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

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<u>Abstract</u>: 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 handling blockadjustment programs is the detection and location of outliers. Dependent on the number and distribution of the observations, errors are shown up only partly by 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 for the development of an automatic procedure has been: to shorten the processing time needed by persons in charge with blockadjustment.

# 3. From least squares to robust adjustment:

The above mentioned problems arising by 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 unsuitable for automatic error detection procedures.

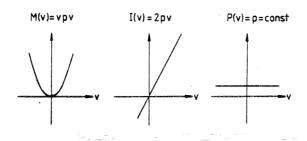


Fig. 1. Least squares: minimum function, influence function and weight function

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.

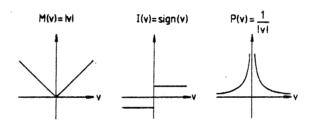


Fig. 2. Minimum Norm: minimum function, influence function and weight function

An adjustment procedure using weight functions eliminating the influence of gross errors completely 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 will have no influence at all onto the result of the adjustment. Their residuals will show up 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 combinations and the different geometry of photo-

grammetric blocks. (Werner, 1984). Because of their effectiveness the following hyperbolic weight function was chosen for the blockadjustment program PAT-M43:

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

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

$$(1)$$

in which: •

$$\alpha_{i} = \frac{1}{1:4 \cdot \hat{\sigma}_{V_{i}}} = \frac{\sqrt{\bar{p}_{i}}}{1.4 \cdot \sqrt{\bar{r}_{i}} \cdot \hat{\sigma}_{0}}$$
 (2)

$$d = 3.5 + \frac{82}{81 + 0^4} \tag{3}$$

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

 $v_i$  = residual of observation i  $P_i$  = a priori weight of observation i  $r_i$  = local redundancy of observation i  $\widehat{\sigma}_{V_i}$  = 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 at 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  $\delta v_i$  (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  $50 \cdot \sigma_0$  cause no problems, even with geometrically very weak configurations, as long as there are still error-free redundant observations.

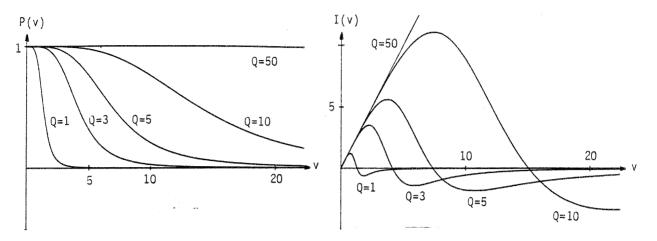


Fig. 3. PAT-M43: weight function Fig. 4. PAT-M43: influence function with  $P_i=1; r_i=1; \sigma_{a\ priori}=1$  and Q=1;3;5;10;50

# 5. New structure of the program:

If no initial values of the orientation parameters 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 bigger gross errors in height do not disturb the orientation of the models too much. After this 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 the geometric stability of the block configuration. Thus the number of iteration steps differs 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 infinite 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 is using a shift in z only. 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 treating 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
- 2. medium-sized gross errors
- 3. large gross errors

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

All gross errors greater  $4 \cdot \sigma$  and less  $50 \cdot \sigma$  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 within the random errors.

All errors between  $50 \cdot \sigma$  and 2-3 base lengths belong to the medium-sized 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 worse 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 location are detected as long as the observations are sufficiently well distributed within the models.

Only for really worse 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-sized and large gross errors:

Medium-sized gross errors and all the more large 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 bigger 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:

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

in which:

SW = starting weight

P; = a priori weight

P = modified a priori weight

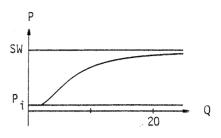


Fig. 5. Modification of a priori weights: with SW=1;  $P_i$ =0.1

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

As long as there is no too worse accumulation of medium-sized 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-sized gross errors the adjustment approachs a different 0-point and the location of the small gross errors will not be correct. The pre-elimination of large and medium-sized gross errors will solve these problems. 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 larger 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 0-approximations of the residuals after the starting least squares iteration are such false 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: 
$$P = P_i \cdot \frac{256}{256 + R_i^3}$$

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

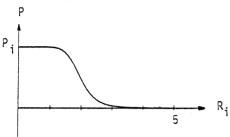
#### in which:

P, = a priori weight of observation i

 $R_i = D_i/D$ 

D; = distance of observation i from center point

D = mean distance



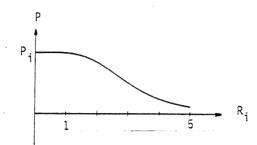


Fig. 6. Modification of weights for the starting least squares planimetric iteration step

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

The effectiveness of the modifications related to medium sized and large gross errors is shown in Example 3. With the relatively bad 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 0-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  $1-2\sigma$ . 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 the class of random errors in the least squares adjustment and also 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  $F(v_i, \hat{\sigma}_{v_i}, 0)$  becomes bigger than the value 0.01, used for elimination, an already eliminated observation will be reinserted in order to stabilize the geometry of the block respectively to be closer to a final result of adjustment.

# 10. Conclusion:

The above described procedure of automatic gross errors localization is a specific developement 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 no redundancy is remaining in a model and the user has to analyse the observations of the specific model.

In most cases the result of the procedure will be only a proposal, but a very good one, and the person in charge with 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, its just the decision whether the adjustment shall be performed with or without automatic error detection.

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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
MODELS IN 1/100 MM
CONTROL IN METER EXAMPLE 2:

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	CONTROL POINT		15401	HV Z			٧Z=	0.554	ELIMINATED	غر
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VERTICAL	CONTROL POINT		21101	HV 1			VZ=	0.544	ELIMINATED	
VERTICAL	CONTROL POINT		21201	HV 2			V Z =	1.494	ELIMINATED	
VERTICAL	CONTROL POINT		31201	HV 4			VZ=	0.435	ELIMINATED	
VERTICAL	CONTROL POINT		36201	HV 2			vz=	0.921	ELIMINATED	
HODEL	112111	POINT	6211	TP 2			vz=	0.381	ELIMINATED	IN HEIGHT
HODEL	212211	POINT	16111	TP 4			v z =	0.352		IN HEIGHT
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MODEL	212211	POINT	26212	TP 4			VZ=	1.074	ELIMINATED	
MODEL	31 23 1 1	POINT	36212	TP 4			VZ=	0.397	ELIMINATED	
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MODEL	102108	POINT	10900	PC 2	VXY=	1.709				IN PLANIMETRY
HODEL	207206	POINT		PC 2	AXA=	1.658				IN PLANIMETRY
MODEL	305305		20612	TP 2			VZ=	0.786	ELIMINATED	IN HEIGHT
	100303	TRICA	35611	TP 4			V Z =	0.905	ELIMINATED	IN HEIGHT

ITERATION STEP 10..... HORIZONTAL ADJUSTMENT

ITERATION STEP FOR ERROR DETECTION

END OF ERROR DETECTION IN PLANIMETRY SIGMA REACHED = 0.3018

	. CONTROL POI	TV	46101	HO 2	YXY=	0.911			ELIMINATED
	ONTROL POINT		31 201	HV 4			VZ=	0.445	RE-INSERTED
MODEL	213212	POINT	16311	TP 2	V X Y =	1.511			ELIMINATED IN PLANIMETRY
MODEL	213212	POINT	16312	TP 2	V X Y=	1.648			ELIMINATED IN PLANIMETRY
MODEL	112111	POINT	6211	TP 2			VZ=	0.395	RE-INSERTED IN HEIGHT
MODEL	112111	POINT	16101	HV Z	VXY=	0.715			ELIMINATED IN PLANISETRY
MODEL	212211	POINT	16111	TP 4			V Z =	0.888	RE-INSERTED IN HEIGHT
HODEL	212211	POINT	16201	HV 3	<b>YXY=</b>	0.793			ELIMINATED IN PLANIMETRY
MODEL	212211	POINT	16211	TP 4	V X Y =	0.303			ELIMINATED IN PLANIFETRY
MODEL	212211	POINT	16212	TP 4	V X Y =	0.808			ELIMINATED IN PLANIMETRY
MODEL	212211	POINT	26111	TP 4			V Z =	0.942	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26112	TP 4			VZ=	1,303	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26211	TP 4			V Z =	0.320	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26212	TP 4			V Z =	0.962	RE-INSERTED IN HEIGHT
MODEL	212211	POINT	26212	TP 4	V X Y =	0.760			SLIMINATED IN PLANIMETRY
MODEL	312311	POINT	36212	TP 4			VŽ≃	0.399	RE-INSERTED IN HEIGHT
MODEL	211216	POINT	21001	HV 2			v z =	0.005	RE-INSERTED IN HEIGHT
MODEL	411410	POINT	46101	HO 2			¥ Z =	0.955	RE-INSERTED IN HEIGHT
MODEL	310309	POINT	26011	TP 4			¥ Z =	0.363	RE-INSERTED IN HEIGHT
MODEL	310309	POINT	26011	TP 4	VXY=	0.984			ELIMINATED IN PLANIMETRY
MODEL	310309	TAICS	26012	TP 4	V X Y =	0.362			ELIMINATED IN PLANIMETRY
MODEL	410409	POINT	30802	HV 4			VZ=	0.384	RE-INSERTED IN HEIGHT
HODEL	308307	POINT	25711	TP 4	V X Y =	0.342			ELIMINATED IN PLANIMETRY
MODEL	207296	POINT	20612	TP 2			¥ Z =	0.373	RE-INSERTED IN HEIGHT
MODEL	207206	POINT	25711	TP 4	4 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			VZ=	0.908	RE-INSERTED IN HEIGHT

	CONTROL POINT	¥T	46101 21101	H0 2 HV 1	<b>Y X Y =</b>	0.933	۷Z=	0.343	RETINSERTED
MODEL	112111	POINT	16101	HV 2	V X Y =	0.719			RE-INSERTED IN PLANIMETRY
MODEL	212211	POINT	16201	HV 3	V X Y =	0.894			RE-INSERTED IN PLANIMETRY
MODEL	212211	POINT	16211	TP 4	VXY=	3.736			RE-INSERTED IN PLANIMETRY
MODEL	212211	POINT	16212	TP 4	<b>V X Y =</b>	0.309			RE-INSERTED IN PLANIMETRY
HODEL	21 2211	POINT	26212	TP 4	VXY=	0.762			RE-INSERTED IN PLANIMETRY
MODEL	308307	POINT	25711	TP 4	<b>V X Y =</b>	0.845			RETINSERTED IN PLANIMETRY

# ITERATION STEP 12....HORIZONTAL ADJUSTMENT

MODEL Model Model	ONTROL POINT 207206 207206 310309	POINT POINT POINT	15401 25711 25712 26011	HV 2 TP 4 TP 3 TP 4	Y X Y = V X Y = V X Y =	0.556 0.611 0.365	¥Z=	0.521	RE-INSERTED RE-INSERTED IN PLANIMETRY RE-INSERTED IN PLANIMETRY RE-INSERTED IN PLANIMETRY
MODEL	310309	POINT	26012	TP 4	Y X Y =	0.572			RE-INSERTED IN PLANIMETRY

# TRANSFORMED PHOTOGRAMMETRIC MODEL COORDINATES AND RESIDUALS

( IN UNITS OF THE TERRAIN SYSTEM ) MODEL NUMBER 213212 SC= 3.53452 19785.406 20170.993 20987.462 20957.464 18519.591 41502.499 43717.534 41513.197 41519.737 657.137 658.197 650.274 -0.076 0.295 -0.001 0 0 0 0 0 0 0 0 0 -0.205 0.005 0.127 -1.079\* 16202 0.001 -0.157 0.037 16211 -0.087 -0.156 10212 650.638 0.013 16311 41480.567 629.558 0.256\* 20 18449.311 41442.252 632.152 0.253\* 16312 43821.375 46404.112 43987.740 4981.772 0.113 -0.002 -0.112 21 200 0.309 0 21 201 19044.651 -0.016 624.532 625.909 4982.950 71 71 1 0.194 21052-156 0.004 0.033 43989.129 21022.216 21300 18457.043 43995.031 650.857 21311 450.543 21312 SP 1 TP 4 TP 4 TP 2 TP 2 46492.155 46492.295 46547.361 46549.505 -0.150 -0.100 0.111 0.217 0.014 -0.025 0.055 0.033 0.099 0.095 -0.100 21002.232 21032.230 600.110 601.323 621.303 000 • • • 26212 13507.633 -0.003 26312 14477.240 621.239 \$ C = MODEL NUMBER 411410 3-46565 0 . . . 31101 23425.562 50495.795 597.317 -0.079 -0.024 0.123 0.023 606.611 0.065 36001 25689.447 26016.967 25986.920 23350.184 23503.741 0.173 0.110 -0.366 -0.170 30011 36012 51446.019 51446.510 602.409 0.220 0.033 -0.072 0 . . 53000.080 570.693 592.963 -0.102 -0-363 36101 0 36111 0.029 0.320 51535.375 51537.630 51613.751 51641.645 53965.705 23473.658 23570.230 593.553 598.551 -0.125 -0.218 0.245 36112 0.057 -0.338 -0.361 ō 30113 -0.013 1.903\* -0.080 0.045 a 36116 23559.386 598.430 -0.153 0.025 20066.920 5040.819 20 535.216 591.393 590.659 5039.734 570.090 ō 56248.214 54037.733 54033.412 41001 24822.185 0.135 41011 +1012 26011.589 0.024 0.234 0.052 ō -0.091 0.043 -0.020 -0.046 -0.029 -0.081 41100 23577.703 53996.041 0.051 a 54014.470 54023.282 23505.941 23477.258 -0.066 41111 570.016 a 26045.235 547.764 0.010 0.024 36582.362 46011 56612.918 56903.700 56277.211 26043.799 547.338 -0.001 0.119 -0.062 46012 0.020 0 0.411 -0.128 40101 TP 2 TP 2 597.472 0.021 0.112 Œ 46131 23790.150 . . 0.093 -0-057 23709.480 56323.343 46132 MODEL NUMBER 109108 \$C= 3-51248 HV 2 HV 2 HV 2 TP 2 TP 2 0.003 -0.033 0.065 0.043 0.069 0.131 -0.119 0.179 -0.126 -0.112 -0.270 0.252 31761.375 36832.748 583-670 0 . . . 38422.50° 36334.185 36477.780 36479.483 36477.794 31.224.345 30.339.372 30.992.108 30.962.101 531.233 537.957 579.263 530.450 5702 5301 0.031 0.030 5311 5312 5312 -0.024 -0.103 -0.235 Q TP 2
TP 2
TP 2
PC 2 -> PC 1/ 2
HV 1
TP 2 -0.138 -0.067 0.255\* 0.034 28582.380 28553.043 31120.157 558.340 539.198 36477.794 36480.071 38970.070 39347.598 39028.096 0.393 -0.136 2.045+ 10800 4939.630 20 29397.094 30977.247 592.489 569.533 0.240 10801 0.096 10311 0.159 TP 2
PC 2 -> PC 1/ 2
HV 2
TP 2
HV 2
TP 4
TP 4
TP 4
TP 4
TP 4 38919.489 40359.040 0.722\* 10900 28007.330 2. 513\* 0.011 10901 28475.439 609-879 -0.795 -0.320 10911 23331.137 38379.300 41539.693 595.718 539.158 0.024 -G.122 0.261 -0.095 0. 195 41533.327 41537.327 41207.073 15811 15312 31031.536 31001.966 561.291 561.320 -0.353 -0.040 -0.649 -0.294 -0.363 0 0 15911 560.629 3.106 -0.057 0.079 0.010 n 28492,602 28495.153 41230.360 VERTICAL CONTROL POINTS 583.800 581.100 5701 HV 2 0.001 5702 9.062 9.062 9.062 -9.063 HV 2 5801 538.200 5901 602.800 10401 548.900 HV 10501 592.300 HV -0.043 609-600 10902 533.400 15703 HV 1 HV 2 587.400 -0-015 -0.346 15301 539.300 15901 578.600 HV 4 623.900 HV 1 0.381 16102 658.600 572.200 2 -> 40 2/ 2 -0.551+ 0.037 16202 20901 592.800 J.154 -0.003 21001 588.900 629.300 21101 -0.026 -1.612\* 25501 HV 2 HV 4 -0.393 -3.351 553,900 25801 609.400 HV 2 HV 4 0.016 537.700 HV 4 HV 2 -> HO 2/ 2 HV 1 36001 606.600 577.300 -1.034\* -0.323 40501 565.900 -0.038 HV Z HV Z 40001 572-200

EXAMPLE 3:	ARTIFICI 4 STRIPS	N MICRON In Meter	MODEL Elap			SUPERPOS MODEL-NO 101 406 HVC		01 01	BASELENGTH: DXY 3 3	0Z 3 3		
ITERATION STEP		ORIZONTAL ADJUS	TMENT	(=1.44T)	The same of the sa		* <u>Luc</u> e = 1 <u></u>	in grand				
ITERATION STEP	FOR ERROR	R DETECTION										
HORIZONTAL CON Model Model	TROL POINT 101 406	T POINT POINT	10901 10201 90701	HV 1 TP 2 TP 2	VXY=	2699.649 2704.637 2702.266			ELIMINATED ELIMINATED ELIMINATED			
		ERTICAL ADJUSTM										
ITERATION STEP	FOR ERROR	R DETECTION										
VERTICAL CONTRA MODEL MODEL	OL POINT 101 406	POINT POINT	10901 10201 90701	HV 1 TP 2 TP 2			V Z =	2699.324 2695.110 2701.269	ELIMINATED ELIMINATED ELIMINATED			
ITERATION STEP	10VE	ERTICAL ADJUSTM	ENT									
ITERATION STEP	FOR ERROR	R DETECTION										
END OF ERROR DE SIGMA REACHED :								e				
MODEL Model Model	101 102 102	POINT POINT POINT	30201 10200 20201	TP 4 PC 2 TP 2			¥Z= ¥Z=	0.537 0.632 0.306	ELIMINATED ELIMINATED ELIMINATED	IN	HEIG	нт
		RTICAL ADJUSTM										
MODEL MODEL MODEL	101 102 102	POINT POINT POINT	30201 10200 20201	TP 4 PC 2 TP 2			VZ= VZ= VZ=	0.346 0.405 0.286	RE-INSERTED RE-INSERTED RE-INSERTED	IN	HEI	GHT
TRANSFORMED PH	***	TRIC MODEL COOR		-	ALS							
MODEL NUMBER		101					sc:	= 100.00530	1			-
10101 10200 90	0.281 0.012 0.763 9.769	-0.053 900.162	1499.976 -0.031 1499.852 2699.508	PC 1 HV 1 PC 2 TP 2 -	> SP 1	-	0.036 -0.043 9.911*	0.010 0.042 -1909.243*	0.01a 0.035 2699.317	5	0.00	8 4 6 9 8 0
20201 90 30101	0.215 0.054 0.356 0.113	900.059 899.941 1300.024 1800.225	0.054 -0.126 -0.052 -0.170	SP 1 TP 2 VE 2 TP 4			-0.010 -0.043 -0.023	-0.008 0.051 -0.054	-J.08J J.016 J.J19	5	0000	6 6 6 0 0 0
MODEL NUMBER		400					s <b>c</b> =	= 100.00958	;			•
40700	0.007 9.796 0.043 9.300 9.923	6300.152 5400.016 5400.076 6299.955 6299.942 7200.099	1500.039 1500.020 0.144 0.074 0.223 0.239 0.137	PC 2 PC 2 TP 4 TP 4 TP 2 TP 2 TP 2			0.025 0.031 -0.043 0.091 -0.003 -0.049 0.009	0.159 0.008 0.059 -0.058 0.014 0.025 -0.040	0.061 0.094 -0.059 -0.006 -0.030 -0.067	) ;	0 0 0 .0	6 0 4 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6 6
	9.120 COORDINATE	9103.984 Es and residual	2699.622	TP 2 -	> SP 1	/ 1 -190	39.235*	-1908.934	-2699.713	*	30	
( IN UNITS OF	THE TERRAL	**************************************										
HORIZONTAL CON	0.033	-0.045		HV 1		•	.0.009	-0.003			2	o .
10901 910: 50101	0.020 9.957 0.037 0.012 0.091 9.979	-0.022 1909.005 3600.061 3600.000 7199.952 7200.067		HV 2 HV 1 = HV 2 HV 2 HV 1 HV 2	> \$P 1	/ 1 -191 - -	0.087 0.117* 0.009 0.015 0.012	0.026 -1908.763* -0.031 0.002 0.005 0.007			22 2 2 2 2 2 2 2 2	0 c
90901 7200 VERTICAL CONTRO	0.001 OL POINTS	7199.981		HV 1		-	.0.063	-0.007			2	• •
10101 10501 10901 30101 30501 30901 50101 50501 50901 70101 70501			-0.009 0.027 2700.038 -0.054 0.003 -0.018 -0.072 0.065 0.032 0.000	HV 1 HV 2 HV 1 - VE 4 VE 2 HV 2 VE 4 HV 2 VE 2 VE 4	> SP 1	/ 1			-0.204 -2700.158 0.309 -0.301 -0.007 0.303 -0.012 3.003 -0.010 0.307	*	222222222	