

AE4878 - Mission Geometry and Orbit Design

Part 6 - Error Analysis

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1 Why analyse errors?

No system can be perfectly designed. Let it be mechanical tolerances or algorithmic limitations, all engineering practices come with limited precision. This precision is quantified with the help of errors. Errors define the accuracy of any given engineering system under consideration. To understand the bounds of perfection of a system, errors present in a system must be analysed. Error analysis of any system come with the following aspects: error sources (where in the system does the error take place?), error characterisation (what type of error is it?), error propagation (what effects does the error have?) and error budget (what's the total error?).

For this assignment, the sources and types of errors are given. The objective is to use different error propagation techniques to evaluate the effect of the errors and obtain the corresponding error budget. This report will look into two different techniques of error propagation, 1. an analytic approach 2. Monte Carlo approach.

For analytic approach, analytic relations between error and effect are assessed and propagated. While for Monte Carlo based approach, N trials are carried out with variations in error assumptions. The errors are defined within a confidence interval and randomly generated, later they are propagated.

First the problem is defined in Section 2. Section 3 generates and verifies analytic error propagation. Section 4 evaluates the impact of systematic errors on analytic error propagation. Section 5 conducts a Monte Carlo error propagation and Section 6 evaluates the impact of systematic error on Monte Carlo based error propagation.

2 Problem Definition

Figure 1[1][Fig.5-11,pg.265] represents a sketch of the problem under consideration. The orbit altitude h , is considered to be 1000km with an elevation angle ε of 21.6052° as seen from the target Earth. Flat Earth model is considered. θ defined the Nadir angle. The distance from sub-satellite point (SSP) to Target is defined as D_{nom} and the distance from the satellite to the target is defined as D .

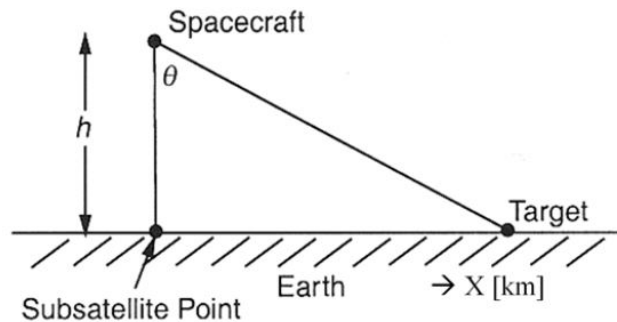


Figure 1: 2D overview of the problem under consideration.

Relations for error propagation are Tabulated with the aid of Figure2[1][Table5.7,pg.254]. Note the Nadir angle η is considered to be equivalent to θ throughout this report.

TABLE 5-7. Mapping and Pointing Error Formulas. ϵ is the observation angle of the spacecraft as seen from the target, λ is the latitude of the target, ϕ is the target azimuth relative to the ground track, λ is the Earth central angle from the target to the satellite, D is the distance from the satellite to the target, R_T is the distance from the Earth's center to the target (typically $\sim R_E$, the Earth's radius), and R_S is the distance from the Earth's center to the satellite. See Fig. 5-5.

Error Source	Error Magnitude (units)	Magnitude of Mapping Error (km)	Magnitude of Pointing Error (rad)	Direction of Error
Attitude Errors: ⁽¹⁾				
Azimuth	$\Delta\phi$ (rad)	$\Delta\phi D \sin \eta$	$\Delta\phi \sin \eta$	Azimuthal
Nadir Angle	$\Delta\eta$ (rad)	$\Delta\eta D / \sin \epsilon$	$\Delta\eta$	Toward nadir
Position Errors:				
In-Track	ΔI (km)	$\Delta I (R_T/R_S) \cos H$ ⁽²⁾	$(\Delta I/D) \sin Y_I$ ⁽⁵⁾	Parallel to ground track
Cross-Track	ΔC (km)	$\Delta C (R_T/R_S) \cos G$ ⁽³⁾	$(\Delta C/D) \sin Y_C$ ⁽⁶⁾	Perpendicular to ground track
Radial	ΔR_S (km)	$\Delta R_S \sin \eta / \sin \epsilon$	$(\Delta R_S/D) \sin \eta$	Toward nadir
Other Errors:				
Target Altitude	ΔR_T (km)	$\Delta R_T / \tan \epsilon$	—	Toward nadir
S/C Clock	ΔT (s)	$\Delta T V_e \cos(\lambda) \cos J$ ⁽⁴⁾	$\Delta T (V_e / D) \cos(\lambda) \sin J$ ⁽⁷⁾	Parallel to Earth's equator
Notes: (1) Includes attitude determination error, instrument mounting error, stability over exposure time (mapping only), and control error (pointing only). The formulas given assume that the attitude is measured with respect to the Earth. (2) $\sin H = \sin \lambda \sin \phi$. (3) $\sin G = \sin \lambda \cos \phi$. (4) $V_e = 464$ m/s (Earth rotation velocity at the equator). (5) $\cos Y_I = \cos \phi \sin \eta$. (6) $\cos Y_C = \sin \phi \sin \eta$. (7) $\cos J = \cos \phi_E \cos \epsilon$, where ϕ_E = azimuth relative to East.				

Figure 2: Relations used for error propagation.

3 Analytic error propagation

The objective is to reconstruct the mapping budget contributions, tabulated with the aid of Figure 3[1][Table 5.18, pg.277]. A script used to produce the results is presented in Section 7. First the distance from Spacecraft to the target is found with the aid of Equation 1.

$$D_{target} = \frac{h}{\cos(\theta)} \quad (1)$$

Impact on mapping budget for errors associated with star sensor (measurement, mounting error and star catalogue accuracy), attitude computation error, payload sensor (measurement and mounting error), target centroiding error and transformation of target location to inertial co-ordinates are estimated with the aid of Equation 2. The results of this is tabulated with the aid of Table 1. Shaded boxes indicate the values which are not identical to the values presented in the literature[1]. However, the offset is quite small in order of 0.1m.

$$\text{impact on mapping budget} = \frac{\Delta \cdot D_{target}}{\sin(\epsilon)} [m] \quad (2)$$

Δ = Assumed error value in radians.

TABLE 5-18. Projection Error Budget. See text for discussion. See Table 5-7 in Sec. 5.3 for formulas.

Error Source	Assumed Error Value	Impact on Mapping Budget (m)	Comments
Star Sensor Measurement Error	0.0015 deg	193.1	Typically larger error in rotation about sensor axis
Star Sensor Mounting Error	0.0020 deg	257.5	May vary from sunlight to eclipse
Star Catalog Accuracy	0.0001 deg	12.9	Have to eliminate double stars
Attitude Computation Error	0.0001 deg	12.9	Computation errors typically small
Payload Sensor Measurement Errors	0.0010 deg	128.8	Sum of many internal error sources
Target Centroiding Error	0.0020 deg	257.5	Depends on size and shape of target
Payload Sensor Mounting Error	0.0010 deg	128.8	Payload sensor opposite side of spacecraft from star sensor
Transformation of Target Location to Inertial Coordinates	0.0001 deg	12.9	Only small arithmetic error (attitude errors accounted for above)
Orbit Determination Error	100 m	100.0	Along-track component much larger than others
Timing Error	50 ms	367.5	LEO spacecraft moving at 7.35 km/s (1,000 km altitude)
Total Pointing Error With Respect to Nadir	—	590.5	RSS of pointing components above; worst in radial direction

Figure 3: Mapping budget to be reproduced and verified with analytic error propagation.

For orbit determination error the maximum among in-track, cross-track and along track is taken. Along track is larger than in-track/cross-track. In-track and cross-track error's impact on mapping budget is calculated with the aid of Equations (in-track, cross-track) presented in Figure 2. The impact of timing error is calculated with the aid of Equation S/C clock presented in Figure 2. V_e is taken to be 7.5 km/s as indicated in Figure 3. While latitude is taken to be 0, i.e. at equator.

Table 1: Reconstructed impact on mapping budget with the aid of analytic error propagation.

Error Source	Impact on Mapping Budget [m]
Star Sensor Measurement	193.09898046
Star Sensor Mounting	257.46530728
Star Catalog Accuracy	12.87326536
Attitude Computation	12.87326536
Payload Sensor Measurement	128.73265364
Target Centroiding	257.46530728
Payload Sensor Mounting	128.73265364
Transformation of target location to inertial co-ordinates	12.87326536
Orbit determination	100.00000000
Timing	367.50000000
Total pointing error with respect to Nadir	590.39104860

Assuming that all errors are random and completely uncorrelated, the total pointing error can be calculated by

taking the "Root Sum Square" RSS of all the individual error impacts. This can be mathematically represented with the aid of Equation 3. Where N is the total number of errors analysed. Total pointing error wrt Nadir is outlined in Table 1.

$$\text{Total pointing error} = \sqrt{\sum_{i=1}^N \text{Error}_i^2} \quad (3)$$

From the above analysis it can be concluded that the analytic model for error propagation is verified to be in line with the results tabulated in literature (Figure 3). A copy of the script used to construct the above results can be found in Section 7.

4 Effect of systematic error on analytic solution

The objective of this Section is to analyse the impact of systematic error on the total pointing error. First the systematic error is generated, later they are propagated for the best case and the worst case scenarios.

An additional systematic error (i.e. constant) of 0.005 degree is introduced in the payload sensor mounting in addition to the random errors. This systematic error in the payload sensor mounting can be calculated with the aid of Equation 2. Next RSS of this systematic error is estimated with the aid of Equation 5.

Next two cases are evaluated. For Case 1, a pessimistic approach is taken. Sum of systematic and random errors are taken to be correlated and linear. In other words, RSS of total random pointing errors and systematic errors are added linearly. For Case 2, an optimistic approach is taken. Sum of systematic random errors are taken to be uncorrelated and RSS is conducted on the both the errors. The result of this is presented in Table 2.

Table 2: Optimistic and pessimistic impact of systematic error on the total mapping budget.

Error Type	Impact on Mapping Budget [m]
RSS of random error	590.3910
RSS of systematic error	643.6633
Case 1	1234.0543
Case 2	873.4209

From Table 2 it can be concluded that in presence of systematic errors, the overall impact on the mapping budget increases, lesser for the optimistic case and more for the pessimistic case.

5 Monte Carlo based error propagation

Monte Carlo based error propagation can be conducted in different ways. One way is to randomise the assumed values of errors, within a desired confidence interval of standard deviation σ . "N" numbers of data points are generated, from which the errors are randomly picked. The picked error values can then be used in the analytic model described in Section 3, to obtain the total pointing error and the total pointing error with systematic errors. However, one immediately realises that this method is not attractive (for the case of 1σ of random errors) for two reasons, 1. the computational effort is N times more than analytic solution, ignoring the computation power consumed in generating N random numbers. 2. The value of total pointing error will in general be smaller and at best equivalent to the analytic solution. There is no point in undertaking this method unless the σ is increased to higher intervals i.e. $\geq 2\sigma$. A copy of this method is presented in the script for the curious readers to try out.

However, another approach which will be presented in the following sections, which is rather attractive for the 1σ case. The objective is to use the nominal model and calculate the nominal target location as indicated in Figure 1. Next the errors are introduced in the model and the corresponding target location is updated. The process is repeated N times, where at every instance a random value of the error is picked in the interval 1σ of error. The difference between the nominal and the new location is stored for every run. And the end of simulation, the mean and rms of the differences is analysed to provide insight into the impact of total pointing error (wrt Nadir) on the mapping budget. Equation 5 outlines the nominal target solution. While, Equation ?? outlines the target location for a new random simulation. Where Δ indicated the random errors introduced,

ss indicates star sensor, sc = star catalogue, att = attitude compensation, ps = payload sensor, tc = target centroiding, tran = transformation of target location to inertial co-ordinates, orb = orbit determination error and time = timing error.

$$x_{\text{nominal target}} = h * \tan(\theta) \quad (4)$$

$$x_{\text{updated target}} = h * \tan(\theta + \Delta_{ss.meas} + \Delta_{ss.mount} + \Delta_{sc.acc} + \Delta_{att} + \Delta_{ps.meas} + \Delta_{ps.mount} + \Delta_{tc} + \Delta_{tran}) + \Delta x_{orb} + \Delta x_{time} \quad (5)$$

The simulation is run for 3 times for 4 different data point sizes (10, 1000, 10000, 100000). Matlab command "randn" is used to generate the data points. Standard deviation of 1σ is used. Expressions for Δx_{orb} and Δx_{time} are the same as the analytic case and can be found in Figure 2. The impact of total random pointing error on the mapping budget is outlined in Table 3. For increasing data points, it can be observed that the rms of total impact on mapping budget Δx is getting closer to the analytic total impact on mapping budget as outlined in Table 1.

Table 3: Monte Carlo simulation of total impact on mapping budget Δx

Δx	Run 1		Run 2		Run 3	
Datapoint	mean [m]	rms[m]	mean [m]	rms [m]	mean [m]	rms [m]
100	0.1903	548.2911	40.1289	648.7165	-69.4567	595.3122
1000	28.7337	585.8341	-5.9516	587.3025	-17.0918	565.3290
10000	-0.2170	591.0601	-2.5325	582.1630	-4.0780	593.4750
100000	2.5344	590.5176	-0.8578	589.0559	1.2926	590.8461

6 Effect of systematic error on Monte Carlo solution

In case of a constant systematic error on the payload mounting, a new "constant" term $\Delta_{systematic} = 0.005$ deg is added to the model. This is indicated with the aid of Equation 6.

$$x_{\text{updated target}} = h * \tan(\theta + \Delta_{systematic} + \Delta_{ss.meas} + \Delta_{ss.mount} + \Delta_{sc.acc} + \Delta_{att} + \Delta_{ps.meas} + \Delta_{ps.mount} + \Delta_{tc} + \Delta_{tran}) + \Delta x_{orb} + \Delta x_{time} \quad (6)$$

Simulation is run 3 times, for 4 different data point sizes and the results are outlined in Table 4. An interesting observation is that with increasing data points the total mapping error budget reaches the optimistic case (case 2) of analytic solution as seen in Table 2.

Table 4: Impact of systematic error on the mapping error budget with Monte Carlo.

Δx	Run 1		Run 2		Run 3	
Datapoint	mean [m]	rms[m]	mean [m]	rms [m]	mean [m]	rms [m]
100	602.9215	856.1260	780.3592	962.4202	592.8838	881.0682
1000	653.8683	873.6976	628.7222	861.1940	662.6669	891.2925
10000	647.0354	869.1803	643.0922	876.2583	653.3000	881.3668
100000	640.5384	870.1111	645.5183	875.2766	645.8447	875.5356
1000000	642.6453	872.3490	644.1882	873.6003	643.7526	873.8586

What can be concluded after all this analysis? It can be concluded that Monte Carlo simulation provides an alluring alternative to analytic approach. Especially for larger systems involving high number of parameters, Monte Carlo can be used to evaluate one simple equation for N data points (very attractive from vectorization perspective, can be solved in one step) rather one evaluating one equation for every single parameter involved.

7 Matlab Script

Following script is used to generate the above presented results.

```

1  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
2  %%% AE4878 Mission Geometry and Orbit Design %%%%%%%%%
3  %%% Assignment 6 - Week 4.2 - DESIGN 7 - Version 2 %%%
4  %%% Author Info: Ali Nawaz; Student ID - 4276477 %%%
5  %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
6  %% Parameters
7  h = 1000; % altitude [km]
8  elevd = 21.6052; %elevation angle [deg]
9  elevr = deg2rad(21.6052); % elevation angle [rad]
10 theta = pi - pi/2 - elevr; % Theta/Eta Nadir angle [rad]
11 lambda = pi/2 - elevr - theta; % Earth central angle [rad] Eqn-9.6 OCDM
12 phi = 0; % target azimuth relative to groundtrack [rad]
13 H = asin(sin(lambda)*sin(phi)); % H angle [rad]
14 G = asin(sin(lambda)*cos(phi)); % G angle [rad]
15 Dtnom = h*tan(theta); % Nominal target distance from SSP [km]
16 Dnom = h/cos(theta); % Nominal distance from S/C to Target [km]
17 RE = 6378.136; % Radius of Earth [km], taken from OCDM/intro lecture slides
18 RT = RE; % Distance from Earth's center to target [km]
19 RS = RE+h; % Distance from Earth's center to the satellite [km]
20
21 %% Part 1: Assumed error values and corresponding impact on mapping budget [m] - Analytical ...
    Approach
22 ss_meas_err = deg2rad(0.0015); %Star Sensor (SS) measurement error [rad]
23 ss_meas_map = (ss_meas_err*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget[m]
24
25 ss_mount_err = deg2rad(0.0020); % SS mounting error [rad]
26 ss_mount_map = (ss_mount_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
27
28 ss_cata_acc = deg2rad(0.0001); % SS catalogue accuracy [rad]
29 ss_cata_map = (ss_cata_acc*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
30
31 att_comp_err = deg2rad(0.0001); % attitude computation error [rad]
32 att_comp_map = (att_comp_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
33
34 ps_meas_err = deg2rad(0.0010); % Payload sensor mounting error [rad]
35 ps_meas_map = (ps_meas_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
36
37 targ_cent_err = deg2rad(0.0020); % Target centroiding error [rad]
38 targ_cent_map = (targ_cent_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
39
40 ps_mount_err = deg2rad(0.0010); % Payload sensor mounting error [rad]
41 ps_mount_map = (ps_mount_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
42
43 trans_err = deg2rad(0.0001); % Transformation of target location to inertial co-ordinates [rad]
44 trans_map = (trans_err*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
45
46 orb_det_err = 100*10^(-3); % Orbit determination error [km]
47 orb_det_map_rad = orb_det_err*(sin(theta)/sin(elevr))*10^(-3); % Radial impact on mapping ...
    budget [m]
48 orb_det_map_int = orb_det_err*(RT/RS)*cos(H)*10^(-3); % Intrack impact on error budget [m] ...
    displacement in the direction of the velocity vector
49 orb_det_map_ct = orb_det_err*(RT/RS)*cos(G)*10^(-3); % Cross track impact on error budget [m]
50 orb_det_map_along = orb_det_err*10^(-3); % Along track impact on error budget [m] transverse ...
    displacement normal to the position vector.
51
52
53 timing_err = 50*10^(-3); % Timing error [s]
54 Ve = 7.35*10^(-3); % [m/s] Velocity at 1000 km altitude
55 lat = deg2rad(0); % Latitude at equator[rad]
56 timing_map = timing_err*Ve*cos(lat); % Impact of timing error on mapping budget [m]
57
58 % total pointing error impact on budget wrt nadir, random and uncorrelated errors i.e. RSS
59 total_pointing_error = sqrt( ss_meas_map^2 + ss_mount_map^2 + ss_cata_map^2 + att_comp_map^2 ...
    + ps_meas_map^2 + targ_cent_map^2 + ps_mount_map^2 + trans_map^2 + orb_det_map_along^2 + ...
    timing_map^2); %[m]
60
61 % Tabulated mapping error/ impact on budget [m]
62 tab_map_part1 = [ss_meas_map, ss_mount_map, ss_cata_map, att_comp_map, ps_meas_map, ...
    targ_cent_map, ps_mount_map, trans_map, orb_det_map_along, timing_map];
63
64 %% Part 2: Addition of systematic payload mounting error
65
66 % Additional systematic error for payload mounting
67 ps_mount_err_sys = deg2rad(0.005); % Payload sensor mounting error [rad]
68 ps_mount_map_sys = (ps_mount_err_sys*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping budget [m]
69

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70 rss_sys = sqrt(ps_mount_map_sys^2); % RSS of systematic error(s)
71
72 % Case 1: Sum of random and systematic errors (correlated)(linear) -
73 % Pessimistic
74 case1_error = total_pointing_error + rss_sys; % [m]
75
76 % Case 2: Sum of random and systematic errors (uncorrelated) (RSS) -
77 % Optimistic
78 case2_error = sqrt(total_pointing_error^2 + rss_sys^2); % [m]
79
80 %% PART 3 [ Monte Carlo Extension to Analytic Approach ] - Assumed error values and ...
81     corresponding impact on mapping budget [m]
82 datapoints = 1000000; % no. of random values generated
83 lb = 0; % Lower bound on error
84
85 % Generate random normally distributed
86 val = randn(datapoints,1);
87 mss_meas_err = deg2rad(0.0015); % Star Sensor (SS) measurement error [rad]
88 val1 = val*mss_meas_err; % Random errors generated for pure monte carlo approach in part 5
89 mss_meas_map = (mss_meas_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
90
91 val = randn(datapoints,1);
92 mss_mount_err = deg2rad(0.0020); % SS mounting error [rad]
93 val2 = val*mss_mount_err; % Random errors generated for pure monte carlo approach in part 5
94 mss_mount_map = (mss_mount_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
95
96 val = randn(datapoints,1);
97 mss_cata_acc = deg2rad(0.0001); % SS catalogue accuracy [rad]
98 val3 = val*mss_cata_acc; % Random errors generated for pure monte carlo approach in part 5
99 mss_cata_map = (mss_cata_acc*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
100
101 val = randn(datapoints,1);
102 matt_comp_err = deg2rad(0.0001); % attitude computation error [rad]
103 val4 = val*matt_comp_err; % Random errors generated for pure monte carlo approach in part 5
104 matt_comp_map = (matt_comp_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
105
106 val = randn(datapoints,1);
107 mps_meas_err = deg2rad(0.0010); % Payload sensor mounting error [rad]
108 val5 = val*mps_meas_err; % Random errors generated for pure monte carlo approach in part 5
109 mps_meas_map = (mps_meas_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
110
111 val = randn(datapoints,1);
112 mtarg_cent_err = deg2rad(0.0020); % Target centroiding error [rad]
113 val6 = val*mtarg_cent_err; % Random errors generated for pure monte carlo approach in part 5
114 mtarg_cent_map = (mtarg_cent_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
115
116 val = randn(datapoints,1);
117 mps_mount_err = deg2rad(0.0010); % Payload sensor mounting error [rad]
118 val7 = val*mps_mount_err; % Random errors generated for pure monte carlo approach in part 5
119 mps_mount_map = (mps_mount_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
120
121 val = randn(datapoints,1);
122 mtrans_err = deg2rad(0.0001); % Transformation of target location to inertial co-ordinates [rad]
123 val8 = val*mtrans_err; % Random errors generated for pure monte carlo approach in part 5
124 mtrans_map = (mtrans_err*val*(Dnom/ (sin(elevr)) ))*10^3; % Impact on mapping budget [m]
125
126 val = randn(datapoints,1);
127 morb_det_err = 100*10^(-3); % Orbit determination error [km]
128 morb_det_map_rad = morb_det_err*val*(sin(theta)/sin(elevr))*10^3; % Radial impact on ...
129     mapping budget [m]
130 morb_det_map_int = morb_det_err*val*(RT/RS)*cos(H)*10^3; % Intrack impact on error budget ...
131     [m] displacement in the direction of the velocity vector
132 morb_det_map_ct = morb_det_err*val*(RT/RS)*cos(G)*10^3; % Cross track impact on error ...
133     budget [m]
134 morb_det_map_along = (-morb_det_err + 2*morb_det_err*val); % Along track impact on error ...
135     budget [m] transverse displacement normal to the position vector.
136 morb_det_map_along = morb_det_err*val*10^3; % Along track impact on error budget [m] ...
137     transverse displacement normal to the position vector.
138 val9 = morb_det_map_along; % Random errors generated for pure monte carlo approach in part 5
139
140 val = randn(datapoints,1);
141 mtiming_err = 50*10^(-3); % Timing error [s]
142 Ve = 7.35*10^3; % [m/s] Velocity at 1000 km altitude
143 lat = deg2rad(0); % Latitude at equator [rad]
144 val10 = val*mtiming_err; % Random errors generated for pure monte carlo approach in part 5
145 mtiming_map = mtiming_err*val*Ve*cos(lat); % Impact of timing error on mapping budget [m]

```

```

140 % total pointing error with monte carlo approach to analytic solution
141 mttotal.pointing.error = sqrt( mss.meas_map.^2 + mss.mount_map.^2 + mss.cata_map.^2 + ...
    matt.comp_map.^2 + mps.meas_map.^2 + mtarg.cent_map.^2 + mps.mount_map.^2 + mtrans_map.^2 ...
    + morb.det_map.along.^2 + mtiming_map.^2); %[m]
142 % mean and rms of corresponding total pointing error
143 mean.mtotal.pointing.error = mean(mtotal.pointing.error); % mean of total pointing error
144 rms.mtotal.pointing.error = rms(mtotal.pointing.error); % rms of total pointing error
145
146 Visualising the behaviour of Monte Carlo generated pointing error
147 figure(1)
148 mean_list = (mean.mtotal.pointing.error*ones(length(mtotal.pointing.error)));
149 rms_list = (rms.mtotal.pointing.error*ones(length(mtotal.pointing.error)));
150 plot(1:length(mtotal.pointing.error), ...
    mtotal.pointing.error, 'ro', 1:length(mtotal.pointing.error), mean_list, 'b-', ...
    1:length(mtotal.pointing.error), rms_list, 'k--');
151 title('Total pointing error behaviour for different Monte Carlo runs');
152 legend('Total pointing error', 'Mean', 'RMS');
153 grid on
154 xlabel('Simulation no.')
155 ylabel('Total pointing error [m]')
156 % Storing individual impact on error budget
157 tab_map.part2 = [mss.meas_map, mss.mount_map, mss.cata_map, matt.comp_map, mps.meas_map, ...
    mtarg.cent_map, mps.mount_map, mtrans_map, morb.det_map.along, mtiming_map];
158
159 %% PART 4 - [ Monte Carlo Extension to Analytic Approach ] Additional systematic error for ...
    payload mounting
160 val = randn(datapoints,1);
161 mps.mount_err.sys = deg2rad(0.005); % Payload sensor mounting error [rad]
162 mps.mount_map.sys = (mps.mount_err.sys*val*(Dnom/ (sin(elevr)) ))*10^3;% Impact on mapping ...
    budget [m]
163
164 mrss.sys = sqrt(mps.mount_map.sys.^2); % RSS of systematic error(s)
165
166 % Case 1: Sum of random and systematic errors (correlated)(linear) -
167 % Pessimistic
168 mcase1.error = mtotal.pointing.error + mrss.sys; % [m]
169
170 % Case 2: Sum of random and systematic errors (uncorrelated)(RSS) -
171 % Optimistic
172 mcase2.error = sqrt(mtotal.pointing.error.^2 + mrss.sys.^2); % [m]
173
174 %% Part 5: Pure Monte Carlo Approach
175 x_nom = h*tan(theta)*10^(3); % Nominal value of target location from SSP [m]
176
177 % Updated target value with random error selection
178 x_target.new = h*tan(theta + val1+ val2 + val3 + val4 + val5 + val6 + val7 + val8)*10^(3) + ...
    val9 + mtiming_map;
179
180 %Updated target value with random and systematic error selection
181 x_target.new.sys = h*tan(theta + val1+ val2 + val3 + val4 + val5 + val6 + val7 + val8 + ...
    mps.mount_err.sys)*10^(3) + val9 + mtiming_map;
182
183 % differences between target and nominal models
184 diff = x_target.new - x_nom; % random error case
185 diff.sys = x_target.new.sys - x_nom; % random error and systematic error case
186
187 % mean and rms of impact on budget for random error case
188 mean.diff = mean(diff);
189 rms_diff = rms(diff);
190
191 % mean and rms of impact on budget for combined random and systematic error
192 % case
193 mean.diff.sys = mean(diff.sys)
194 rms_diff.sys = rms(diff.sys)
195
196 %%%% END OF SCRIPT %%%%

```

References

[1] James R. Wertz. *Orbit Constellation Design Management*. Springer, 2009.