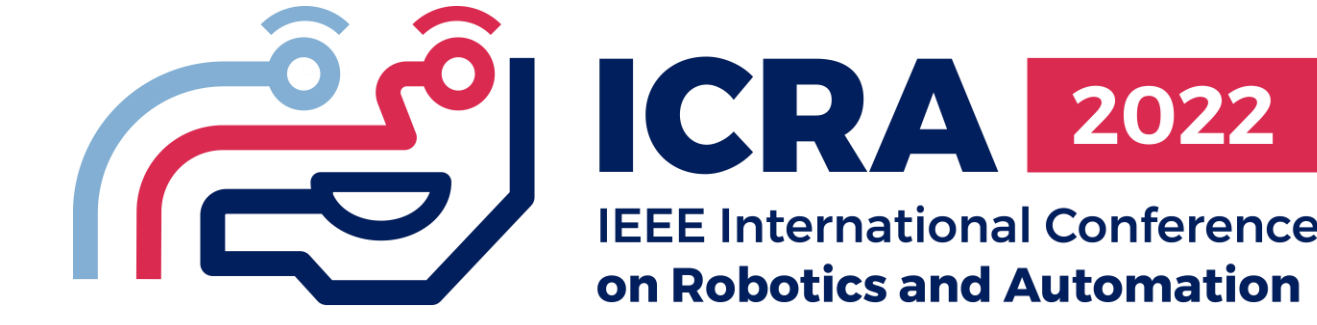


Extrinsic Calibration of Multiple Inertial Sensors From Arbitrary Trajectories

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INTRODUCTION

- The simultaneous use of multiple IMUs—as compared with a single IMU of the same total size, weight, power, and cost—may have three benefits:
 - Higher measurement accuracy
 - Increased bandwidth
 - Better fault tolerance
- It is necessary to perform *extrinsic calibration*—to estimate the relative position and orientation of each IMU—in order to realize these benefits
- We propose a method of extrinsic calibration for multiple IMUs that does not require instruments (e.g., rate tables) or aiding sensors (e.g., a camera) and that can be applied to data collected in-flight along arbitrary trajectories
- Our method is based on solving a nonlinear least-squares problem that penalizes inconsistency between measurements from pairs of IMUs

METHOD

- Each IMU's measurement is modeled as:

$$\begin{aligned} \text{Accelerometer: } {}^I\tilde{\mathbf{a}} &= {}^I\mathbf{a}_{WI} - {}^I\mathbf{g} + \mathbf{b}_a + \mathbf{n}_a \\ \text{Gyroscope: } {}^g\tilde{\boldsymbol{\omega}} &= \mathbf{C}({}^g\mathbf{q}) {}^I\boldsymbol{\omega}_{WI} + \mathbf{b}_g + \mathbf{n}_g \end{aligned}$$

$\mathbf{a}_{a,k+1} - \mathbf{b}_{a,k} \sim \sigma_{a_g} \sqrt{\Delta t} \cdot \mathcal{N}(\mathbf{0}, \mathbf{I})$
 $\mathbf{n}_a \sim \sigma_a / \sqrt{\Delta t} \cdot \mathcal{N}(\mathbf{0}, \mathbf{I})$
 $\mathbf{b}_{g,k+1} - \mathbf{b}_{g,k} \sim \sigma_{b_g} \sqrt{\Delta t} \cdot \mathcal{N}(\mathbf{0}, \mathbf{I})$
 $\mathbf{n}_g \sim \sigma_g / \sqrt{\Delta t} \cdot \mathcal{N}(\mathbf{0}, \mathbf{I})$
 Δt : sampling interval; (σ_a, σ_g) : square root of noise spectral density; $(\sigma_{b_a}, \sigma_{b_g})$: bias instability

- Given: ${}^{I_0}\tilde{\mathbf{a}}_k, {}^{g_0}\tilde{\boldsymbol{\omega}}_k, {}^{I_1}\tilde{\mathbf{a}}_k, {}^{g_1}\tilde{\boldsymbol{\omega}}_k$ for each time $k \in \{1, \dots, K\}$
- To find: ${}^{I_0}\mathbf{p}_{I_0 I_1} := \mathbf{p} \in \mathbb{R}^3, {}^{I_1}\mathbf{q} := \mathbf{q} \in \mathbb{H}, {}^{g_0}\mathbf{q} \in \mathbb{H}, {}^{g_1}\mathbf{q} \in \mathbb{H},$

$$\mathbf{b}_{a_0, k}, \mathbf{b}_{g_0, k}, \mathbf{b}_{a_1, k}, \mathbf{b}_{g_1, k}, {}^{I_0}\boldsymbol{\alpha}_{W I_0, k} := {}^{I_0}\boldsymbol{\alpha}_k$$

slack variables; their estimation is not main focus

$$\min \sum_k \left(\|\mathbf{r}_a\|_{\Sigma_a}^2 + \|\mathbf{r}_g\|_{\Sigma_g}^2 + \|\mathbf{r}_{b_{a_0}}\|_{\Sigma_{b_a}}^2 + \|\mathbf{r}_{b_{a_1}}\|_{\Sigma_{b_a}}^2 + \|\mathbf{r}_{b_{g_0}}\|_{\Sigma_{b_g}}^2 + \|\mathbf{r}_{b_{g_1}}\|_{\Sigma_{b_g}}^2 \right)$$

$$\mathbf{r}_a = {}^{I_1}\hat{\mathbf{a}}_k - ({}^{I_1}\tilde{\mathbf{a}}_k - \mathbf{b}_{a_1, k})$$

$\hat{\mathbf{a}}_k = \mathbf{C}({}^{g_1}\mathbf{q})^{-1} ({}^{g_1}\tilde{\boldsymbol{\omega}}_k + [{}^{g_1}\boldsymbol{\omega}_k]_x \mathbf{p} + [{}^{g_1}\boldsymbol{\alpha}_k]_x \mathbf{p})$, where ${}^{g_1}\tilde{\boldsymbol{\omega}}_k = {}^{g_1}\boldsymbol{\omega}_k - \mathbf{b}_{g_1, k}$ and ${}^{g_1}\boldsymbol{\omega}_k = \mathbf{C}({}^{g_0}\mathbf{q})^{-1} ({}^{g_0}\tilde{\boldsymbol{\omega}}_k - \mathbf{b}_{g_0, k})$

$$\mathbf{r}_g = {}^{g_1}\hat{\boldsymbol{\omega}}_k - ({}^{g_1}\tilde{\boldsymbol{\omega}}_k - \mathbf{b}_{g_1, k})$$

$\hat{\boldsymbol{\omega}}_k = \mathbf{C}({}^{g_1}\mathbf{q}) \mathbf{C}({}^{g_0}\mathbf{q})^{-1} \mathbf{C}({}^{I_0}\mathbf{q})^{-1} ({}^{I_0}\tilde{\boldsymbol{\omega}}_k - \mathbf{b}_{g_0, k})$

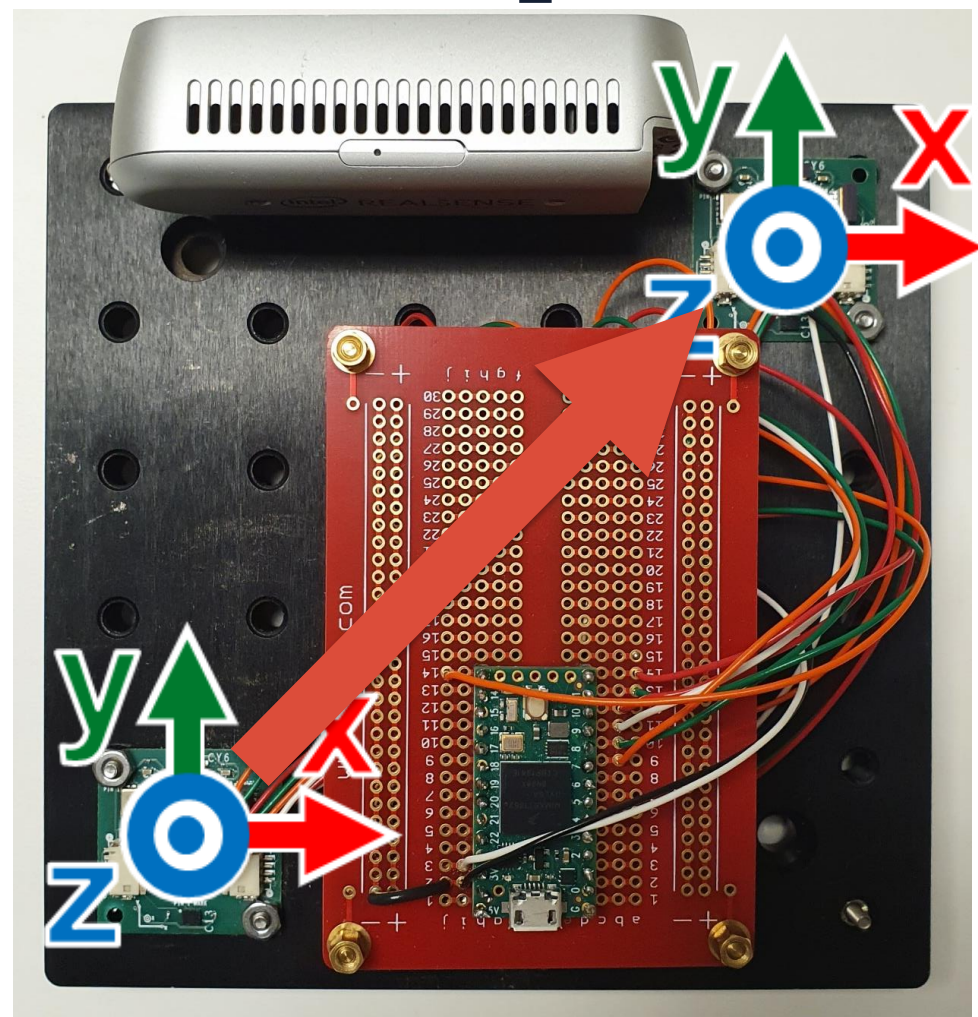
$$\mathbf{r}_{b_{a_0}} = \mathbf{b}_{a_0, k+1} - \mathbf{b}_{a_0, k}, \quad \mathbf{r}_{b_{a_1}} = \mathbf{b}_{a_1, k+1} - \mathbf{b}_{a_1, k}$$

$$\mathbf{r}_{b_{g_0}} = \mathbf{b}_{g_0, k+1} - \mathbf{b}_{g_0, k}, \quad \mathbf{r}_{b_{g_1}} = \mathbf{b}_{g_1, k+1} - \mathbf{b}_{g_1, k}$$

$$\Sigma_a = \{2\sigma_a^2/\Delta t + (\sigma_g^2/\Delta t)^2\} \cdot \mathbf{I}_{3 \times 3}, \quad \Sigma_g = 2\sigma_g^2/\Delta t \cdot \mathbf{I}_{3 \times 3},$$

$$\Sigma_{b_a} = \sigma_{b_a}^2 \Delta t \cdot \mathbf{I}_{3 \times 3}, \quad \Sigma_{b_g} = \sigma_{b_g}^2 \Delta t \cdot \mathbf{I}_{3 \times 3}$$

IMU1 (\mathcal{F}_{I_1}): ${}^{I_1}\tilde{\mathbf{a}}_k, {}^{g_1}\tilde{\boldsymbol{\omega}}_k$



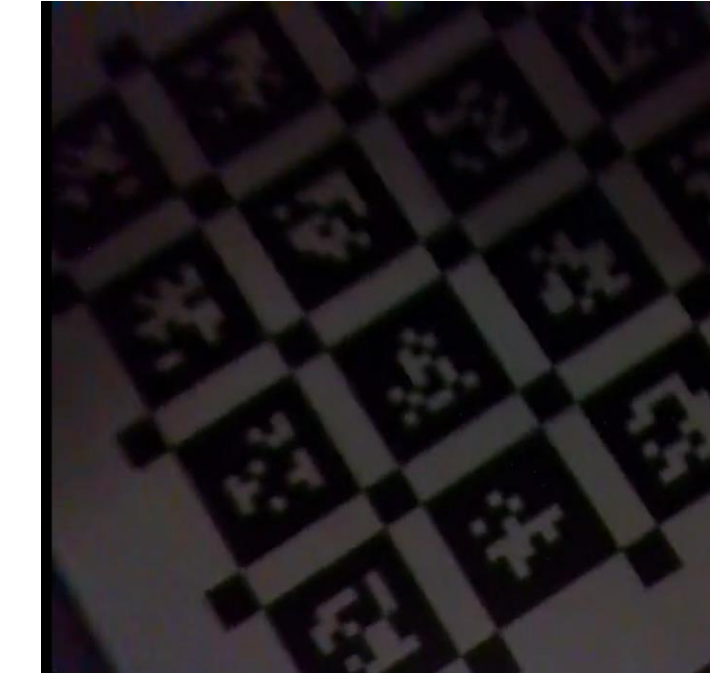
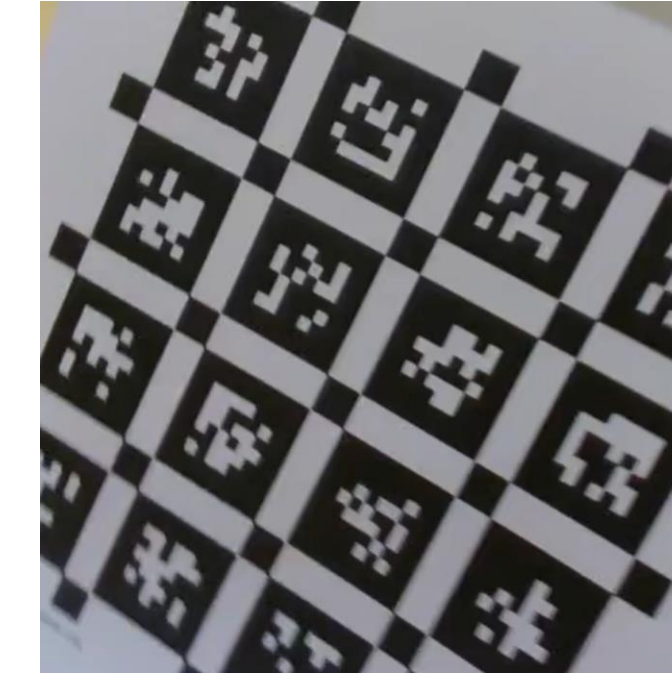
IMU0 (\mathcal{F}_{I_0}): ${}^{I_0}\tilde{\mathbf{a}}_k, {}^{g_0}\tilde{\boldsymbol{\omega}}_k$



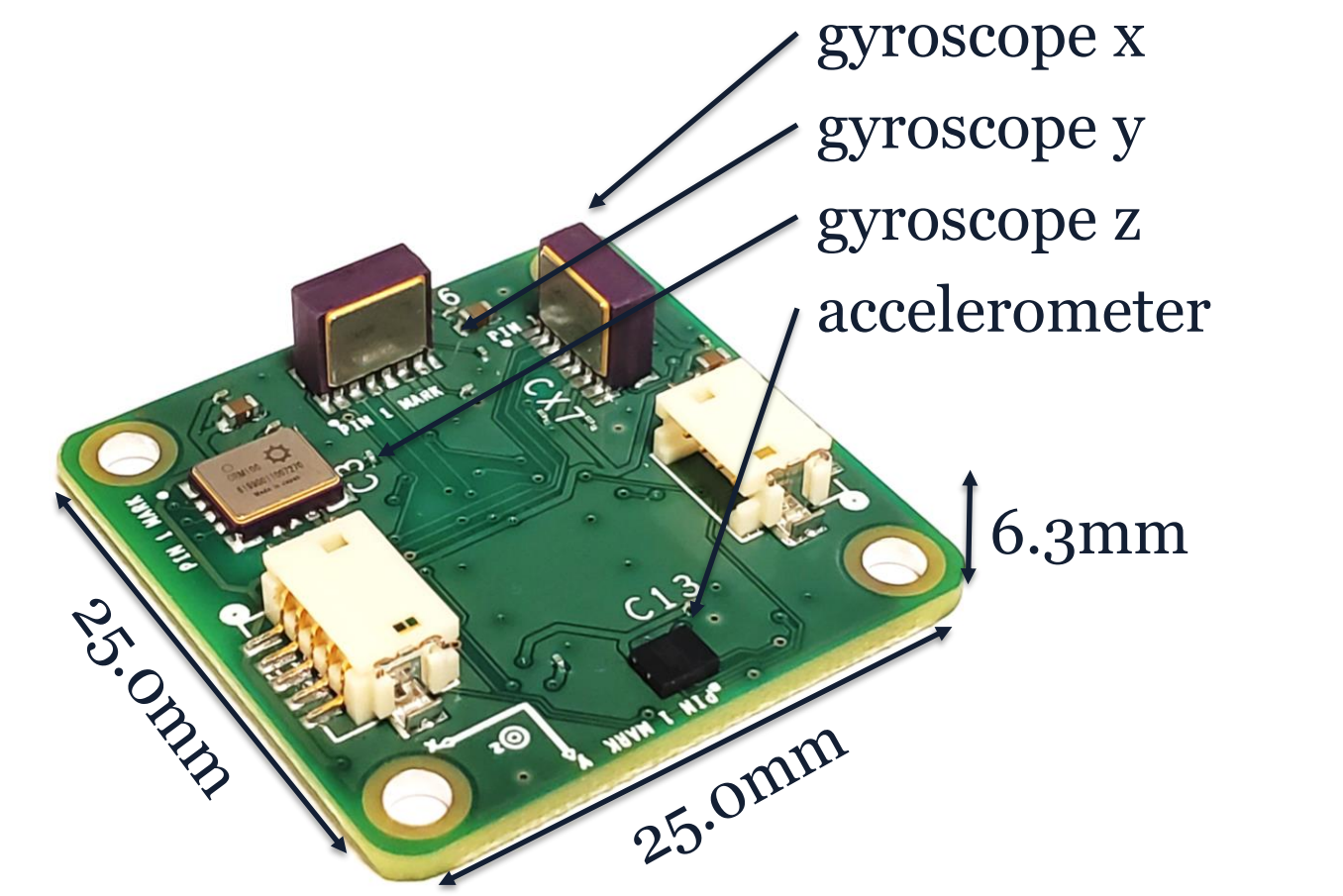
RESULTS

- We compare our method to Kalibr, a convenient benchmark for multi-IMU extrinsic calibration
- Kalibr needs a camera and fiducial marker, which accompany failure cases depending on visual conditions
- Both methods are tested over 65 trajectories (60 seconds for each) collected in three different conditions:

baseline (21 trajectories) blurry (23 trajectories) ill-lit (21 trajectories)



condition		Kalibr	Our Method
baseline	RMSE in \mathbf{p} [mm]	4.39	1.64
	RMSE in \mathbf{q} [deg]	1.35	2.91
	RMSE in ${}^g\mathbf{q}$ [deg]	2.06	2.50
	computation time [ms]	106.96 \pm 16.34	4.61 \pm 0.36
	success rate	19 / 21	21 / 21
blurry	RMSE in \mathbf{p} [mm]	70.41	2.02
	RMSE in \mathbf{q} [deg]	26.98	2.86
	RMSE in ${}^g\mathbf{q}$ [deg]	30.67	2.05
	computation time [ms]	98.99 \pm 12.27	4.74 \pm 0.58
	success rate	14 / 23	23 / 23
ill-lit	RMSE in \mathbf{p} [mm]	107.42	1.37
	RMSE in \mathbf{q} [deg]	1.81	4.19
	RMSE in ${}^g\mathbf{q}$ [deg]	1.03	3.51
	computation time [ms]	7.83 \pm 0.28	5.16 \pm 0.55
	success rate	2 / 21	21 / 21



IMU used for the experiments

\mathbf{p} : relative position, \mathbf{q} : relative orientation, ${}^g\mathbf{q}$: gyroscope misalignment

$\mathbf{p}_{\text{ref}}: [100, 100, 0] \pm [25.0, 25.0, 6.3] \text{ mm}$, $\mathbf{q}_{\text{ref}}, {}^g\mathbf{q}_{\text{ref}}: (\mathbf{e}, 0^\circ) \pm (\mathbf{e}, 9.5^\circ)$ for $\forall \mathbf{e} \in \mathbb{R}^3 \setminus \{\mathbf{0}\}$ (angle-axis)

- We acknowledge that the reference value we used— \mathbf{p}_{ref} and \mathbf{q}_{ref} derived from the optical breadboard on which sensors are mounted and zero gyroscope misalignment ${}^g\mathbf{q}_{\text{ref}}$ —can be a source of error themselves
- Hence, we focus on establishing that both our method and Kalibr produce “comparable” errors with respect to these reference values, rather than on rigorously establishing that our method produces “lower” error
- Under the baseline condition, our method matches the performance of Kalibr
- Under the blurry and ill-lit conditions, Kalibr often fails, whereas our method is unaffected

CONCLUSIONS AND FUTURE WORK

- We proposed a method of extrinsic calibration for multiple IMUs that only uses measurements collected by the IMUs themselves along arbitrary trajectories
- We used all available measurements for extrinsic calibration—in future, it may be helpful to choose a subset of measurements in order to decrease computation time and increase robustness
- We assumed measurements were time-synchronized—in future, it may be possible to include time synchronization as part of the extrinsic calibration process

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