RESEARCH ON CONSTRUCTING DEM WITH POINT CLOUD FILTERING ALGORITHM CONSIDERING SPECIAL TERRAIN

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ABSTRACT:

With the development of airborne LiDAR, the use of LiDAR point cloud to construct DEM model has become a hot topic in recent years. The characteristics of point cloud filtering and non-ground point filtering are significant. This paper proposes a method of constructing DEM based on the point cloud filtering algorithm of airborne Lidar point cloud data considering special terrain. The experiment proves that the algorithm of this paper is effective for establishing DEM model, and the quality of DEM model is good.

1. INTRODUCTION

Airborne Lidar Technology (LiDAR) is an active measurement method that can quickly acquire high-precision, high-density three-dimensional spatial information in large-area measurement areas[1]. The degree of automation is high, and it is less affected by the weather. It can block through the woods, and the spatial and temporal resolution of the data is relatively high. The digital elevation model contains a wealth of surface and landform feature information, which is a digital description of the ground shape through the ruled discrete grid sampling data. It can simulate the surface undulations well using the limited data with elevation attributes[2]. The use of point cloud technology to generate DEM models can be a development trend in the future. It is an important step in point cloud data processing, which plays an indispensable role in the generation and application of subsequent digital products. The amount of point cloud data is highly discrete. How to effectively use the interpolation method to achieve high-precision DEM model has become an urgent problem in the case of reducing manual intervention.

2. PRINCIPLE

At present, there are few researches on special terrain. At present, most of the mature algorithms are effective for constructing DEM models in cities and flat areas. However, the accuracy of DEM model generation for both mountains and flat areas is not very good. This paper proposes a point cloud processing method that takes into account special terrain. First, the last echo is filtered out in the original data as the basic point cloud data. The filtered cloud data is grossly eliminated and the wrong data is eliminated. The elevation value of the point cloud is extracted, and the gradient of the adjacent point cloud is calculated by the elevation, and the slope is divided into two types of 0-30, 30-60. The point cloud data is classified according to angles and divided into flat areas and mountains. The point cloud data of each area is separately filtered to extract ground points and non-ground points. Manually rejecting a small number of rough points that have not been filtered makes the modeling accuracy higher. The method of neighbor search is used to establish DEM models in cities and flat areas. However, the present, most of the mature algorithms are effective for constructing DEM models in cities and flat areas. However, the accuracy of DEM model construction with both mountains and flat areas is not very good. This paper proposes a method of constructing DEM based on the point cloud filtering algorithm of airborne Lidar point cloud data considering special terrain. The experiment proves that the algorithm of this paper is effective for establishing DEM model, and the quality of DEM model is good.

irregular grid of the area. According to the Delaunay triangulation algorithm, the point-by-point insertion method is used to establish an irregular triangular network TIN, and the TIN is used to approach the ground to make it more in line with the field fluctuation state. Then use the Kriging interpolation method and the inverse distance weighting method to perform elevation interpolation, and finally generate the DEM model.

2.1 Extract slope

At present, the accuracy of DEM model construction with both mountains and flat areas is not too good. The complexity of DEM model terrain surface is another factor that must be considered when sampling terrain data. The complexity of terrain can be described by roughness and irregularities. The terrain surface can be completely formed by the slope. The first-order differential function of the slope's topographic surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance. The variability of the slope is the second-order differential of the surface function, expressing the ratio of elevation to distance.
A simplified form of differential publicity:

\[ \text{Slope} = \arctan \sqrt{\frac{x^2}{y^2} + \frac{y^2}{x^2}} \times 180/\pi \]

Figure 1 slope.

Fx represents the rate of change of the elevation in the X direction, and fy represents the rate of change of the elevation in the Y direction. The terrain relief is the difference between the maximum and minimum elevations at a given location, and the terrain relief can visually screen the surface features. The expression is:

\[ \text{RF}_i = \text{H}_{\text{max}} - \text{H}_{\text{min}} \]

RFi refers to the terrain relief, Hmax is the maximum elevation value in the analysis window, and Hmin refers to the minimum elevation value in the analysis window.

2.2 KdTree

KdTree is a data structure that divides the k-dimensional data space. It is mainly used for the search of key data in multi-dimensional space, which is a special case of binary space segmentation tree. It has great advantages in the neighborhood search. However, in the case of large data volume, if the partition granularity is small, the overhead of building a tree is relatively large, and it will be more flexible than the octree. In the case of small data volume, the search efficiency is compared high. KdTree is a recursive proximity search strategy that consists of two main aspects: building KdTree and finding it in KdTree. The neighboring point of a point only needs to search in its parent node and child node, which greatly reduces the search scale of the neighboring point. KdTree can effectively judge the insertion point and its nearest point is at that position, which is very important for low-level vision.

2.3TIN

The regular moment shopping net data is equally spaced in the horizontal and vertical directions. The plane coordinates of the grid points are hidden in the row and column numbers, and are often stored in a matrix structure, or the elevation values of each grid unit are recorded one by one. In addition, data structures such as a run length coding structure, a block coding structure, and a quadtree coding structure are used to store DEM data.

An irregular model is created based on the undulating state of the terrain. The data obtained by the regular grid sampling method and the progressive sampling method, especially the square grid data, is most suitable for grid-based surface modeling[3]. Usually grid modeling is often used to process authority data covering a flat area. For special terrain, a method of establishing an irregular grid should be used.

Irregular grids are much more complicated than regular grids. Due to the irregularity of triangles, the definition of triangles and their adjacent relationships are to be expressed. The TIN model not only stores the elevation of each vertex, but also stores the plane coordinates of the vertices of the triangle, the connection relationship between the vertices, and the topological relationship such as the adjacent triangle. Irregular triangulation uses fewer points to return higher precision and has a good topological relationship. Irregular grids are constructed by fitting multiple triangles to discrete data to fit the true ground. It is a good reflection of the undulating shape of the earth's surface, especially in valleys and ridges. The more common method for constructing irregular triangulation is the Delaunay triangle. The method of splitting.

1, divide and conquer the algorithm, the points are refined according to the region, the triangulation is merged, and the local optimization is performed;

2, the triangle is first established by the neighboring method, and then the triangulation is established by using the point-by-point insertion method. Define the minimum outer rectangle range to include as a simple bulge. The search data point p contains a triangle of points, and p is connected with three vertices of the triangle to form three triangles, and the entire area is optimized from the inside to the outside;

3. The implementation process starts from an arbitrary point and finds the shortest distance point. Connected to the line, according to the method of triangulation, find the point containing another triangle edge, form a new triangulation, and then extend to the entire area.

2.4Interpolation

The inverse distance weighted interpolation algorithm calculates the distance between the coordinates of the point and all the discrete points as the weight influence value. The farther from the sampling point, the smaller the influence of the sampling, that is, the inverse relationship, the reciprocal can have the second constraint. The weakening of the weight is declining. Its formula is:

\[ D_Z = \sum_{i=0}^{n} D_i \times V_i \]

\[ D_i = \frac{1}{c + d_{ij}^m} \]

\[ \sum_{i=1}^{n} \frac{1}{c + d_{ij}^m} \]

The inverse distance interpolation method is based on the similar principle, that is, the closer the two objects are, the more similar their properties are. Centering on the interpolation point, when the variable Z value is interpolated at the points X and Y to be interpolated, the weighted average of each data point is calculated in the local neighborhood. The formula is:

Where \( Z_{i,0} \) is the measured value of the interpolated element at the i-th point, and
of nearest neighbors set `MaximumNearestNeighbors` (100). It sets `Mu` (2.5) of the neighbor distance, and the maximum number set `SearchRadius` (1.5) of the k-nearest neighbor, the multiplier the connection point to determine the spherical radius KdTree proximity search needs to set the maximum distance of neighbor's sphere radius and a neighbor distance multiplier. The search neighbor important parameter such as a k-nearest obtained point cloud data, and it is necessary to set a near-

The neighbor search performs gross error rejection on the point cloud data. The elevation value of the point cloud is extracted, and the gradient of the adjacent point cloud is calculated by the elevation, and the slope is divided into two types of 0-30, 30-60. The point cloud data is classified according to angles and elevation, and the slope is divided into two types of terrain.

The main purpose of noise filtering is to separate points below the ground or points that are not in the desired area, and points that exceed the threshold are separated from the point cloud data. Point cloud classification to distinguish between ground and non-ground points. The bare ground point has only one return band, and the position of the echo reflection is the position of the ground point. The ground point of the coverage area is the point at which the last echo of the multiple echoes is reflected. The position of the callback point is the elevation corresponding to the feature point. To establish the DEM, it is necessary to extract the accurate ground point. If the accuracy of the ground point extraction is insufficient, the quality of the generated DEM model is poor, and the effect is not ideal.

Manual editing, because of the influence of the algorithm, not all areas can complete the point cloud filtering process, which will generate some faulty areas, resulting in the processed data can not meet the quality requirements of DEM data results, so human-computer interaction is required. It is necessary to manually delete the missing point cloud data, so that the factors affecting the DEM model are minimized.

In this paper, the point cloud data is respectively used by Kriging interpolation method and inverse distance weighting method. The artificially edited data is generated into DEM model, and the experimental comparison is made. The area of 0-30° is suitable for using Kriging interpolation method, 30-60° area is suitable for the inverse distance weighting method. The Kriging interpolation method is applicable to flat areas, and the inverse distance weighting method is applicable to mountains. For special terrains, we should use the combination of Kriging and inverse distance weighting methods to establish DEM. The DEM effect of this method is better.

In this paper, the process of rapid processing of airborne LiDAR point cloud data to generate DEM is designed, and the theory is used to demonstrate the theory.
The theory obtained by using the analysis of point cloud data is better in using the Kriging interpolation method in the flat terrain. It is suitable to use the inverse distance weighted interpolation method in the mountains. This method takes into account flat terrain and mountains. This method works better than using a single algorithm. In the experiment, the effect of point cloud classification is not very good and needs further optimization.

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