

# Robust Estimation of the Outliers in GPS Baseline Components

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**Key Words:** Outlier, Robust, Estimation, GPS

## SUMMARY

Several definitions can be made for the term ‘outlier’ such as a single observation far away from the rest of the observation set. The outlier is the observation that not includes systematic and gross errors but random errors and differs from the other data with its distribution. It is assumed that geodetic observations free from systematic and gross errors always have random errors which are in normal distribution. If an observation is separated from the other data with respect to its random error which is not in normal distribution, that observation is named ‘outlier’.

To date several approaches have been developed to identify the outliers more accurately. The first proposed approaches, conventional methods, used the statistical tests to detect outlying observations and removed them from the data set. The main drawback of such an application is the shape defect of the network. Unlike the conventional methods, the robust estimations do not remove any observation from the network. They reduce the effects of the possible outliers on the adjustment model as making the weights of the outlying observations smaller even zero. In this study, some robust estimations (Danish method, Huber’s, Andrew’s, and IGGIII M-Estimations) are applied to a 17-point GPS network in order to make decision about outliers and the baseline components ( $\Delta X_{i-j}, \Delta Y_{i-j}, \Delta Z_{i-j}$ ) of that network are taken into account as measurements. The used GPS network consists of 34 baselines ranging from 302 meters to 1684 meters.

In robust estimations, it is crucial to define a proper critical value to form the weight functions. This critical value can be calculated or a constant value, which is defined by the a priori standard deviation of the unit weight, can be assigned to it. In this study, both choices are tested in examining the robust estimations. Danish method and Andrew’s M-Estimation with constant critical values yield similar results by making zero the weights of the suspicious observations. All estimations with calculated critical values and IGGIII M-Estimation indicate same results by reducing the weights of the possible outlying observations.

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## 1. INTRODUCTION

The methods developed for outlier detection are classified in two models. One of these models is “The mean-shift model”, which determines the outliers by means of statistical tests, and the other is “The variance-inflation model”, which reduces the effects of the outlying observations on the adjustment model assigning to them small weights (Kubik and Wang, 1991). In this study, some robust estimations are applied to the GPS baseline components which are considered as measurements in order to determine the outlying measurements in a GPS network. The robust estimations belong to the second model and they use the iteratively reweighted least squares procedure.

Unlike conventional statistical test methods, there is not the drawback of network distortion in the robust methods. No measurement is removed from the network after the iterations, only the weights are changed. The residuals/standardized residuals are compared with a critical value and robust weights between 0-1 are assigned to them in order to obtain the final weights of the measurements. It is observed that the weights of the outlying measurements reduce, even become zero after the iteration steps. In this way, the robust methods decrease the influences of the outliers on the parameter estimation.

The robust estimations used in this study and the constant critical values connected with them are given below Table 1.

**Table 1.** The weight functions of the M-Estimations

M-Estimation	Weight Function	Critical Value (c)
Andrews	$w_i = \begin{cases} \sin(v_i/c)/(v_i/c) &  v_i  \leq c\pi \\ 0 &  v_i  > c\pi \end{cases}$	$1.5s_0 - 2s_0$
Huber (Gui and Zhang, 1998; Hekimoglu and Berber, 2003)	$w_i = \begin{cases} 1 &  v_i  \leq c \\ \frac{c}{ v_i } &  v_i  > c \end{cases}$	$1.5s_0 - 2s_0$
Danish (Berberan, 1992; Berberan, 1995)	$w_i = \begin{cases} \exp\left(-\frac{v_i^2}{c^2}\right) &  v_i  > c \\ 1 &  v_i  \leq c \end{cases}$	$1.5s_0 - 2s_0$
IGGIII (Yang, et al., 2001)	$w_i = \begin{cases} 1 &  \tilde{v}_i  \leq c_0 \\ \frac{c_0}{ \tilde{v}_i } & c_0 <  \tilde{v}_i  \leq c_1 \\ 0 &  \tilde{v}_i  > c_1 \end{cases}$	$c_0 = 2.0 \sim 3.0$ $c_1 = 4.5 \sim 8.5$

In Table 1,  $s_0$  is the a priori standard deviation of the unit weight,  $\tilde{v}_i$  is the normalized residual and is equal to  $v_i/s_{v_i}$ . Since it is very crucial to define a proper the value to form the weight functions, the formulation related to median of the absolutes of the residuals with weights can be used to calculate  $s_0$  (Yang et al., 1999).

$$s_0 = \text{med} \left\{ \left| \sqrt{P_i} v_i \right| \right\} / 0.6745 \quad (1)$$

where “*med*” denotes median,  $P_i$  and  $v_i$  are the weight and residual of the observation  $\ell_i$ , respectively. After  $s_0$  is obtained, it should not be changed in the iteration steps of the robust estimation.

As stated above, besides assigning constant values to the critical value, it can be calculated and the calculation procedure can be applied as follows:

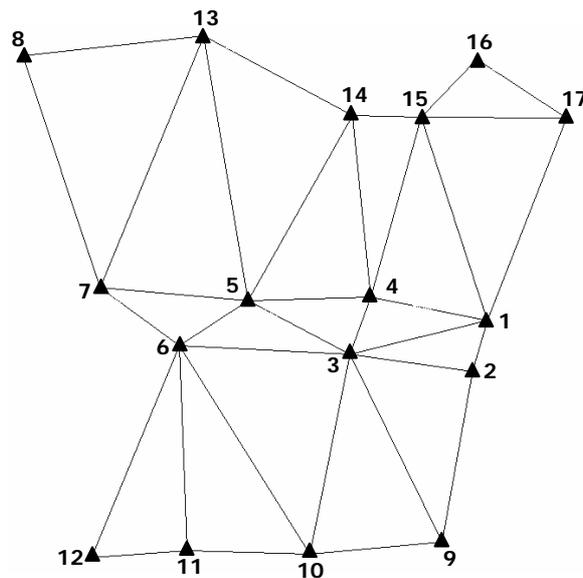
$$c_i = s_0 \sqrt{Q_{vv_{ii}}} \sqrt{P_{ii}} t_{f, 1-\alpha_0/2} \quad (2)$$

Here  $c$  represents the critical value,  $s_0$  is the a priori standard deviation,  $Q_{vv}$  is the cofactor matrix of the residuals,  $P$  is the weight matrix of the observations,  $f$  is the degree of freedom,  $\alpha_0$  is the significance level, and  $t$  represents the t table.

The desired critical value is calculated by taking the average of the critical values calculated for each observation:

$$c = \sum_{i=1}^n c_i/n \quad (3)$$

## 2. THE WRITTEN COMPUTER PROGRAM AND ITS EXECUTION USING THE DATA OF A GPS NETWORK



**Figure 1.** The 17-point GPS network

The GPS network used for outlier investigation consists of 17 points and 34 baselines ranging from 302 meters to 1684 meters (Figure 1). The baseline components of the network have been considered as measurements. In order to adjust the measurements free adjustment has been performed. The adjustment process and the estimation procedures have been written in MATLAB technical computing language (MATLAB, 2001). The input data such as approximate coordinates of the network points, the variance-covariance matrix of the measurements etc. are provided from the computer program GeoGenius (GeoGenius, 2000).

The a priori standard deviation of the unit weight  $s_0$  is obtained as 4.05 mm. from equation (1). The constant critical values used in the robust estimations have been taken into account as  $2s_0$ , thereby they are obtained as 8.1 mm. The calculated critical value from equation (2) is equal to 6.9 mm.

## 2.1 The Measurements (Baseline Components) and the Numbers Assigned to Them

Table 2. The baseline components in the GPS network

Baselines		Baseline Components					
		$\Delta X$ (m)	MN	$\Delta Y$ (m)	MN	$\Delta Z$ (m)	MN
2	1	-233.7982	<b>1</b>	82.3269	<b>2</b>	172.6048	<b>3</b>
3	1	-166.2416	<b>4</b>	864.5049	<b>5</b>	131.2265	<b>6</b>
3	2	67.5540	<b>7</b>	782.1781	<b>8</b>	-41.3778	<b>9</b>
4	1	143.9372	<b>10</b>	737.2146	<b>11</b>	-90.8069	<b>12</b>
3	4	-310.1788	<b>13</b>	127.2902	<b>14</b>	222.0348	<b>15</b>
4	5	44.0973	<b>16</b>	-773.2000	<b>17</b>	-37.9013	<b>18</b>
3	5	-266.0820	<b>19</b>	-645.9108	<b>20</b>	184.1373	<b>21</b>
6	5	-219.1057	<b>22</b>	433.3547	<b>23</b>	160.2469	<b>24</b>
3	6	-46.9721	<b>25</b>	-1079.2660	<b>26</b>	23.8936	<b>27</b>
7	5	104.9636	<b>28</b>	936.4827	<b>29</b>	-77.6581	<b>30</b>
7	6	324.0713	<b>31</b>	503.1278	<b>32</b>	-237.9039	<b>33</b>
7	8	-1156.9079	<b>34</b>	-470.0131	<b>35</b>	837.9788	<b>36</b>
9	2	-843.1286	<b>37</b>	206.7179	<b>38</b>	623.6820	<b>39</b>
3	10	1033.6841	<b>40</b>	-286.6628	<b>41</b>	-760.4310	<b>42</b>
9	10	122.9529	<b>43</b>	-862.1080	<b>44</b>	-95.4264	<b>45</b>
9	3	-910.7314	<b>46</b>	-575.4424	<b>47</b>	665.0048	<b>48</b>
11	10	18.6660	<b>49</b>	796.8231	<b>50</b>	-10.0875	<b>51</b>
6	10	1080.6767	<b>52</b>	792.6002	<b>53</b>	-784.3027	<b>54</b>
6	11	1062.0076	<b>55</b>	-4.2209	<b>56</b>	-774.2169	<b>57</b>
11	12	46.9800	<b>58</b>	-593.6542	<b>59</b>	-34.1216	<b>60</b>
6	12	1108.9881	<b>61</b>	-597.8759	<b>62</b>	-808.3387	<b>63</b>
13	8	79.7232	<b>64</b>	-1121.1667	<b>65</b>	-61.2690	<b>66</b>
13	7	1236.6348	<b>67</b>	-651.1432	<b>68</b>	-899.2448	<b>69</b>
5	14	-1043.7486	<b>70</b>	674.1378	<b>71</b>	769.5501	<b>72</b>
13	14	297.8383	<b>73</b>	959.4782	<b>74</b>	-207.3410	<b>75</b>
13	5	1341.5867	<b>76</b>	285.3401	<b>77</b>	-976.8956	<b>78</b>
15	14	-54.2960	<b>79</b>	-415.9632	<b>80</b>	37.0028	<b>81</b>
4	14	-999.6738	<b>82</b>	-99.0542	<b>83</b>	731.6512	<b>84</b>
4	15	-945.3777	<b>85</b>	316.9089	<b>86</b>	694.6486	<b>87</b>
15	1	1089.3219	<b>88</b>	420.2947	<b>89</b>	-785.4428	<b>90</b>
17	1	1041.4928	<b>91</b>	-490.9504	<b>92</b>	-755.9327	<b>93</b>
17	15	-47.8309	<b>94</b>	-911.2453	<b>95</b>	29.5093	<b>96</b>
15	16	-282.8851	<b>97</b>	428.9426	<b>98</b>	219.5607	<b>99</b>
17	16	-330.7117	<b>100</b>	-482.3059	<b>101</b>	249.0796	<b>102</b>

MN: Measurement Number

## 2.2 Results with Constant Critical Value

Danish method and Andrews's M-Estimation tend to produce many outliers making zero the weights of the measurements in the iteration steps of the adjustment. In Danish method, the measurements with the numbers 9, 33, 34, 40, 42, 58, 59, 60, 73, 90 and in Andrews's M-Estimation, the measurements 7, 9, 18, 33, 34, 40, 42, 58, 59, 60, 73, and 90 got zero weight after 7 iterations. Huber's M-Estimation caused small variations in almost all measurements after 5 iterations.

### 2.3 Results with Calculated Critical Value

The weights of the measurements with the numbers 58 and 60 became smaller after 3 iterations in Danish method and after 4 iterations in Huber's M-Estimation. In addition to these measurements, 3<sup>rd</sup> and 9<sup>th</sup> measurements got smaller weights after 3 iterations in Andrews's M-Estimation. In IGGIII M-Estimation, only the weight of the 60<sup>th</sup> measurement became smaller after 2 iterations.

### 3. CONCLUSIONS AND RECOMMENDATIONS

The behaviors of Danish method and Andrews's M-Estimation tend to reduce the weights of outlying measurements to zero in case of using constant critical values. In this condition, it is a difficult process to obtain the unknown parameters of the adjustment and sometimes it is also impossible to reach a result. All of the estimations with calculated critical value and also IGGIII M-Estimation reduce the weights of suspicious measurements throughout the iterations and indicate almost the same measurements as outliers. Considering this result, it is the better choice to use the robust estimations with calculated critical values in detection of the outliers.

### REFERENCES

- Berberan, A., 1992. Outlier Detection and Heterogeneous Observations a simulation case study. *Australian Journal of Geodesy, Photogrammetry, and Surveying* **56**, 49-61.
- Berberan, A., 1995. Multiple Outlier Detection. A Real Case Study. *Survey Review* **33**, 255, 41-49.
- GeoGenius, 2000, Spectra Precision Terrasat GmbH.
- Gui, Q., Zhang, J., 1998. Robust Biased Estimation and Its Applications in Geodetic Adjustments. *Journal of Geodesy* **72**, 430-435.
- Hekimoglu, S., Berber, M., 2003. Effectiveness of Robust Methods in Heterogeneous Linear Models. *Journal of Geodesy* **76**, 706-713.
- Kubik, K., Wang, Y., 1991. Comparison of Different Principles for Outlier Detection. *Australian Journal of Geodesy, Photogrammetry, and Surveying* **54**, 67-80.
- MATLAB, 2001, The Language of Technical Computing, Version 6.1.0.450 Release 12.1.
- Yang, Y., Cheng, M. K., Shum, C. K., and Tapley, B. D., 1999. Robust Estimation of Systematic Errors of Satellite Laser Range *Journal of Geodesy* **73**, 345-349.
- Yang, Y., Song, L., and Xu, T., 2001. Robust Estimator for the Adjustment of Correlated GPS Networks. *First International Symposium on Robust Statistics and Fuzzy Techniques in Geodesy and GIS* March 12-16, Zurich / Switzerland.

## **BIOGRAPHICAL NOTES**

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