AIMC

Attitude Independent Magnetometer Calibration in the Earth's Parking Orbit

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For UUM517E - Spacecraft Dynamics | Dr. Demet Cilden-Guler





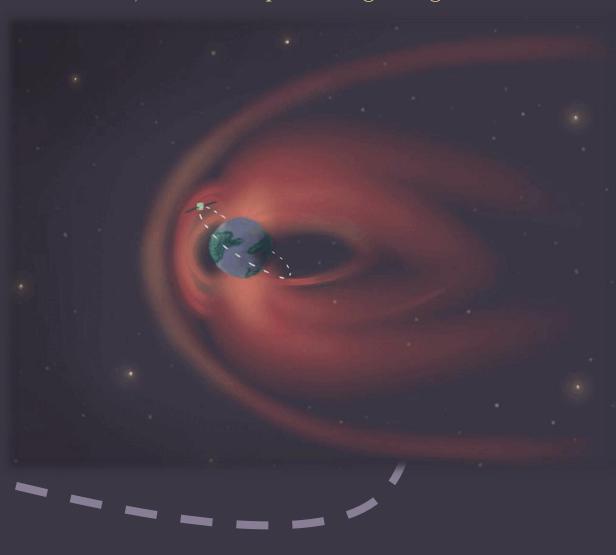
Classical problem in navigation: Where are we *headed*?

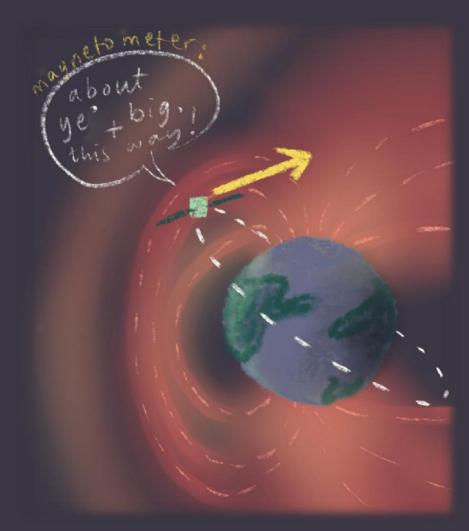


North (a reference vector)

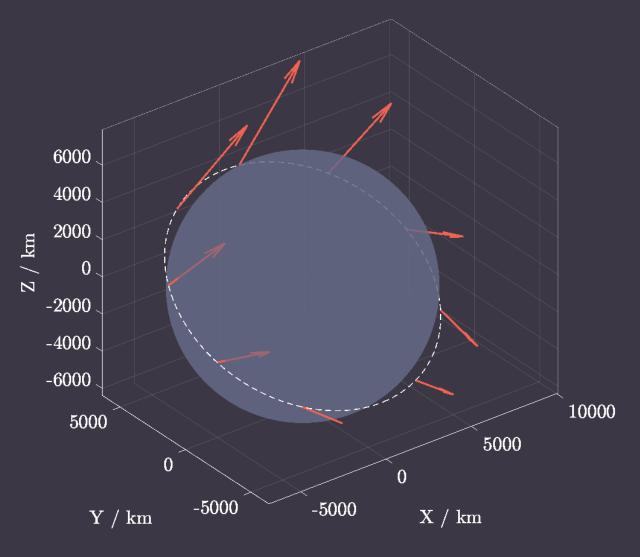
Compasses navigate in 2D (Earth's surface) using the geomagnetic field.

Spacecraft also need a reference to navigate in 3D, and also exploit the geomagnetic field.





Inform of geomagnetic field strength and 3D direction, like 3D compasses!



True Magnetic Field (IGRF) along the Earth's parking orbit every 10 minutes

Magnetometer Calibration

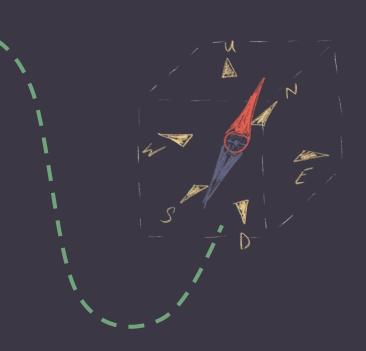
Two manufacturing defects:

- Noisy measurements •
- Non-orthonormal magnetometer axes -

Error propagates in measurements.

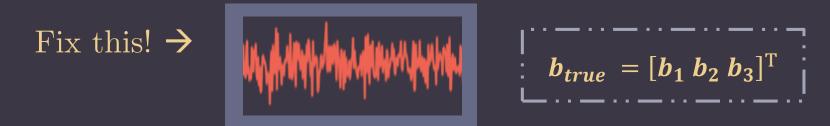
Calibration compensates for this.



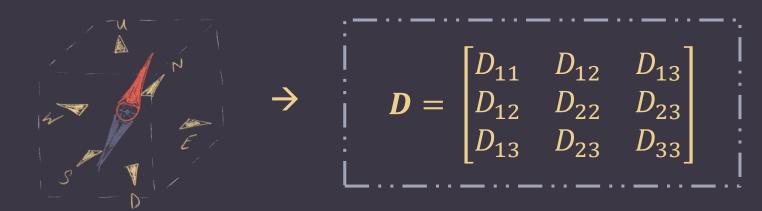


Goal of Calibration

• Find bias *compensation* to add to reading of each component.

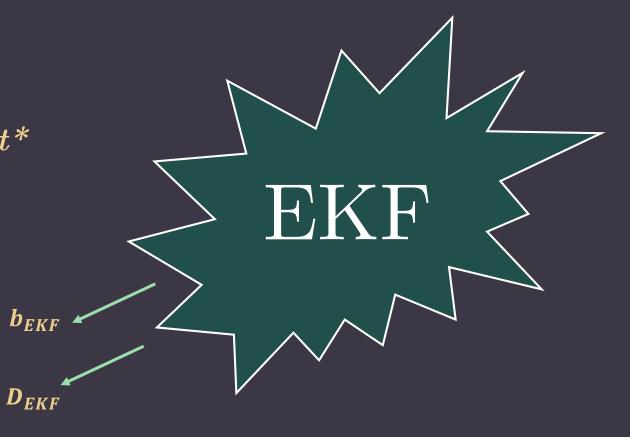


- Realign the magnetometer axes...
 - Find the orthonormality compensation matrix (symmetric 3x3).

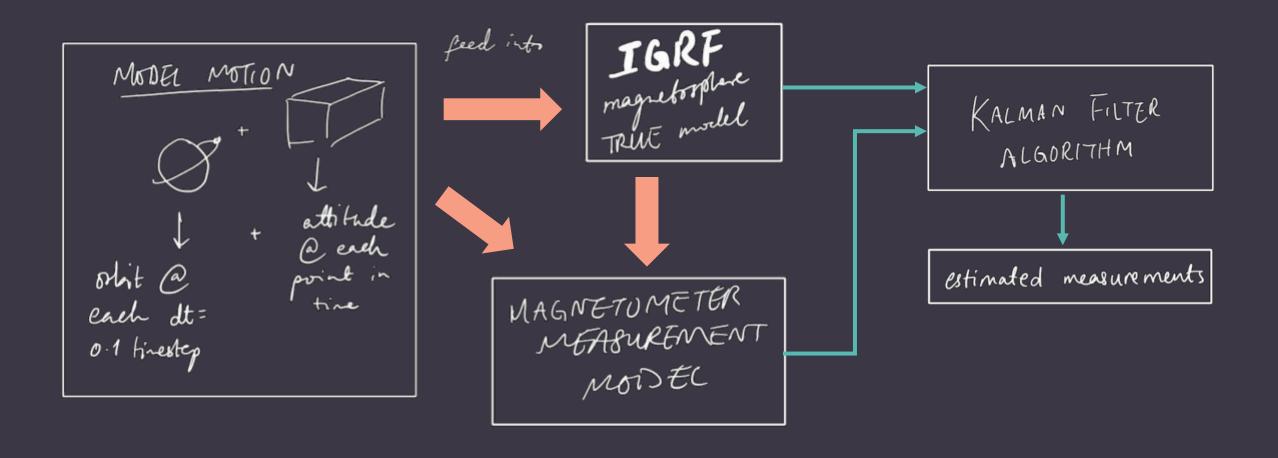


EKF for AIMC

- Current goal for modern-day spacecraft is:
 - onboard and
 - autonomous calibration
 - in real-time
 - without ground support*



My EKF 4 AIMC Algorithm



EKF Algorithm

Time Update ("Predict")

(1) Project the state ahead

$$\hat{x}_k = A\hat{x}_{k-1} + Bu_{k-1}$$

(2) Project the error covariance ahead

$$P_k = AP_{k-1}A^T + Q$$



(1) Compute the Kalman gain

$$K_{k} = P_{k}^{T}H^{T}(HP_{k}^{T}H^{T} + R)^{-1}$$

(2) Update estimate with measurement z_k

$$\hat{x}_k = \hat{x}_k + K_k(z_k - H\hat{x}_k)$$

(3) Update the error covariance

$$P_k = (I - K_k H) P_k$$



Calibration Scenarios

- Scenario 1: constant true bias vector
 - Usual case in calibrations

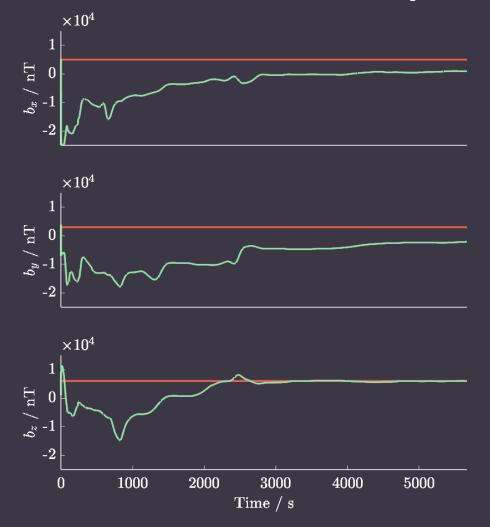
$$b_{true} = [3000\ 5000\ 6000]^{\mathrm{T}}$$

- Scenario 2: time-dependent true bias vector
 - Arises when interference with magnetometer due to other onboard electronics

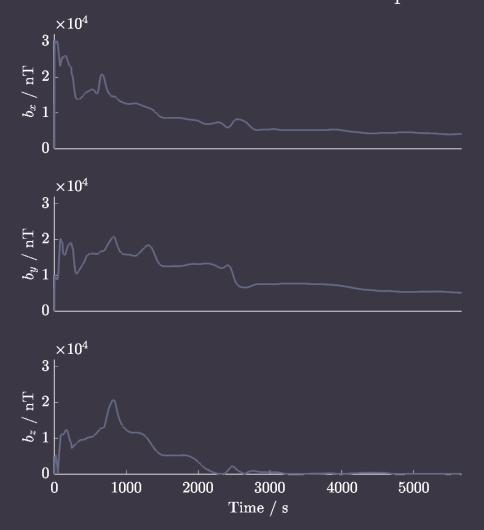
$$b_{true} = v_{TAM,bias}$$

where $v_{TAM,bias}$ can be modelled by a Gaussian distribution with a standard deviation of 300 nT/s

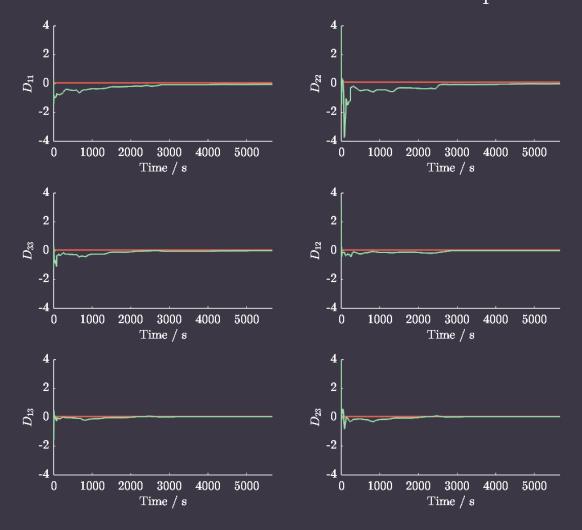
True and Estimated Bias Vector Components



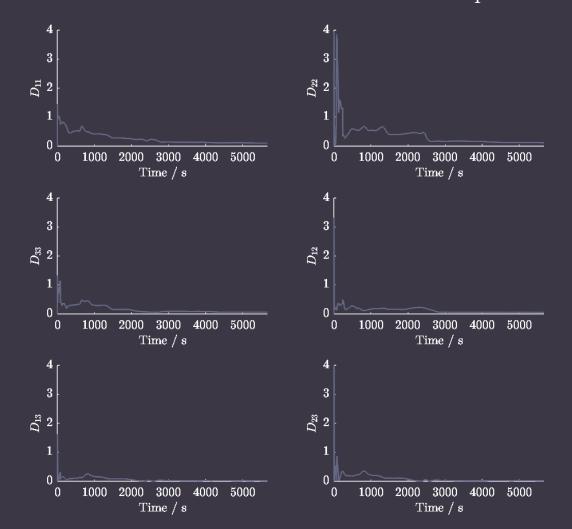
Error in Estimated Bias Vector Components



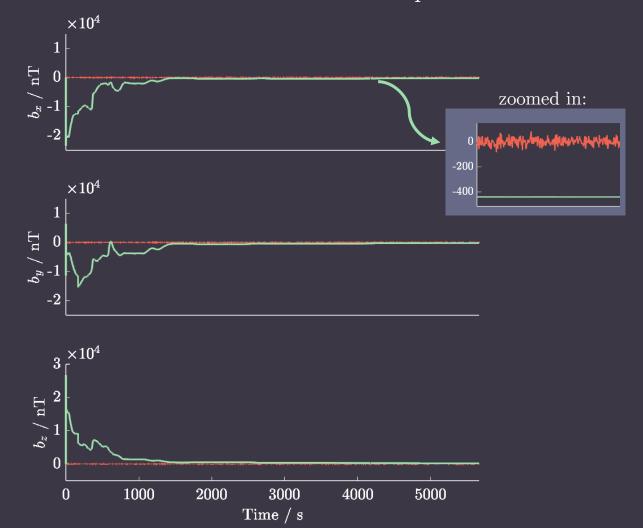
True and Estimated Correction Matrix Components



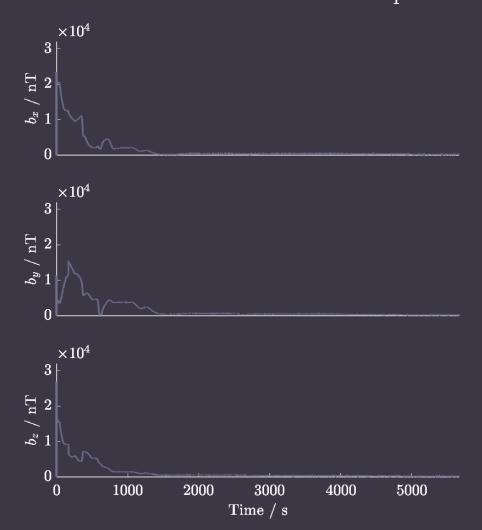
Error in Estimated Correction Matrix Components



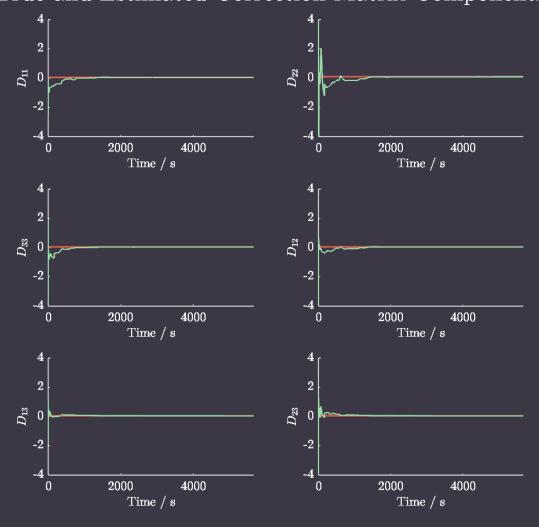
True and Estimated Bias Vector Components



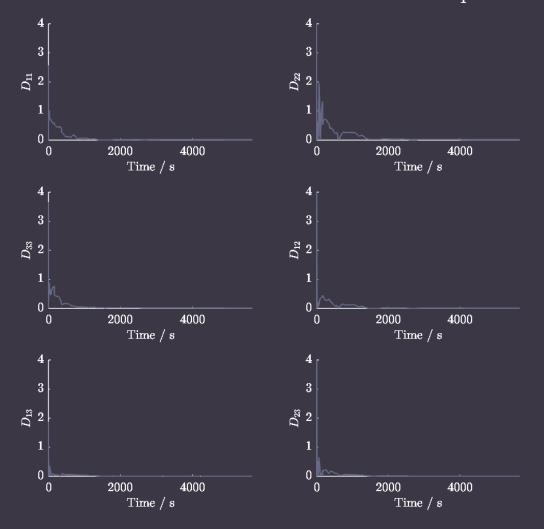
Error in Estimated Bias Vector Components



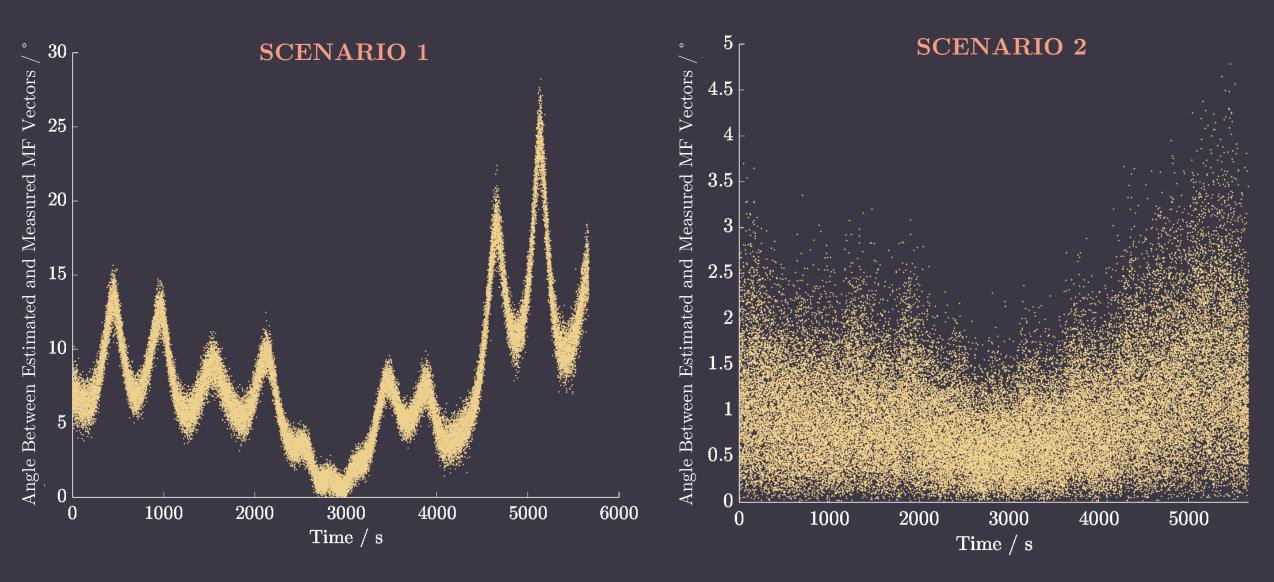
True and Estimated Correction Matrix Components



Error in Estimated Correction Matrix Components



Angle Residuals



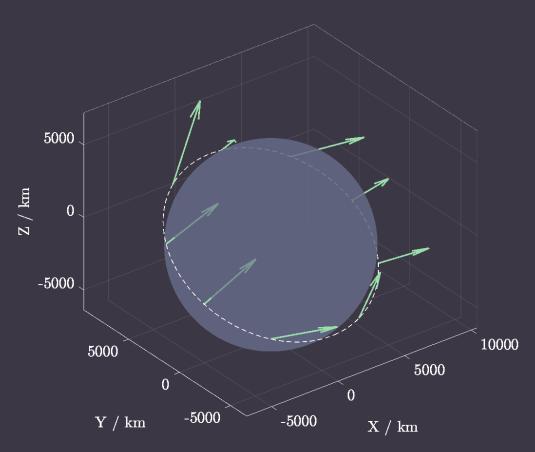
Results Summary (Vanilla)

RMS	$\boldsymbol{b}_{x} (nT)$	$oldsymbol{b_y} (nT)$	$b_z (nT)$
Scenario 1	6323	8400	5815
Scenario 2	3553	3100	2677

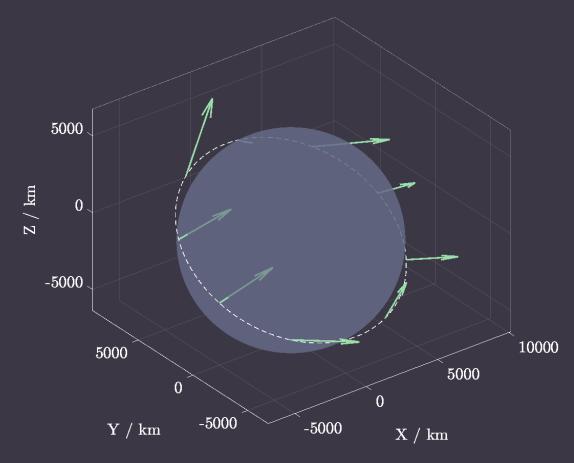
RMS	D_{11}	D_{22}	D_{33}	D_{12}	D_{13}	D_{23}
Scenario 1	0.2765	0.4235	0.1721	0.1046	0.0603	0.1198
Scenario 2	0.1421	0.2310	0.1295	0.0839	0.0676	0.1001

RMS	$B_x (nT)$	B_y (nT)	$B_z (nT)$
Scenario 1	22836	19810	21093
Scenario 2	20443	19190	20157

Visualizing the Measurements



Scenario 1 TAM Measurements every 10 minutes



Scenario 2 TAM Measurements every 10 minutes

Parameter Tuning (bonus)

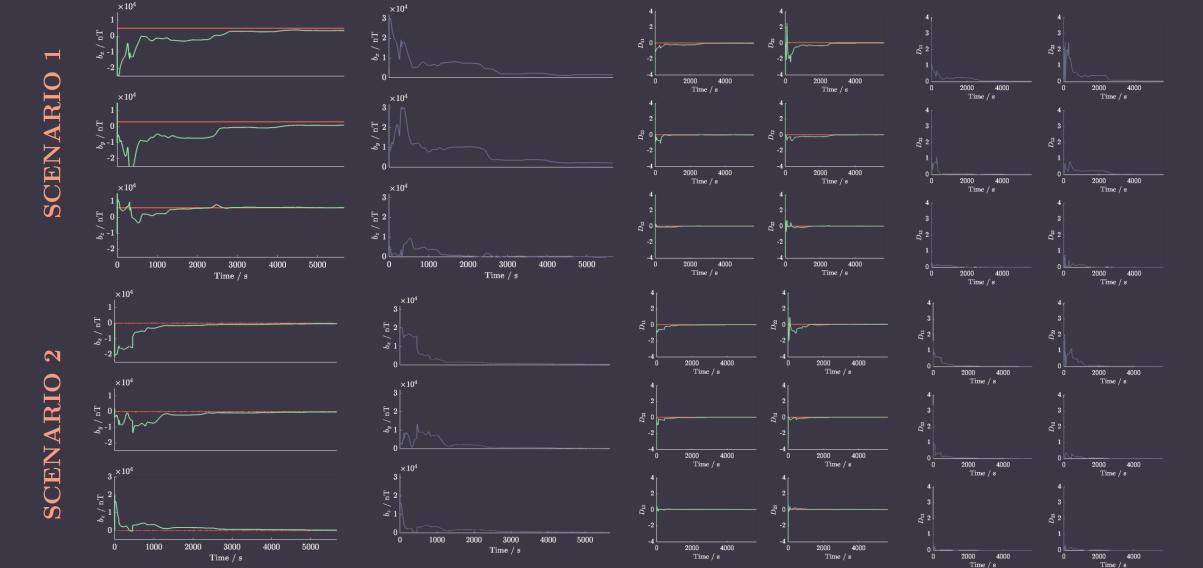
We can improve the results of our simulation by:

- 1. Tuning the initial parameters
- 2. Tuning the covariance matrix guess
- 3. Using a different true magnetic field model

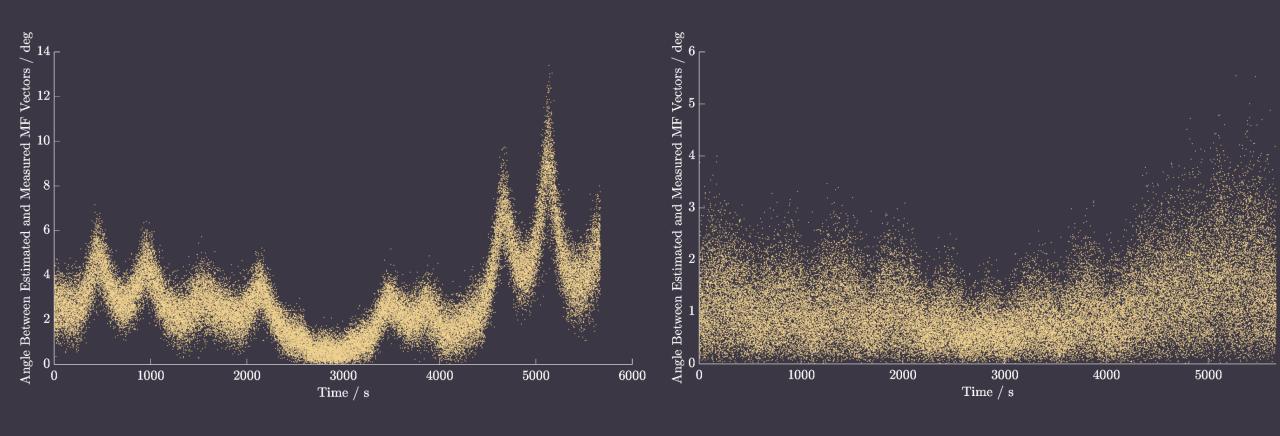
In case 3: IGRF → WMM

```
for i = 1:Nsteps
        [Btrue(i, :), ~] = wrldmagm(LLAorbit(i, 3), LLAorbit(i, 1), ...
        LLAorbit(i, 2), decyear(2024,11,22));
end
```

Employing WMM



WMM angle residuals



Parameter Tuning - \hat{P}_0

• Couldn't find the time ③

Parameter Tuning - b_{true0}

• Couldn't find the time ③

