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REALIZATION OF AUTOMATIC ERROR DETECTION IN THE BLOCK ADJUSTMENT PROGRAM PAT-M43 USING ROBUST ESTIMATORS

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Abstract: The detection of outliers can be automated using robust estimators. The principle is to interpret the residuals v_i of the observations after each iteration as errors in order to calculate new weights based on a weight function $p(v_i)$. The new weights $p_i = p(v_i)$ are then used in the following iteration step.

The paper reports on the realization of this error detection strategy in PAT-M43. Main topic is the extension of the method, especially the choice of a proper weight function, the iteration sequence and the stopping rule. The significant facilitation in handling the program is explained.

1. The original program:

The computer program PAT-M43 performs a blockadjustment by independent photogrammetric models. This approach implies a spatial similarity transformation for each model. The adjustment is based on a least squares solution. The nonlinear observational equations are linearized with respect to the orientation parameters. Because of computational economy the program iterates sequential horizontal and vertical adjustments, applying 4-parameter and 3-parameter transformations, respectively. For each iteration the partially reduced normal equations that contain only the unknown orientation parameters are formed directly from the model and control coordinates and are solved by a modified Cholesky method (Ackermann et. al., 1970). An extension allows the combined adjustment of photogrammetric models with APR and/or statoscope data, including photogrammetric height measurements of shorelines of lakes (Ackermann et. al. 1972).

2. Manual data cleaning:

One of the main problems of practical blockadjustment is the detection and location of outliers. Dependent on the number and distribution of the observations, errors show up only partly in the residuals of the corresponding observations, the other parts falsify the absolute orientation of the photogrammetric models (Förstner, 1978). The mutual interference of outliers, especially of different size, is a further handicap. For that reason several adjustments for a step by step location and elimination of outliers in accordance with the size of the errors, and some further adjustments in order to avoid wrong decisions are necessary. Nevertheless the quality of manual data cleaning is sufficiently good and comparable with most of the more sophisticated procedures (Förstner, 1982), but in general it requires a great deal of time by fully qualified persons. Thus the main argument of the development of an automatic procedure has been: to shorten the processing time needed by persons in charge of blockadjustment.

3. From least squares to robust adjustment:

The above mentioned problems which arise in the adjustment of data with gross errors are not a specific attribute of the manual data cleaning procedure, but a bad point of the method of least squares. Applying a constant weight p = const for each observation the influence function (first derivative of the minimum function by the residual) shows, that the influence of a defective observation onto the result of the adjustment is directly proportional to the size of the error. Thus as a matter of fact the method of least squares is applicable for errorfree data only and unsuited for automatic error detection procedures.

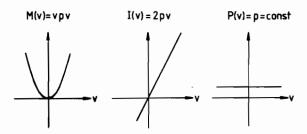


Fig. 1. Least squares: minimum function, influence function and weight function

Logically erroneous observations have to be handled with reduced weights and can not be treated with the same weights as errorfree data. All the observations must be introduced into the adjustment with weights chosen in correspondence with their errors. The problem of locating gross errors is therefore identical with the determination of proper weights for the observations. An alternative to least squares is the minimum norm method (Huber, 1981). Thereby the weights of the observations are progressively determined in an iterative process. After each iteration step new weights for the observations are calculated as a function of the residuals with $P(v) = \frac{1}{|v|}$. The influence function shows, that after convergency of the procedure the influence of all the |v| observations onto the result is equal. Observations with gross errors have the same influence onto the result as errorfree data. This is better than with least squares but still not sufficient.

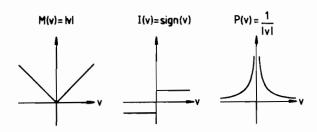


Fig. 2. Minimum Norm: minimum function, influence function and weight function

An adjustment procedure which uses weight functions for complete elimination of the influence of gross errors is the so-called method of robust estimators (robust against the influence of gross errors) (Krarup, 1980; Kubik, 1984). After convergency of the iterative process proper weights are determined for all observations and erroneous data will get weights approximately equal to zero and thus will have no influence at all onto the result of the adjustment. Their residuals will show the true errors. The method of robust estimators can be interpreted as an a posteriori estimation of the variances. Many simple weight functions can be found which meet the conditions of robust estimators, but most of them cover only a small range of gross errors and will fail with the variety of gross errors occuring in practical cases. The reason for the failure in these cases is the assumption of linearity by the robust estimators (Huber, 1981).

4. Weight function for PAT-M43:

Thus, a lot of research was necessary to find a weight function and to develop a procedure which covers the wide range of gross errors, their multiple combinations and the varying geometry of photogrammetric blocks (Werner, 1984). Because of their effectiveness the following hyperbolic weight function was chosen for the blockadjustment program PAT-M43:

$$P = P_{i} \cdot F \left(v_{i}, \sigma_{v_{i}}, Q\right)$$

$$= P_{i} \cdot \frac{1}{1 + \left(\alpha_{i} \cdot |v_{i}|\right)^{d}}$$
(1)

in which:

$$\alpha_{\mathbf{i}} = \frac{1}{1.4 \cdot \hat{\sigma}_{\mathbf{v}_{\mathbf{i}}}} = \frac{\sqrt{P_{\mathbf{i}}}}{1.4 \cdot \sqrt{r_{\mathbf{i}}} \cdot \hat{\sigma}_{\mathbf{0}}}$$
 (2)

$$d = 3.5 + \frac{82}{81 + Q^4}$$
 (3)

$$Q = \frac{\hat{\sigma}_0}{\sigma_{\text{a priori}}} \tag{4}$$

v_i= residual of observation i

 P_i = a priori weight of observation i

 r_i = local redundancy of observation i

 $\hat{\sigma}_{v_i}$ = estimated sigma of the residual v_i

 $\hat{\sigma}_0$ = estimated sigma-naught

worth mentioning are two attributes of the weight function expanding the range of gross errors locatable with this function.

The first is the dependence on Q (see formula 3 and 4). At the start of the iteration process the value of Q is relatively big and it will become smaller with convergency. At the end of the procedure Q will reach approximately the value one. Thus the curve of the weight function is flat in the beginning and will become steeper and steeper with the disappearing influence of the gross errors and the final orientation of the models. This attribute of the weight function allows the correction of wrong decisions caused by false O-approximations of the residuals at the beginning and makes it easier to distinguish between errorfree and erroneous observations at the end of the iteration process.

The second attribute is the dependence on the estimated standard deviation of the particular residual $\hat{\sigma}v_i$ (see formula 2). Even with the simplification of using the value one as local redundancy for all the observations this feature allows the determination of small gross errors in the critical range of localization.

Without any further modifications the localization of locatable gross errors up to $50 \cdot \sigma_0$ causes no problems, even with geometrically very weak configurations, as long as there are still error-free redundant observations.

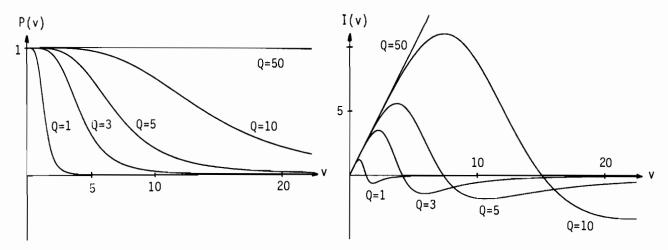


Fig. 3. PAT-M43: weight function Fig. 4. PAT-M43: influence function with $P_i=1; r_i=1; \sigma_{a priori}=1$ and Q=1;3;5;10;50

5. New structure of the program:

If no initial values of the orientation parameters are available the program begins with a least squares horizontal adjustment that does not require approximate values. The resulting transformed model coordinates enter into a vertical least squares adjustment using a shift in z only. Thus big gross errors in height do not disturb the orientation of the models too much. After these first two iteration steps initial O-approximations for the residuals are calculated, needed to start consecutive robust iteration steps. Robust estimators can relatively easy be realized using the least squares algorithm and modifying the weights after each iteration step by means of the weight function. Robust iteration steps are repeated until sufficient convergence is reached. The convergence is quite good but highly correlated to the number and the size of gross errors and to the geometric stability of the block configuration. Thus the number of iteration steps may differ from 6 to 20. If the change of $\hat{\sigma}_0^2$ between two iteration steps becomes less than $2 \cdot \hat{\sigma}(\hat{\sigma}_0^2)$ in planimetry or height after a corresponding iteration step the final elimination of erroneous observations is performed. All observations being used in that iteration step and getting $F(v_i, \hat{\sigma}_{v_i}, Q) < 0.01$ (see formula 1) will be marked as erroneous observations and will get an infinitely small weight. The others receive their original a priori weight. Some least squares iteration steps complete the procedure to reach the final result.

Treating errorfree data in a least squares adjustment the favourable sequence of iteration steps is a consecutive alteration between planimetry and height. Handling erroneous data the succession of iteration steps depends on the existing gross errors. That means the succession is data dependent and therefore must be directed by the program itself. The robust iteration steps always begin with horizontal adjustments in order to reduce the influence of the very big gross errors in planimetry because big gross errors in the planimetric coordinates would disturb the levelling of the models completely. Due to the same effect the first robust iteration step in height applies only a shift in z. The sequence of all further iteration steps is chosen properly in order to keep the reduced influence of gross errors in planimetry and height approximately on the same level because of the mutual interference of erroneous data.

6. Classification of gross errors:

Regarding the different effects of gross errors related to their size we can group them into 3 different classes:

- 1. small gross errors
- medium-size gross errors.
- 3. large gross errors

The classification bounds are not fixed, they depend on the geometry and may vary for different photogrammetric blocks.

All gross errors greater than $4 \cdot \sigma$ and less than $50 \cdot \sigma$ can be designated as small gross errors. They have no significant influence onto the orientation of the models and do not disturb the domain of linearity of the adjustment. Gross errors of the stochastical model and systematic errors are not taken into account but can be considered as small gross errors. Errors less than $4 \cdot \sigma$ are integrated with the random errors.

All errors between $50 \cdot \sigma$ and 2-3 base lengths belong to the medium-size gross errors. They have no big influence onto the geometry of the photogrammetric block and don't disturb the convergence of the adjustment but they are not within the range of the linearization and the solution may tend to a different 0-point. Errors bigger than 3 base lengths are named large gross errors. They change the geometry of the block severely and cause poor convergence or even divergence. Especially for blocks with bad geometry the adjustment must be stopped before reaching the point of convergence.

7. Location of small gross errors:

The location of small gross errors poses no problems for the robust adjustment with the chosen weight function. Even small gross errors at the limit of possible location are detected as long as the observations are sufficiently well distributed within the models.

Only for really bad distributions the consideration of the local redundancy (see formula 2) would improve the effectiveness of the procedure. The check for the inherent limit of localization can be performed only with artificial data. Example 1 shows that the introduced errors greater than the lower limit of $5\,\sigma$ are located without any wrong decision. This lower limit is even better than the theoretical expectation for the statistical test (R. Schroth, 1980).

Example 2 shows a practical photogrammetric block and is demonstrating the effectiveness of the robust estimators. At first data cleaning has been performed manually and the cleaned data have been submitted to the automatic procedure. Although the residuals after the manual procedure did not indicate remaining gross errors, the automatic procedure located further ones.

8. Modifications of the procedure with respect to medium-size and large gross errors:

Medium-size gross errors and all larger gross errors do not belong to any normal distribution of observations, they are independent from the a priori weights introduced into the adjustment. Thus as long as large gross errors have still an influence onto the adjustment all photogrammetric observations are treated with the starting weight 1, used as a priori weight in the weight function. This starting weight tends to the introduced specific a priori weight in dependency on the value of Q by a weight function. The same is true for all non-photogrammetric observations, but for them the starting weight 1/100 is used. The weight function is as follows:

Weight function for modified a priori weights:

$$P = SW + (P_i - SW) \cdot \frac{37}{36 + (0-1)^2}$$

in which:

SW = starting weight

 $P_i = a$ priori weight

P = modified a priori weight

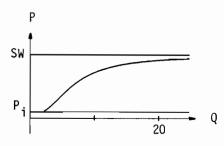


Fig. 5. Modification of a priori weights: with SW=1; P_i =0.1

The ratio of the two starting weights has the effect to reduce the influence of gross errors of photogrammetric observations always a little earlier than for control. This supports the location of gross errors at control points in case of weak control point distributions.

As long as there is no bad accumulation of medium-size gross errors in relation to the geometry their location is no problem. But there are two effects to be avoided:

The bigger the gross errors the more falsified are the O-approximations of the residuals. Sometimes this results in so-called "swimming" models. By means of false O-approximations the weights for all the observations of a model will be reduced too much in spite of a flat weight function and the model will not be able to get oriented. Nevertheless the calculated weights point to the biggest gross errors.

An other effect is, that after location of the medium-size gross errors the adjustment approaches a different 0-point and the location of the small gross errors will not be correct. The pre-elimination of large and medium-size gross errors will solve this problem. As soon as the value of F $(v_i, \hat{\sigma}_{v_i}, Q)$ reaches a certain lower limit the corresponding observation will get the minimum weight for elimination, all other observations receive their a priori weights to start a new robust adjustment. The lower limit of the weight function for preelimination starts with 10^{-18} and is increased for each iteration step by a factor 10 up to the value 10^{-9} . This modification results in a step by step elimination of all large errors down to gross errors of approximately $50 \cdot \sigma$. Thus "swimming" models will be reincluded into the block and linearity for the final elimination of small gross errors in provided.

Large gross errors may disturb the geometry of the block completely. Already the O-approximations of the residuals after the starting least squares iteration are false to the extent that the point of convergence will not be reached. Therefore large gross errors have to be introduced with already reduced weights into the starting least squares iteration step. The problem can be solved by calculating a center point for each model and the distances to this point for all observations. The ratio of distance and mean distance is used in order to reduce weights by a weight function.

The coordinates of the center point are the arithmetic mean of the coordinates of observations, as long as there are more than 5 observations, otherwise the median is used. The same is true for the mean distance, but the median is used already for 20 and less observations. The calculations are done seperately for planimetry and height. The weight functions used are as follows:

planimetry:
$$P = P_i \cdot \frac{256}{256 + R_i^8}$$

height:
$$P = P_i \cdot \frac{81}{81 + R_i^4}$$

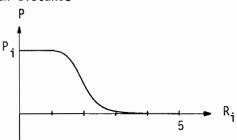
in which:

P; = a priori weight of observation i

 $R_i = D_i/D$

 D_i = distance of observation i from center point

D = mean distance



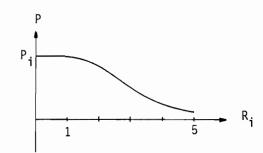


Fig. 6. Modification of weights for the starting least squares planimetric iteration step

Fig. 7. Modification of weights for the starting least squares height iteration step

The effectiveness of the modifications related to medium size and large gross errors is shown in Example 3. With the relatively poor geometry the gross error of three base lengths at point 10201 would not be locatable in planimetry without the reduction of the weight in the first least squares iteration step.

9. Reinsertion of observations:

In two cases it is required to reinsert already eliminated observations. Due to falsified O-approximations it may happen that an observation is wrongly eliminated. After orientation of the model the residuals of this observation will become small and a reinsertion is advisable.

Secondly the result of a least squares adjustment differs from the result of an adjustment with robust estimators in the range of $1\text{-}2\sigma$. After the final elimination of the small gross errors at the end of the robust adjustment some iteration steps with least squares are performed and small gross errors just at the limit of localization will tend to move to the class of random errors in the least squares adjustment and thus should be reinserted.

Therefore the weight function (formula 1) is used in the final least squares iteration steps to check for reinsertion of eliminated observations.

During the whole procedure of adjustment, as soon as the value of $F(v_i, \hat{\sigma}_{v_i}, Q)$ becomes larger than the value 0.01, used for elimination, an already eliminated observation will be reinserted in order to stabilize the geometry of the block and to contribute to a final result of adjustment.

10. Conclusion:

The above described procedure of automatic gross error localization is a specific development for the blockadjustment by the method of independent models and is not transferable to other problems without modifications.

The procedure covers the full range of occurring gross errors from the small ones, just at the limit of localization, up to the big ones with several base lengths and shows the power of robust estimators.

When the worst comes to the worst the procedure results in the elimination of a complete model or in the elimination of observations up to the point where no redundancy is remaining in a model. Then the user has to analyse the observations of the specific model and to take a decision.

In most cases the result of the procedure will be only a proposal, although a very good one, and the person in charge of the project has to judge the proposal and to decide about the final corrections of the gross errors.

The program is in an operational stage and the automatic error detection procedure is easy to handle. No parameters with respect to the procedure have to be changed by the operator, except for the decision whether the adjustment shall be performed with or without automatic error detection.

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EXAMPLE 1:	ARTIFICI	FROM PRINTOUT AL BLOCK WITH ; ~'5 POINTS PE			JPERPOSED (FROSS ERR			METER:	D Z	z	
	SCALE= 1 SIGMA= 1 MODELS I CONTROL	/10000 O Micron N Micron			101 102 107 304 402 404 407 407 HC HC HC VC VC VC	20203 10203 30803 50503 80503 80503 80703 70803 10103 50103 50103 50103 50103 50903	80.0 0.0 30.0 -70.0 -40.0 0.0 30.0 -1.0 -0.3	100 -100 00 00 00 00 00 00 00 00 00 00 00 00	0.000 00.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000 0.000	-80.0 100.0 30.0 -30.0 100.0 -40.0 50.0 30.0	000 000 000 000 000 000 000 000	
END OF ERROR SIGMA REACHES												
MODEL	101 401 102 102 207 208 308 408	POINT POINT POINT POINT POINT POINT POINT POINT POINT	10103 80203 10203 20203 30701 50903 90903	HV 1 TP 2 TP 2 TP 2 TP 4 HV 2 HV 2 HV 1			V Z = V Z =	0.955 0.469 0.799 0.392 0.417 0.467 0.451 0.521	ELIMINATED ELIMINATED ELIMINATED ELIMINATED ELIMINATED ELIMINATED ELIMINATED	IN HE IN HE IN HE IN HE IN HE	E I G H T E I G H T E I G H T E I G H T E I G H T	T T T T T
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40°00 7' 70°801 6' 70°803 6' 70°901 7' 80°801 6'	299.539 199.497 300.100 300.100 200.106 300.189 200.172	6299.904 6299.810 5399.757 5399.757 5399.878 6299.996	1499.887 1500.097 -0.070 -0.070 0.023 -0.238 0.050	PC 2 PC 1 TP 4 TP 4 VE 2 TP 2 SP 1		-0 0 0 -0	• 142 • 013 • 062 • 013 • 068	0.147 0.036 0.036 -0.065 -0.022	-0.02 -0.05 0.01 0.03 -0.01	3 C 7 C 0 C		::
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EXAMPLE 2:

EXTRACT FROM PRINTOUT:
PRACTICAL BLOCK WITH 32 MODELS
4 STRIPS; ~18 POINTS PER MODEL
SCALE= 1/28000; 20% SIDELAP
SIGMA PLANIMETRY = 5.6 MICRON
SIGMA HEIGHT = 9.3 MICRON SIGMA HEIGHT = MODELS IN 1/100 MM

CONTROL IN METER

ADJUSTMENT WITH AUTOMATIC ERROR DETECTION AFTER MANUAL DATA CLEANING

ITERATION STEP 9.....VERTICAL ADJUSTMENT

ITERATION STEP FOR ERROR DETECTION

END OF ERROR DETECTION IN ELEVATION SIGMA REACHED = 0.5511

VERTICAL	TRIOR JCRTROD		15401	ΗV	2			VZ=	0.554	ELIMINATED	
VERTICAL	CONTROL POINT		16202	HV	2			VZ=	0.405	ELIMINATED	
VERTICAL	CONTROL POINT		21101	HV				vz=	0.544	ELIMINATED	
VERTICAL	CONTROL POINT		21201	н٧				VZ=	1.494	ELIMINATED	
VERTICAL	CONTROL POINT		31201	HV				VZ=	0.435		
	CONTROL POINT		36201	н٧						ELIMINATED	
MODEL								V Z =	0.921	ELIMINATED	
	112111	POINT	6211	TP				V Z =	0.881	ELIMINATED IN	HEIGHT
HODEL	212211	POINT	16111	TΡ	4			VZ=	0.352	ELIMINATED IN	HEIGHT
MODEL	212211	POINT	26111	TΡ	4			VZ=	0.799	ELIMINATED IN	
MODEL	21 22 11	POINT	26112	TΡ	4			V Z =	1.008	ELIMINATED IN	
MODEL	212211	POINT	26211	TP	4			vz=	0.844	ELIMINATED IN	
MODEL	212211	POINT	26212	TP	Ĺ			V Z =	1.074		
MODEL	31 2311	POINT	36212	TP	-					ELIMINATED IN	
MODEL	211210							V Z =	0.397	ELIMINATED IN	
		TRICA	21001	HV				V Z =	0.863	ELIMINATED IN	HEIGHT
MODEL	411410	POINT	4100 0	PC	2	VXY=	1.365			ELIMINATED IN	PLANIMETRY
MODEL	4 1141 0	POINT	46101	но	2			vz=	0.945	ELIMINATED IN	
MODEL	3 1 0309	POINT	26011	TΡ	4			V Z =	0.850	ELIMINATED IN	
MODEL	410409	POINT	30802	HV	4			V Z =	0.882	ELIMINATED IN	
MODEL	109108	POINT	10800	PC		VXY=	1.709		0.302		
MODEL	109108	POINT	10900	PC	_	VXY=				ELIMINATED IN	
MODEL	207206	POINT	20612	70	_	W A T -	1.658			ELIMINATED IN	PLANIMETRY

V Z = V Z =

0.786 0.905

0.521

RE-INSERTED

RE-INSERTED IN PLANIMETRY RE-INSERTED IN PLANIMETRY

RE-INSERTED IN PLANIMETRY RE-INSERTED IN PLANIMETRY

vz=

ELIMINATED IN HEIGHT ELIMINATED IN HEIGHT

ITERATION STEP 10.....HORIZONTAL ADJUSTMENT

ITERATION STEP 12....HORIZONTAL ADJUSTMENT

310309

POINT

POINT POINT POINT

VERTICAL CONTROL POINT

MODEL 207206 MODEL 207206 MODEL 310309

MODEL

MODEL

POINT

POINT

20612

35611

ITERATION STEP FOR ERROR DETECTION

207206 30**63**05

MODEL

MODEL

END OF ERROR DETECTION IN PLANIMETRY
SIGMA REACHED = 0.3618

SIGMA REA	CHED = 0.3	618							
HORIZONTA	L CONTROL POI	NT	46101	но 2	V X Y =	0.911			ELIMINATED
VERTICAL	CONTROL POINT		31 201	HV 4		20/11	V Z =	0.445	RE-INSERTED
MODEL	213212	POINT	16311	TP 2	VXY=	1.511	• • •	0.445	ELIMINATED IN PLANIMETRY
MODEL	213212	POINT	16312	TP 2	V X Y =	1.648			ELIMINATED IN PLANIMETRY
MODEL	112111	POINT	6211	TP 2			¥ Z =	0.895	RE-INSERTED IN HEIGHT
MODEL	112111	POINT	16101	HV 2	VXY=	0.715	• •	4.373	ELIMINATED IN PLANIMETRY
MODEL	212211	POINT	16111	TP 4			v z =	0.388	RE-INSERTED IN HEIGHT
HODEL	212211	POINT	16201	HV 3	VXY=	0.793			SLIMINATED IN PLANIMETRY
MODEL	212211	POINT	16211	TP 4	VXY=	0.803			ELIMINATED IN PLANIFETRY
MODEL	212211	POINT	16212	TP 4	VXY=	0.808			ELIMINATED IN PLANIMETRY
MODEL	212211	POINT	26111	TP 4			Vz=	0.942	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26112	TP 4			VZ=	1.003	RE-INSERTED IN HEIGHT
MODEL	212 211	POINT	26211	TP 4			vz=	0.320	RE-INSERTED IN HEIGHT
MODEL	21 2 21 1	POINT	26212	TP 4			vz=	0.962	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26212	TP 4	VXY=	0.760			ELIMINATED IN PLANIMETRY
MODEL	312311	POINT	36212	TP 4			V Z =	0.399	RE-INSERTED IN HEIGHT
MODEL	211216	POINT	21001	HV 2			v z =	0.065	RE-INSERTED IN HEIGHT
MODEL	411410	POINT	46101	HO 2			v z =	0.955	RE-INSERTED IN HEIGHT
MODEL	310309	POINT	26011	TP 4			v z =	0.663	RE-INSERTED IN HEIGHT
MODEL	310309	POINT	26011	TP 4	V X Y =	0.984			ELIMINATED IN PLANIMETRY
MODEL	310309	POINT	26012	TP 4	V X Y =	0.362			ELIMINATED IN PLANIMETRY
MODEL	410409	POINT	30802	HV 4			V Z =	0.385	RE-INSERTED IN HEIGHT
MODEL	308307	POINT	25711	TP 4	VXY=	0.342			ELIMINATED IN PLANIMETRY
MODEL	207206	POINT	20612	TP 2			V Z =	0.873	RE-INSERTED IN HEIGHT
MODEL	207206	POINT	25711	TP 4	V X Y =	0.922			ELIMINATED IN PLANIMETRY
MODEL	207206	POINT	25712	TP 3	V X Y =	0.990			ELIMINATED IN PLANIMETRY
MODEL	306305	POINT	35611	TP 4			v z =	0.908	RE-INSERTED IN HEIGHT
ITERATION	STEP 11	VERTICAL ADJI	JSTMENT						
HODI TONTA	L CONTROL POI	N.T.	46101	но 2	vxY=	0.933			RE-INSERTED
	CONTROL POINT	•••	21101	HV 1	V A 1 -	0.733	¥ Z =	0.348	RE-INSERTED
MODEL	112111	POINT	16101	HV 2	V X Y =	0.719	47+	0.348	
MODEL	212211	POINT	16201	HV 3	VXY=	0.894			RE-INSERTED IN PLANIMETRY. RE-INSERTED IN PLANIMETRY
MODEL	212211	POINT	16211	TP 4	VXY=	0.736			RE-INSERTED IN PLANIMETRY
MODEL	21 2 2 1 1	POINT	16212	TP 4	VXY=	0.809			RE-INSERTED IN PLANIMETRY
MODEL	212211	POINT	26212	TP 4	VXY=	0.762			RE-INSERTED IN PLANIMETRY
MODEL	308307	POINT	25711	TP 4	VXY=	0.845			RE-INSERTED IN PLANIMETRY
HOVEL	300301	FOINT	23/11	17 4	A Y 1 -	0.043			KE-INSERIED IN PLANIMEIRY

YXY=

VXY=

YXY=

0.556

0.611

0.865

HV 2 TP 4 TP 3 TP 4 TP 4

15401 25711 25712

26011

26012

TRANSFORMED PHOTOGRAMMETRIC MODEL COORDINATES AND RESIDUALS

1 10 00113 01	THE TERRETA STATEM				
MODEL NUMBER	213212			sc= 3.53452	
16202 20171 16211 20981 16212 20951 16311 18514 16312 18467 21200 21100 21201 19944 21211 2105 21212 2102 21300 18511 21311 18452 21312 18452	5.406 41662.499 0.998 43717.534 7.462 41518.197 7.464 41519.737 9.591 41482.252 0.479 43821.375 4.651 4504.112 2.150 43989.129 2.329 43989.129 2.329 43995.038 7.154 44000.347	657-137 658-197 650-274 650-688 629-558 632-152 4981-772 633-600 624-532 625-999 4982-950 650-543	HV 3 HV 2 -> HO 2/ 2 TP 4 TP 4 TP 2 -> TP 1/ 2 TP 2 -> TP 1/ 2 PC 2 HV 2 -> HO 2/ 2 TP 2 TP 2 PC 1 SP 1 SP 1	-0.001 -0.076 -0.205 0.001 0.005 -0.157 0.127 0.037 -1.079* 0.266* -1.212* 0.253* 0.309 0.113 -0.016 -0.013 0.004 0.083 0.015 0.049	0.295
26212 2103. 26311 1350	2.232	600.110 601.323 621.303 621.239	TP 4 TP 2 TP 2	-0.150 0.014 -0.100 -0.025 0.111 0.055 0.217 0.033 SC= 3.46565	0.099 0 0.095 0 -0.100 0 -0.003 0
36001 2568; 36011 2674; 36012 2574; 360101 2335; 36111 2350; 36112 2357; 36114 2355; 41000 2606; 41001 2432; 41011 2601; 41012 2598; 41102 2357; 41111 2350; 41112 2347; 46011 2604; 46012 2604; 46012 2604; 46131 2379;	5.502 50895.996 9.497 52144.309 6.907 51446.519 6.920 53.628 51641.649 6.920 53965.705 6.188 54037.733 1.542 54037.733 1.542 54037.735 1.542 54038.412 7.258 54023.282 5.236 5682.862 5.799 56612.918 6.9150 56277.211 6.480 56328.343	597.317 606.611 602.409 602.578 570.693 592.963 593.553 598.551 598.480 5040.819 545.516 591.393 590.659 5039.783 570.090 570.016 547.764 547.338 608.497 597.472 601.201	HC 4 HV 4 TP 4 TP 4 HO 2 TP 4 TP 4 TP 4 TP 4 TP 2	-0.079	0.023
MODEL NUMBER	109108			SC= 3.51248	
5702 3122 5301 3065 5311 3096 5312 3096 5911 2858 5912 2655 10800 3112 10801 2939 10811 3097 10900 2867 10901 2867 10911 2833 15301 2937 15811 3103 15812 3100 15911 2849	1.875 36832.748 4.645 38422.500 9.872 36334.185 2.101 36477.780 2.101 36477.794 3.043 36480.071 0.157 38970.670 7.094 39347.598 7.247 39028.095 7.247 39028.095 7.247 39028.095 1.187 38879.800 4.812 41539.893 4.1536 41537.827 2.602 41207.073 5.158 41230.860	583.670 581.233 537.957 579.263 580.450 538.840 539.198 4989.630 592.489 569.533 4986.889 609.879 595.718 539.158 561.291 561.320 566.629 567.405	HV 2 HV 2 HV 2 TP 2 TP 2 TP 2 TP 2 TP 2 PC 2 -> PC 1/ 2 HV 1 TP 2 PC 2 TP 2 HV 2 TP 2 HV 2 TP 2 HV 2 TP 4 TP 4 TP 4	0.003	0.131 0
VERTICAL CONTR	OL POINTS				
5701 5702 5801 5901 10401 10501 10902 15703 15301 15901 16102 16201 16202 20501 20901 21101 21101 21201 25501 25801 35901 36901		583.800 581.100 538.200 602.800 548.900 592.800 609.600 583.400 587.400 589.300 657.600 658.600 658.600 658.900 629.800 635.100 553.900 609.400 658.7.700 606.600	HV 2 HV 2 HV 1 HV 1 HV 1 HV 2 HV 2 HV 2 HV 1 HV 1 HV 2 HV 1 HV 2 HV 1 HV 3 HV 2 HV 1 HV 2 HV 1 HV 2 HV 1 HV 2 HV 1 HV 2 HV 2 HV 1 HV 2 HV 2 HV 4 HV 1 HV 2 HV 4		0.001 2 0.065 20.064 2 0.062 2 0.023 20.068 20.048 20.063 20.015 20.015 20.055 2 0.081 20.118 20.551* 12 0.087 2 0.0154 20.154 20.154 20.008 20.154 20.008 20.008 20.008 20.008 20.008 20.008 20.008 20.008 20.008 20.008 20.008 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 20.009 2
36201 40501 40502 40601		577.300 541.000 565.900 572.200	HV 2 -> HO 2/ 2 HV 1 HV 2 HV 2		-1.034* 12 -0.028 2 . -0.038 2 . 0.044 2 .

EXAMPLE 3:	ARTIFICI 4 STRIPS SCALE= 1 SIGMA= 1 MODELS I	FROM PRINTOUT AL BLOCK WITH 3 6 POINTS PE 7109003 20% S O MICRON N MICRON IN METER	32 MODELS R MODEL					SUPERPOSED 100 101 406 406 400		NT-NO. 01 01	BASELENGTH: DXY 3 3 3	9Z 3 3			
ITERATION STEP												•			
ITEPATION STEP	FOR ERRO	R DETECTION													
HORIZONTAL CON Model Model	ETROL POIN 101 4-3さ	T POINT POINT	10901 10201 90701	TP 2		V X Y =		2699.649 2704.637 2702.264			ELIMINATED ELIMINATED ELIMINATED				
ITERATION STEP SVERTICAL ADJUSTMENT															
ITERATION STEP FOR ERROR DETECTION															
VERTICAL CONTR MODEL MODEL	OL POINT 101 406	POINT POINT	10961 10201 20701	TP 2					v z =	2699.824 2695.110 2701.269	ELIMINATED				
ITERATION STEP															
ITERATION STEP	FOR ERROI	R DETECTION													
END OF EPROR D Sigma reached															
MODEL MODEL MODEL	191 192 192	TVIC9 TVIC9 TVIC9	30201 1020) 20201	PC Z					v z = v z = v z =	0.632		IN	HEIG	HT	
ITERATION STEP	12v	ERTICAL ADJUST	TMENT												
MODEL MODEL	1⊎1 1⊎2 1⊎2		30201 10200 20201						V Z = V Z = V Z =	0.405	RE-INSERTED	IN	H E 1	GHT	
THANSFORMED PH	******	********													
MODEL NUMBER		101							sc:	= 100.0053	o .				
10101 - 10200 -	0.2s1 -0.012 10.7o3 0.7o9	1909.157	1499.976 -0.031 1499.852 2699.508	₽ С	1 2	SP	1/	- O. :	036 083 911*	0.642			0000	::	:
20201 90 30101).215 }).054).356 }).113	909.059 399.941 1809.024 1809.225	0.054 -0.126 -0.062 -0.175	TP VE	2			-0.0 -0.0 0.0	343	-0.008 0.051 -0.054	0.016		0	: :	:
MODEL NUMBER		435							3 C =	= 100.0095	3				
40700 533 70601 450 70701 533 30601 450	00.055 90.963 00.007 99.795 00.043	6299.407 6300.157 5400.016 5400.075 6299.955 6299.942	1500.089 1500.020 0.144 0.074 0.723 0.239	PC PC TP TP TP	2 4 4 2			-8.1 0.0 +0.0 -0.0 -0.0	061 043 091 003	0.159 0.008 0.059 -0.058 0.014	0.061 0.059 +0.059 -0.006 -0.030 -0.067		00000		:
90501 449	99. °23 39.120	7200.00° 9108.984	0.137 2699.622	TP TP	2 ->	SP	1/	າ. ∙)() o	-0.040 -1908.934	0.005		0 3.0		•
CONTROL POINT	******	*******	ALS ***												
MCD JATHOLISON	NTROL POIN	T S													
10901 910 50101 - 50901 720 90101 - 90501 359	0.033 00.030 09.957 -0.037 -0.912 -0.991 09.070	-).u45 -0.022 1909.005 3600.061 3600.000 7199.952 7260.067 7199.981		H V H V H V H V H V H V H V	? 1 -> 2 2 1	SP	1/	1 -1913.1 -0.0 0.0 0.0 -0.0 -0.0	007 117* 009 016 012	-0.003 0.026 -1908.763 -0.031 0.002 0.005 0.007			2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2		
VERTICAL CONTR	ROL POINTS		-0.000										_		
10101 10501 10501 30101 30501 30501 50101 50501 50901 70101 70501			-0.009 0.027 2700.038 -0.054 0.024 0.093 -0.018 -0.072 0.065 0.032	# # * V V V H V V V V V V V V V V V V V V V	2 1 -> 2 4 2 2 4 2 2 4 2 2 2	SP	1/	1			-0.004 0.304 -2702.158 0.309 -0.301 -0.007 0.203 -0.012 0.003 -0.010 0.307	*	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2		