

ICT4SS - Lab Exercise #1, 16th April 2020

# Lab Exercise #1: Evaluation of PVT

Laboratory on Least Mean Square positioning solution



POLITECNICO  
DI TORINO  
Dipartimento  
di Elettronica  
e Telecomunicazioni



# Least Mean Square Solution Computation



- In the most general case the PVT solution is given by the LMS equation

$$\Delta \mathbf{x} = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \Delta \rho$$

- One possible way to solve for the PVT is to use the LMS algorithm not applied a single time, but iterated in order to provide a more precise solution
- At each iteration, the estimated position is expected to be closer to the real user location

# Recursive Least Mean Square

- For each time instant  $n$ , a number of iterations  $k = 1, \dots, K$  is performed for the same measurements vector  $\boldsymbol{\rho}_n$

INITIAL POSITION/APPROXIMATION POINT	
Initialization	$\hat{\mathbf{x}}_n^0$
Procedure for $k^{th}$ iteration	$\Delta \hat{\boldsymbol{\rho}}_n^k = \hat{\boldsymbol{\rho}}_n^k - \boldsymbol{\rho}_n$
	$\Delta \hat{\mathbf{x}}_n^k = ((\mathbf{H}_n^k)^T \mathbf{H}_n^k)^{-1} (\mathbf{H}_n^k)^T \Delta \hat{\boldsymbol{\rho}}_n^k$
	$\hat{\mathbf{x}}_n^k = \hat{\mathbf{x}}_n^{k-1} + \Delta \hat{\mathbf{x}}_n^k$
RANGE VECTOR GEOMETRICALLY EVALUATED FROM $\hat{\mathbf{x}}_n^0$ AND EPHEMERIS	
VECTOR OF MEASURED PSEUDORANGES	

- $\mathbf{H}$  is updated at each iteration  $k$
- The initial approximation point,  $\hat{\mathbf{x}}_n^0$ , could be the same at each time  $n$
- $K$  has to be chosen in order to have “stable” solution (generally  $K < 10$ )

# Recursive Least Mean Square

- $\hat{\mathbf{x}}_n^k = [\hat{x}_n^k \quad \hat{y}_n^k \quad \hat{z}_n^k \quad \hat{b}_n^k]$  is the estimated position/state
- The number of visible satellites  $J$  is not constant over observation time  $n$
- $\Delta \hat{\rho}_n$  is the vector of the  $J_n$  pseudorange differences (between measured and calculated) at the  $k$ -th iteration
- At each iteration  $k$ ,  $\mathbf{H}_n^k$  is the geometrical matrix obtained from the previous estimated position:

Satellite coordinates are fixed for any  $k$

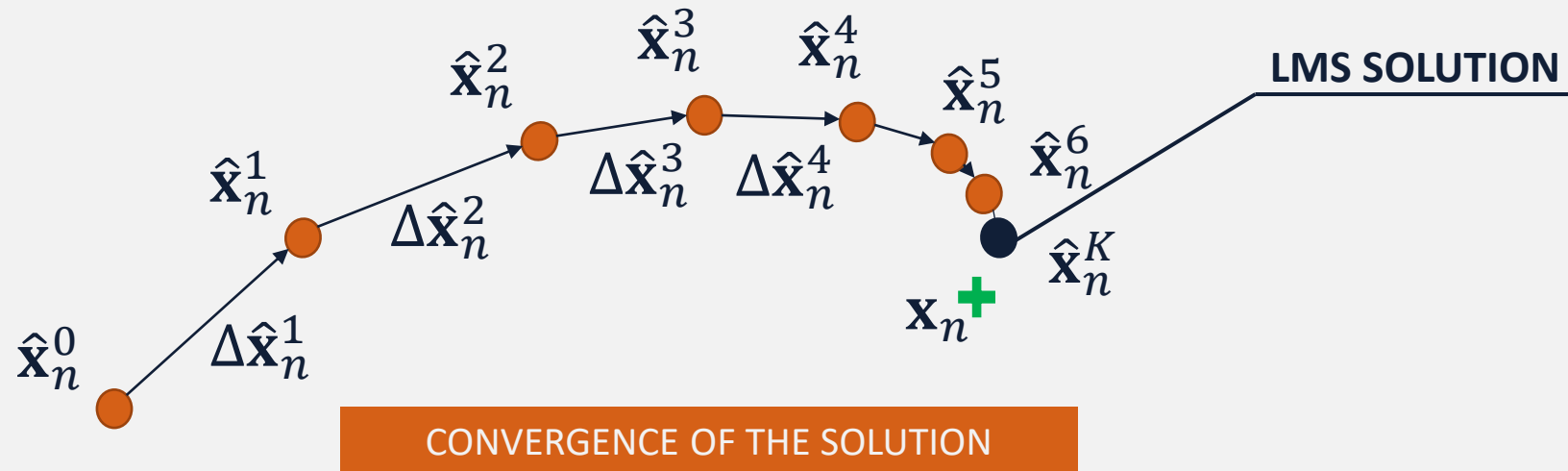
$$\mathbf{H}_n^k = \begin{bmatrix} a_{x,1} & a_{y,1} & a_{z,1} & 1 \\ a_{x,2} & a_{y,2} & a_{z,2} & 1 \\ a_{x,3} & a_{y,3} & a_{z,3} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ a_{x,J_n} & a_{y,J_n} & a_{z,J_n} & 1 \end{bmatrix} \quad a_{x,j} = \left[ \frac{x_{j,n} - \hat{x}_n^k}{\hat{r}_{j,n}} \right], \quad a_{y,j} = \dots, \quad a_{z,j} = \dots$$

$$\hat{r}_{j,n} = \sqrt{(x_{j,n} - \hat{x}_n^k)^2 + (y_{j,n} - \hat{y}_n^k)^2 + (z_{j,n} - \hat{z}_n^k)^2}$$

# Recursive Least Mean Square

- Adding  $\Delta\hat{\mathbf{x}}_n^k$  to the estimated position at the previous iteration, a new estimation of the position is obtained

$$\hat{\mathbf{x}}_n^k = \hat{\mathbf{x}}_n^{k-1} + \Delta\hat{\mathbf{x}}_n^k$$



# From raw to corrected pseudorange



- The measurement of the pseudoranges must be corrected of all the predictable contributions to the errors
- After the correction there will be still a random contribution that is modeled by the User Equivalent Range Error (UERE)

$$UERE \sim \mathcal{N}(0, \sigma_{UERE})$$

# Different errors for different satellites

Basic LMS algorithm attributes the same relevance to all the measurements but actually pseudoranges are not equally precise.

- In real cases, each pseudorange may be characterized by a different value of standard deviation,  $\sigma_{j, \text{UERE}}$
- They can still be considered uncorrelated such that:

$$\mathbf{R} = \text{Cov}(\boldsymbol{\rho}) = \begin{bmatrix} \sigma_{1, \text{UERE}}^2 & 0 & 0 & \dots & 0 \\ 0 & \sigma_{2, \text{UERE}}^2 & 0 & \dots & 0 \\ 0 & 0 & \sigma_{3, \text{UERE}}^2 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \sigma_{J, \text{UERE}}^2 \end{bmatrix}$$

# Different errors for different satellites



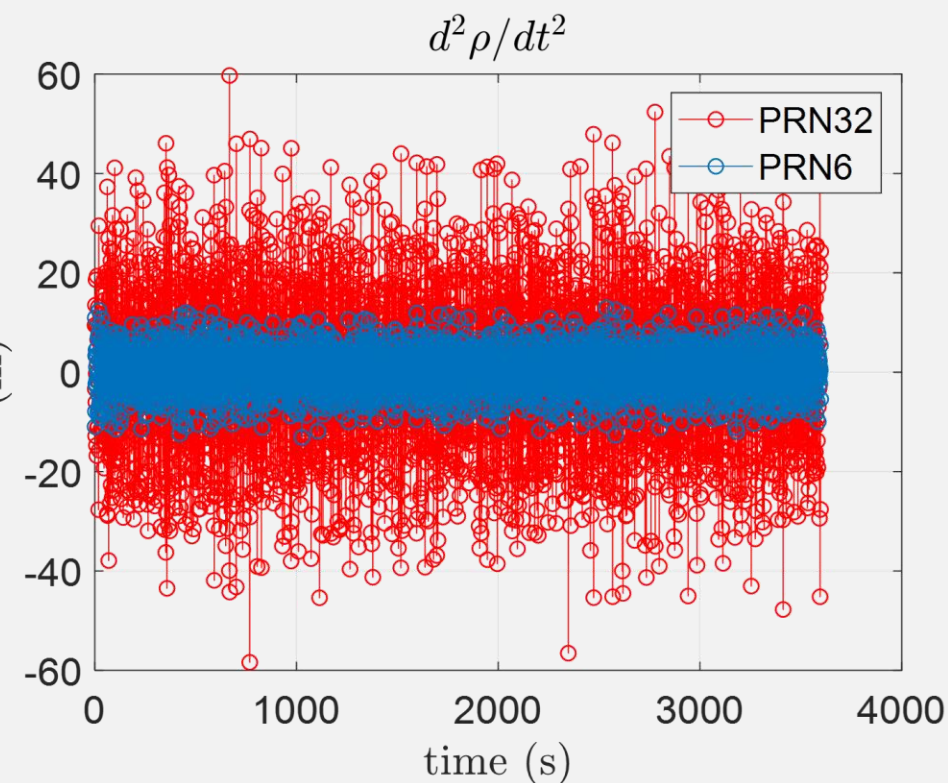
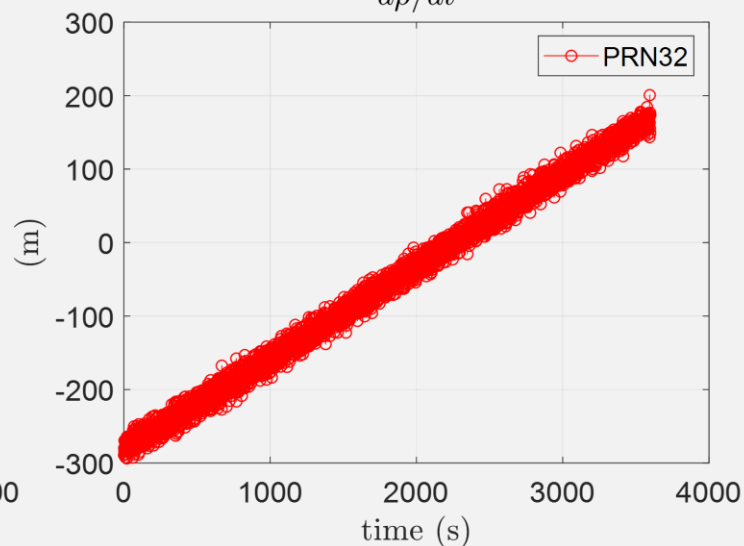
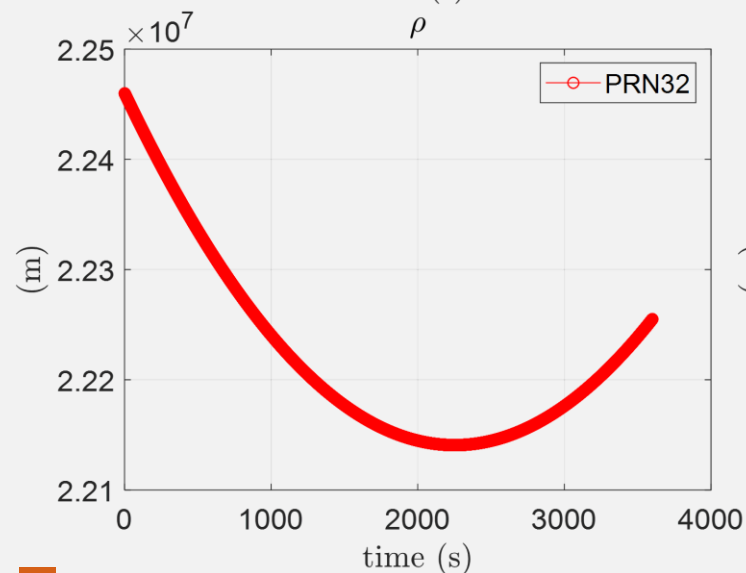
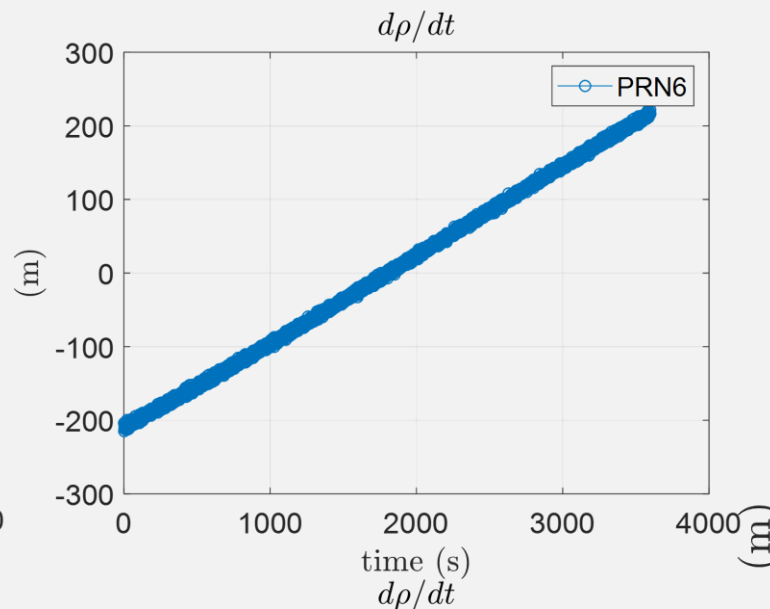
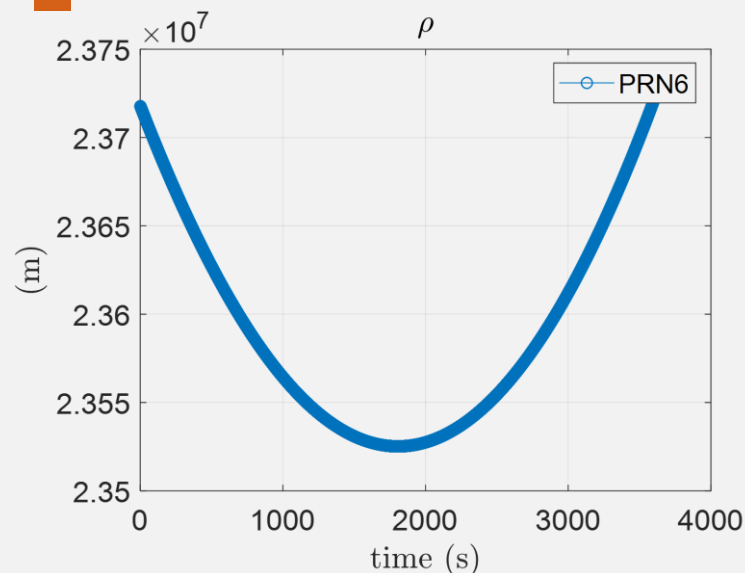
- The direct calculation of the covariance matrix is not possible since the  $\rho_{j,n}$  are time dependent (multiple realizations for the same instant  $n$  are not available)
- We can estimate the  $\sigma_{j, \text{UERE}}$  for each pseudorange  $\rho_j$  analyzing the error on the pseudorange itself along the time (supposing it is ergodic)
- Dependency on time must be removed
- Since pseudorange measurement follows a quadratic trend, the second order derivative can be applied to remove the variation along the time



Use MATLAB function `diff(X, order)`



# Pseudorange de-trending by differentiation



# Weighted Least Mean Square



- A second approach foresees to solve the system by means of a **Weighted Least Squares** (WLS) algorithm, characterized by the introduction of the weighting matrix **W**, which is a positive definite matrix
- Since some measurements may be known to be more accurate than others, the measurement accuracy is known to be characterized by the inverse of the measurement errors covariance matrix **R**
- It is natural to select  $\mathbf{W} = \mathbf{R}^{-1}$  to give the least weighting to the most uncertain measurements.

# Uncorrelated measurements



- The weight matrix can be estimated from the measurements, thus designing a new **weighted geometrical matrix**  $\bar{\mathbf{H}}_n^k$
- For each time instant  $n$ , a number of iterations  $k = 1, \dots, K$  is performed for the same measurements vector  $\boldsymbol{\rho}_n$

Initialization	$\hat{\mathbf{x}}_n^0$
Procedure for $k^{th}$ iteration	$\Delta \hat{\boldsymbol{\rho}}_n^k = \hat{\boldsymbol{\rho}}_n^k - \boldsymbol{\rho}_n$
	$\bar{\mathbf{H}}_n^k = \left( (\mathbf{H}_n^k)^T \mathbf{W} \mathbf{H}_n^k \right)^{-1} (\mathbf{H}_n^k)^T \mathbf{W}$
	$\Delta \hat{\mathbf{x}}_n^k = \bar{\mathbf{H}}_n^k \Delta \hat{\boldsymbol{\rho}}_n^k$
	$\hat{\mathbf{x}}_n^k = \hat{\mathbf{x}}_n^{k-1} + \Delta \hat{\mathbf{x}}_n^k$



CCNE - Lab Exercise #1, 22nd November 2019

# Lab Exercise #1: Evaluation of PVT

Laboratory on Least Mean Square positioning solution



POLITECNICO  
DI TORINO  
Dipartimento  
di Elettronica  
e Telecomunicazioni



# Variables for GPS constellation



## USER and SATELLITE STATE VECTORS (position and clock bias)

Unknown User Position

$$\mathbf{x}_n = [x_n \quad y_n \quad z_n \quad b_{n,GPS}]$$

Known Satellite Position

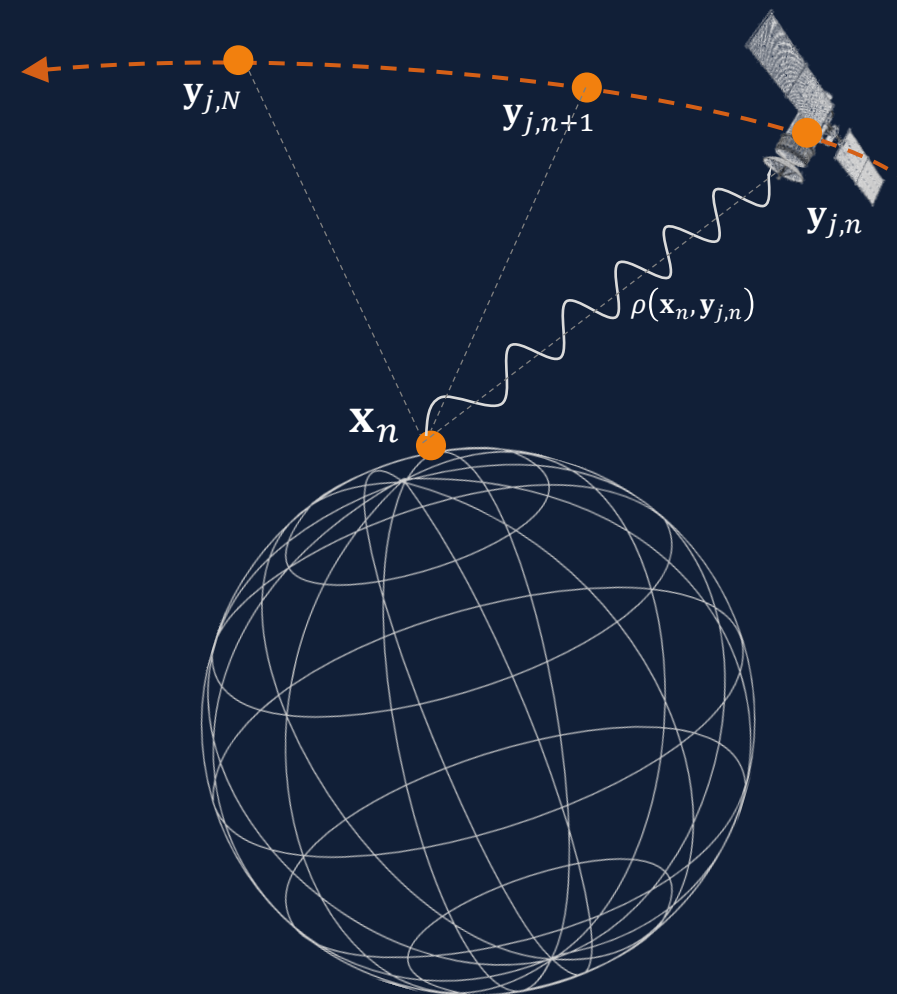
$$\mathbf{y}_{j,n} = [x_{j,n} \quad y_{j,n} \quad z_{j,n}]$$

## PSEUDORANGE EQUATION

Measured range distance from the satellite

$$\rho(\mathbf{x}_n, \mathbf{y}_{j,n}) = \sqrt{(x_n - x_{j,n})^2 + (y_n - y_{j,n})^2 + (z_n - z_{j,n})^2} + b_{n,GPS}$$

- $n \in (1, 2, \dots, N)$  is the time index
- $j \in (1, 2, \dots, J)$  is the satellite identifier
- Data collections include 3600 seconds of satellites observations from a static position  $\mathbf{x}_n$ :  $\mathbf{x}_n = \mathbf{x}_{n+1} = \dots = \mathbf{x}_N$ , where  $N = 3600$





# Observables Data Structures



## PSEUDORANGE MEASUREMENTS (GPS)

RHO.GPS				
	$n = 1$	$n = 2$	..	$n = N$
GPS.PRN_1	$\rho(\mathbf{x}_1, \mathbf{y}_{1,1})$	$\rho(\mathbf{x}_2, \mathbf{y}_{1,2})$	...	$\rho(\mathbf{x}_N, \mathbf{y}_{1,N})$
GPS.PRN_2	$\rho(\mathbf{x}_1, \mathbf{y}_{2,1})$	$\rho(\mathbf{x}_2, \mathbf{y}_{2,2})$	...	$\rho(\mathbf{x}_N, \mathbf{y}_{2,N})$
GPS.PRN_3	$\rho(\mathbf{x}_1, \mathbf{y}_{3,1})$	$\rho(\mathbf{x}_2, \mathbf{y}_{3,2})$	...	$\rho(\mathbf{x}_N, \mathbf{y}_{3,N})$
...	...	...	...	...
GPS.PRN_J	$\rho(\mathbf{x}_1, \mathbf{y}_{J,1})$	$\rho(\mathbf{x}_2, \mathbf{y}_{J,2})$	...	$\rho(\mathbf{x}_N, \mathbf{y}_{J,N})$

## SATELLITE POSITIONS FROM EPHEMERIS (GPS)

SV_ECEF.GPS		Ex: SV Position State Vector History			
SV ID	$\mathbf{Y}_j$	$\mathbf{Y}_1$	$x$	$y$	$z$
GPS.PRN_1	$\mathbf{Y}_1$	$n = 1$	$x_{1,1}$	$y_{1,1}$	$z_{1,1}$
GPS.PRN_2	$\mathbf{Y}_2$	$n = 2$	$x_{1,2}$	$y_{1,2}$	$z_{1,2}$
GPS.PRN_3	$\mathbf{Y}_3$	$n = 3$	$x_{1,3}$	$y_{1,3}$	$z_{1,3}$
...	...	...	...	...	...
GPS.PRN_J	$\mathbf{Y}_J$	$n = N$	$x_{1,N}$	$y_{1,N}$	$z_{1,N}$

- The number of visible satellites is not constant over observation time  $n$
- The folder named «NominalUERE» contains pseudorange measurements with the same  $\sigma_{\text{UERE}}$
- The folder named «RealisticUERE» contains pseudorange measurements with satellite-dependent  $\sigma_{\text{UERE}}$
- Pseudorange measurements can be considered an ergodic random process over short time periods
- It is possible to select different constellations using strings GPS, GLO, BEI, GAL in the data structure

# Lab 1 session A1



TASK

A-1

Load data from the *NominalUERE* folder. Check the satellites visibility at each time instant  $n$  for all the constellations. Plot the number of the visible satellites as well as the measured pseudoranges along the time.

TASK

A-2

Choose a dataset and a constellation (e.g. GPS, Galileo) and develop a Least Mean Square (LMS) positioning algorithm and estimate the user state,  $\hat{\mathbf{x}}_n^K$ , at each time instant  $n$ .  
Verify the position on Google Earth: convert from ECEF to LLA and use *writeKML\_GoogleEarth.m*.

TASK

A-3

Compute the standard deviation of the position error over time  $n = 1, 2, \dots, N$  and compare the quality of the obtained solution for different datasets and constellations. Motivate the results.

TASK

A-4

Estimate the  $\sigma_{j, \text{UERE}}$  for all satellites using the dataset from the *realisticUERE* folder. Implement the **Weigthed Least Mean Square (WLMS)** repeating Tasks 2 and 3.  
Compare the performance of LMS and WLMS on *realisticUERE* dataset.

# OPTIONAL TASKS



TASK

**B-5**

Replace the weighting strategy of the Google PVT with the one developed for the **Weigthed Least Mean Square (WLMS)** in Task A-4

TASK

**B-6**

Replace the PVT computation with your own code developed in the session A

# Simulation hint



For each epoch  $n = 1:N$

- Find the available pseudorange measurements
- Build the measurement vector  $\boldsymbol{\rho}_n = [\rho_{1,n} \quad \dots \quad \rho_{J,n}]$
- Retrieve the corresponding satellites coordinates  $\mathbf{y}_{j,n}$
- Compute PVT solution ( $K$  iterations)

end

Number of LMS/WLMS iterations

Use a reasonable number of iterations, (i.e.  $K < 10$ )

# How to use refMap

- Install the Mapping toolbox
- `refMap(mean(latitude), mean(longitude), 'Here I am!');`

