# **Assignment 02 - Navigation**

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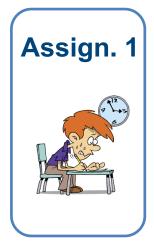
This teaching material was prepared by F. Topputo, P. Di Lizia, A. Morselli and G. Purpura.

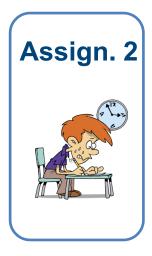
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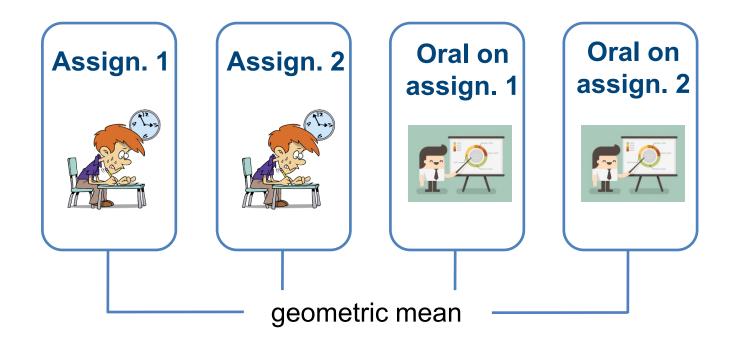
Spacecraft Guidance and Navigation

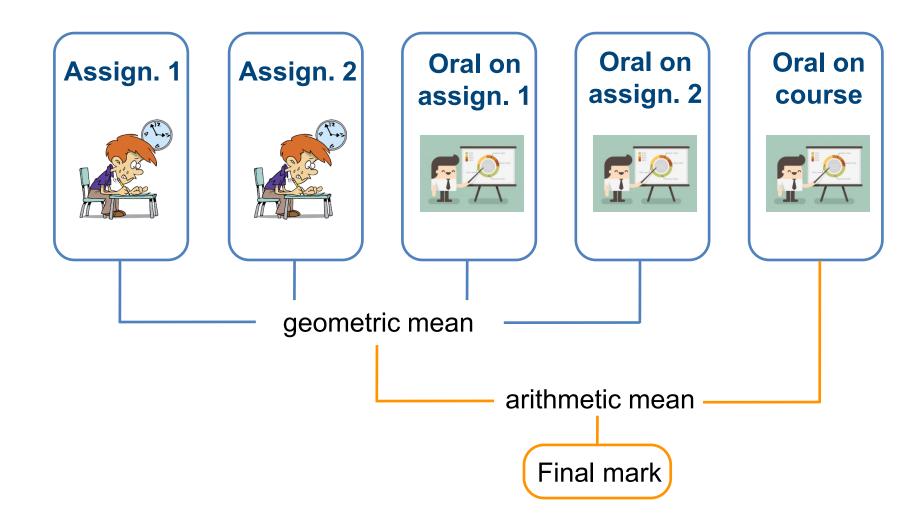
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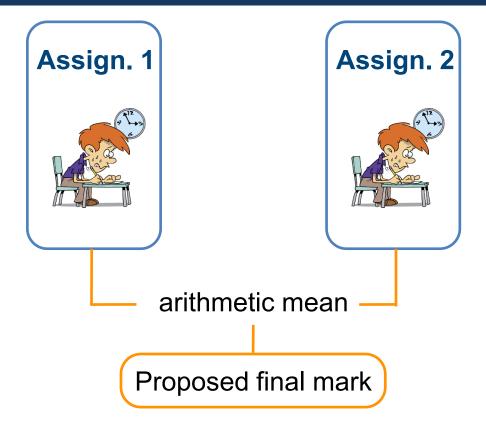




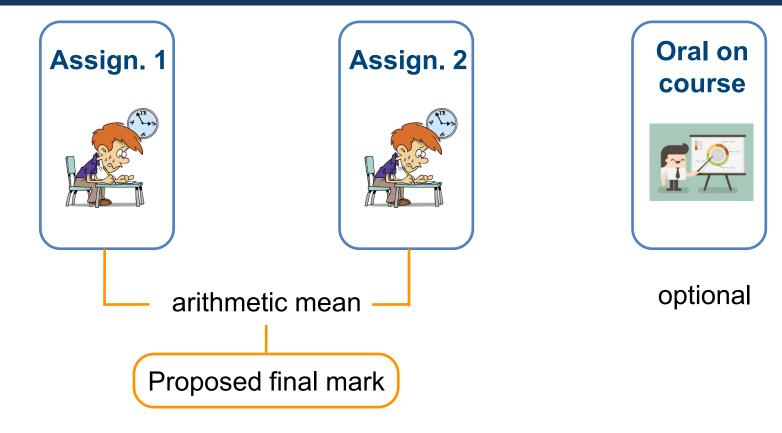




# **JUST JOKING!**



➤ All of you will be asked to come in person during the official exam sessions to check our evaluation and discuss any doubts you/we might have



- ➤ All of you will be asked to come in person during the official exam sessions to check our evaluation and discuss any doubts you/we might have
- In case you are not satisfied with the proposed final mark, you can take an optional oral exam (which will take place during the official exam sessions)

#### > Remarks:

- For those who will not submit the report by the deadlines, you'll be asked to do it before the official exam sessions
- Once you receive the evaluation of the assignments, please wait for the in-person meeting during the exam sessions to discuss about the evaluation

# **Laboratory sessions**

- ➤ Laboratory sessions <u>will not be recorded</u>
  - > we are here to give you answers while you are working
- ➤ When writing emails:
  - ➤ Use the following object structure: [SGN – Assignment2] Your email object here
  - > Do not attach/send code (or screenshot of the code) asking for debugging

Timetable		Room			
24 Nov	16.15-18.15	BL28.12 ( + Morselli virtual room)			
26 Nov	10.15-13.15	L11 ( + Morselli virtual room)			
30 Nov	14.15-16.15	BL27.14 ( + Morselli virtual room)			
01 Dec	16.15-18.15	BL28.12 ( + Morselli virtual room)			
03 Dec	10.15-13.15	L11 ( + Morselli virtual room)			
16 Dec	23.30 CET	Assignment 2 deadline			

# **Delivery of assignments**

Assignment will be delivered through WeBeep:

1) Click on the link to load Assignment 2 in your Overleaf

https://bit.ly/SGN 21 Assignment2



- 2) Fill the report and be sure it is compiled properly
- 3) Download the PDF and merge it in a zipped file with MATLAB code.

Rename it lastname123456\_Assign2.zip ←———

Do **NOT** put spaces in file names

4) Submit the compressed file by uploading it on WeBeep

# **Assignment 02 - Topics**

### 3 Exercises

- Uncertainty propagation
- Batch filters
- Sequential filters

**Suggested MATLAB Version: R2021b** 

### Report

#### 1 Impulsive guidance

#### Exercise 1

Let  $\mathbf{x}(t) = \varphi(\mathbf{x}_0, t_0; t)$  be the flow of the geocentric two-body model. 1) Using one of Matlab's built-in integrators, implement and validate a propagator that returns  $\mathbf{x}(t)$  for given  $\mathbf{x}_0, t_0, t$ , and  $\mu$ . 2) Given the pairs  $\{\mathbf{r}_1, \mathbf{r}_2\}$  and  $\{t_1, t_2\}$ , develop a solver that finds  $\mathbf{v}_1$  such that  $\mathbf{r}(t_2) = \mathbf{r}_2$ , where  $(\mathbf{r}(t), \mathbf{v}(t))^{\top} = \varphi((\mathbf{r}_1, \mathbf{v}_1)^{\top}, t_1; t_2)$  (Lambert's problem). To compute the derivatives of the shooting function, use either a) finite differences or b) the state transition matrix  $\Phi = d\varphi/d\mathbf{x}_0$ . Validate the algorithms against the classic Lambert solver. 3) Using the propagator of point 1) in the heliocentric case, and reading the motion of the Earth and Mars from SPICE, solve the shooting problem

$$\min_{\mathbf{x}_1, t_1, t_2} \Delta v \quad \text{s.t.} \begin{cases} \mathbf{r}_1 = \mathbf{r}_E(t_1) \\ \mathbf{r}(t_2) = \mathbf{r}_M(t_2) \\ t_1^L \leq t_1 \leq t_1^U \\ t_2^L \leq t_2 \leq t_2^U \\ t_2 \geq t_1 \end{cases}$$
(1)

where  $\Delta v = \Delta v_1 + \Delta v_2$ ,  $\Delta \mathbf{v}_1 = \mathbf{v}_1 - \mathbf{v}_E(t_1)$ ,  $\Delta \mathbf{v}_2 = \mathbf{v}(t_2) - \mathbf{v}_M(t_2)$ .  $\mathbf{x}_1 = (\mathbf{r}_1, \mathbf{v}_1)^\top$ , and  $(\mathbf{r}(t), \mathbf{v}(t))^\top = \varphi(\mathbf{x}_1, t_1; t_2)$ . Define lower and upper bounds, and make sure to solve the problem stated in Eq. (1) for different initial guesses.

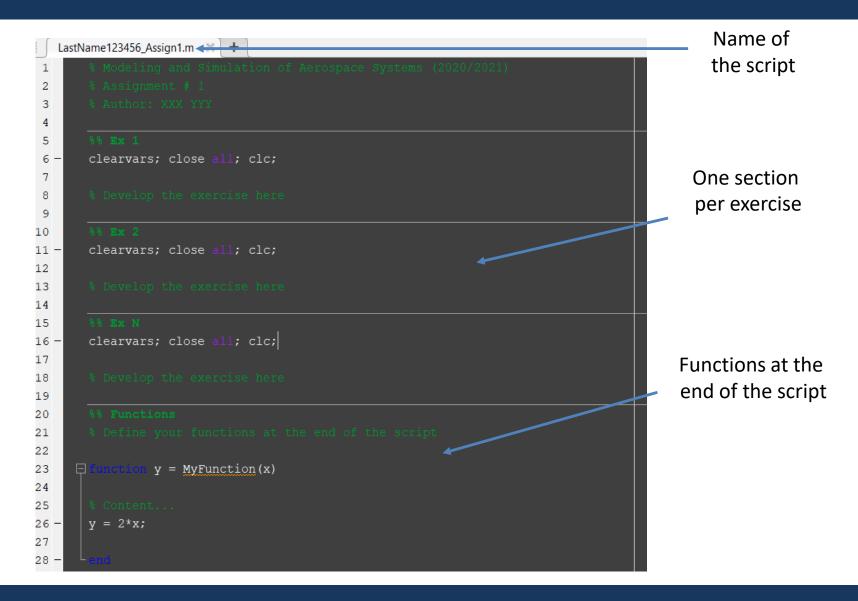
Write your answer here

- Develop the exercises in one Matlab script; name the file lastname123456\_Assign1.m
- Organize the script in sections, one for each exercise; use local functions if needed.
- Download the <u>PDF</u> from the Main menu.
- Create a single .zip file containing both the report in PDF and the MATLAB file. The name shall be lastname123456\_Assign1.zip.
- Red text indicates where answers are needed; be sure there is no red stuff in your report.
- . In your answers, be concise: to the point.
- Deadline for the submission: Nov 11 2021, 23:30.
- Load the compressed file to the Assignments folder on Webeep.

Exercise

Answer here

# Script



# **Exercise 1 – Uncertainty propagation**

#### Exercise 1: Uncertainty propagation

You are the operator of the sensor network composed by the stations reported in Table  $\boxed{1}$ . You have been assigned the task of tracking the spacecraft GIOVE-A (NORAD ID: 28922, INT.DES.: 2005-051A) and you have been provided with an estimate of the spacecraft state at time  $t_0$  in terms of its mean and covariance, as reported in Table  $\boxed{2}$ . Assume Keplerian motion can be used to model the spacecraft dynamics.

- 1. By using the mean state to predict the reference trajectory over a uniform time grid (one point per minute), compute all the visibility windows from the available stations in the time interval from  $t_0$  to  $t_f = 2021-11-23T14:00:00$  (UTC).
- 2. Propagate the initial mean and covariance to the last epoch of each visibility window (as computed with the uniform time grid in the previous point) using both a linearized approach and the unscented transform. Elaborate on the results of the uncertainty propagation, including possible differences between the two approaches.
- 3. Perform a Monte Carlo simulation (using at least 100 samples drawn from the initial covariance) to predict the spacecraft state distribution at the very last visibility epoch (the closest to  $t_f$ , as resulting from the uniform time grid used at point 1). Compute the sample mean and sample covariance, and compare them with the results obtained at the previous point. Then, for each sample, compute the angle between the topocentric direction of the satellite and the topocentric direction of the center of the field-of-view (FoV) of the sensor. Provide the percentage of the samples that lie inside the sensor FoV.

Table 1: Sensor network: list of stations, including their features.

Station name	Milano	Wellington	La Silla
Coordinates	${ m LAT} = 45.50122^{\circ} \ { m LON} = 9.15461^{\circ} \ { m ALT} = 20 \ { m m}$	$LAT = -41.28437^{\circ}$ $LON = 174.76697^{\circ}$ ALT = 117  m	$\begin{split} LAT &= \text{-}29.26117^{\circ} \\ LON &= \text{-}70.73133^{\circ} \\ ALT &= 2400 \text{ m} \end{split}$
Туре	Radar (monostatic)	Optical	Optical
Provided measurements	Az, El [deg] Range (one-way) [km]	Ra, Dec [deg]	Ra, Dec [deg]
Measurements noise (diagonal noise matrix R)	$\sigma_{Az,El} = 0.1 \text{ deg}$ $\sigma_{range} = 0.01 \text{ km}$	$\sigma_{Ra,Dec}=0.0005$	$\sigma_{Ra,Dec}=0.001$
FoV (circular)	$6 \deg$	$2 \deg$	1 deg
Minimum elevation	30 deg	20 deg	10 deg

**Table 2:** Estimate of the spacecraft state at  $t_0$  provided in ECI J2000.

Parameter	Value							
Ref. epoch $t_0$ [UTC]	2021-11-19T14:25:39.652							
Mean state $\hat{x}_0$ [km, km/s]	$\hat{r}_0 = [23721.610, 17903.673, -49.918]$ $\hat{v}_0 = [-1.150987, 1.529718, 3.122389]$							
	+1.4e - 2	+1.4e - 2 +1.8e - 2	+2.3e - 3	0	0 0	$\begin{bmatrix} 0 \\ 0 \end{bmatrix}$		
Covariance $P_0$ [km <sup>2</sup> , km <sup>2</sup> /s, km <sup>2</sup> /s <sup>2</sup> ]	-1.8e - 3	+2.3e - 3		0 + 1.6e - 7	0 0	0		
	0	0 0	0 0	0 0	+1.6e - 7	$\begin{bmatrix} 0 \\ +1.6e - 7 \end{bmatrix}$		

### **Exercise 2 - Batch filters**

#### Exercise 2: Batch filters

The Two-Line Elements (TLE) set of the spacecraft GIOVE-A of Exercise 1 is reported in Table 3 (and in WeBeep as 28922.tle).

- 1. Simulate measurements. Use SGP4 and the provided TLE to simulate the measurements acquired by the sensor network in Table [1] by:
  - (a) Computing the spacecraft position over the same visibility windows identified in Exercise 1 and deriving the associated expected measurements.
  - (b) Simulating the measurements by adding a random error to the expected measurements (assume a Gaussian model to generate the random error, with noise provided in Table 1).

**Table 3:** TLE of spacecraft GIOVE-A.

```
1_28922U_05051A___21323.60115338_-.00000088__00000-0__00000+0_0__9998
2_28922__58.3917__37.3090_0004589__75.2965_284.7978__1.69421077_98469
```

Do not copy-paste from PDF

### **Exercise 2 - Batch filters**

- 2. Solve the navigation problem. Using the measurements simulated at the previous point:
  - (a) Find the least squares (minimum variance) solution to the navigation problem without a priori information using
    - the mean state provided in Table 2 as first guess;
    - the simulated measurements up to epoch  $t_{f,1} = 2021-11-21T14:00:00$ ;
    - pure Keplerian motion and J2-perturbed motion to model the spacecraft dynamics.
  - (b) Repeat step 2a by solving the navigation problem with the a priori information reported in Table 2.
  - (c) Repeat step 2a by adding incrementally the simulated measurements up to  $t_{f,2}$  =2021-11-22T14:00:00 and  $t_{f,3}$  =2021-11-23T14:00:00.

# **Supporting material**

- Available on WeBeep, inside the folder Laboratories
- Compressed folder Assignment02.zip contains:
  - Subfolder kernels with all required SPICE kernels
  - Subfolder sgp4 with all sgp4 functions (Lab03)
  - Subfolder tle with two-line elements
  - Subfolder **tdm** with Tracking Data Message files
  - Meta-kernel assignment02.tm

#### 

### How to prepare your script

- Unzip the folder and work on your Matlab script inside it
- Remember to load the meta-kernel and to add the path to sgp4 subfolder in your script
  - Suggestion: use relative paths
- Obs: no need to upload the supporting material when preparing your delivery on WeBeep

# A quick recap

#### Usage of SGP4

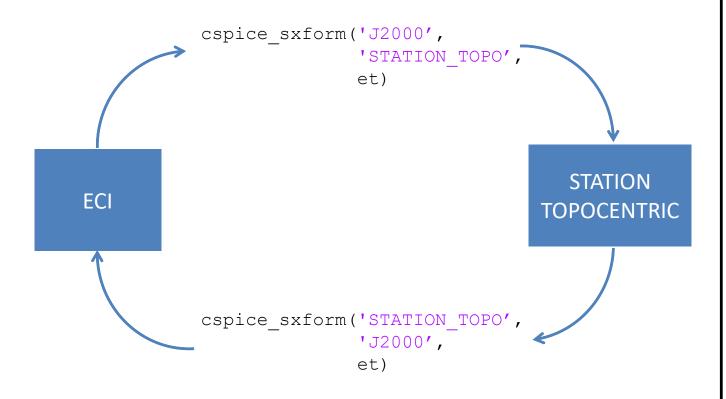
- a. Initialize the satellite record with
  - twoline2rv (requires tle strings)
  - read TLE (requires TLE file name or sat id)
- b. Call SGP4
  - Provide time since reference epoch (minutes)
- c. Convert from/to TEME (if needed)
  - Compute ttt
    - Centuries from TDT 2000 January 1 00:00:00
  - TEME acceleration can be set to [0,0,0]
  - Get offsets dPsi and dEpsilon
    - Convert them in rad if provided in arcsec
  - Call functions teme2eci or eci2teme

```
= 'u'; % user-provided inputs to SGP4 Matlab function
typerun
opsmode = 'a'; % afspc approach ('air force space command')
whichconst = 72; % WGS72 constants (radius, gravitational parameter)
% initialize the satrec structure, using the function twoline2rv
satrec = twoline2rv(longstr1, longstr2, typerun, 'e', opsmode, whichconst)
satrec = read TLE(123456, whichconst);
% Evaluate the TLE
[satrec, rteme, vteme] = sgp4(satrec, 0.0);
% Centuries from TDT 2000 January 1 00:00:00.000
et0 = cspice str2et('2004-04-06T07:51:28.386009');
ttt = cspice unitim(et0, 'ET', 'TDT')/cspice_jyear()/100;
% ------ teme2eci - transform teme to eci vectors
ateme = [0;0;0];
[reci, veci, aeci] = teme2eci(rteme, vteme, ateme, ttt, ddpsi, ddeps);
% ----- eci2teme - transform eci to teme vectors
[rteme, vteme] = eci2teme(reci, veci, aeci, ttt, ddpsi, ddeps);
```

# A quick recap (2)

### Reference frame conversions

With station kernel generated with pinpoint



#### Without station kernel

```
% Define station coordinates
lat = 45.50122; % deg
lon = 9.15461; % deg
alt = 20; % m
% Get Earth radii (equatorial and polar)
% (requires pck00010.tpc kernel)
radii = cspice bodvrd('EARTH', 'RADII', 3);
re = radii(1); rp = radii(3);
% Compute flattening
flat = (re - rp) / re;
% Convert to radians and km
lat rad = deg2rad(lat); lon rad = deg2rad(lon);
alt km = alt / 1000;
% Compute station pos wrt Earth center (ECEF)
pos ECEF = cspice pgrrec(...
    'EARTH', lon rad, lat rad, alt km, re, flat);
% Compute ECEF2TOPO rotation matrix
ECEF2TOPO = cspice eul2m(...
    lat rad - pi, pi - lon rad, pi / 2, 2, 1, 2); 21
```

# A quick recap (3)

#### Use mynrnd to generate multivariate normal random numbers

Generate sample points for Monte Carlo simulations

```
Example: generate 100 vectors with given mean [1 3] and covariance diag([0.01 0.02]):
n = 100; mu = [1 3]; Sigma = diag([0.01 0.02]);
R = mvnrnd(mu, Sigma, n); % R size: 100x2
```

Add random noise to measurements

```
Example: add noise with covariance diag([0.01 0.02]) to a set of angular measurements:
mu = [1 3; 1.5 2; 2 1]; Sigma = diag([0.01 0.02]);
R = mvnrnd(mu, Sigma); % R size: 3x2
```

#### Notes:

- When Sigma is diagonal, you can directly pass the diagonal (e.g. Sigma = [0.01 0.02];)
- It is possible to define a different Sigma for each measurement (using the 3rd dimension)

# A quick recap (4a)

#### Solve nonlinear least-squares problems in MATLAB

```
[x, resnorm, residual, exitflag, \sim, \sim, jac] = lsqnonlin(fun, x0, lb, ub, opt);
```

#### Inputs:

- fun: handle of the cost function to be minimized
  - fun must take as input only the x and must return the residual
  - the residual can be a vector, lsqnonlin implicitly computes the norm
- x0: initial point
- lb, ub: lower and upper bounds
- opt: to be defined using the optimoptions function

#### Outputs:

- x: solution
- resnorm: squared norm of the residual
- residual: value of fun(x)
- exitflag: reason the solver stopped (convergence if exitflag > 0)
- jac: jacobian matrix of fun with respect to x

# A quick recap (4b)

end

### Solve nonlinear least-squares problems in MATLAB - example

```
% Variables of the problem
x 0 = ...; var 1 = ...; var 2 = ...;
% Encapsuate the variables in the function handle
fun = @(x) costfunction(x, var 1, var 2, ...);
% Optimization options
opt = optimoptions('lsgnonlin', 'Algorithm', 'levenberg-marquardt', 'Display', 'iter');
% Execute least squares
[x, resnorm, residual, exitflag, ~, ~, jac] = lsqnonlin(fun, x 0, [], [], opt);
% Compute resulting covariance
Jac = full(jac);
P ls = resnorm / (length(residual)-length(x 0)) .* inv(Jac'*Jac);
function residual = costfunction(x, var 1, var 2, ...)
    residual = []; % Initialize output variable
                                                                           Weight for «a priori»
    % If least squares with "a priori" append it to the residual
                                                                               W = inv(sqrtm(P 0));
    diff apriori weighted = W ap * (x - x0);
    residual = [residual; diff apriori weighted(:)];
    % Propagate x to the epochs of the measurements
                                                                           Weight for measurements:
    x prop = ...;
    % Compute predicted measurements
                                                                              W m = diag(1 ./ sigma meas \mathcal{M});
    meas pred = ...;
    % Compute the residual of the measurements and append it to the output
    diff meas weighted = W m * (meas pred - meas real);
    residual = [residual; diff meas weighted(:)];
```

### **General hints**

#### LaTeX

- Never put multiplication symbols, especially as asterisks
- Pay attention to distinguish between vectors and scalar: write vector quantities using underline or in bold
  - Latin alphabet: \mathbf{x} or \mathbf{r}
  - Greek symbols: \boldsymbol { \lambda }
  - Obs: No need to use norm with this notation  $\|\lambda_x\| \leftrightarrow \lambda_x$
- Write dot product using \cdot
- Prefer the use of \dfrac{}{} over\frac{}{} when writing equations
- When inserting text in an equation wrap it in \mathrm{}

#### **Plots**

- Make sure that axis labels and titles are clearly readable, even without zooming
- Remember to put the (correct!) units