

## 6. (Geodetic) Observation equations

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- Linearization of nonlinear observation equations
- Adjustment by elements with nonlinear observation equations
- Observation equations of common geodetic measurements
- Numerical example

## Nonlinear observation equations

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- n measurements,
- m necessary observations,
- m unknown parameters
- each observation is a non-linear function of m unknown parameters

$$\left. \begin{aligned} \tilde{\ell}_1 &= f_1(x_1, x_2, \dots, x_m) \\ \tilde{\ell}_2 &= f_2(x_1, x_2, \dots, x_m) \\ &\dots\dots\dots \\ \tilde{\ell}_i &= f_i(x_1, x_2, \dots, x_m) \\ &\dots\dots\dots \\ \tilde{\ell}_n &= f_n(x_1, x_2, \dots, x_m) \end{aligned} \right\}$$

## Unknown parameters

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Approximate values of unknown parameters:

$$X_{m \cdot 1}^0 = \begin{bmatrix} x_1^0 \\ x_2^0 \\ \dots \\ x_m^0 \end{bmatrix}$$

Corrections to the approximate values:

$$\delta X_{m \cdot 1} = \begin{bmatrix} \delta x_1 \\ \delta x_2 \\ \dots \\ \delta x_m \end{bmatrix}$$

Correct values of unknown parameters:

$$X_{m \cdot 1} = \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_m \end{bmatrix} = X_{m \cdot 1}^0 + \delta X_{m \cdot 1}$$

## Linearization of one nonlinear function

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$$\tilde{\ell}_i = \ell_i - \varepsilon_i = f(x_1^0 + \delta x_1, x_2^0 + \delta x_2, \dots, x_m^0 + \delta x_m)$$

$$\begin{aligned} &= f(x_1^0, x_2^0, \dots, x_m^0) + \frac{1}{1!} \left[ \frac{\partial f}{\partial x_1} \delta x_1 + \frac{\partial f}{\partial x_2} \delta x_2 + \dots + \frac{\partial f}{\partial x_m} \delta x_m \right] + \\ &+ \frac{1}{2!} \left[ \frac{\partial^2 f}{\partial x_1^2} \delta x_1^2 + \frac{\partial^2 f}{\partial x_1 \partial x_2} \delta x_1 \delta x_2 + \dots + \frac{\partial^2 f}{\partial x_1 \partial x_m} \delta x_1 \delta x_m \right] + \dots \end{aligned}$$

If corrections  $\delta x_k$  are sufficiently small, terms of order 2 and higher can be neglected. *Keeping linear terms = linearization !*

$$\ell_i - \varepsilon_i = a_{i1} \delta x_1 + a_{i2} \delta x_2 + \dots + a_{im} \delta x_m + c_i$$

$$a_{ij} = \frac{\partial f_i}{\partial x_j}, \quad c_i = f_i(x_1^0, x_2^0, \dots, x_m^0), \quad i = 1, 2, \dots, n; \quad j = 1, 2, \dots, m$$

## Linearization of all $n$ measurements

$$\left. \begin{aligned} \tilde{\ell}_1 &= f_1(x_1, x_2, \dots, x_m) \\ \tilde{\ell}_2 &= f_2(x_1, x_2, \dots, x_m) \\ &\dots\dots\dots \\ \tilde{\ell}_i &= f_i(x_1, x_2, \dots, x_m) \\ &\dots\dots\dots \\ \tilde{\ell}_n &= f_n(x_1, x_2, \dots, x_m) \end{aligned} \right\} \quad \begin{aligned} L_{n \cdot 1} - \varepsilon_{n \cdot 1} &= A_{n \cdot m} \delta X_{m \cdot 1} \\ L_{n \cdot 1} &= L'_{n \cdot 1} - c_{n \cdot 1}, \quad L'_{n \cdot 1} = \begin{bmatrix} \ell_1 \\ \ell_2 \\ \dots \\ \ell_n \end{bmatrix}, \quad c_{n \cdot 1} = \begin{bmatrix} c_1 \\ c_2 \\ \dots \\ c_n \end{bmatrix}, \\ c_i &= f_i(x_1^0, x_2^0, \dots, x_m^0) \\ A_{n \cdot m} &= \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{bmatrix} \\ a_{ij} &= \frac{\partial f_i}{\partial x_j} \quad (1 \leq i \leq n; 1 \leq j \leq m) \end{aligned}$$

## Estimate of the unknown parameters

Least squares estimate of parameter correction:  $\delta \hat{X} = (A^\top P A)^{-1} A^\top P L$

Least squares estimate of the total parameters:  $\hat{X} = X^0 + \delta \hat{X}$

Variance-covariance matrix:  $C_{\hat{X}\hat{X}} = C_{\delta \hat{X} \delta \hat{X}} = \sigma_0^2 (A^\top P A)^{-1}$

Unbiased estimate of the variance factor:  $\hat{\sigma}_0^2 = \frac{\hat{\varepsilon}^\top P \hat{\varepsilon}}{n - m}$

## Estimates of other quantities

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Least squares estimates of

- the errors (residuals),
- the reduced measurements and
- the original measurements:

$$\hat{\varepsilon} = L - A\delta\hat{X} = [I - A(A^T P A)^{-1} A^T P] L$$

$$\hat{L} = L - \hat{\varepsilon} = A \delta\hat{X} = A(A^T P A)^{-1} A^T P L$$

$$\hat{L}' = \hat{L} + c$$

..... and their variance-covariance matrices:

$$C_{\hat{\varepsilon}\hat{\varepsilon}} = \sigma_0^2 [P^{-1} - A(A^T P A)^{-1} A^T]$$

$$C_{\hat{L}\hat{L}} = C_{\hat{L}'\hat{L}'} = \sigma_0^2 A(A^T P A)^{-1} A^T$$

## Selection of unknown parameters

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- Most often, coordinates of unknown points are chosen as unknown parameters
- Sufficient initial data (datum) must exist.  
Otherwise the network has datum defect.  
Then the normal equation matrix will be singular.
- A network without any initial data is called a *free network*.
- Adjustment of networks with datum defect requires the use of generalized matrix inverses.  
See Chapter 4 !

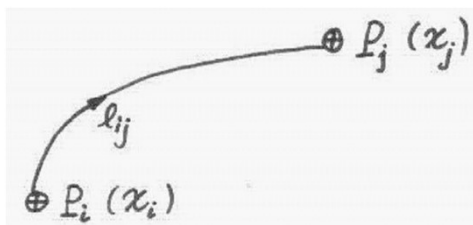
## Observation equations of common geodetic measurements

- height differences from levelling
- 3D or 2D distances
- horizontal angles
- vertical angles
- GPS measurements – code pseudo-range
- GPS measurements – phase pseudo-range

Most often, coordinates of points are chosen as unknown parameters

## A height difference from levelling

Heights of unknown benchmarks are chosen as unknown parameters

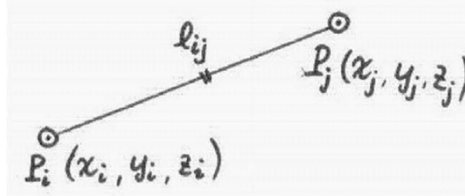


$$l_{ij} - \varepsilon_{ij} = x_j - x_i$$

Linear observation equation !

## 3D distance measurement

Coordinates are chosen as the unknown parameters



$$l_{ij} - \varepsilon_{ij} = \sqrt{(x_j - x_i)^2 + (y_j - y_i)^2 + (z_j - z_i)^2}$$

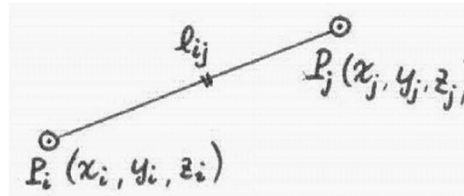
Nonlinear equation !

To linearize the above equation, approximate coordinates are needed.

From approximate coordinates, approximate distance can be derived:

$$s_{ij}^0 = \sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2 + (z_j^0 - z_i^0)^2}$$

## Linearization of a 3D distance



$$(l_{ij} - s_{ij}^0) - \varepsilon_{ij} = a \cdot \delta x_i + b \cdot \delta y_i + c \cdot \delta z_i + d \cdot \delta x_j + e \cdot \delta y_j + f \cdot \delta z_j$$

$$a = \frac{\partial l_{ij}}{\partial x_i} = -\frac{x_j^0 - x_i^0}{s_{ij}^0}$$

$$d = \frac{\partial l_{ij}}{\partial x_j} = \frac{x_j^0 - x_i^0}{s_{ij}^0} = -a$$

$$b = \frac{\partial l_{ij}}{\partial y_i} = -\frac{y_j^0 - y_i^0}{s_{ij}^0}$$

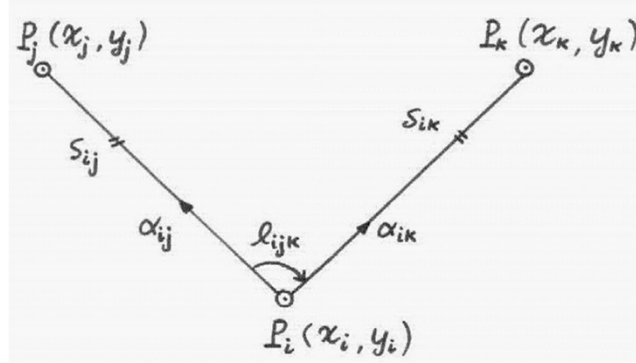
$$e = \frac{\partial l_{ij}}{\partial y_j} = \frac{y_j^0 - y_i^0}{s_{ij}^0} = -b$$

$$c = \frac{\partial l_{ij}}{\partial z_i} = -\frac{z_j^0 - z_i^0}{s_{ij}^0}$$

$$f = \frac{\partial l_{ij}}{\partial z_j} = \frac{z_j^0 - z_i^0}{s_{ij}^0} = -c$$

For 2D horizontal distances, terms related to z are to be omitted.

## A 2D horizontal angle



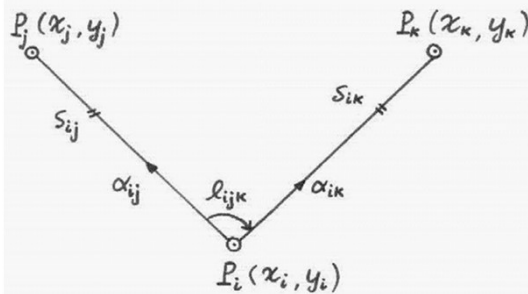
$$l_{ijk} - \varepsilon_{ijk} = \arctan\left(\frac{y_k - y_i}{x_k - x_i}\right) - \arctan\left(\frac{y_j - y_i}{x_j - x_i}\right)$$

Nonlinear equation !

## Quantities derived from approximate coordinates

To linearize the nonlinear observation equation, approximate coordinates are introduced.

From the approximate coordinates, approximate values of angles, azimuths and distances can be derived.



$$\beta_{ijk}^0 = \alpha_{ik}^0 - \alpha_{ij}^0$$

$$\alpha_{ik}^0 = \arctan\left(\frac{y_k^0 - y_i^0}{x_k^0 - x_i^0}\right)$$

$$\alpha_{ij}^0 = \arctan\left(\frac{y_j^0 - y_i^0}{x_j^0 - x_i^0}\right)$$

$$s_{ij}^0 = \sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2}$$

$$s_{ik}^0 = \sqrt{(x_k^0 - x_i^0)^2 + (y_k^0 - y_i^0)^2}$$

## Linearization of a horizontal angle

$$(\ell_{ijk} - \beta_{ijk}^0) - \varepsilon_{ijk} = a \cdot \delta x_i + b \cdot \delta y_i + c \cdot \delta x_j + d \cdot \delta y_j + e \cdot \delta x_k + f \cdot \delta y_k$$

$$a = \frac{\partial \ell_{ijk}}{\partial x_i} = \frac{y_k^0 - y_i^0}{(s_{ik}^0)^2} - \frac{y_j^0 - y_i^0}{(s_{ij}^0)^2} = \frac{\sin \alpha_{ik}^0}{s_{ik}^0} - \frac{\sin \alpha_{ij}^0}{s_{ij}^0}$$

$$b = \frac{\partial \ell_{ijk}}{\partial y_i} = -\frac{x_k^0 - x_i^0}{(s_{ik}^0)^2} + \frac{x_j^0 - x_i^0}{(s_{ij}^0)^2} = -\frac{\cos \alpha_{ik}^0}{s_{ik}^0} + \frac{\cos \alpha_{ij}^0}{s_{ij}^0}$$

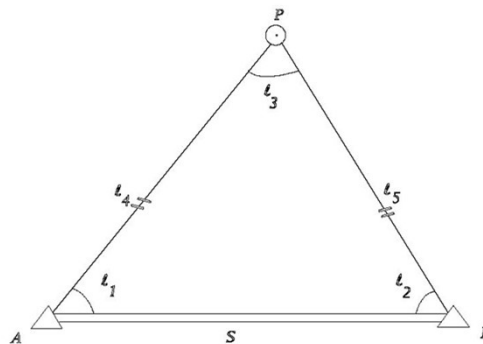
$$c = \frac{\partial \ell_{ijk}}{\partial x_j} = \frac{y_j^0 - y_i^0}{(s_{ij}^0)^2} = \frac{\sin \alpha_{ij}^0}{s_{ij}^0}$$

$$d = \frac{\partial \ell_{ijk}}{\partial y_j} = -\frac{x_j^0 - x_i^0}{(s_{ij}^0)^2} = -\frac{\cos \alpha_{ij}^0}{s_{ij}^0}$$

$$e = \frac{\partial \ell_{ijk}}{\partial x_k} = -\frac{y_k^0 - y_i^0}{(s_{ik}^0)^2} = -\frac{\sin \alpha_{ik}^0}{s_{ik}^0}$$

$$f = \frac{\partial \ell_{ijk}}{\partial y_k} = \frac{x_k^0 - x_i^0}{(s_{ik}^0)^2} = \frac{\cos \alpha_{ik}^0}{s_{ik}^0}$$

## An example of adjustment by elements



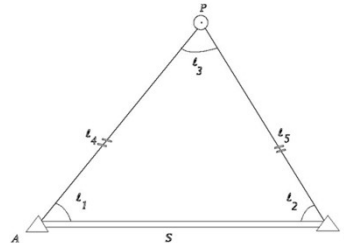
In a 2D triangulation network, P is an unknown point and A, B are fixed points with given coordinates :

$$\begin{aligned} x_A &= 6\,500\,000.000 \text{ m}, & x_B &= 6\,500\,060.000 \text{ m} \\ y_A &= 1\,500\,000.000 \text{ m}, & y_B &= 1\,500\,080.000 \text{ m} \end{aligned}$$



## The measurements

- 3 angles and 2 distances
- all uncorrelated.



$$L' = \begin{bmatrix} l_1 \\ l_2 \\ l_3 \\ l_4 \\ l_5 \end{bmatrix} = \begin{bmatrix} 60^0 & 00' & 05'' \\ 60^0 & 00' & 05'' \\ 59^0 & 59' & 58'' \\ 100.008 & m \\ 99.997 & m \end{bmatrix}, \quad \sigma = \begin{bmatrix} \sigma_1 \\ \sigma_2 \\ \sigma_3 \\ \sigma_4 \\ \sigma_5 \end{bmatrix} = \begin{bmatrix} \pm 6'' \\ \pm 6'' \\ \pm 6'' \\ \pm 3 \text{ mm} \\ \pm 3 \text{ mm} \end{bmatrix}$$

If we chose variance factor as  $\sigma_0 = \pm 3 \text{ mm}$ , then the weight matrix becomes:

$$p_i = \frac{\sigma_0^2}{\sigma_i^2} \rightarrow P_{5 \times 5} = \begin{bmatrix} \frac{1}{4} & & & & \\ & \frac{1}{4} & & & \\ & & \frac{1}{4} & & \\ & & & 1 & \\ & & & & 1 \end{bmatrix}$$

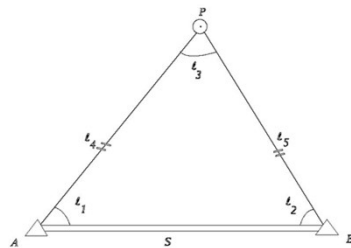
## Analysis of the network

- $n=5$  measurements,  $\rightarrow n=5$  observation equations
- To determine the position of P,  $m=2$  necessary measurements are needed
- $\rightarrow m=2$  unknown parameters in adjustment by elements
- We choose the coordinates of P,  $(x_p, y_p)$ , as parameters
- There are  $n-m=3$  over-determinations (redundancies)

## Procedures to compute linearized observation equations

- Find approximate coordinates for the unknown point P
- Calculate approximate distances and angles
- Calculate linearized observation equations for angles
- Calculate linearized observation equations for distances

## Find approximate coordinates of P



$$S = s_{AB} = \sqrt{(x_B - x_A)^2 + (y_B - y_A)^2} = \sqrt{60.000^2 + 80.000^2} = 100.000 \text{ (m)}$$

$$\alpha_{AB} = \arctan\left(\frac{y_B - y_A}{x_B - x_A}\right) = \arctan\left(\frac{80.000}{60.000}\right) = 53.1301024^0 = 53^0 7' 48.37''$$

$$\alpha_{BA} = \alpha_{AB} + 180^0 = 233.1301024^0 = 233^0 7' 48.37''$$

$$\alpha_{AP}^0 = \alpha_{AB} - \ell_1 = 353^0 7' 43.35''$$

$$x_P^0 = x_A + \ell_4 \cos \alpha_{AP}^0 = x_A + 99.2897 \text{ m} = 6\ 500\ 099.2897 \text{ m}$$

$$y_P^0 = y_A + \ell_4 \sin \alpha_{AP}^0 = y_A - 11.9649 \text{ m} = 1\ 499\ 988.0351 \text{ m}$$

## Calculate approximate angles & distances

From	To	Distance $s^0$ (m)	Azimuth $\alpha^0$ ( $^\circ$ ' ")	Remark
A	B	100.0000	$53^0 07' 48.37''$	fixed
A	P	100.0080	$353^0 07' 43.35''$	approximate
B	A	100.0000	$233^0 07' 48.37''$	fixed
B	P	100.0061	$293^0 08' 00.18''$	approximate
P	A	100.0080	$173^0 07' 43.35''$	<b>Approximate</b>
P	B	100.0061	$113^0 08' 00.18''$	approximate

## Linearization of a horizontal angle

$$(\ell_{ijk} - \beta_{ijk}^0) - \varepsilon_{ijk} = a \cdot \delta x_i + b \cdot \delta y_i + c \cdot \delta x_j + d \cdot \delta y_j + e \cdot \delta x_k + f \cdot \delta y_k$$

$$a = \frac{\partial \ell_{ijk}}{\partial x_i} = \frac{y_k^0 - y_i^0}{(s_{ik}^0)^2} - \frac{y_j^0 - y_i^0}{(s_{ij}^0)^2} = \frac{\sin \alpha_{ik}^0}{s_{ik}^0} - \frac{\sin \alpha_{ij}^0}{s_{ij}^0}$$

$$b = \frac{\partial \ell_{ijk}}{\partial y_i} = -\frac{x_k^0 - x_i^0}{(s_{ik}^0)^2} + \frac{x_j^0 - x_i^0}{(s_{ij}^0)^2} = -\frac{\cos \alpha_{ik}^0}{s_{ik}^0} + \frac{\cos \alpha_{ij}^0}{s_{ij}^0}$$

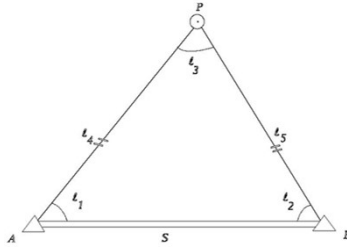
$$c = \frac{\partial \ell_{ijk}}{\partial x_j} = \frac{y_i^0 - y_j^0}{(s_{ij}^0)^2} = \frac{\sin \alpha_{ji}^0}{s_{ij}^0}$$

$$d = \frac{\partial \ell_{ijk}}{\partial y_j} = -\frac{x_j^0 - x_i^0}{(s_{ij}^0)^2} = -\frac{\cos \alpha_{ij}^0}{s_{ij}^0}$$

$$e = \frac{\partial \ell_{ijk}}{\partial x_k} = -\frac{y_k^0 - y_i^0}{(s_{ik}^0)^2} = -\frac{\sin \alpha_{ik}^0}{s_{ik}^0}$$

$$f = \frac{\partial \ell_{ijk}}{\partial y_k} = \frac{x_k^0 - x_i^0}{(s_{ik}^0)^2} = \frac{\cos \alpha_{ik}^0}{s_{ik}^0}$$

## Linearized equations for angles



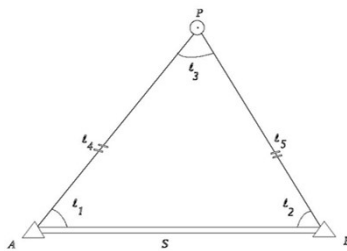
If one of the 2 points is fixed, terms related to that point do *not* exist.

$$\ell_1 - \varepsilon_1 = (\alpha_{AB} - \alpha_{AP}^0) + \frac{\rho''}{1000} \frac{\sin \alpha_{AP}^0}{s_{AP}^0} \delta x_p - \frac{\rho''}{1000} \frac{\cos \alpha_{AP}^0}{s_{AP}^0} \delta y_p$$

$$\ell_2 - \varepsilon_2 = (\alpha_{BP}^0 - \alpha_{BA}^0) - \frac{\rho''}{1000} \frac{\sin \alpha_{BP}^0}{s_{BP}^0} \delta x_p + \frac{\rho''}{1000} \frac{\cos \alpha_{BP}^0}{s_{BP}^0} \delta y_p$$

$$\ell_3 - \varepsilon_3 = (\alpha_{PA}^0 - \alpha_{PB}^0) + \frac{\rho''}{1000} \left( \frac{\sin \alpha_{PA}^0}{s_{PA}^0} - \frac{\sin \alpha_{PB}^0}{s_{PB}^0} \right) \delta x_p - \frac{\rho''}{1000} \left( \frac{\cos \alpha_{PA}^0}{s_{PA}^0} - \frac{\cos \alpha_{PB}^0}{s_{PB}^0} \right) \delta y_p$$

## Linearized equations for angles



$$\frac{\rho''}{1000}$$



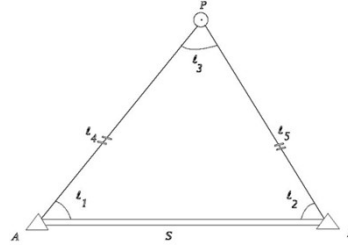
coordinate corrections  
in mm and residuals in "

$$\ell_1 - \varepsilon_1 = (\alpha_{AB} - \alpha_{AP}^0) + \frac{\rho''}{1000} \frac{\sin \alpha_{AP}^0}{s_{AP}^0} \delta x_p - \frac{\rho''}{1000} \frac{\cos \alpha_{AP}^0}{s_{AP}^0} \delta y_p$$

$$\ell_2 - \varepsilon_2 = (\alpha_{BP}^0 - \alpha_{BA}^0) - \frac{\rho''}{1000} \frac{\sin \alpha_{BP}^0}{s_{BP}^0} \delta x_p + \frac{\rho''}{1000} \frac{\cos \alpha_{BP}^0}{s_{BP}^0} \delta y_p$$

$$\ell_3 - \varepsilon_3 = (\alpha_{PA}^0 - \alpha_{PB}^0) + \frac{\rho''}{1000} \left( \frac{\sin \alpha_{PA}^0}{s_{PA}^0} - \frac{\sin \alpha_{PB}^0}{s_{PB}^0} \right) \delta x_p - \frac{\rho''}{1000} \left( \frac{\cos \alpha_{PA}^0}{s_{PA}^0} - \frac{\cos \alpha_{PB}^0}{s_{PB}^0} \right) \delta y_p$$

## Linearized equations for distances



$$\ell_4 - \varepsilon_4 = s_{AP}^0 + \cos \alpha_{AP}^0 \delta x_p + \sin \alpha_{AP}^0 \delta y_p$$

$$\ell_5 - \varepsilon_5 = s_{PB}^0 + \cos \alpha_{PB}^0 \delta x_p + \sin \alpha_{PB}^0 \delta y_p$$

## Numerical values of observation equations

$$\begin{aligned} 0'' - \varepsilon_1 &= -0.246\,754\,45 \delta x_p - 2.047\,670\,68 \delta y_p \\ -8.81'' - \varepsilon_2 &= +1.896\,681\,98 \delta x_p + 0.810\,309\,87 \delta y_p \\ 14.83'' - \varepsilon_3 &= -1.649\,927\,53 \delta x_p + 1.237\,360\,81 \delta y_p \\ 0^{mm} - \varepsilon_4 &= +0.992\,817\,41 \delta x_p - 0.119\,639\,41 \delta y_p \\ -9.12^{mm} - \varepsilon_5 &= +0.392\,872\,97 \delta x_p - 0.919\,592\,75 \delta y_p \end{aligned}$$



$$\underset{5 \cdot 1}{L} - \underset{5 \cdot 1}{\varepsilon} = \underset{5 \cdot 2 \ 2 \cdot 1}{A} \underset{2 \cdot 1}{X}$$



$$L = \begin{bmatrix} 0 \\ -8.81 \\ +14.83 \\ 0 \\ -9.12 \end{bmatrix}, \quad \varepsilon = \begin{bmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \varepsilon_3 \\ \varepsilon_4 \\ \varepsilon_5 \end{bmatrix}, \quad A = \begin{bmatrix} -0.246\,754\,45 & -2.047\,670\,68 \\ +1.896\,681\,98 & +0.810\,309\,87 \\ -1.649\,927\,53 & +1.237\,360\,81 \\ +0.992\,817\,41 & -0.119\,639\,41 \\ +0.392\,872\,97 & -0.919\,592\,75 \end{bmatrix}, \quad \delta X = \begin{bmatrix} \delta x_p \\ \delta y_p \end{bmatrix}$$

## Calculation of least squares estimates

$$A^T P A = \begin{bmatrix} +2.735\ 173 & -0.479\ 909 \\ -0.479\ 909 & +2.455\ 119 \end{bmatrix}, \quad |A^T P A| = 6.484\ 863$$

$$(A^T P A)^{-1} = \frac{1}{6.484\ 863} \begin{bmatrix} +2.455\ 119 & +0.479\ 909 \\ +0.479\ 909 & +2.735\ 173 \end{bmatrix} = \begin{bmatrix} +0.378\ 592 & +0.074\ 005 \\ +0.074\ 005 & +0.421\ 778 \end{bmatrix}$$

$$A^T P L = \begin{bmatrix} -13.890\ 199 \\ +11.199\ 127 \end{bmatrix}$$

$$\delta \hat{X} = (A^T P A)^{-1} A^T P L = \begin{bmatrix} -4.43 \\ +3.70 \end{bmatrix} \text{ (mm)}$$

$$\hat{X} = X^0 + \delta \hat{X} = \begin{bmatrix} 6\ 500\ 099.2897 \\ 1\ 499\ 988.0351 \end{bmatrix} + \begin{bmatrix} -0.00443 \\ +0.00370 \end{bmatrix} = \begin{bmatrix} 6500\ 099.2853 \\ 1499\ 988.0388 \end{bmatrix} \text{ (m)}$$

## Error estimates of parameters (coordinates)

$$\hat{\sigma}_0^2 = \frac{\hat{\varepsilon}^T P \hat{\varepsilon}}{n-m} = \frac{1}{5-2} \sum_{i=1}^n \{p_i \hat{\varepsilon}_i^2\} \approx 18.1885 \text{ mm}^2$$

$$\hat{\sigma}_0 \approx \pm 4.26 \text{ mm}$$

$$C_{\hat{X}\hat{X}} = \hat{\sigma}_0^2 (A^T P A)^{-1} = \hat{\sigma}_0^2 \begin{bmatrix} +0.378\ 592 & +0.074\ 005 \\ +0.074\ 005 & +0.421\ 778 \end{bmatrix}$$

$$\hat{\sigma}_{x_P} = \hat{\sigma}_0 \sqrt{0.378572} \approx \pm 2.62 \text{ mm};$$

$$\hat{\sigma}_{y_P} = \hat{\sigma}_0 \sqrt{0.421778} \approx \pm 2.77 \text{ mm}.$$

$$\sigma_P = \sqrt{\sigma_{x_P}^2 + \sigma_{y_P}^2} = \hat{\sigma}_0 \sqrt{0.378572 + 0.421778} \approx \pm 3.82 \text{ mm}$$

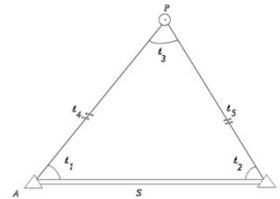
## Calculation of least squares estimates

$$\hat{\varepsilon} = L - A \delta \hat{X} = \begin{bmatrix} +6.45'' \\ -3.40'' \\ +2.95'' \\ +4.82\text{ mm} \\ -3.98\text{ mm} \end{bmatrix}$$

$$\hat{L}' = L' - \hat{\varepsilon} = \begin{bmatrix} 60^0 & 00' & 05'' \\ 60^0 & 00' & 03'' \\ 59^0 & 59' & 58'' \\ 100.008\text{ m} \\ 99.997\text{ m} \end{bmatrix} - \begin{bmatrix} +6.45'' \\ -3.40'' \\ +2.95'' \\ +4.82\text{ mm} \\ -3.98\text{ mm} \end{bmatrix} = \begin{bmatrix} 59^0 & 59' & 58.55'' \\ 60^0 & 00' & 06.40'' \\ 59^0 & 59' & 55.05'' \\ 100.0032\text{ m} \\ 100.0010\text{ m} \end{bmatrix}$$

## Error estimates of adjusted measurements

$$C_{\hat{L}\hat{L}} = \hat{\sigma}_0^2 A(A^T P A)^{-1} A^T =$$



$$= \hat{\sigma}_0^2 \begin{bmatrix} +1.866\ 333 & -1.179\ 235 & -0.687\ 098 & -0.137\ 684 & +0.714\ 774 \\ -1.179\ 235 & +1.866\ 364 & -0.687\ 129 & +0.714\ 765 & -0.137\ 698 \\ -0.687\ 098 & -0.687\ 129 & +1.374\ 227 & -0.577\ 081 & -0.577\ 076 \\ -0.137\ 684 & +0.714\ 765 & -0.577\ 081 & +0.361\ 630 & +0.123\ 031 \\ +0.714\ 774 & -0.137\ 698 & -0.577\ 076 & +0.123\ 031 & +0.361\ 639 \end{bmatrix}$$

$$\hat{\sigma}_1 = \hat{\sigma}_0 \sqrt{1.866333} \approx \pm 5.83''$$

$$\hat{\sigma}_2 = \hat{\sigma}_0 \sqrt{1.866364} \approx \pm 5.83''$$

$$\hat{\sigma}_3 = \hat{\sigma}_0 \sqrt{1.374227} \approx \pm 5.00''$$

$$\hat{\sigma}_4 = \hat{\sigma}_0 \sqrt{0.361630} \approx \pm 2.56\text{ mm}$$

$$\hat{\sigma}_5 = \hat{\sigma}_0 \sqrt{0.361639} \approx \pm 2.56\text{ mm}$$

$$\begin{matrix} \rightarrow \\ \leftarrow \end{matrix} \sigma = \begin{bmatrix} \sigma_1 \\ \sigma_2 \\ \sigma_3 \\ \sigma_4 \\ \sigma_5 \end{bmatrix} = \begin{bmatrix} \pm 6'' \\ \pm 6'' \\ \pm 6'' \\ \pm 3\text{ mm} \\ \pm 3\text{ mm} \end{bmatrix}$$

## Observation equation of the vertical angle

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Local 3D measurements:

- ✓ slope distance (3D)
- ✓ horizontal angle (2D)
- ✓ vertical angle

$$\ell_{ij} - \varepsilon_{ij} = \arctan \left( \frac{z_j - z_i}{\sqrt{(x_j - x_i)^2 + (y_j - y_i)^2}} \right)$$

approximate coordinates  $(x_i^0, y_i^0, z_i^0)$ ,  $(x_j^0, y_j^0, z_j^0)$  of  $P_i$  and  $P_j$

$$\ell_{ij}^0 = \arctan \left( \frac{z_j^0 - z_i^0}{\sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2}} \right)$$

$$(\ell_{ij} - \ell_{ij}^0) - \varepsilon_{ij} = a \delta x_i + b \delta y_i + c \delta z_i + d \delta x_j + e \delta y_j + f \delta z_j$$

## Coefficients for vertical angle measurement

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$$a = \frac{1}{s_{ij}^0} \frac{(x_j^0 - x_i^0) (z_j^0 - z_i^0)}{\sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2}}$$

$$b = \frac{1}{s_{ij}^0} \frac{(y_j^0 - y_i^0) (z_j^0 - z_i^0)}{\sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2}}$$

$$c = \frac{1}{s_{ij}^0} \sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2}$$

$$d = -a, \quad e = -b, \quad f = -c$$

$$s_{ij}^0 = \sqrt{(x_j^0 - x_i^0)^2 + (y_j^0 - y_i^0)^2 + (z_j^0 - z_i^0)^2}$$



## GPS code (pseudo-range) measurement

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$$s_{ik} = c_0 \cdot \Delta t_{ik} = \sqrt{(x_k - x_i)^2 + (y_k - y_i)^2 + (z_k - z_i)^2}$$

approximate coordinates  $(x_i^0, y_i^0, z_i^0)$  of the receiver  $P_i$

$$s_{ik}^0 = \sqrt{(x_k - x_i^0)^2 + (y_k - y_i^0)^2 + (z_k - z_i^0)^2}$$

$$(\ell_{ik} - s_{ik}^0) - \varepsilon_{ik} = a \delta x_i + b \delta y_i + c \delta z_i + d \delta t_{ik}$$

$$a = \frac{\partial \ell_{ik}}{\partial x_i} = -\frac{x_k - x_i^0}{s_{ik}^0}$$

$$b = \frac{\partial \ell_{ik}}{\partial y_i} = -\frac{y_k - y_i^0}{s_{ik}^0}$$

$$c = \frac{\partial \ell_{ik}}{\partial z_i} = -\frac{z_k - z_i^0}{s_{ik}^0}$$

$$d = c_0$$

## GPS phase measurement

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$$s_{ik} = \lambda \cdot (\phi_{ik} + N_{ik})$$

$$\phi_{ik} - \varepsilon_{ik} = \frac{1}{\lambda}(s_{ik} + c_0 \delta t_{ik}) - N_{ik} = \frac{1}{\lambda} \sqrt{(x_k - x_i)^2 + (y_k - y_i)^2 + (z_k - z_i)^2} + \frac{c_0}{\lambda} \delta t_{ik} - N_{ik}$$

approximate coordinates  $(x_i^0, y_i^0, z_i^0)$  of the receiver  $P_i$

$$\phi_{ik}^0 = \frac{1}{\lambda} s_{ik}^0 = \frac{1}{\lambda} \sqrt{(x_k - x_i^0)^2 + (y_k - y_i^0)^2 + (z_k - z_i^0)^2}$$

$$(\phi_{ik} - \phi_{ik}^0) - \varepsilon_{ik} = a \delta x_i + b \delta y_i + c \delta z_i + d \delta t_{ik} + e N_{ik}$$

$$a = \frac{\partial \ell_{ik}}{\partial x_i} = -\frac{x_k - x_i^0}{s_{ik}^0} \frac{1}{\lambda}$$

$$b = \frac{\partial \ell_{ik}}{\partial y_i} = -\frac{y_k - y_i^0}{s_{ik}^0} \frac{1}{\lambda}$$

$$c = \frac{\partial \ell_{ik}}{\partial z_i} = -\frac{z_k - z_i^0}{s_{ik}^0} \frac{1}{\lambda}$$

$$d = \frac{1}{\lambda} c_0$$

$$e = -1$$