ABSTRACT

A digital elevation model (DEM) created by automatic image matching or laser scanning – also named as LIDAR, includes a not negligible number of points, not located on the terrain surface but on buildings or vegetation or even miscorrelation. The manual refinement of such a DEM is time consuming. In general, points located not on a continuous surface which can be differentiated, have to be identified and removed. A Digital Filtering Technique to be applied to the automatically acquired DEM-data is presented. The strategy is based on Linear Prediction of stationary random function after trend removal. The filtering was applied to Photogrammetrically acquired DEM-data via automatic digital correlation techniques. Different terrain compositions like buildings and canopy density; terrain roughness; grid sizes; negative scales have been investigated. The system can also be used in connection with DEM acquired through LIDAR. For both applications very acceptable results have been achieved.

INTRODUCTION

Digital Elevation Models (DEM) acquired using photogrammetric correlation techniques or LIDAR have the disadvantage that the resulting modelation surface may not represent the bare terrain but the visible surface including vegetation and buildings, the so called Digital Surface Model (DSM). The first step to be done after data acquisition is the removal of the points not belonging to the terrain. An automated method using linear prediction has been applied. The method and results are explained.

DEM FILTERING

The elimination of points not belonging to the ground is known as Filtering. There are several methods or procedures for interpolation and filtering. Among them are:

- Splines approximation
- Shift Invariant Filters
- Linear Prediction
- Morphological Filters
Although Morphological Filters are the most frequently being applied, Linear Prediction, also named as Linear Least Squares Interpolation, is a very robust methodology for the filtering of Digital Surface Models. It is based on the dependency of one point in relation to the neighbored points described by the covariance function.

THE PROGRAM DTMCOR

Developed at the Institute for Photogrammetry and Engineering Survey of the University of Hannover, the software analyses and filters a Digital Elevation Models. The programs locates firstly possible Spikes or Blunders by the introduction of a tolerable minimum and maximum height.

The elimination of the Trend is carried out via the use of a moving plane.

\[ Z_i = a_0 + a_1 X_i + a_2 Y_i \]  

(1)

The area covered by the DEM is divided into a mesh of equal size. The dimensions of the square grid is suggested by the program based on the distribution of the points. For the processing of the points of a particular mesh (Processing Unit = 1 mesh), the points located in the 8 surrounding meshes are considered. In this way, the moving plane coefficients are computed using the points located in the 9 contiguous meshes (Area of Consideration), via least squares. See Fig. 1

![Figure 1](image)

The trend removal is based on the tilted plane in the Area of Consideration resulting in the centered measurement values \( l_i \).

Assuming a normal distribution of those \( l_i \) values their standard deviation (\( \sigma_z \)) is computed by the program and a multiplication factor (\( fac \)) is entered into the program for the computation of a threshold or tolerance factor (\( T_z \)).

\[ T_z = fac \sigma_z \]  

(2)

All those points within the corresponding processing unit whose deviations (i.e., centered measurement values \( l_i \)) are exceeding the above tolerance (\( T_z \)) are excluded. The trend separation is repeated in a loop until no more defective heights are recognized by the system. The erroneous heights of the processing mesh are deleted from the records. The erroneous heights of the neighboring patches (i.e., area of consideration) remains for the computations of the next patch. They only remains unconsidered for the current patch.
Figure 2 shows the above explained iterative process of trend removal and elimination of erroneous height values.

The figure 2 shows a typical terrain height-profile. It consists of 10 points, three of which are blunders. The inclination of the moving plane represented by the colored straight lines, varies considerably with each iteration. Four iterations are needed in the example, until the moving plane stabilizes and fits the terrain surface and no more blunders are identified. The standard deviation decreases drastically from iteration to iteration with the removal of the largest blunders to the smallest.

In relation with the Linear Prediction DTMCOR makes use of a Covariance Function with the following form:

$$C(P_iP_k) = A \cdot e^{-1.30103\left(\frac{PP_k}{B}\right)^2}$$  \hspace{1cm} (3)

A and B are parameters of the function. A is the vertex-value of the signal-covariance function. It is a filter factor for the normed covariance function. Its specifies the relationship between random and systematic components of the height discrepancies (i.e., centered measured heights $l_i$). A value $A = 1.0$ (in the program limited to 0.99), means no random errors are available. The parameter B is the slope of the function and represents the distance at which the influence of points is reduced to 5%. On the other hand its value also limits the width of the mesh for the local prediction.

The interpolated surface is defined as in equation (1) where the main diagonal elements of the covariance matrix contain variances $V_{ll} = 1$, meaning all measurements are regarded as being of the same accuracy. As the vertex-value A have been found to be appropriate at 0.7, interpolation and filtering are possible. The higher the vertex-value of the function is, the smaller the variance $\sigma_l^2$ is, and smaller is the filtering effect (See Figure 3).
The differences between the centered measured values at each DEM point and the predicted value according to formulae (1), are computed. Once again assuming normal distribution of the above discrepancies, their corresponding standard deviation ($\sigma_{zp}$) is computed. A multiplication factor is introduced in the program for the calculation of a threshold or tolerance value ($T_{zp}$)

$$T_{zp} = \text{fac} \sigma_{zp}$$  \hspace{1cm} (4)

If the computed discrepancies are exceeding the threshold, the corresponding DEM points are also eliminated in a loop fashion.

Figure 3. **Covariance Function**

Figure 4. **surface of prediction**

Figure 5. **profile through area including a point not belonging to the surface**
(left: left hand profile, center: center profile, right: right hand profile
red line = surface of prediction in these profiles based on the area,
points = real height point)
EXPERIMENTAL TESTS

DEM's automatically acquired using digital correlation techniques have been filtered using DTMCOR. Three different terrain areas have been used, namely:

Area Type A: Urban flat area with heavy building density, buildings of different heights and dimensions, moderate canopy.

Area Type B: Urban Flat area with moderate building density, regular building size and height, moderate to heavy canopy.

Area Type C: Flat urban / Industrial area. Large open spaces, mixture of low altitude, small and big size buildings, low to moderate density of canopies.

Three different image scales, 3 DEM spacing and a DEM in a TIN fashion have been analyzed. The results are shown in Table 1:

<table>
<thead>
<tr>
<th>AREA TYPE</th>
<th>IMAGE SCALE</th>
<th>DEM Spacing (ft)</th>
<th>TIN</th>
<th>&lt;min &gt;max Z (%)</th>
<th>Tilted Plane (%)</th>
<th>Linear Predict. (%)</th>
<th>Total Excluded (%)</th>
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<tr>
<td>B</td>
<td>1:6000</td>
<td>30</td>
<td>16.7</td>
<td>29.6</td>
<td>6.1</td>
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<td></td>
<td>50</td>
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<td>28.7</td>
<td>5.74</td>
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<td>28.9</td>
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</table>

Table 1 percentage of excluded points

A close look of Table 1 reveals:

a For given terrain coverage there is a very small increase of the eliminated points with smaller DEM spacing. This is more noticeable in Area B with regular building size and heights.

b Regardless the type of terrain coverage, the largest percentage of eliminated points is carried out by the iterative trend removal procedure. This corresponds to non surface terrain points such as correlated points on building roofs. A key parameter for the good performance of the iterative trend removal procedure is the chose of the weight factor for lower points. In all cases a numerical value of $3\sigma_h$ was used.

c In general the performance of the method do not change considerably with the use of Triangular Irregular Networks (TIN's).
d The Image Scale does not influence the performance of the procedure.
e Linear prediction works better in cases of very abrupt changes of heights. This can be observed in terrain cover type A.
f The performance of the minimum - maximum height filter depends on the actual knowledge the minimum and maximum terrain heights, which in the majority of cases is not known, but it is supported by a frequency distribution.
Figures 6 and 7 are showing a piece of the terrain cover type C with overlaid contour lines. Figure 6 shows the contour lines before filtering the Digital Elevation Model and Figure 7 after filtering. Contour lines of Figure 6 honors correlated points on roof top of buildings and tree canopies. Contour lines of Figure 7 are representing the terrain.

**CONCLUSIONS**

The above digital filtering strategy based on an iterative trend elimination plus linear prediction has proven to be a very effective tool for removing non terrain points of a Digital Elevation Model based on digital correlation techniques in a soft copy photogrammetry environment and also by LIDAR.

The filtered Digital Elevation Models can be used without excitation for production of digital orthophotos. Contour lines shall be derived with the aid of breaklines and digital graphical editing.

Further investigation is required using other terrain coverage and other DEM geometry including breaklines.

**ACKNOWLEDGEMENTS**

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**REFERENCES**


