



Article Adjustment of an Integrated Geodetic Network Composed of GNSS Vectors and Classical Terrestrial Linear Pseudo-Observations

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Abstract: The paper proposes a new method for adjusting classical terrestrial observations (total station) together with satellite (GNSS-Global Navigation Satellite Systems) vectors. All the observations are adjusted in a single common three-dimensional system of reference. The proposed method does not require the observations to be projected onto an ellipsoid or converted between reference systems. The adjustment process follows the transformation of a classical geodetic network (distances and horizontal and vertical angles) into a spatial linear (distance) network. This step facilitates easy integration with GNSS vectors when results are numerically processed. The paper offers detailed formulas for calculating pseudo-observations (spatial distances) from input terrestrial observations (horizontal and vertical angles, horizontal distances, height of instrument and height of target). The next stage was to set observations equations and transform them into a linear form (functional adjustment model of geodetic observations). A method was provided as well for determining the mean errors of the pseudo-observations, necessary to assess the accuracy of the values following the adjustment (point coordinates). The proposed algorithm was verified in practice whereby an integrated network made up of a GNSS vector network and a classical linear-angular network was adjusted.

Keywords: GNSS vector network; classical terrestrial measurements; linear pseudo-observations; adjustment of observations; method of least squares

1. Introduction

Integrated measurement methods are usually employed for various surveying engineering jobs, such as monitoring land surface displacements or structure deformation [1-7]. Classical (terrestrial) surveying techniques are usually based on control networks referred to as a local (national) system of coordinates [8,9]. Survey results are usually processed by a simultaneous adjustment of classical observations (angles and distances) and GNSS vectors in a common mathematical space [5]. Integrated networks may be adjusted on the GRS'80 (Geodetic Reference System '80) reference ellipsoid surface or a horizontal projection plane in a local system. It is necessary to pre-process the observations in both cases. This process can include the projection of GNSS vectors onto an ellipsoid (calculating the length of a geodetic line and its original azimuth), projection of classical observations (horizontal distances) onto the surface of an ellipsoid (calculating the projection corrections), or transformation of GNSS vectors (ΔX , ΔY , ΔZ) onto a horizontal plane [10–13]. The determination of the height of the GNSS network points (e.g., calculation of ellipsoidal heights and their conversion into values referenced to the local model of geoid) is a separate computational stage [14–17]. All the pre-processing is rather labour-intensive and requires practical experience and knowledge [13,18,19]. What is more, one cannot avoid errors resulting from the transforming of the original observations into pseudo-observations on a common plane of reference (such as projection errors or errors of the geoid model).

Integrated networks are proposed to be used if satellite signal exposure is insufficient (in forests or difficult topography). It is then that classical observations can be used to improve the GNSS vector network. Detailed investigations into integrated geodetic networks can be found in many available publications [6,14,20–22]. For example, Kutoglu [8] showed a method for adjusting a GNSS network as a linear (trilateration) network. He proved that slope distances measured with any surveying method can be adjusted in any reference system (cf. [23]). Gargula [24] proposed an alternative adjustment method whereby both distances and angles between GNSS vectors are calculated in a geocentric spatial system (*XYZ*). A resulting set of linear-angular non-reduced pseudo-observations can be adjusted in reference to a local reference system.

Land-surveying measurements are adjusted because they are burdened with random errors (cf. [8,25]). The errors are treated like normally distributed random variables (Gauss distribution, Figure 1).

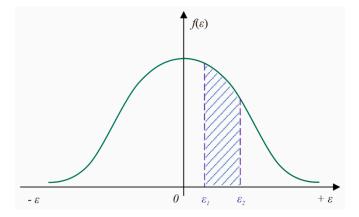


Figure 1. Normal distribution of measurement error (density function).

The error of measurement (ε) exhibits normal distribution if the error's density function $f(\varepsilon)$ can be expressed as:

$$f(\varepsilon) = \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{\varepsilon^2}{2}} = \frac{1}{\sqrt{2\pi}} \cdot \exp\left(-\frac{\varepsilon^2}{2}\right)$$
(1)

where: *e*—the base of a natural logarithm.

The probability that the measurement error ε (as a random variable) will take a value from the interval (ε_1 ; ε_2) can be expressed in the form (Figure 1):

$$P(\varepsilon \in \langle \varepsilon_1; \varepsilon_2 \rangle) = \frac{1}{\sqrt{2\pi}} \cdot \int_{\varepsilon_1}^{\varepsilon_2} \exp\left(\frac{-\varepsilon^2}{2}\right) d\varepsilon$$
(2)

where: $d\varepsilon$ —the differential of ε .

As the actual values of errors of measurement ε are unknown, they are replaced with observation corrections v, to be determined during adjustment. The goal of adjusting observations with the least-squares method is to select such correction values v that the sum of their squares multiplied by the weights (p) is the smallest.

The principal idea behind the adjustment method proposed in this paper involves the transformation of a classical geodetic network (distances and horizontal and vertical angles) into a spatial linear (distance) network independent of the local reference system. The objective is to adjust a classical network and a GNSS vector network together in a common, geocentric *XYZ* coordinate system (referenced to the GRS'80 ellipsoid).

2. Materials and Methods

2.1. Creating Linear Pseudo-Observations from the Classical Terrestrial Measurements

Mathematical equations needed for the task are developed using the principle of indirect levelling (Figure 2) used in the traditional topographic survey (total station). This

way one can calculate (using the Pythagorean theorem) the spatial (actual) distance between two points of a geodetic network (*j*-instrument station, *k*-target position):

$$d^{2} = \overline{d}^{2} + (i - s + h)^{2}$$
(3)

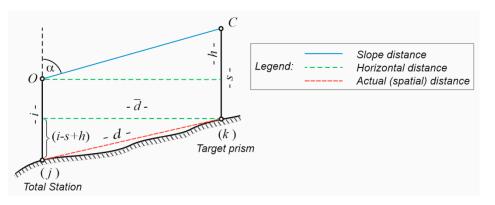


Figure 2. The principle of indirect levelling (*O*—tilt axis; *C*—target point; \overline{d} —horizontal distance; *d*—spatial distance; *i*—height of instrument; *s*—height of signal/target; *h*—the difference in elevation; α —vertical angle).

Next, the formulation to calculate the height difference h (see Figure 2) is substituted into Equation (3).

k

$$a = d \times \cot \alpha \tag{4}$$

This yields a general equation for the spatial distance between point *j* (station) and the measured point *k* as a function of the initial observations (\overline{d} , α , *i*, *s*):

$$d_{jk} = \sqrt{\overline{d}_{jk}^2 + (i_j - s_k)^2 + 2 \cdot (i_j - s_k) \cdot \overline{d}_{jk} \cdot \cot \alpha_{jk} + \overline{d}_{jk}^2 \cdot \cot^2 \alpha_{jk}}$$
(5)

Apart from this type of distance (5), the adjustment of the spatial network will require the length of the section between two points (*L*–left target and *R*–right target), measured from station *S* (Figure 3).

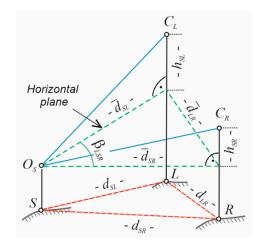


Figure 3. Determination of the horizontal distance \overline{d}_{LR} from total station measurement (*S*, *L*, *R*—points on the ground; *C*—target points; β —horizontal angle).

First, the horizontal distance \overline{d}_{LR} is determined with the law of cosines:

$$\overline{d}_{LR}^2 = \overline{d}_{SL}^2 + \overline{d}_{SR}^2 - 2 \cdot \overline{d}_{SL} \cdot \overline{d}_{SR} \cdot \cos\beta_{LSR}$$
(6)

The relationship between the horizontal distance \overline{d}_{LR} and the spatial distance d_{LR} is shown in Figure 4 and Equation (7).

$$d_{LR}^2 = \overline{d}_{LR}^2 + \left[(s_L - s_R) + (h_{SR} - h_{SL}) \right]^2$$
(7)

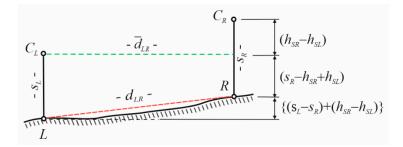


Figure 4. Determination of the height difference between two ground points L and R.

Substitution of Equation (6) to Equation (7) and employment of the general Equation for height difference (4) yields an equation for distance d_{LR} as a function of results of a classical field survey (horizontal distances \overline{d}_{SL} , \overline{d}_{SR} ; horizontal angle β_{LSR} ; vertical angles α_{SL} , α_{SR} and signal heights s_L , s_R):

$$d_{LR}^{2} = \overline{d}_{SL}^{2} + \overline{d}_{SR}^{2} - 2 \cdot \overline{d}_{SL} \cdot \overline{d}_{SR} \cdot \cos \beta_{LSR} + (s_{L} - s_{R})^{2} + 2 \cdot (s_{L} - s_{R}) \cdot \left(\overline{d}_{SR} \cdot \cot \alpha_{SR} - \overline{d}_{SL} \cdot \cot \alpha_{SL}\right) + \left(\overline{d}_{SR} \cdot \cot \alpha_{SR} - \overline{d}_{SL} \cdot \cot \alpha_{SL}\right)^{2}$$

$$(8)$$

The distances calculated with (5) and (8) will be considered linear pseudo-observations.

2.2. Stochastic Adjustment Model

Proper numerical processing of geodetic survey data involves the adjustment of observations (according to the method of least squares) and assessment of the accuracy of the results. To this end, it is necessary to transform the mean errors of the original observations (\overline{d} , α , β , I, s—see Figures 2 and 3) into mean errors of the pseudo-observations. To do this, one can employ the propagation of mean error [7,25].

The mean error of the pseudo-observations d_{jk} (5), which are distances between the instrument station *j* and signal *k* (target), is expressed as:

$$m_{jk}^{(d)} = \sqrt{\left(\frac{\partial d_{jk}}{\partial \overline{d}_{jk}}\right)^2 \cdot \left(m_{jk}^{(\overline{d})}\right)^2 + \left(\frac{\partial d_{jk}}{\partial \alpha_{jk}}\right)^2 \cdot \left(m_{jk}^{(\alpha)}\right)^2 + \left(\frac{\partial d_{jk}}{\partial i_j}\right)^2 \cdot \left(m_j^{(i)}\right)^2 + \left(\frac{\partial d_{jk}}{\partial s_k}\right)^2 \cdot \left(m_k^{(s)}\right)^2} \tag{9}$$

Partial derivatives (∂) of each variable (survey result) are determined as follows:

$$\frac{\partial d_{jk}}{\partial \overline{d}_{jk}} = \frac{\overline{d}_{jk} + (i_j - s_k) \cdot \cot \alpha_{jk} + \overline{d}_{jk} \cdot \cot^2 \alpha_{jk}}{d_{jk}}$$
$$\frac{\partial d_{jk}}{\partial \alpha_{jk}} = \frac{-\overline{d}_{jk}^2 \cdot \cot \alpha_{jk} - \overline{d}_{jk} \cdot (i_j - s_k)}{d_{jk} \cdot \sin^2 \alpha_{jk}}$$
$$\frac{\partial d_{jk}}{\partial i_j} = \frac{\overline{d}_{jk} \cdot \cot \alpha_{jk} + i_j - s_k}{d_{jk}}$$
$$\frac{\partial d_{jk}}{\partial s_k} = \frac{-\overline{d}_{jk} \cdot \cot \alpha_{jk} - (i_j - s_k)}{d_{jk}}$$

The mean error of the pseudo-observations d_{LR} (4), which are distances between the left (*L*) and right (*R*) target (Figure 3), is determined as shown below (also using the propagation of mean error):

$$m_{LR}^{(d)} = \sqrt{ \left(\frac{\partial d_{LR}}{\partial \overline{d}_{SL}} \right)^2 \cdot \left(m_{SL}^{(\overline{d})} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \overline{d}_{SR}} \right)^2 \cdot \left(m_{SR}^{(\overline{d})} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SL}} \right)^2 \cdot \left(m_{SR}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{SR}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{SR}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{SR}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 \cdot \left(m_{R}^{(\alpha)} \right)^2 + \left(\frac{\partial d_{LR}}{\partial \alpha_{SR}} \right)^2 + \left(\frac{\partial$$

Partial derivatives of each variable (survey results) are expressed with the following equations:

$$\frac{\partial d_{LR}}{\partial d_{SL}} = \frac{d_{SL} - d_{SR} \cos \beta_{LSR} - (s_L - s_R) \cot \alpha_{SL} - (d_{SR} \cot \alpha_{SR} - d_{SL} \cot \alpha_{SL}) \cot \alpha_{SL}}{d_{LR}};$$

$$\frac{\partial d_{LR}}{\partial d_{SR}} = \frac{\overline{d_{SL}} - \overline{d_{SL}} \cos \beta_{LSR} + (s_L - s_R) \cot \alpha_{SR} + (\overline{d_{SR}} \cot \alpha_{SR} - \overline{d_{SL}} \cot \alpha_{SL}) \cdot \cot \alpha_{SR}}{d_{LR}};$$

$$\frac{\partial d_{LR}}{\partial \alpha_{SL}} = \frac{(s_L - s_R) \cdot \overline{d}_{SL} - (\overline{d_{SR}} \cdot \cot \alpha_{SR} - \overline{d}_{SL} \cdot \cot \alpha_{SL}) \cdot \overline{d}_{SL}}{\sin^2 \alpha_{SL} \cdot d_{LR}};$$

$$\frac{\partial d_{LR}}{\partial \alpha_{SR}} = \frac{-(s_L - s_R) \cdot \overline{d}_{SR} - (\overline{d}_{SR} \cdot \cot \alpha_{SR} - \overline{d}_{SL} \cdot \cot \alpha_{SL}) \cdot \overline{d}_{SR}}{\sin^2 \alpha_{SR} \cdot d_{LR}};$$

$$\frac{\partial d_{LR}}{\partial \beta_{LR}} = \frac{\overline{d}_{SL} \cdot \overline{d}_{SR} \sin \beta_{LSR} \cos \beta_{LSR}}{d_{LR}};$$

$$\frac{\partial d_{LR}}{\partial \delta_{LR}} = \frac{s_L - s_R + \overline{d}_{SR} \cdot \cot \alpha_{SR} - \overline{d}_{SL} \cdot \cot \alpha_{SL}}{d_{LR}};$$

The formulas for partial derivatives appearing in Equations (9) and (10) were verified also for units.

The law of propagation of variance and covariance can be used to determine the mean errors instead of the propagation of mean error (e.g., [8,25]) because consecutive pseudo-observations can depend on the same angles and distances. Nevertheless, based on previous tests (cf. [24]), the actual effect of correlation of angles and distances on the values of calculated pseudo-observations is negligible.

The stochastic model for the integrated network is complemented with information on a priori mean errors $(m^{(\Delta x)}; m^{(\Delta y)}; m^{(\Delta z)})$ of components of the GNSS vector $(\Delta x_{jk}; \Delta y_{jk}; \Delta z_{jk})$, which can be obtained in post-processing [26].

2.3. Functional Adjustment Model

The creation of the functional model of the adjustment of a geodetic network involves the listing of observation equations and transforming them into linear equations of correction. The formulas below are general equations for three types of observations (in an integrated spatial geodetic network): (1) the station–target distance; (2) the left target–right target distance; (3) components of the GNSS vector.

(1) The station-target spatial distance (cf. Equation (5)):

$$d_{jk} + v_{jk}^{(d)} = \sqrt{\left(x_k - x_j\right)^2 + \left(y_k - y_j\right)^2 + \left(z_k - z_j\right)^2}$$
(11)

$$v_{jk}^{(d)} = \frac{\partial d_{jk}}{\partial x_j} \cdot \delta x_j + \frac{\partial d_{jk}}{\partial y_j} \cdot \delta y_j + \frac{\partial d_{jk}}{\partial z_j} \cdot \delta z_j + \frac{\partial d_{jk}}{\partial x_k} \cdot \delta x_k + \frac{\partial d_{jk}}{\partial y_k} \cdot \delta y_k + \frac{\partial d_{jk}}{\partial z_k} \cdot \delta z_k + l_{jk}^{(d)}$$
(12)

$$I_{jk}^{(d)} = d_{jk}^{(0)} - d_{jk}$$
(13)

where:

$$l_{jk}^{(d)}$$
—the absolute term in the correction Equation (12);
$$d_{jk}^{(0)} = \sqrt{\left(x_k^{(0)} - x_j^{(0)}\right)^2 + \left(y_k^{(0)} - y_j^{(0)}\right)^2 + \left(z_k^{(0)} - z_j^{(0)}\right)^2}$$
—the approximate distance;

$$\begin{cases} \frac{\partial d_{jk}}{\partial x_j} = -\frac{x_k^{(0)} - x_j^{(0)}}{d_{jk}^{(0)}}; & \frac{\partial d_{jk}}{\partial y_j} = -\frac{y_k^{(0)} - y_j^{(0)}}{d_{jk}^{(0)}}; & \frac{\partial d_{jk}}{\partial z_j} = -\frac{z_k^{(0)} - z_j^{(0)}}{d_{jk}^{(0)}} \\ \frac{\partial d_{jk}}{\partial x_k} = \frac{x_k^{(0)} - x_j^{(0)}}{d_{jk}^{(0)}}; & \frac{\partial d_{jk}}{\partial y_k} = \frac{y_k^{(0)} - y_j^{(0)}}{d_{jk}^{(0)}}; & \frac{\partial d_{jk}}{\partial z_k} = \frac{z_k^{(0)} - z_j^{(0)}}{d_{jk}^{(0)}} & -\text{partial derivatives}; \end{cases}$$

 $x^{(0)} y^{(0)} z^{(0)}$ —approximate coordinates.

(2) The left target-right target spatial distance (cf. Equation (8)):

$$d_{LR} + v_{LR}^{(d)} = \sqrt{(x_R - x_L)^2 + (y_R - y_L)^2 + (z_R - z_L)^2}$$
(14)

$$v_{LR}^{(d)} = \frac{\partial d_{LR}}{\partial x_L} \cdot \delta x_L + \frac{\partial d_{LR}}{\partial y_L} \cdot \delta y_L + \frac{\partial d_{LR}}{\partial z_L} \cdot \delta z_L + \frac{\partial d_{LR}}{\partial x_R} \cdot \delta x_R + \frac{\partial d_{LR}}{\partial y_R} \cdot \delta y_R + \frac{\partial d_{LR}}{\partial z_R} \cdot \delta z_R + l_{LR}^{(d)}$$
(15)

$$l_{LR}^{(d)} = d_{LR}^{(0)} - d_{LR} \tag{16}$$

where:

$$l_{LR}^{(d)} - \text{the absolute term in the correction Equation (15);} \\ d_{LR}^{(0)} = \sqrt{\left(x_R^{(0)} - x_L^{(0)}\right)^2 + \left(y_R^{(0)} - y_L^{(0)}\right)^2 + \left(z_R^{(0)} - z_L^{(0)}\right)^2} - \text{the approximate distance.}$$

The partial derivatives in Equation (15) are calculated similarly as for Equation (12). (3) The GNSS vector (Δx , Δy , Δz) between two points *j* and *k*:

$$\begin{cases} \Delta x_{jk} + v_{jk}^{(\Delta x)} = x_k - x_j \\ \Delta y_{jk} + v_{jk}^{(\Delta y)} = y_k - y_j \\ \Delta z_{jk} + v_{jk}^{(\Delta z)} = z_k - z_j \end{cases}$$
(17)

$$\begin{cases} v_{jk}^{(\Delta x)} = \frac{\partial (\Delta x)_{jk}}{\partial x_k} \cdot \delta x_k - \frac{\partial (\Delta x)_{jk}}{\partial x_j} \cdot \delta x_j + l_{jk}^{(\Delta x)} = \delta x_k - \delta x_j + l_{jk}^{(\Delta x)} \\ v_{jk}^{(\Delta y)} = \frac{\partial (\Delta y)_{jk}}{\partial y_k} \cdot \delta y_k - \frac{\partial (\Delta y)_{jk}}{\partial x_j} \cdot \delta y_j + l_{jk}^{(\Delta y)} = \delta y_k - \delta y_j + l_{jk}^{(\Delta y)} \\ v_{jk}^{(\Delta z)} = \frac{\partial (\Delta z)_{jk}}{\partial z_k} \cdot \delta z_k - \frac{\partial (\Delta z)_{jk}}{\partial z_j} \cdot \delta z_j + l_{jk}^{(\Delta z)} = \delta z_k - \delta z_j + l_{jk}^{(\Delta z)} \\ \begin{cases} l_{jk}^{(\Delta x)} = \Delta x_{jk}^{(0)} - \Delta x_{jk} \\ l_{jk}^{(\Delta y)} = \Delta y_{jk}^{(0)} - \Delta y_{jk} \\ l_{jk}^{(\Delta z)} = \Delta z_{jk}^{(0)} - \Delta z_{jk} \end{cases} \end{cases}$$
(18)

where:

 $l^{(\Delta x)}$, $l^{(\Delta y)}$, $l^{(\Delta z)}$ —absolute terms in the correction Equation (18);

 $v^{(\Delta x)}$, $v^{(\Delta y)}$, $v^{(\Delta z)}$ —corrections for the GNSS vector components;

 δx , δy , δz —the increments (corrections) to be determined for approximate coordinates; $\Delta x^{(0)}$, $\Delta y^{(0)}$, $\Delta z^{(0)}$ —approximate values of the GNSS vector calculated as:

$$\begin{cases} \Delta x_{jk}^{(0)} = x_k^{(0)} - x_j^{(0)} \\ \Delta y_{jk}^{(0)} = y_k^{(0)} - y_j^{(0)} \\ \Delta z_{jk}^{(0)} = z_k^{(0)} - z_j^{(0)} \end{cases}$$
(20)

The partial derivatives in Equation (18) assume values 1 or -1 because the observations in Equation (17) are linear (partial derivatives of linear equations calculated for unknowns are equal to the coefficients at these unknowns).

2.4. The Procedure for Adjusting the Integrated Network

The adjustment of the integrated geodetic network (using the method of least squares) will be based on an overdetermined system of equations of correction type (12), (15) and (18), expressed as the following matrix form:

$$\mathbf{V} = \mathbf{A} \cdot \mathbf{X} - \mathbf{L} \tag{21}$$

where:

 $\mathbf{V} = \left[\left\{ v_{jk}^{(d)}; v_{LR}^{(d)}; \left(v_{jk}^{(\Delta x)}; v_{jk}^{(\Delta y)}; v_{jk}^{(\Delta z)} \right) \right\} \right]^{\mathrm{T}}$ —the vector of corrections type (12), (15) and (18) to be determined (curly brackets { . . . } stand for all elements of a type);

A—the matrix of coefficients of the unknowns (partial derivatives) in Equations (12), (15) and (18); **X** = $[\{(\delta x_L; \delta y_L; \delta z_L; \delta x_R; \delta y_R; \delta z_R); (\delta x_j; \delta y_j; \delta z_j; \delta x_k; \delta y_k; \delta z_k)\}]^T$ —the vector of the unknowns—increments to approximate coordinates;

$$\mathbf{L} = \left[\left\{ l_{jk}^{(d)}; l_{LR}^{(d)}; \left(l_{jk}^{(\Delta x)}; l_{jk}^{(\Delta y)}; l_{jk}^{(\Delta z)} \right) \right\} \right]^{1}$$
 the vector of absolute terms type (13), (16) and (19).

The estimated vector of the unknowns $\hat{\mathbf{X}}$ is calculated with the method known from the adjustment calculus [13], which stems from the imposition of the least square condition $(\mathbf{V}^{T} \cdot \mathbf{P} \cdot \mathbf{V} = \min.)$ on the system (21):

$$\hat{\mathbf{X}} = \left(\mathbf{A}^{\mathrm{T}} \cdot \mathbf{P} \cdot \mathbf{A}\right)^{-1} \mathbf{A}^{\mathrm{T}} \cdot \mathbf{P} \cdot \mathbf{L}$$
(22)

where:

P—the matrix of weights set up from mean errors of the linear pseudo-observations (9), (10) and mean errors of GNSS vector measurements $(m^{(\Delta x)}; m^{(\Delta y)}; m^{(\Delta z)})$:

$$diag\{\mathbf{P}\} = \left\{ \frac{1}{\left(m_{jk}^{(d)}\right)^{2}}; \ \frac{1}{\left(m_{LR}^{(d)}\right)^{2}}; \ \left(\frac{1}{\left(m_{jk}^{(\Delta x)}\right)^{2}}; \ \frac{1}{\left(m_{jk}^{(\Delta y)}\right)^{2}}; \ \frac{1}{\left(m_{jk}^{(\Delta z)}\right)^{2}}\right) \right\}$$
(23)

The next step is to substitute the vector of unknowns X (21) with the calculated vector \hat{X} (22) and calculate the vector of observation corrections V, which are used to adjust the observations—the left sides of the observation Equations (11), (14) and (17).

Information on mean errors of the adjusted coordinates (m_x , m_y , m_z) can be found on the diagonal of the covariance matrix (Q_x) of the vector **X**:

$$\mathbf{Q}_{\mathbf{x}} = m_0 \cdot \left(\mathbf{A}^{\mathbf{T}} \cdot \mathbf{P} \cdot \mathbf{A}\right)^{-1}$$
(24)

$$m_0 = \sqrt{\frac{\mathbf{V}^{\mathbf{T}} \cdot \mathbf{P} \cdot \mathbf{V}}{r}}$$
(25)

where:

 m_0 —the standard error of unit weight;

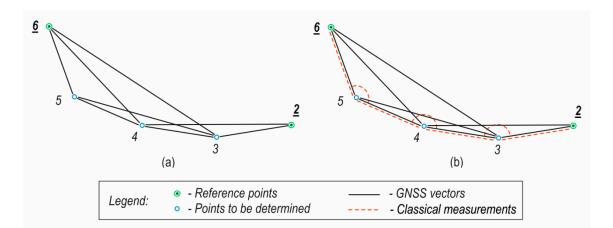
r—the number of redundant observations.

Next, a single parameter characterising the point's accuracy is calculated for each point—the error of position (m_P) in a three-dimensional Cartesian system:

$$m_P = \sqrt{m_x^2 + m_y^2 + m_z^2} \tag{26}$$

3. Results and Discussion (Numerical Example)

The proposed method for adjusting an integrated network was verified with a simple practical example (Figure 5b). Calculations were performed for a GNSS vector network (without classical linear-angular observations, Figure 5a) as well, to compare the results.



The test calculations involved actual survey results for a section of a control network for monitoring ground displacement in an active mining area.



Vectors between vertices of the network were measured with a static GNSS method (two receivers, survey duration 60 min). GNSS vectors were used in both test variants: GNSS network (Figure 5a) and integrated network (Figure 5b).

Classical terrestrial measurements (horizontal and vertical angles, horizontal distances) were performed with a precise total station. The surveying instrument and the signal (reflector) were placed on carefully levelled and centred tripods with tribrachs. The height of both the instrument (*i*) and the signal (*s*) was measured repeatedly with a special device. The classical observations were used in the integrated network (Figure 5b).

For calculation purposes, points 2 and 6 were assumed to be fixed (reference points), while the other points (3, 4 and 5) were to be determined—in both cases (Figure 5a,b).

Tables 1 and 2 show the input data necessary to adjust the vector network (Figure 5a). Values of components of the GNSS vectors (ΔX , ΔY and ΔZ) and their mean errors (Table 1) were obtained from post-processing of GNSS data. Approximate coordinates of the points to be determined (Table 2) were calculated from measured (non-adjusted) vectors. For the purpose of the comparative analyses, coordinates and distances were recorded down to 0.0001 m.

Vector **Observations (Components of GNSS Vectors) Mean Observation Error** Labels [m][m] From То ΔX ΔΥ ΔZ $m_{\Delta X}$ $m_{\Delta Y}$ $m_{\Delta Z}$ 2 3 9.7354 -22.9314-1.60570.0019 0.0016 0.0020 2 4 16.9362 -46.7425-0.69960.0018 0.0019 0.0016 3 4 7.2020 -23.81020.9088 0.0021 0.0016 0.0016 5 3 -8.792447.6362 -6.09450.0038 0.0029 0.0028 5 4 -1.589823.8237 -5.18550.0033 0.0026 0.0026 6 3 5.3467 61.6613 -20.59540.0024 0.0018 0.00186 4 12.5497 37.8504 -19.68650.00240.0019 0.0019 6 5 14.0259 -14.50220.0029 0.0031 14.1397 0.0026

Table 1. Measured GNSS vectors and their mean errors.

Point	<i>X</i> [m]	Y [m]	Z [m]
2	3,871,857.1432	1,345,974.9571	4,870,463.1848
3 *	3,871,866.8786	1,345,952.0257	4,870,461.5791
4 *	3,871,874.0806	1,345,928.2155	4,870,462.4879
5 *	3,871,875.6704	1,345,904.3918	4,870,467.6734
6	3,871,861.5368	1,345,890.3711	4,870,482.1739

Table 2. Coordinates of reference points (2 and 6) and approximate coordinates (*) of points to be determined (*3*, *4*, *5*)—geocentric system ETRF'89.

Linear pseudo-observations, that is, actual spatial distances, and their mean errors were calculated from the classical observations (Tables 3 and 4). Note the values of errors $m_{jk}^{(d)}$. They are identical to the mean error of the measured horizontal distance (Table 3) due to small differences between horizontal distances \overline{d} (Table 3) and spatial distances d_{jk} (Table 4).

Table 3. Classical measurements (total station) and their mean errors.

Station (j)	Target (k)	Height of the Instrument <i>i</i> [m]	Height of the Signal <i>s</i> [m]	Horizontal Angle β [Grad]	Vertical Angle α [Grad]	Horizontal Distance \overline{d} [m]
5	6	1.733	1.882	144.35765	100.63750	24.6360
	4		1.858		99.48384	24.4400
4	5	1.858	1.733	181.80672	100.51717	24.4434
	3		1.821		100.20077	24.8923
3	4	1.821	1.858	190.63125	99.80366	24.8924
	2		1.630		100.07838	24.9649
A priori m	ean errors	0.002	0.002	0.0030	0.0020	0.004

Table 4. Linear pseudo-observations of types j-k (5) and L-R (8) and their mean errors (9) (10).

Edge	j-k	<i>d_{ik}</i> [m]	$m_{ik}^{(d)}$ [m]	Edge	L-R	\overline{d}_{LR} [m]	<i>d</i> _{<i>LR</i>} [m]	$m_{LR}^{(d)}$ [m]
From (j)	To (<i>k</i>)	Equation (5)	Equation (9)	From (L)	To (<i>R</i>)	Equation (6)	Equation (8)	Equation (10)
5	6	24.6374	0.0040	6	4	44.4639	44.4663	0.0051
5	4	24.4412	0.0040	5	3	48.8329	48.8329	0.0056
4	5	24.4444	0.0040	4	2	49.7224	49.7225	0.0056
4	3	24.8924	0.0040					
3	4	24.8925	0.0040					
3	2	24.9656	0.0040					

Table 5 shows part of a table with a matrix of coefficients **A** (20), built as the integrated network is being adjusted. The remaining part of matrix **A** is filled with coefficients from equations of classical GNSS vectors (18). The absolute terms (the last table column) in the equations of corrections (12) and (15) are necessary to create the matrix **L** (20).

Type	Edge			Point No. 3			Point No. 4			Point No. 5		
of Pseudo- Observation	From	То	X	Ŷ	Ζ	X	Ŷ	Ζ	X	Ŷ	Ζ	- Term <i>l</i> [m]
	5 5	6 4	0 0	0 0	0 0	0 -0.065	0 0.975	0 -0.212	0.574 0.065	$0.569 \\ -0.975$	-0.589 0.212	$-0.0081 \\ -0.0080$
j-k	$4 \\ 4$	5 3	0 -0.289	0 0.957	0 -0.037	$-0.065 \\ 0.289$	$0.975 \\ -0.957$	-0.212 0.037	0.065 0	$-0.975 \\ 0$	0.212 0	$-0.0111 \\ -0.0002$
	3 3	4 2	$-0.289 \\ 0.390$	$0.957 \\ -0.919$	$-0.037 \\ -0.064$	0.289 0	-0.957 0	0.037 0	0 0	0 0	0 0	$-0.0004 \\ -0.0015$
L-R	6 5 4	4 3 2	$0 \\ -0.180 \\ 0$	0 0.976 0	$\begin{array}{c} 0\\ -0.125\\ 0\end{array}$	0.282 0 0.341	$0.851 \\ 0 \\ -0.940$	$-0.443 \\ 0 \\ -0.014$	0 0.180 0	$0 \\ -0.976 \\ 0$	0 0.125 0	$-0.0019 \\ -0.0126 \\ -0.0019$

Table 5. Coefficients at the unknowns (partial derivatives) and the absolute terms (13), (16) in the equations of corrections for the pseudo-observations (12), (15).

The final results of the calculations made using Equations (22)–(26) are summarized in Tables 6 and 7 (for the two test variants, respectively).

Table 6. Adjusted coordinates and mean errors (GNSS vector network).

Point	Coordinate	es of Points to Be I [m]	Determined	Mean I	Error of Coor [m]	Error of Position [m]	
	X	Ŷ	Ζ	m_X	m_Y	m_Z	m _P
3	3,871,866.8806	1,345,952.0287	4,87,0461.5783	0.0017	0.0014	0.0015	0.0026
4	3,871,874.0824	1,345,928.2179	4,870,462.4867	0.0016	0.0013	0.0015	0.0026
5	3,871,875.6742	1,345,904.3947	4,870,467.6723	0.0027	0.0022	0.0024	0.0042

Table 7. Adjusted coordinates and	mean errors (integrated network).
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Point	Coordinate	es of Points to Be I [m]	Determined	Mean I	Error of Coor [m]	Error of Position [m]	
	X	Ŷ	Ζ	m_X	m_Y	m_Z	m_P
3	3,871,866.8807	1,345,952.0287	4,870,461.5782	0.0016	0.0013	0.0014	0.0025
4	3,871,874.0825	1,345,928.2182	4,870,462.4865	0.0016	0.0012	0.0014	0.0025
5	3,871,875.6753	1,345,904.3924	4,870,467.6723	0.0025	0.0019	0.0023	0.0039

A comparison of results of the adjustment of the GNSS vector network and the integrated network (Table 8) reveals only minor differences in the coordinates *X*, *Y*, *Z* (up to about 5 mm). The difference in the point location is illustrated by the resultant linear discrepancy (δ_{XYZ}), which assumes values from about 3 mm to about 5 mm. The values demonstrate the impact of the classical measurements on the adjustment of the GNSS vector network (this order of differences in coordinates can be significant when the absolute ground displacements are measured, for example). Note also the mean errors of the coordinates (m_X , m_Y , m_Z) and the error of position (m_P). They are almost identical for both variants. This might be indicative of the correctness of equations of the pseudo-observation mean errors (9) and (10), always assuming that both the procedures employed similar observation weighting principles.

Point	Differences in Coordinates [mm]			Resultant Linear Discrepancy [mm]	Differ	ences in Mean [mm]	Differences in Error of Position [mm]	
_	X	Y	Ζ	δ_{XYZ}	m_X	m_Y	mZ	m _P
3	-2.1	-3.0	0.9	3.8	0.0	0.1	0.0	0.1
4	-1.9	-2.7	1.4	3.6	0.0	0.1	0.0	0.1
5	-4.9	-0.6	1.1	5.0	0.2	0.3	0.1	0.3

Table 8. Comparison of results of adjustments of the GNSS network and the integrated network.

For ground displacement monitoring, the relative positions of points (so-called relative displacement) are measured as well. Therefore, the adjusted coordinates (Tables 6 and 7) were used to calculate spatial distances between the points (Table 9). Despite differences of several millimetres in point coordinates (Table 8), the computed values are practically identical for both variants (the differences are below 1 mm in most cases). This fact may indicate cumulative GNSS survey errors (caused by antenna phase centre variations, for example).

Table 9. Comparison of spatial distances following the adjustment.

Sid	le	GNSS Vector Network [m]	Integrated Network [m]	Difference 1 [mm]	Classical Measurement [mm]	Difference 2 [mm]	Difference 3 [mm]
From	То	(<i>d</i> _{<i>G</i>})	(d _{IN})	$(d_G - d_{IN})$	(<i>d</i> _{CL})	$(d_G - d_{CL})$	$(d_{IN}-d_{CL})$
2	3	24.9621	24.9623	-0.1	24.9650	-2.9	-2.8
3	4	24.8927	24.8924	0.3	24.8924	0.3	-0.1
4	5	24.4329	24.4356	-2.7	24.4419	-9.0	-6.3
5	6	24.6339	24.6331	0.8	24.6397	-5.9	-6.7

The calculated distances (Table 9) were juxtaposed with reference distances (d_{CL}), which were obtained from the additional precise (repeated several times) classical surveys: the horizontal distances were measured with a precise total station, and the height differences were measured using the precise geometric levelling method. Allowing for the hypothetical assumption that spatial distances d_{CL} are free of errors, one can note a positive impact of the additional classical measurements (Difference 3) on most distances measured with GNSS (Difference 2).

4. Summary and Conclusions

The paper proposes a new method for adjusting an integrated network made up of GNSS vectors and classical terrestrial observations. The first computing stage involves a list of linear pseudo-observations that are original linear-angular observations converted into spatial distances. The next step is to transform (a priori) the mean errors of the classical measurements into mean errors of the pseudo-observations (the stochastic model). The functional model of the adjustment is made up of a set of observation equations for the GNSS vectors and for the new pseudo-observations (expressed as a function of the original linear-angular measurements). The objective of the pre-processing is the concurrent adjustment of the pseudo-observations and GNSS vectors in a common mathematical space *XYZ*.

The second part of the paper presented a practical application of the method to adjust an integrated geodetic network. The results (coordinates of points, their mean errors and spatial distances between the points) were juxtaposed with results of the adjustment of the GNSS vector network.

The comparative analysis demonstrated that the new method for adjusting integrated networks yields similar results (coordinates) as an adjustment of a vector network. It has

further been demonstrated that an integrated network provides more accurate (close to real) spatial distances between points (in relation to reference distances obtained from precise classical surveys).

The primary advantage of the proposed adjustment method is the ease of integration of GNSS vectors with linear pseudo-observations obtained from classical measurements. The preparation of pseudo-observations is much easier than for other available methods for adjusting integrated networks (where it is necessary to determine the lengths of geodetic lines and their azimuths on the ellipsoid, project classical observations onto the ellipsoid, or convert ellipsoidal heights into orthometric values, etc.). The proposed method can be employed to adjust periodic measurements of control networks for ground displacement monitoring. However, the use of this calculation method is limited to short lines. If distances between the points are longer than 200–300 m, the effect of refraction and the curvature of the earth should be considered [27].

The research reported here will be continued. A detailed computing algorithm for the method and its implementation as a computer application is planned. Furthermore, an attempt will be made to test the new adjustment method on a network with much longer GNSS vectors and pseudo-observations (about a few hundred metres).

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