

COMP0130 Robot Vision and Navigation

Week 1 Seminar

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Friday 13 January 2023



Today's Session

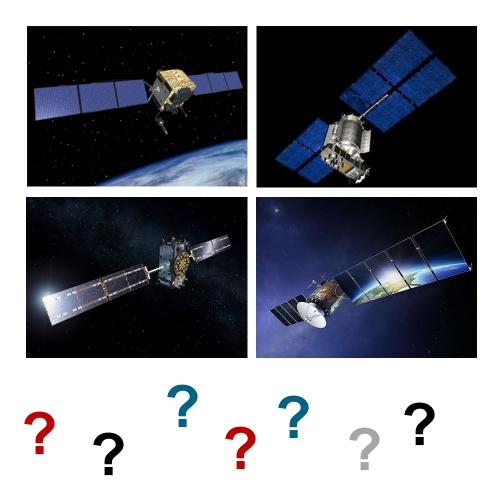
- Week 1 Quiz and Exercises
- Week 1 Summary
- Questions from Students on the Lecture
- Preparing for Next Week



Which of these is not part of GNSS? (Select one)

- 1. GPS
- 2. Galileo
- 3. GLASNOST
- 4. QZSS
- 5. EGNOS



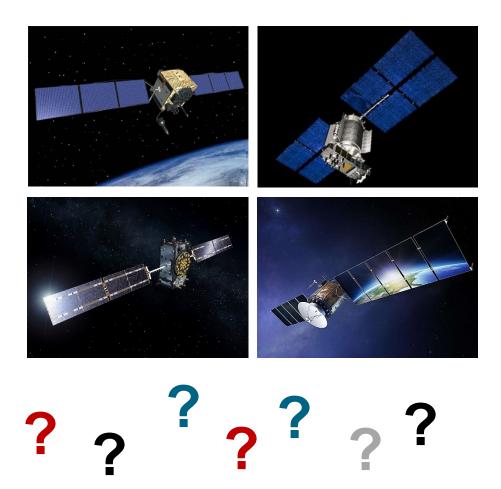




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Which of these best describes the operation of GNSS? (Select one)

- GNSS receivers transmit signals to the satellites which compute their position
- Satellites broadcast signals to receivers, which compute position using range measurements
- 3. Satellites broadcast signals to receivers, which compute position using signal timing measurements
- 4. Satellites broadcast signals to receivers, which compute position by triangulation





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Which of these pairs of measurements have independent errors and which have correlated errors?

- A range measurement and an angle measurement
- A north position solution and an east position solution
- GNSS pseudo-ranges measured at different times 3.





Which of these pairs of measurements have independent errors and which have correlated errors?

- A range measurement and an angle measurement independent
- A north position solution and an east position solution correlated
- GNSS pseudo-ranges measured at different times it depends 3.
 - Successive measurements from the same satellite are highly correlated, while measurements made hours apart are independent





What is GNSS?

GNSS = Global Navigation Satellite System(s)

A generic term covering GPS and similar satellite navigation systems

Four global systems:



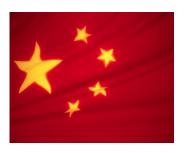
GPS (United States)



GLONASS (Russia)



Galileo (European Union)



Beidou (China)

And various regional augmentation systems

Most modern receivers use two or more of these systems



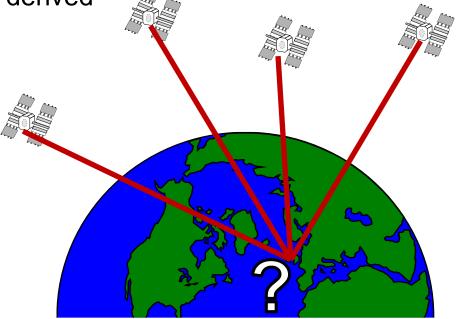
How does GNSS work?

- Each satellite continuously transmits ranging signals
- GNSS user equipment measures signal arrival time, t_{sa} , from 4 (or more) satellites

Each transmission time, t_{st} , is derived from the signal modulation

$$\rho = c(t_{sa} - t_{st})$$

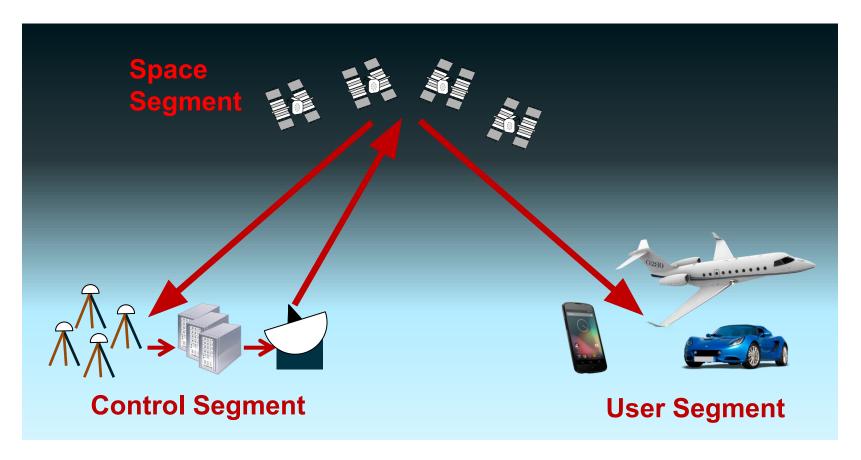
- This is pseudo-range, not range, because of unknown receiver clock offset
- With four pseudo-range measurements, the 3D user position and clock offset may be determined





The Segments of GNSS

Each GNSS comprises three segments





GNSS User Equipment









User **Antenna**

Converts GNSS signals from radio to electrical

Receiver **Hardware**

Downconverts, Samples & Correlates Signals

Ranging **Processor**

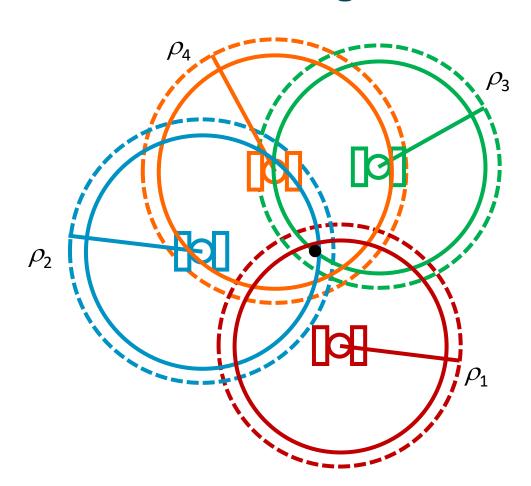
Acquires and tracks the signals

Navigation Processor

Computes position solution



GNSS Positioning Geometry



The position solution is determined by the ranges

With GNSS, we have pseudo-ranges, due to the receiver clock error

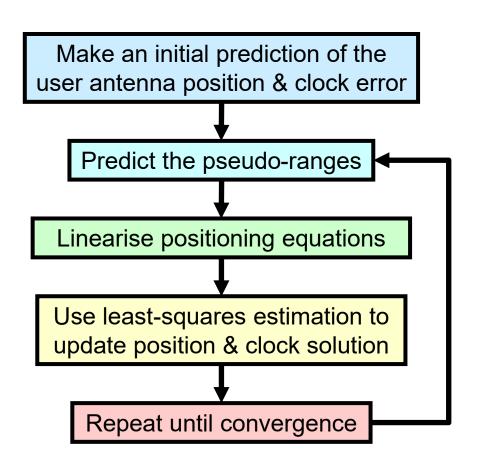
A 4th satellite is needed to determine this

An unknown correction must be applied to the pseudo-ranges to obtain spheres that intersect

In practice, the receiver solves nonlinear simultaneous equations



Single-Epoch Positioning using Least-Squares



- The more accurate the predicted antenna position, the smaller the linearisation errors will be
- Iteration may be needed if the error in the predicted position is more than a few km
 - New prediction = old solution
- Large predicted clock errors do not affect the position solution



Positioning Equations

Predict pseudo-ranges:

$$\hat{\rho}_{a,C}^{s-} = \sqrt{\left(x_{Is}^{I} - \hat{x}_{ea}^{e-}\right)^{2} + \left(y_{Is}^{I} - \hat{y}_{ea}^{e-}\right)^{2} + \left(z_{Is}^{I} - \hat{z}_{ea}^{e-}\right)^{2}} + \delta\hat{\rho}_{c}^{a-} \qquad s \in 1, 2....m$$
Satellite positions (known)

Linearise positioning equations:

$$\begin{pmatrix} \tilde{\rho}_{a,C}^{1} - \hat{\rho}_{a,C}^{1-} \\ \tilde{\rho}_{a,C}^{2} - \hat{\rho}_{a,C}^{2-} \\ \vdots \\ \tilde{\rho}_{a,C}^{m} - \hat{\rho}_{a,C}^{m-} \end{pmatrix} \approx \mathbf{H} \begin{pmatrix} \hat{x}_{ea}^{e+} - \hat{x}_{ea}^{e-} \\ \hat{y}_{ea}^{e+} - \hat{y}_{ea}^{e-} \\ \hat{z}_{ea}^{e+} - \hat{z}_{ea}^{e-} \\ \delta \hat{\rho}_{c}^{a+} - \delta \hat{\rho}_{c}^{a-} \end{pmatrix}$$

$$\begin{pmatrix} \tilde{\rho}_{a,C}^{1} - \hat{\rho}_{a,C}^{1-} \\ \tilde{\rho}_{a,C}^{2} - \hat{\rho}_{a,C}^{2-} \\ \vdots \\ \tilde{\rho}_{a,C}^{m} - \hat{\rho}_{a,C}^{m-} \end{pmatrix} \approx \mathbf{H} \begin{pmatrix} \hat{x}_{ea}^{e+} - \hat{x}_{ea}^{e-} \\ \hat{y}_{ea}^{e+} - \hat{y}_{ea}^{e-} \\ \hat{z}_{ea}^{e+} - \hat{z}_{ea}^{e-} \\ \delta \hat{\rho}_{c}^{a+} - \delta \hat{\rho}_{c}^{a-} \end{pmatrix} \qquad \mathbf{H} = \begin{pmatrix} -\hat{u}_{a1,x}^{e-} & -\hat{u}_{a1,y}^{e-} & -\hat{u}_{a1,z}^{e-} & 1 \\ -\hat{u}_{a2,x}^{e-} & -\hat{u}_{a2,y}^{e-} & -\hat{u}_{a2,z}^{e-} & 1 \\ \vdots & \vdots & \vdots & \vdots \\ -\hat{u}_{am,x}^{e-} & -\hat{u}_{am,y}^{e-} & -\hat{u}_{am,z}^{e-} & 1 \end{pmatrix}$$

$$\mathbf{u}_{as}^{e-} = \frac{\mathbf{C}_{e}^{I} \mathbf{r}_{es}^{e} - \hat{\mathbf{r}}_{ea}^{e-}}{\hat{r}_{as}^{-}} \quad s \in 1, 2....m$$

Predicted Predicted receiver

user position clock offset



Least-Squares Position Solution (Unweighted)

$$\begin{pmatrix} \hat{x}_{ea}^{e+} \\ \hat{y}_{ea}^{e+} \\ \hat{z}_{ea}^{e+} \\ \delta \hat{\rho}_{c}^{a+} \end{pmatrix} = \begin{pmatrix} \hat{x}_{ea}^{e-} \\ \hat{y}_{ea}^{e-} \\ \hat{z}_{ea}^{e-} \\ \delta \hat{\rho}_{c}^{a-} \end{pmatrix} + (\mathbf{H}^{\mathrm{T}} \mathbf{H})^{-1} \mathbf{H}^{\mathrm{T}} \begin{pmatrix} \tilde{\rho}_{a,C}^{1} - \hat{\rho}_{a,C}^{1-} \\ \tilde{\rho}_{a,C}^{2} - \hat{\rho}_{a,C}^{2-} \\ \vdots \\ \tilde{\rho}_{a,C}^{m} - \hat{\rho}_{a,C}^{m-} \end{pmatrix}$$

We can use a similar approach for the velocity:

$$\begin{pmatrix} \hat{\mathbf{v}}_{ea,x}^{e+} \\ \hat{\mathbf{v}}_{ea,y}^{e+} \\ \hat{\mathbf{v}}_{ea,z}^{e+} \\ \hat{\mathbf{v}}_{ea,z}^{e+} \\ \hat{\mathbf{v}}_{ea,z}^{e+} \\ \hat{\mathbf{v}}_{ea,z}^{e+} \end{pmatrix} = \begin{pmatrix} \hat{\mathbf{v}}_{ea,x}^{e-} \\ \hat{\mathbf{v}}_{ea,y}^{e-} \\ \hat{\mathbf{v}}_{ea,z}^{e-} \\ \hat{\mathbf{v}}_{ea,z}^{e-} \\ \hat{\mathbf{v}}_{ea,z}^{e-} \\ \hat{\mathbf{v}}_{ea,z}^{e-} \end{pmatrix} + (\mathbf{H}^{\mathrm{T}}\mathbf{H})^{-1} \mathbf{H}^{\mathrm{T}} \begin{pmatrix} \tilde{\boldsymbol{\rho}}_{a,C}^{1} - \hat{\boldsymbol{\rho}}_{a,C}^{1-} \\ \tilde{\boldsymbol{\rho}}_{a,C}^{2} - \hat{\boldsymbol{\rho}}_{a,C}^{2-} \\ \vdots \\ \tilde{\boldsymbol{\rho}}_{a,C}^{m} - \hat{\boldsymbol{\rho}}_{a,C}^{m-} \end{pmatrix}$$
 Difference between measured and predicted pseudorange rates

$$\hat{\dot{\rho}}_{aj}^{-} = \hat{\mathbf{u}}_{as}^{e-\mathrm{T}} \left[\mathbf{C}_{e}^{I} \left(\hat{\mathbf{v}}_{es}^{e} + \mathbf{\Omega}_{ie}^{e} \hat{\mathbf{r}}_{es}^{e} \right) - \left(\hat{\mathbf{v}}_{ea}^{e-} + \mathbf{\Omega}_{ie}^{e} \hat{\mathbf{r}}_{ea}^{e-} \right) \right] + \delta \hat{\dot{\rho}}_{c}^{a-}$$



Outlier Detection using Normalised Residuals

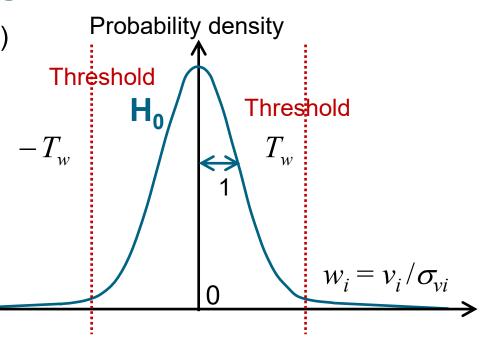
1. Calculate Residuals (unweighted)

$$\mathbf{v} = \left[\mathbf{H} \left(\mathbf{H}^{\mathrm{T}} \mathbf{H}\right)^{-1} \mathbf{H}^{\mathrm{T}} - \mathbf{I}_{m}\right] \delta \mathbf{z}^{-}$$

2. Calculate Residual Covariance

$$\mathbf{C}_{\mathbf{v}} = \left[\mathbf{I}_{m} - \mathbf{H} \left(\mathbf{H}^{\mathrm{T}} \mathbf{H} \right)^{-1} \mathbf{H}^{\mathrm{T}} \right] \sigma_{\rho}^{2}$$

Measurement error variance



- 3. Calculate Normalised Residuals
- 4. Compare them with a threshold

$$w_i = \frac{v_i}{\sigma_{vi}}$$
 $|w_i| \le T_w$: No fault assumed $|w_i| > T_w$: Fault assumed – reject measurement



Questions on the Lectures from the Audience



Preparing for Next Week Monday Workshop

In Workshop 1, you will compute your own GNSS position solution using least-squares estimation

- Simulated data and some source code is available on Moodle
- You will need to write the rest of the source code
- A full mathematical description of the algorithms and answers for testing your code are provided

You can start work immediately if you want to

Help will be available during the Monday lab session

Check your timetable to see if you have been allocated to the morning group or the afternoon group

Preparing for Next Week Monday Workshop

In Workshop 1, you will compute your own GNSS position solution using least-squares estimation

There are four tasks

- Compute a position solution from the first set of measurements at time 0, working step by step
- Compute single-epoch position solutions for the remaining times
- 3. Detect any errors in the measurement data
- Compute a velocity solution

Preparing for Next Week Lectures 2A and 2B

Before Friday 20 January: Watch the recordings of Lectures 2A & 2B

Lecture 2A: GNSS Errors and Advanced Techniques

- Understand GNSS error sources, limitations and performance
- Show how Advanced Techniques can be used to improve performance

Lecture 2B: The Kalman Filter and its use for GNSS

- Introduce sequential least-squares estimation to efficiently process measurements made at different times
- Introduce the Kalman filter for estimating time-varying states
- Apply the Kalman filter to GNSS positioning



Final Questions and Comments

