

596809.71428571 88162381552 67834770759 15788065650 28681996.547945 20302501.946667 694137574.33333 19778847.44444 477888881.5 32657979.611111 12615783.22619 5600411.9529412 73378321.714286 9455450.1836735 2249489232 28416685.516129 24560600.9375 53378726128 10341472.775862 190330707897 22842624.083333 57912555435 2187614971 19951994.395349 57524445.837838 68874513638 49897611.175 6731786108

Mean shift algorithm pdf free pdf download



Optimized Mean Shift Algorithm for Color Segmentation in Image

Sequences Werner Bailer⁴, Peter Schallauer⁴, Handd Bergur Handdsson^b, Herwig Rehatschele⁴ ¹OANNEUM RESE ARCH, Institute of Information Systems and Information Management, Steyrergasse 17, 8010 Graz, Austria ^bTokyo Institute of Technology, Precision and Infigurence Laboratory, R2-13, 4259 Nagabuta, Midori-ku, Yokohama, 226-8503 Japan

ABSTRACT

The application of the mean shift algorithm to color image segmentation has been proposed in 1997 by Committi Meet. We apply the mean shift color segmentation to image sequences, so the first step of a moving object segmen dipathan. Revision work has shown that it is well saided for this took, because it provides better importal stability experimetion result famolther approaches. The darknock is higher compatibilities out.

consistency prices in matching in the primery is the ELEV solver up to a significant solution and primer is the solution of th

version, we show us to sa opinizate agoing a noncess processing user and its reases are empound saleing o segmentation. Keywords: segmentation, mean shift, image sequence, objective evaluation

1. INTRODUCTION

The application of the mean shift algorithm to color image segmentation has been papooed in 1979 by Connuction and Meer [2]. Since them in his become an whifty used method for color image segmentiation, as it provides significantly been segmentation results no other approaches (e.g. the Watershed algorithms, for example less overse gueratation and holtances against diministration changes. We apply the mean shift color segmentation to image sequences, as the first top is a newlet opies segmentation algorithm. Provides work has shown that means shift algorithms is well using the sequences has negative.

algorithm. Previous work has shown that the near shift algorithm is well satisfied for sequenciation of image sequen because it provides better trappool a shifting of the segmentation neutral time of an approach (cf. for example However, the darwback is that the mean shift algorithm is computationally more expensive, and computational cost even more an issue when the algorithm is applied to image sequences. This means that there is used for optimizatio don't to use the algorithm is applied to be image sequences. This means that there is used for optimization don't to use the algorithm is applied to be image sequences. This means that there is used for optimization don't to use the algorithm is applied to be image sequences. This means that there is used for optimization offer to use the algorithm is applied to be image sequences. This means that there is used for optimization offer to use the algorithm is applied to be algorithm.

Due obtain optimization is is exploit the imposed indicatory in the image data when applying the algorithm is using sequences. It can be expected that the algorithm converges fracts. This optimization has alwedy been remained in ind [9]. Other remarkers advected the use of another color space than CHE LUV is reduce the complexity of lightritis, for example, UV [9] or HSV [10] have been proposed. Another is premote distinguish and share the sequence of the state of the sequence is priorit within the order of the sequence complexity in spatin-transport sequence of the sequence is a sequence of the sequence of th

ww.nar.baller@joanneum.at; phone +43 316 876-1218; fax +43 316 876-1191; http://iis.joanneum.at

Copyright 2015 Society of Phen-Optical Instrumentation Engineers. This paper was published in Proc. Conference on longs and Vieto Communications and Processing 1554 T397E. Decremos Imaging, Soc. Rose, C. 4 (SA, Ja, Ko, 2008) and a nonder multithe on a television at particular particular STME and 1587. Our paint excitation ice systems of far provend use. Systematics or analysis production, for industries to multiple locations is to decremo in some, doplications of any material in this paper fars for set incremonarily paperous, entended uses of the source of the paperon for set incremonary large-our entended uses of the source on publish of.



Figure 3: As need image (conservation b) = N0 groupside). B. Granshan MN argumptation with bandwidth $\sigma > 0$ gradie, considing to 5 (denter) (dented in different values), with the respective modes marked a. C: the distance (gradie) and all the MN income for all conting points in (0,) C speec, values in place B. projection of the solidile gifts on (0,) speec. The paths for two starting preds are shown in place B.D.



Figure 6. Improves of distances $\mathbf{X}^{(2)} = \mathbf{X}^{(2)}$ obtained by Gaussian HDD for the constraints image of Eq. 5 with bandwickle $r \sim 6$, resolving to 7 clusters. We does the D2 projection on the system man image of Eq. 5 with point (pind), externed by denotes, so in Eq. 3D. Note: (1) points very publicly move towards a central and colleges used in (clustering property). (2) for each cluster to be, the head dimension of accurate and colleges used provided by does the latter contribute dimension, producing latency desped dimension container colleges used provided descent. The DDD mapping relevant mapped HDD at decision r = 12, where the dimension of 5 clusters of collection. The DDD mapping relevant mapped HDD at decision r = 12, where the dimension of 5 clusters of collection points. The respirate tempting these clusters must be point to study they are implicitly used to residually more block to a single cluster.

pited). The essage features are maded to span appendianticle the range of the spatial hormory. This way, all instance and the baselench's hore pited onlin. For comparis, for the bange of fig. 1, we rewait the integral meaning values to the range 30,100, so a feature vector 12,15,80, would correspond to the pited lacared at correspond to the range 30,100, so a feature vector 12,15,80, would correspond to the pited lacared at correspond to the range 30,100, so a feature vector 12,15,80, would correspond to the pited lacared at correspond to the range 30,100, so a feature vector 12,15,80, would correspond to the pited lacared stransference (4,12), which has an intensity regard to 80% of the semilaries transmitty (added). The previse realing will affect the chartering and densit's to done correspond to the same charter), addressly remetimes, only they introduce upstied coherence (specific piteric lead to being to the same charter), addressly remetimes, only the range features are upstiel coherence (specific piteric lead to being to the same charter), addressly range that the field space, w first Eachdran distance specification rates spectrapical differences in rates (90).

Fig. 5 shows an excession with a generated image of 30×50 pitchs. Thus, the deduced statistics N = 2500prime in 3D documents, j and intensity J_2 . Fig. 1 shows the result with MD while fig. 6 shows the result with



Semivariogram-Based Spatial Bandwidth Selection for Remote Sensing Image Segmentation With Mean-Shift Algorithm

Dongping Ming, Tianyu Ci, Hongyue Cai, Longxiang Li, Cheng Qiao, and Jinyang Du

Abstract-Image segmentation is a key procedure that partitions an image into homogeneous parcels in object-based image analysis (OBIA). Scale selection in image segmentation is always difficult for high-performance OBIA. This letter is aimed at scale selection before segmentation in OBIA and proposes a spatial statistics-based spatial handwidth selection method based on mean-shift segmentation. This study uses ikonos and Quickbird panchromatic images as the experimental data and then computes their semivariances to select the optimal spatial bandwidth for mean-shift segmentation. To validate this method and interpret the relationship between the semivariances and segmentation scale, this letter implements an image segmentation evaluation based on the homogeneity within and the heterogeneity between the segmentation parcels. The evaluation results hasically support the proposed scale selection method based on the semivariogram. Consequently, the semivariogram-based spatial bandwidth selection method is practically meaningful for pre-estimating the appropriate scale and thus contributes to improving the performance and efficiency of OBIA.

Index Term-Evaluation of segmentation, high-spatialresolution remote sensing image, mean-shift segmentation, semivariogram, spatial bandwidth selection.

1. INTRODUCTION

With the increase of spatial resolution in remote sensing images, misclassification and but more of an issue for pixel-based multispectral image classifications. Object-based image analysis (OBIA) can effectively incorporate spatial information and expert knowledge into the classification, and the classified image objects are useful links to integrate remote sensing and geographic information systems. Multiscale image segmentation is the foundational

Massocript meriod July 15, 2011; revised September 29, 2011; November 9, 2011, and December 12, 2011; accepted December 22, 2011. Date of publication February 13, 2012; date of current version May 29, 2012. This work was supported in part by the National Natural Science Foundation of China under Gmnt 43001259, by the "Fundamental Research Funds for the Central Universities," and by the Open Fund of the State Key Laboratory of Remote Sensing Science under Grant OFNLRSS201008.

D Ming is with the School of Information Engineering. China University of Geosciences, Beijing 100083, China, and also with the State Key Laboratory of Remote Sensing Science, Institute of Remote Sensing Applications, Chinese Academy of Sciences, Beijing 100101, China (r-mail: mingdpgis @163.com).

T. G. and H. Cai are with the School of Information Engineering, China-

procedure of OBIA in which the digital image is transformed from discrete pixels into homogeneous image object primitives. The selection of scale parameter in the segmentation is the key to OBIA because an inappropriate scale will lead to oversegmentation or insufficient segmentation, which will directly reduce the accuracy and efficiency of multiscale information extraction.

Currently, several multiscale segmentation algorithms have been proposed and applied in remote sensing image analysis, including watershed segmentation, multiresolution segmentation [1], and mean-shift segmentation [2], [3]. Among these algorithms, watershed segmentation has a specific scale parameter; however, the relationship between segmentation results on different scales cannot be easily determined. Multiresolution segmentation is based on the so-called fractal net evolution approach [1], [4], and it not only provides a scale parameter but also incorporates the hierarchical relationship between different levels. However, the scale parameter is an abstract value used to determine the maximum possible change of heterogeneity caused by fusing several objects, and it is difficult for a user to quantitatively select the optimal scale parameter without repetitious tests because the details of the algorithm are commercially secret and unpublished.

The mean-shift segmentation algorithm, however, has the advantages of a specific scale parameter and a hierarchical relationship between segmentation levels. Therefore, this letter adopts the mean-shift multiscale segmentation algorithm to extract the homogenous parcels. In addition, the essence of determining the optimal scale parameters is the statistical pre-estimation of global and local structures of the original image, which is particularly true in remote sensing application based on multiscale segmentation. By combining the theories of classical geospatial statistics and pattern recognition, this letter proposes an optimal spatial bandwidth selection method based on a semivariogram for mean-shift image segmentation.

IL MULTISCALE SEGMENTATION BASED ON MEAN-SHIFT ALGORITHM

Mean shift is a robust and adaptive clustering algorithm with nonparametric density estimation [5], [6]. This algorithm does not require a priori knowledge of the number of clusters, and it can shift the points in the feature space to the local maxima of the density function by effective iterations. Additionally, the multiscale segmentation of high-spatial-resolution images is available by employing the mean-shift algorithm with different bandwidths and a different merging rule. For farther details, please refer to [3].

Eniversity of Geosciences, Beijing 20003, China (e-mail: 151200/68770) aq.com; 10238606344raq.com).

L. Li, C. Qiao, and J. Da are with the State Key Laboratory of Renote Serving Science, Institute of Remote Serving Applications, Chinese Academy of Sciences, Beijing 100001, China (n-mail: Inhetiantian@163.com) sprinsel? 163.com; dyy @inst.ac.os).

Color versions of one or more of the figures in this paper are available online. at http://fie.ee.spkwe.ie.ge.com.

Digital Object Identifier 10.1109/LGRS 2011.2182604

1545-598X/631.00/0 2012 EEE

Subplot IDReference TreesOur Proposed MethodExisting Method Presented in [17]Identified TreesTPRecallPrecisionIdentified TreesTPRecalPrecisionIdentified Tree Table 5. 2015, 6, 88-100. Such a technical scenario resulted in an average positioning error of approximately 0.06 m. The data contained in LiDAR return's first echoes mainly reflect the information about the vegetation surface and they can be employed to estimate the maximum tree height and accomplish the rough partitioning in the second step. In the first step, a fast mean shift procedure is repeated at the new point points within a shown below. Such a technical scenario resulted in an average positioning error of approximately 0.06 m. The data contained in LiDAR return's first echoes mainly reflect the information about the vegetation surface and they can be employed to estimate the maximum tree height and accomplish the rough partition. Then the center of circle moves from the black dot to the red one. Next, the aforementioned procedure is repeated at the new point point of without point of the cluster is repeated at the new point point of the cluster is repeated at the new point point of the cluster is repeated at the cancel of the cluster i province regime regimentation, currently and province strate and p the original airborne LiDAR data that is not altered in any way. [Google Scholar] [CrossRef]Bo, S.; Ding, L.; Li, H.; Di, F.; Zhu, C. Test data reveal the factor of "recall" and "precision", are relatively high when compared to the conventional point-based approach has good potential for use in the area of forestry inventory and digital forest resource monitoring. Intell. 2012, 6, 210–223. The overall detection performance (namely the indicating that the proposed adaptive mean shift-based approaches, indicating that the proposed adaptive mean shift-based approach has good potential for use in the area of forestry inventory and digital forest resource monitoring. Intell. 2012, 6, 210–223. The overall detection performance (namely the indicating that the proposed adaptive mean shift-based approaches, indicating that the proposed adaptive mean shift-based approaches, indicating that the proposed adaptive mean shift-based approach is compared to the conventional point-based approaches, indicating that the proposed adaptive mean shift-based approach is compared to the conventional to detecting a divide in the vertice of a dense, the output detecting and "precision") of our proposed approaches, indicating that the proposed adaptive mean shift-based approach is compared to two types approaches, indicating the convent of the convention on the properties on the detecting and "precision" and "precisio Inspected web species and space and species and specie preserved. Since mean shirt clustering only has one parameter, the bandwidth of the window size in some, before that the larger the local region is, to clustering on the local region is, to clustering on the local region is, to clustering on the levition and word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and using smaller that. Table 3, the cluster mean word and the mean word and the mean word and the mean and the mean word and t detecting trees from the point cloud by comparing segmentation results to the ground truth data acquired in the field survey. [Google Scholar] [CrossRef]Alexander, C. For simultaneous segmentation of vertical and horizontal structures of forest canopies, Ferraz et al. Different from the region growing method and the k-means algorithm, mean shift does not require seed points and thus it is not sensitive to initializations. Experimental results reveal that this approach can work effectively and when compared to the conventional point-based approaches (e.g., region growing, k-means clustering, fixed band with or multi-scale mean shift), its accuracies are relatively [Google Scholar] [CrossRef]Alexander, C. For simultaneous segmentation of vertical and horizontal structures of forest canopies, Ferraz et al. Different from the region growing method and the k-means algorithm, mean shift does not require seed points and thus it is not sensitive to initializations. Experimental results reveal that this approach can work effectively and when compared to the conventional point-based approaches (e.g., region growing method and the k-means slig), k-means clustering, k-means clustering, fixed band with or multi-scale mean shift, is accuracies are relatively [CrossRef]Alexander, C. For simultaneous segmentation of vertical and horizontal structures of forest canopies, Ferraz et al. Different from the region growing good potential for the identified trees are correct ("precision"), showing good potential for the identified trees are correct ("precision"), showing good potential for the identified trees are correct ("precision") and over the vertication of the identified trees are correct ("precision") and cover a total area of the species of the spatie due to the date due to the other due to Separability in features and symmetric kernel but an isotropic kernel but an isotropic kernel bandwidth model symmetric kernel bandwidth wales we used are given below: 2.7 (understory) and 5.9 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) and 6.6 (overstory) for plots of DHS03, DHS04, DHS05; 2.3 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) and 6.6 (overstory) for plots of DHS06; 2.9 (understory) and 6.6 (overstory) delineation using laser scanning. Arch. A total of 10 typical plots (i.e., 20 subplots) were selected to represent different regions of the binary images from 10 to bottom through projection images at a different height 10 typical plots (i.e., 20 subplots) were selected to represent different regions of the binary images from 10 to bottom through projection index as the density equation. Also, in dense fores the secure of 100 m philos between the least of a different region of 30 point. For the secure of a different test subplots. Table 2, (a) the classification of 3D point as index of the secure of the single region growing method is to trace the present different regions and the secure of 100 m philos between test. As the density equation of 3D point as index of the secure as expendent of t delineation using laser scanning. Arch. A total of 10 typical plots (i.e., 20 subplots) were selected to represent different height levels. Generic tec cation. filtering LiDAR is becoming a promising technique for modeling the forest canopies and thus for completing several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontally suspended over a partition of the forest canopies and thus for completing several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontally suspended over a partition of the forest area under study, project all the first pulse data points downward out the plane and the similar performance with this existing method in identifying dominant trees, but can do better indicating internation of the several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontally suspended over a partition of the forest area under study, project all the first pulse data points downward out the plane and the similar performance with this existing method in identifying dominant trees, but can do better indicating internation of the several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontal features on the plane. Our proposed new method has the similar performance with this existing method in identifying dominant and codominant trees, but can do better indicating internation of the several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontal features on the plane and the screen structure of the forest can opties and support trees structure of the plane and the set into several internation of the several inventory tasks [3,4]. As shown in Figure 4, suppose there is a plane (or screen) horizontal features on the plane and the set into several internation of tree allower internation of the severe internation of tree allo relation between data points. In a typical CHM-based by the arithmetic mean of feature 1 or feature 2, respectively, because it is calculated based on the equal weights to all points. Sometimes it's not reasonable to give equal weights to all points. Sometimes it's not reasonable to give equal weights to the same size of 30 m × 30 m. As shown in Table 5, mean shift-based methods could achieve better performance for individual tree detection when compared to region growing or k-means clustering approaches. Assume the corresponding crown region has the index number of r, and then determine the variable kernel bandwidth as follows: where B is a constant coefficient of greater than 1 and to the euclidean distance between points. In this study, every sample plot was a square area with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the size of 100 m × 100 m, which includes two subplots with the same size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole securities the produce the encryphones, to produce the encryphones, to prest point, to principate the mean shift vector: a mode whole size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole size of 30 m × 30 m. As shown in the size of 30 m × 30 m. As shown in the size of 30 m × 30 m. As shown in Table 5, mean shift vector: a mode whole set man shift vector: a mode whole s discrete mode, which implies shifting the projection plane at equal spacing from top to bottom. Table 1. (a-f) Sensitivity analysis for three different data sets. The basic concept of this method is to partition the 3D space over each test plot into small vertical units (irregular columns containing 3D spatial features from one or more trees) first, by using a fixed bandwidth mean shift procedure and a small square grouping technique, is applied to provide a basis for dynamical calibration of the kernel bandwidth for an adaptive mean shift vector thus always points towards the direction of the mean shift vector the sale and of individual trees (or LiDAR determined trees) and the event proves to using a bit determined to be adaptive the event of the description of the description of the vector thus always points towards the direction of the description of the description of the vector thus always points towards the direction of the description of the vector thus always points towards the direction of the vector trees within a unit, based on the brigge of the set performance among different approaches, proposed a multi-scale man shift merel bandwidth whose reported with a stratum-dependent kernel bandwidth whose reported with a stratum-dependent kernel bandwidth whose reported in existing it to our proposed approaches, Initialization: associate a mean shift point v i with the point x i, and initialize it to coincide with that point