

Geomagnetic Storms in the Context of Spacecraft Attitude Estimation under Different Noise Levels

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Abstract

This study examines geomagnetic storm conditions for attitude estimation of a spacecraft having magnetometers with various measurement noise levels at low Earth orbit. The geomagnetic field models are introduced by taking the external field effects into account. It is important to discuss the limitations of the models and the measurements that are being used. Therefore, the external effects are evaluated under different noise levels of magnetometer to determine and discuss the suppression level caused by the noise levels in a general framework. External magnetic field effects are examined by the ratio to noise under various magnetometer standard deviations up to 300 nT. An analysis is carried out for attitude estimation using Kalman-type filter in order to determine what noise level is acceptable on a particular sensor with a specified attitude requirement, magnetic field model, and space weather conditions.

This study indicates that external magnetic fields are vital for establishing attitude processes based on magnetometers with varying noise on which the spacecraft relies for accuracy.

Keywords: Geomagnetic storm, Attitude estimation, Magnetometer noise, Spacecraft.

1. Introduction

Spacecraft or their instruments are controlled for maintaining their orientation to specified points or directions. To control the craft or the device, the attitude needs to be estimated during flight. There are several sensors that can be used for satellites in low Earth orbit (LEO). One of the most widely used attitude sensor is the magnetometers as they are cheap, commercial off-the-shelf, light, and reliable sensors. However, they may be affected by the electronic devices on the satellite and, therefore, placed away from the satellite body on a boom. The magnetometers implemented on a LEO satellite to measure the internal geomagnetic field sources caused by the Earth's dynamo and crust as well as the external sources such as those created by interplanetary magnetic field and solar

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1 wind. In attitude determination and control systems, the International Geomagnetic Reference Field
2 (IGRF) (Thébault et al., 2015) is frequently used as the major magnetic field model. Yet, it might
3 remain incapable when geomagnetic activities occur, which can produce an error in the magnetic
4 field model in comparison with the sensor measurements. At this point, it is important to study the
5 effects of the geomagnetic storms on the spacecraft (Cui et al., 2020; Lu et al., 2019).

6 The external magnetic field variations caused by solar wind and magnetic storms and
7 magnetospheric substorms are generally treated as bias on the measurements, and removed from
8 the measurements by estimating them in the augmented states (Inamori et al., 2016; Inamori and
9 Nakasuka, 2012). The measurements in this case deviate from the real case after the elimination.
10 Another approach to reduce the errors resulting from the external fields is to consider the external
11 field in the geomagnetic model and not treat it as an error source (Cilden-Guler et al., 2021, 2018).
12 By this way, the geomagnetic model can represent the magnetic field closer to the reality. If the
13 magnetic field model used for the satellite attitude control does not consider the external fields, it
14 can misevaluate that there is more noise on the sensor, while actually the variations are caused by
15 a physical phenomenon (e.g. a magnetospheric substorm event), not from the sensor itself. That is
16 why, in this study, another geomagnetic field model in addition to the IGRF model is introduced,
17 which is called T89 (Tsyganenko, 1989) for modeling the geomagnetic field and simulating the
18 magnetometer better within the context of the spacecraft attitude estimation.

19 Many of the error sources such as bias on the magnetometer measurements may be compensated
20 during the ground tests prior to the spacecraft launch or during in-orbit calibration after launch. In
21 this research, the magnetometers are assumed to be calibrated against biases and scaling errors. On
22 the other hand, the measurement noise is present in the magnetometers as in all other sensors. The
23 standard deviation of magnetometer measurement noise in small satellites may vary depending on
24 the chosen sensor such as ~ 3 nT (Olsen et al., 2020), 40 nT (Cui et al., 2020), 100 nT (Zhang et al.,
25 2015), ~ 200 nT (Schulz et al., 2019), 200 nT (Carletta et al., 2020), 300 nT (Soken and Sakai,
26 2020). According to the literature, the standard deviations of magnetometers might vary by as much
27 as two orders of magnitude. This could be due to the magnetometer's required parameters such as
28 cost, size, weight, accuracy, and so on for a certain mission. This variation may result in having a
29 large error source from the sensor itself. If the noise on the magnetometer is large enough, the
30 external field might be suppressed and not revealed in the magnetic field observations. Therefore,
31 the critical noise levels masking the external fields are discussed for one magnetically active event
32 in (Cilden-Guler et al., 2019), and for an attitude estimation algorithm in the follow-up study
33 (Cilden-Guler et al., 2020). The purpose of this paper, on the other hand, is to evaluate these issues
34 in detail and to give a general framework in determining which model to be used at which noise
35 level case for attitude estimation applications. For this purpose, an extensive analysis is carried out
36 under various magnetic activity and magnetometer noise levels.

37 The rest of the paper is composed of,

- 38 • Mathematical model for the spacecraft's rotational motion,

- 1 • Description of the geomagnetic field models,
- 2 • Mathematical model for the magnetometer measurements,
- 3 • Attitude estimation algorithm based on Kalman type filtering,
- 4 • Analysis and results of case studies,
- 5 • Discussion and conclusions based on the analysis and results.

6 **2. Mathematical Models**

7 The mathematical models used in the estimation filter is given in this section. The spacecraft's
 8 rotational motion is used in the process model of the estimation filter and in modeling the actual
 9 dynamics of the spacecraft within the simulation. The geomagnetic field model is utilized in filter
 10 as a reference model and in modeling the magnetometer measurements.

11 **2.1. Spacecraft's Rotational Motion**

12 The spacecraft's attitude angles and angular rates are composing the state vector considered in
 13 this study. The attitude angles are represented by Euler angles by,

$$14 \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} = \begin{bmatrix} 1 & s(\phi)t(\theta) & c(\phi)t(\theta) \\ 0 & c(\phi) & -s(\phi) \\ 0 & s(\phi)/c(\theta) & c(\phi)/c(\theta) \end{bmatrix} \begin{bmatrix} p \\ q \\ r \end{bmatrix}. \quad (1)$$

15 Here cosine, sine and tangent functions are expressed by $c(\cdot)$, $s(\cdot)$, and $t(\cdot)$ respectively, vector
 16 of the angular velocity (ω_{BR}) in body coordinates with respect to the orbit coordinates has the
 17 components of p , q , r . Here, attitude problem is formulated using Euler angles as it is easy to
 18 visualize the three angles of rotation. However, it should be noted that Euler angles might be subject
 19 to singularity in some cases. Other attitude representations including quaternions and modified
 20 Rodriguez parameters can be used for a solution to the singularity issue. The angular velocity vector
 21 (ω_{BI}) in body coordinates with respect to the inertial frame can be expressed as,

$$22 \quad \omega_{BI} = \begin{bmatrix} \omega_x & \omega_y & \omega_z \end{bmatrix}^T, \quad (2)$$

23 The angular velocities (ω_{BI}, ω_{BR}) in body frame have a relationship,

$$24 \quad \omega_{BR} = \omega_{BI} - \mathbf{A} \begin{bmatrix} 0 & -\omega_o & 0 \end{bmatrix}^T, \quad (3)$$

25 where ω_o is the orbital angular velocity and can be computed for the circular orbits as,

$$26 \quad \omega_o = \left(\mu / r_0^3 \right)^{1/2} \quad (4)$$

27 with μ gravitational constant, r_0 spacecraft-Earth distance. \mathbf{A} is the attitude matrix transforming

1 the vectors from orbit to body coordinates.

2 Using the principle of the conservation of angular momentum, the dynamic equations are shown
3 as,

$$4 \quad J_x \frac{d\omega_x}{dt} = N_x + (J_y - J_z) \omega_y \omega_z, \quad (5)$$

$$5 \quad J_y \frac{d\omega_y}{dt} = N_y + (J_z - J_x) \omega_z \omega_x, \quad (6)$$

$$6 \quad J_z \frac{d\omega_z}{dt} = N_z + (J_x - J_y) \omega_x \omega_y. \quad (7)$$

7 Here, J_x , J_y and J_z are the elements of the moment of inertia matrix whereas the external
8 disturbances are shown as N_x , N_y and N_z .

9

10 2.2. Geomagnetic Field Models

11 To determine the attitude angles of a satellite using magnetometer measurements, the estimation
12 filter uses one of the geomagnetic field models. IGRF (International Geomagnetic Reference Field)
13 is a commonly used geomagnetic field model and it computes the geomagnetic field vector using
14 the inputs of date and position of the spacecraft orbiting around Earth (Thébault et al., 2015; Wertz,
15 2002). In the model, the magnetic field at a point in space is predicted using the spherical
16 components as stated in Eq. (8) below:

$$17 \quad \mathbf{B}_{\text{IGRF}}(\bar{r}, \text{colat}, \text{lon}, t) = -\nabla \left\{ a \sum_{n=1}^N \sum_{m=0}^n \left(\frac{a}{r} \right)^{n+1} [g_n^m(t) c(m \text{lon}) + h_n^m(t) s(m \text{lon})] \times P_n^m(c(\text{colat})) \right\}, \quad (8)$$

18 where \mathbf{B}_{IGRF} represents the magnetic field vector prediction in nT, $\text{colat}(t)$ is the co-latitude,
19 $\text{lon}(t)$ is the longitude, \bar{r} is spacecraft-Earth mass centers' distance with $a = 6371.2$ km being the
20 mean radius of Earth, P_n^m is the Schmidt quasi-normalized associated Legendre polynomials of
21 degree n and order m , g_n^m and h_n^m are the Gaussian coefficients in nT (Thébault et al., 2015).

22 As IGRF model considers only the internal magnetic field, the real magnetometer measurements
23 might be under predicted in the simulations since the external magnetic fields are not considered.
24 In order to simulate the real physical space environment, the T89 model that consists both internal
25 and external magnetic field sources is selected (Tsyganenko, 1989). T89 is composed of two parts
26 as:

$$27 \quad \mathbf{B}_{\text{T89}_k} = \mathbf{B}_{\text{IGRF}_k} + \mathbf{B}_{\text{ext}_k}, \quad (9)$$

$$28 \quad \mathbf{B}_{\text{ext}} = \mathbf{B}_{\text{ring}} + \mathbf{B}_{\text{tail}} + \mathbf{B}_{\text{mp}} + \mathbf{B}_{\text{FC}}, \quad (10)$$

1 where \mathbf{B}_{T89} represents magnetic field vector of T89 model including the IGRF outputs (\mathbf{B}_{IGRF}) and
 2 external magnetic field contribution (\mathbf{B}_{ext}). IGRF model only considers the internal part of the
 3 geomagnetic field of Earth and updates its constants every five year (Thébault et al., 2015) whereas
 4 T89 model uses large data sets from variety of satellites at different orbits ranging from LEO up to
 5 40 Re (1 Re = 6371 km) in the magnetosphere behind the Earth. As seen in Eq. (10), \mathbf{B}_{ext} includes
 6 magnetic fields generated from different external sources like magnetospheric ring (\mathbf{B}_{ring}), tail
 7 (\mathbf{B}_{tail}), magnetopause (\mathbf{B}_{mp}), and field aligned (\mathbf{B}_{FC}) currents (Tsyganenko, 2002, 1995, 1989) in
 8 the magnetosphere.

9 Since the external magnetic field (\mathbf{B}_{ext}) is superimposed on the internal geomagnetic field, T89
 10 gives the total geomagnetic field at the specified position. The magnetic activity level in T89 model
 11 is determined using the K_p index. It is calculated globally using magnetic field data from
 12 midlatitude magnetic stations at every 3-hours and expressed as the thirds of a unit, e.g. 2- is 1 2/3,
 13 5o is 5 and 5+ is 5 1/3 (“Geomagnetic Kp and Ap Indices | NCEI,” n.d.; Siebert and Meyer, 1996).
 14 One of the current issues related to T89 model in using it for spacecraft attitude purposes is about
 15 its inputs. T89 model acquires the magnetic index data such as K_p , AE , D_{st} etc. on board the
 16 spacecraft. It can be used during especially geomagnetically active times, for more accurate attitude
 17 estimations. These indices processed under T89 can be obtained from magnetic stations, and then
 18 sent to the spacecraft by telecommand signals. In this paper, we assumed to have this information
 19 every 3-hour without any delays. However, this might not be possible when the telecommand is
 20 not available for a portion of the orbital period. In such case, one solution is to prepare look-up
 21 tables for specific inputs as an alternative. Also, since T89 computes the external fields in addition
 22 to the internal fields from IGRF and sums those, its computational burden is heavier than the IGRF
 23 model.

24 2.3. Magnetometer Measurements

25 The magnetometers are the basic sensors at low Earth orbiting satellites for attitude determination
 26 purposes. The magnetometer measurements are modelled by using the geomagnetic field model
 27 vector that is transformed to the body frame and the measurement noise vector as,

$$28 \quad \mathbf{B}_{m_k} = \mathbf{A}_k \mathbf{B}_{o_k} + \mathbf{v}_{B_k}, \quad (11)$$

29 where \mathbf{B}_o is the magnetic field vector in orbit frame obtained from a geomagnetic field model, \mathbf{B}_m
 30 is the magnetometer measurements in the spacecraft’s body frame, \mathbf{v}_B is the zero-mean Gaussian
 31 noise of the measurements. In this study, magnetometers are modeled so as to sense the external
 32 field based on the K_p index defined. For this purpose, T89 model is used for the magnetic field
 33 vector as $\mathbf{B}_{o_k} = \mathbf{B}_{o_k}^{T89}$ to generate the magnetometer measurements.

3. Attitude Estimation Algorithm based on Kalman Type Filtering

An attitude estimation algorithm using Kalman type filter is seen in this section. The attitude states of the spacecraft can be estimated by using magnetometer measurements based filtering methods. For this purpose, a conventional extended Kalman filter is used in this study to show the effects of the external field and measurement noise levels on the attitude estimations. Fig. 1 shows the filtering procedure using geomagnetic field models and magnetometer measurements. The geomagnetic activity index is also included for taking the magnetic field anomalies into account. An EKF algorithm is used in this paper as the dynamics of the satellite's rotational motion and measurement models are nonlinear. We kept the filtering algorithm as a conventional EKF for focusing on the external effects and not the filtering extensions.

The attitude estimation problem can be expressed in terms of discrete-time nonlinear state-space model,

$$\mathbf{x}_k = f(\mathbf{x}_{k-1}) + \mathbf{w}_k, \quad (12)$$

$$\mathbf{y}_k = h_k(\mathbf{x}_k) + \mathbf{v}_k, \quad (13)$$

Here, $f(\cdot)$ is the system function and $h(\cdot)$ is the measurement function, \mathbf{x} is composed of the state elements, \mathbf{w} is zero-mean Gaussian noise with \mathbf{Q} , \mathbf{y} is measurement vector, and \mathbf{v} is a zero-mean Gaussian noise with \mathbf{R} . The state vector is composed of attitude angles and angular rates in this study. The measurement model used in the filter differentiates by scenarios in this study as \mathbf{B}_{IGRF} for IGRF and \mathbf{B}_{T89} for T89. However, the magnetometer measurements do not vary case to case and always sense the external field by using T89 model for the geomagnetic field vector.

The prediction and the update stages should be processed under EKF algorithm (Psiaki et al., 1990). The state vector can be estimated by,

$$\hat{\mathbf{x}}_k = \hat{\mathbf{x}}_{k|k-1} + \mathbf{K}_k \left\{ \mathbf{y}_k - h(\hat{\mathbf{x}}_{k|k-1}) \right\} \quad (14)$$

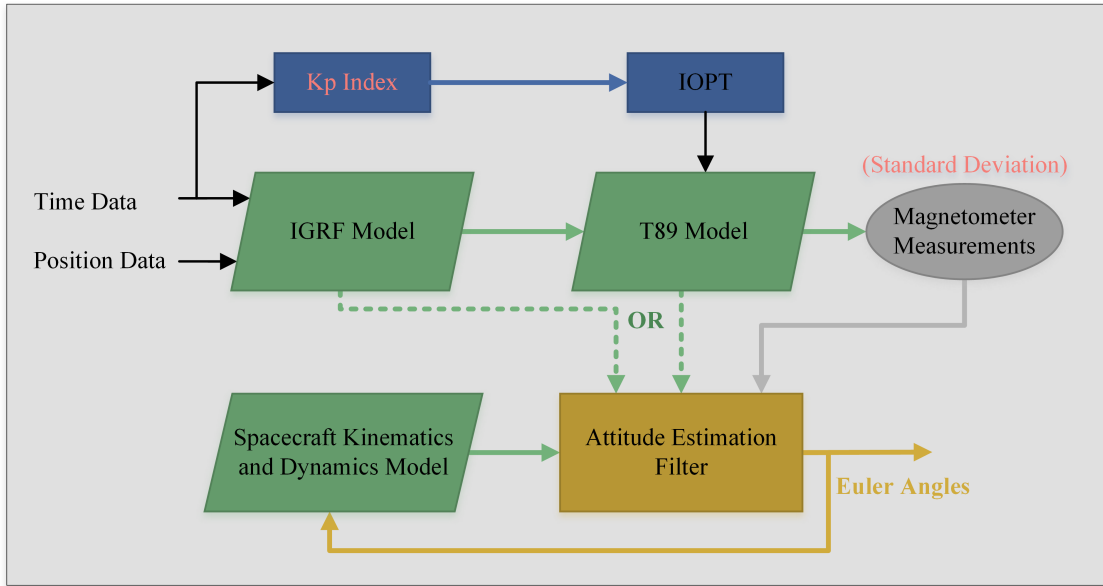
The predicted vector is,

$$\hat{\mathbf{x}}_{k|k-1} = f(\mathbf{x}_k) \quad (15)$$

The EKF's filter gain is,

$$\mathbf{K}_k = \mathbf{P}_{k+1|k} \mathbf{H}_k^T \left[\mathbf{H}_k \mathbf{P}_{k+1|k} \mathbf{H}_k^T + \mathbf{R} \right]^{-1} \quad (16)$$

where the measurement matrix is created using $\mathbf{H}_k = \frac{\partial h(\hat{\mathbf{x}}_{k|k-1})}{\partial \hat{\mathbf{x}}_{k|k-1}}$.



1

2 **Fig. 1.** Attitude estimation algorithm flow chart using magnetometer measurements.

3 The prediction error covariance matrix is,

4
$$\mathbf{P}_{k|k-1} = \frac{\partial f(\hat{\mathbf{x}}_k)}{\partial \hat{\mathbf{x}}_k} \mathbf{P}_{k-1|k-1} \frac{\partial f^T(\hat{\mathbf{x}}_k)}{\partial \hat{\mathbf{x}}_k} + \mathbf{Q} \quad (17)$$

5 The covariance matrix of the estimation error is,

6
$$\mathbf{P}_{k|k} = [\mathbf{I} - \mathbf{K}_k \mathbf{H}_k] \mathbf{P}_{k|k-1} \quad (18)$$

7 In this study, only magnetometer measurements are used as defined in (11),

8
$$\mathbf{y}_k = \mathbf{B}_{m_k} \quad (19)$$

9 The magnetic field model is used to compose the measurement matrix \mathbf{H}_k within the filter.

10 **4. Analysis and Results of the Case Studies**

11 In the first setup, the simulations are run for a spacecraft with
 12 $\mathbf{J} = \text{diag}\left(\left[2.1 \times 10^{-3} \quad 2.0 \times 10^{-3} \quad 1.9 \times 10^{-3}\right]\right)$ kg.m² principle mass moment of inertia that is
 13 tumbling along a low Earth circular orbit in the equatorial plane with 500 km altitude. The diagonal
 14 elements of the measurement noise covariance matrix are set to the square of standard deviation of
 15 the related sensor, and system noise covariance is set to

1 $\mathbf{Q} = \text{diag}\left(\left[10^{-10} \quad 10^{-10} \quad 10^{-10} \quad 10^{-12} \quad 10^{-12} \quad 10^{-12}\right]\right)$. The simulation is initialized with the state
2 vector of $\mathbf{x}_0^{\text{true}} = [0.03 \text{ deg} \quad 0.02 \text{ deg} \quad 0.01 \text{ deg} \quad 0.001 \text{ deg/s} \quad 0.0015 \text{ deg/s} \quad 0.002 \text{ deg/s}]^T$
3 and the filter is initialized as $\mathbf{x}_0^{\text{EKF}} = 2\mathbf{x}_0^{\text{true}}$. The simulations are performed for two orbital periods
4 around 3 hours in total, and repeated for each magnetometer measurement noise standard deviation
5 (σ) case and each K_p value. The orbit is propagated by employing the Simplified General
6 Perturbation Version 4 (SGP4) model introduced by (Vallado and Crawford, 2008). The
7 magnetometers are processed at 1 Hz and corrupted by Gaussian zero-mean noise with various
8 standard deviations in nT. The simulations take the external field effects into account by
9 implementing the various levels of K_p index.

10 The first part of this section focuses on the ratio between the external magnetic field and
11 magnetometer measurement noise. The ratio is calculated using the norms of these vectors as
12 $\|\mathbf{B}_{\text{ext}}\|/\|\mathbf{v}\|$ using predicted external field given by Eq. (10) and measurement noise given in Eq.
13 (13). The ratio is presented as box plots for different activity levels and standard deviation value in
14 different panels of Fig. 2. The boxplot has the data distribution of minimum, first quartile (bottom
15 of the box), median (thin line in the box), third quartile (top of the box), and maximum. If the ratio
16 is larger than 1, it means that the external field is greater than the noise, whereas the noise might
17 cover the external effects for the rates that are smaller than 0.5. As reference lines, y-axis is marked
18 with green at 0.5 and 1 in Fig. 2. The mean ratio is shown with a dark blue line. Selection of 0.5 is
19 arbitrary just to give an easy reference guide.

20 The red box in Fig. 2 is created for determining the standard deviation level, where the noise mean
21 is twice as large than the external field mean value. From this reference red box, one can determine
22 the noise level that makes the external field less significant for attitude estimation purposes.
23 However, it should be noted that the factor is arbitrarily chosen as 0.5 to be used as a guide. The
24 white filled ellipse shape marks the first ratio value under 1.

25 Fig. 2.a shows the simulation performed under $K_p = 0$ that indicates absence of external magnetic
26 field variations. The external field ratio over noise is reduced fast, and is masked after 60 nT
27 magnetometer measurement noise standard deviation. The mean ratio follows an exponentially
28 decreasing line.

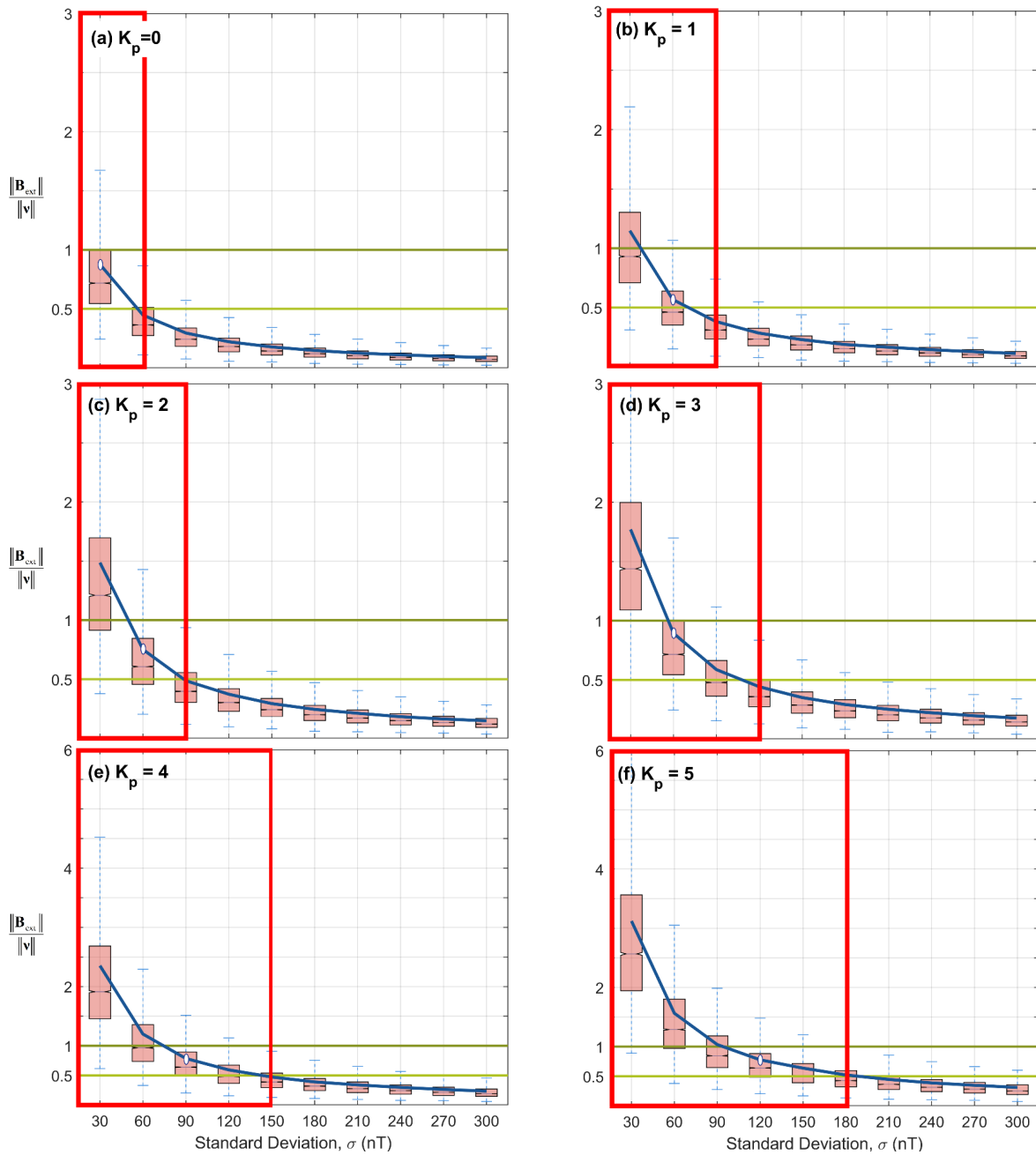
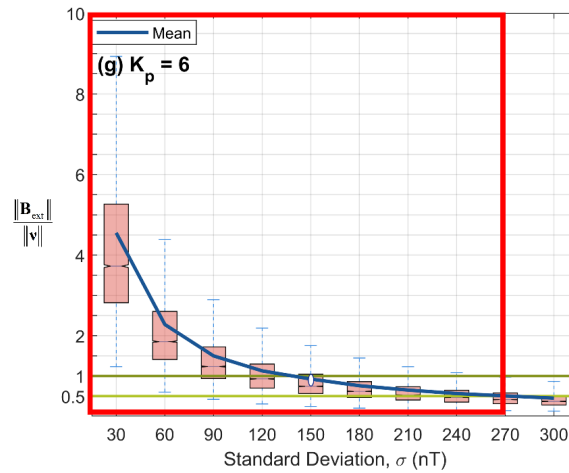
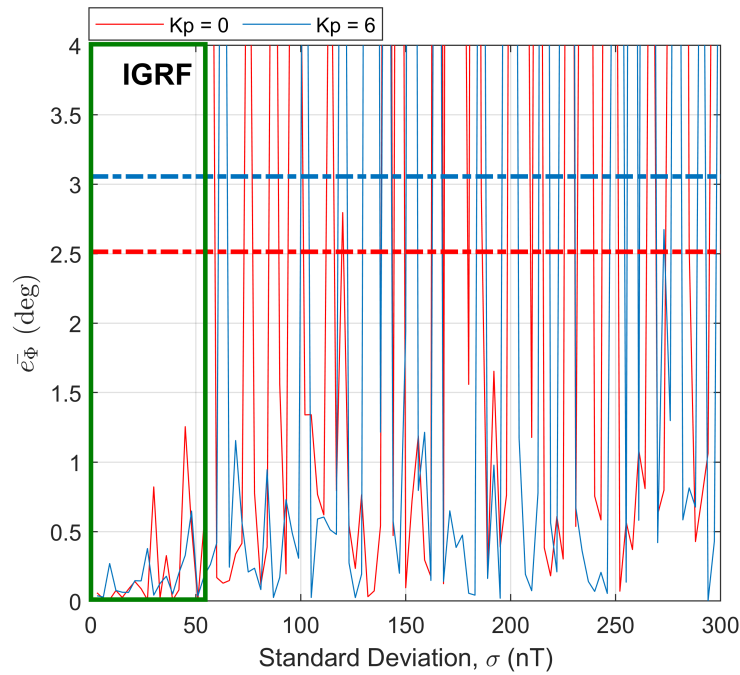


Fig. 2. Rate of the external field over noise on the magnetometer measurements. Note that the vertical scale is not the same in each panel (continued on the next page).

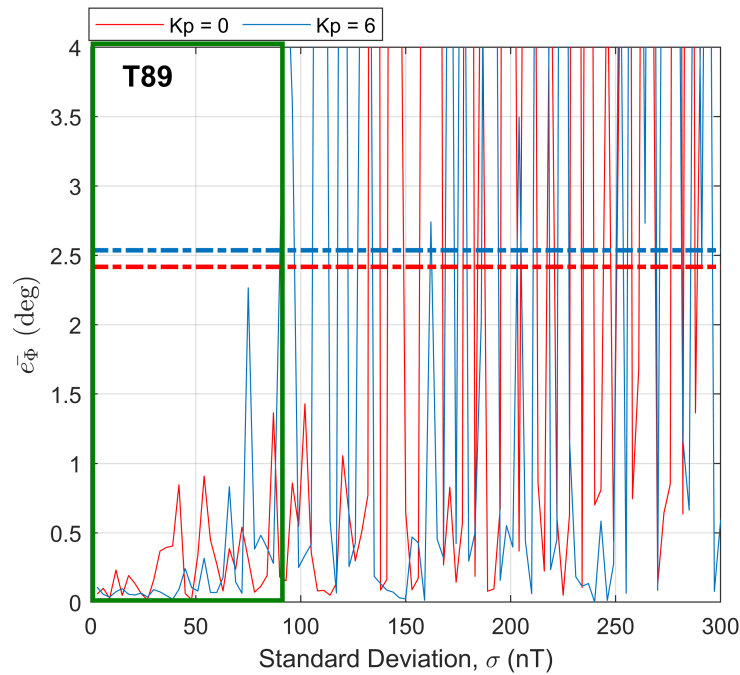


1 **Fig. 2.** Rate of the external field over noise on the magnetometer measurements. Note that the
 2 vertical scale is not the same in each panel (continued from the previous page).

3 The following panels give other cases with different K_p levels up to 6 which corresponds to the
 4 maximum magnetospheric activity level. The external effects are more effective up to the white
 5 filled ellipse than the measurement noise. The reference line 0.5 and the region in the red box in
 6 our case indicates that the external field still have an impact in the measurements. The region
 7 expands with larger K_p values. The corresponding standard deviation values are 60 nT, 90 nT, 90
 8 nT, 120 nT, 150 nT, 180 nT, and 270 nT for $K_p = \{0 \ 1 \ 2 \ 3 \ 4 \ 5 \ 6\}$ respectively. This
 9 indicates that the external field can affect the magnetometer measurements that have noise levels
 10 with standard deviation up to 270 nT. It should be noted that the y axis limits are set differently for
 11 active times ($K_p \geq 4$) and $K_p = 6$ in the panels of Fig. 2. The top quartiles of the box plots start
 12 to exceed the value 1 when $K_p \geq 4$. The mean value of the ratio also gets greater than 2. The
 13 external field is larger for 90 nT, 120 nT, and 150 nT standard deviation under $K_p = \{4 \ 5 \ 6\}$
 14 respectively. From this result, it can be said that external magnetic field should be taken into
 15 account especially if the magnetometer measurement noise standard deviation is smaller than 150
 16 nT. External field that is greater than the sensor noise will not be masked in the attitude
 17 determination procedure as an insignificant term. The differences in the strength of the magnetic
 18 field are found much smaller than those of magnetic field components (Cilden-Guler et al., 2018).
 19 Therefore, it can be stated that the geomagnetic storms have an effect on the attitude determination
 20 that is based on the angle between the directions of the reference and measurement vectors. The
 21 magnitude variations caused by geomagnetic storms without any directional changes would not
 22 affect the attitude performance as the attitude determination methods mostly use only the direction
 23 and not the magnitude in their computations.



1



2

3 **Fig. 3.** The absolute error mean of the attitude estimation with two activity levels and varying noise
4 using IGRF (top) and T89 (bottom).

1 The absolute mean attitude estimation errors after convergence along the orbit are calculated as

2
$$\bar{e}_{\Phi} = \sum_{k=1}^N \left\| \mathbf{x}_{\Phi_k}^{true} - \mathbf{x}_{\Phi_k}^{est} \right\| / N$$
 where $\mathbf{x}_{\Phi_k}^{true}$ is the true attitude angle vector, $\mathbf{x}_{\Phi_k}^{est}$ is the estimated attitude

3 vector, and N is the number of samples after convergence are presented in Fig. 3. Two cases are
4 considered for the simulation as IGRF or T89 implementation for the magnetic field model in the
5 filter. There aren't any additional rules defined in the EKF design for the external field. The filter
6 is tested against this case in its conventional form. Two activity levels are selected to see the
7 difference in the estimation errors. Also, the mean values of the absolute mean attitude estimation
8 errors are indicated with dotted dashed lines for both $K_p = \{0, 6\}$ as around 2.51 and 3.05 degrees
9 for IGRF, 2.41 and 2.53 degrees for T89. The external field is only considered under the
10 magnetometer measurements when the IGRF model is used. The mean value line of the attitude
11 errors is only used as a limiting value to determine the green box's width before the first jump
12 larger than the mean values of each figure. The corresponding standard deviations are around 50
13 nT and 100 nT for IGRF and T89 cases. These are critical points of the magnetometer noise levels
14 that affect the attitude estimations abruptly. Another important point is that the mean error
15 difference between $K_p = 0$ and $K_p = 6$ cases becomes smaller when using T89, as it considers
16 the activity levels within the model. When the standard deviation is approaching to zero, in case of
17 $K_p = 6$, the attitude estimation errors are under 0.1 deg. From the results, the measurement noise
18 has more significant effect on the attitude estimation than the external field depending on the
19 standard deviation level. The most significant conclusion from this figure is that one can decide
20 the acceptable noise level for a given sensor with specified pointing requirement. For example, if
21 T89 model is used on a spacecraft with 1-degree pointing requirement during a quiet time, then
22 one can use a magnetometer with a standard deviation up to ~ 70 nT. For active times, it is ~ 80 nT
23 for T89 and ~ 50 nT for IGRF. Here, T89 gives more room for the magnetometer selection. The
24 variations in the standard deviation affect the estimation results significantly as seen in the figure
25 as an abrupt change at the critical point that starts to mask the external field effects.

26 5. Conclusions and Discussion

27 In this study, the geomagnetic field models used in estimating the spacecraft's attitude angles are
28 presented during geomagnetically active and quiet times. Magnetic field models commonly used
29 for spacecraft attitude determination and control systems do not consider the external fields and
30 thus the system might misinterpret the field anomalies as noise/ bias on the sensor, whereas the
31 variations are caused by storm events, and not from the sensor itself. The external magnetic fields
32 are detected by the magnetometer sensors; therefore, one of the geomagnetic models is selected so
33 that it would take the magnetic anomalies into account. Here, the important point is that the level
34 of the external magnetic field disturbance or the magnetometer noise that can vary depending on
35 the sensor or strength of the external event. In order to see the effects of the external field and the
36 measurement noise, a series of magnetometer sensor noise levels are applied under various
37 magnetic activity levels.

1 A critical magnetometer noise level was previously presented in (Cilden-Guler et al., 2020) as 70
2 nT based upon one simulation case. However, in this study, it is shown that the external magnetic
3 fields should be taken into account for the standard deviation of the measurement noise up to 150
4 nT. The external field can also be exposed and affect the results for the sensors having larger
5 standard deviation value than 150 nT but not at the same level with those of smaller ones. The
6 attitude estimation is influenced by the noise abruptly after approximately 50 nT and 90 nT for
7 IGRF and T89 respectively. The mean error difference between $K_p = 0$ and $K_p = 6$ cases
8 decreases when using T89 and comparing IGRF.

9 The presented results for determining the critical noise levels can be implemented on other systems,
10 which are close enough to Earth and use magnetometer measurements for attitude determination
11 purposes. Conventional EKF algorithm is used for attitude estimation in this study. For further
12 work, other filtering algorithms can also be tested such as UKF, adaptive filters, particle filters etc.

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