

Three-dimensional modeling method and apparatus

5 Cross reference to related applications

The present application claims priority of the Chinese Patent Application No. 201510117117.2, filed on 17 March 2015 and entitled “THREE-DIMENSIONAL MODELING METHOD AND APPARATUS”, which is incorporated herein by reference in its entirety.

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Technical Field

The present application relates to the technical field of computers and, in particular to a three-dimensional modeling method and apparatus.

15 Background Art

In the field of computer vision, three-dimensional modeling for an object based on 2D image data has been widely applied. Information included in a 2D image, such as light streams, shadows and texture, can all be used for determining the depth of an object, thereby constructing a three-dimensional model.

20 However, the amount of calculation in acquiring a depth value by means of information, such as light streams, shadows and texture, is extremely large. Taking a mobile device such as a mobile phone as an example, a built-in processor has difficulty in performing real-time calculation on the information above, thereby limiting the application of some services or functions based on three-dimensional modeling on the mobile device such as a mobile phone.

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On the other hand, if the amount of processing of information is reduced to accommodate the computation capability of the processor of a mobile device, such as a mobile phone, the accuracy of three-dimensional modeling may possibly be reduced.

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Summary of the Invention

The embodiments of the present application provide a three-dimensional modeling method and apparatus, in order to solve the problem of three-dimensional modeling in the prior art requiring a great amount of calculations which are not easily implemented on a mobile device, such as a mobile phone.

A three-dimensional modeling method is provided in an embodiment of the present application, comprising:

respectively capturing a first image in a first direction and a second image in a second direction of an object;

10 acquiring a plurality of feature points on the first image;

determining matching feature points on the second image that correspond to the plurality of feature points;

calculating similarities between the plurality of feature points and the corresponding matching feature points, and depth values of the plurality of feature points;

15 performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points; and

performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

A three-dimensional modeling method is provided in another embodiment of the present application, comprising:

capturing images of an object in a plurality of directions;

25 acquiring a plurality of feature points on one of the plurality of images;

respectively determining matching feature points on the remaining images that correspond to the plurality of feature points;

respectively calculating similarities between the plurality of feature points and the corresponding matching feature points on the remaining images and depth values of the matching feature points;

30 calculating depth values of the plurality of feature points according to direction

information about the plurality of images and the depth values of the matching feature points;

performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points on the remaining images; and

performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

A three-dimensional modeling apparatus is provided in an embodiment of the present application, comprising:

an image capture module for respectively capturing a first image in a first direction and a second image in a second direction of an object;

a feature detection module for acquiring a plurality of feature points on the first image;

a matching module for determining matching feature points on the second image that correspond to the plurality of feature points;

a calculation module for calculating similarities between the plurality of feature points and the corresponding matching feature points and depth values of the plurality of feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points; and

a model generation module for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

A three-dimensional modeling apparatus is provided in another embodiment of the present application, comprising:

an image capture module for capturing images of an object in a plurality of directions;

a feature detection module for acquiring a plurality of feature points on one of the plurality of images;

a matching module for respectively determining matching feature points on the

remaining images that correspond to the plurality of feature points;

a calculation module for respectively calculating similarities between the plurality of feature points and the corresponding matching feature points on the remaining images and depth values of the matching feature points, calculating depth values of the plurality of feature points according to direction information about the plurality of images and the depth values of the matching feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points on the remaining images; and

a model generation module for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

The embodiments of the present application provide a three-dimensional modeling method and apparatus. The method involves: capturing images in at least two different directions of an object, acquiring feature points on one of the images, and determining corresponding matching feature points on the images in the remaining directions, so as to calculate depth values of the various feature points; and performing weighted calculation on the depth values of the various feature points with similarities between the feature points and the matching feature points, and performing three-dimensional modeling on the object by utilizing the depth values of the various feature points after being subjected to the weighted calculation, so that the method effectively reduces the amount of calculation of three-dimensional modeling, is easily implemented on a mobile device, such as a mobile phone, and has a relatively high accuracy.

Brief Description of the Drawings

The drawings described herein are used to provide further understanding of the present application, which constitute a part of the present application. The illustrative embodiments of the present application and the description thereof are used to explain the present application and do not constitute a limitation to the present application. In the drawings:

Fig. 1 is a schematic flow chart of three-dimensional modeling provided in an embodiment of the present application;

Fig. 2 is a flow chart of acquiring feature points provided in an embodiment of the present application;

5 Fig. 3 is a flow chart of determining matching feature points on a second image that correspond to feature points on a first image provided in an embodiment of the present application;

Fig. 4 is a flow chart of calculating similarities between feature points and matching feature points provided in an embodiment of the present application;

10 Fig. 5 is a schematic diagram of coordinates when a human face rotates by an angle α of elevation clockwise from the Y axis to the Z axis on a YZ plane provided in an embodiment of the present application;

Fig. 6 is a schematic diagram of coordinates when a human face rotates by an angle β of deflection anti-clockwise from the X axis to the Z axis on an XZ plane provided in an embodiment of the present application;

15 Fig. 7 is a schematic diagram of coordinates when a human face rotates by an angle γ of rotation anti-clockwise from the X axis to the Y axis on an XY plane provided in an embodiment of the present application;

Fig. 8 is a schematic diagram of initial feature points acquired on a first image provided in an embodiment of the present application;

20 Fig. 9 is a schematic diagram of screening initial feature points utilizing a first grid provided in an embodiment of the present application;

Fig. 10 is a schematic diagram of searching, on a second image, for a matching feature point A' corresponding to a feature point A on a first image provided in an embodiment of the present application;

25 Fig. 11 is a schematic flow chart of three-dimensional modeling provided in another embodiment of the present application;

Fig. 12 is a schematic diagram of a front face image, a front face elevated image, a front face left-inclined image and a front face right-inclined image of a human face captured utilizing a mobile phone provided in another embodiment of the present application; and

Fig. 13 is a schematic diagram of modules of a three-dimensional modeling apparatus provided in an embodiment of the present application.

Detailed Description of Embodiments

5 In order to make the object, technical solution and advantages of the present application more clear, the technical solution of the present application will be described clearly and completely in conjunction with particular embodiments of the present application and corresponding drawings hereinafter. Obviously, the described embodiments are merely some embodiments of the present application rather than all embodiments. Based on the embodiments of the present application, all other embodiments obtained by those of ordinary skill in the art without any inventive effort all fall within the scope of protection of the present application.

Fig. 1 is a process of three-dimensional modeling provided in an embodiment of the present application, specifically comprising the steps of:

15 S11: respectively capturing a first image in a first direction and a second image in a second direction of an object.

In the embodiment of the present application, the object comprises various objects having a three-dimensional construction, e.g., a human face. The first image and the second image may have specific differences in terms of angle of elevation, angle of deflection and angle of rotation from each other. Both the first image and the second image contain RGB channels (i.e., the first image and the second image are color images), and the pixels of both images are above 392×440 . Of course, in substituted embodiments, the first image and the second image may also be grayscale maps.

25 In cooperation with the reference to Figs. 5-7, taking a human face as an example, the angle of elevation, angle of deflection and angle of rotation as described in the present application are explained. Taking a front face image of the human face as a reference, the X axis is defined as a horizontal axis of the front face of the human face, the Y axis is defined as a vertical axis of the front face of the human face, while the Z axis is perpendicular to the X axis and the Y axis, and the Z axis indicates the depth of features on the human face. As shown in Fig. 5, the angle

α of elevation is an angle for the human face to rotate clockwise from the Y axis to the Z axis on the YZ plane; as shown in Fig. 6, the angle β of deflection is an angle for the human face to rotate anti-clockwise from the X axis to the Z axis on the XZ plane; and as shown in Fig. 7, the angle γ of rotation is an angle for the human face to rotate anti-clockwise from the X axis to the Y axis on the XY plane.

S12: acquiring a plurality of feature points on the first image.

In the embodiment of the present application, the acquisition of the feature points may use the method of edge detection, so as to acquire points with significant changes in brightness in the first image. At the same time, image edge detection may greatly reduce the amount of processing of data, while retaining important structural attributes of the image, such as: discontinuity in depth, discontinuity in a surface direction, a material property change, a scenario illumination change.

With reference to Fig. 2, the steps of acquiring a plurality of feature points on the first image in step S12 are introduced, comprising:

S121: acquiring several initial feature points on the first image.

With reference to Fig. 8, the acquisition of the initial feature points (points indicated by cross marks in Fig. 8) herein uses the method of edge detection, and these initial feature points reflect feature profiles of edges, five sense organs on the human face.

S122: matching the first image with a pre-set first grid.

With reference to Fig. 9, in the embodiment of the present application, grid cells in the first grid are defined as squares, and of course the grid cells of the first grid may also be pre-defined as a plurality of shapes such as a triangle or a regular hexagon according to requirements. By means of this step, the several initial feature points on the first image may respectively fall into various grid cells of the first grid. It needs to be noted that in practical application scenarios, for example, in a mobile phone, the matching of the first image and the first grid is accomplished by operation processing of a data processor within the mobile phone, rather than being necessarily fed back visually.

S123: screening the several initial feature points until the numbers of initial feature points retained in grid cells of the first grid are equal, and determining them

as the feature points.

Still taking a human face as an example, since the features thereon are not distributed uniformly, when edge detection is utilized to acquire the initial feature points, the distribution of initial feature points in some regions on the first image may obviously be more intensive relative to the other regions. However, in applications of a mobile device, e.g., a mobile phone, performing depth calculation on all of these acquired initial feature points is obviously hard to achieve and unnecessary. Utilizing a first grid to screen these initial feature points may ensure that the initial feature points after screening can still reflect basic shapes of the human face, while the amount of calculation of data is greatly reduced.

It needs to be noted that although in this embodiment, the standard is to ensure that the numbers of initial feature points retained in the grid cells of the first grid are equal to screen the several initial feature points, in some substituted embodiments, other screening standards may also be used, for example: setting that the difference between numbers of initial feature points in various grid cells is not greater than a pre-determined threshold.

With reference to Fig. 10, in the embodiment of the present application, the specific screening process for the initial feature points is as follows: first, calculating the distances from the several initial feature points to centers of the grid cells of the first grid where these points are located, and then determining, in the grid cells of the first grid, an initial feature point closest to the center of the grid cell where this point is located as the feature point. In this process, a case where certain grid cells of the first grid are devoid of initial feature points may arise, and in this case, the centers of these grid cells (points indicated by hollow circles in Fig. 10) are determined as feature points therein.

Of course, in some substituted embodiments, the average values of the coordinates of the initial feature points of the various grid cells may also be solved, and the points corresponding to the obtained average coordinate values are taken as feature points retained in the corresponding grid cells in the first grid.

S13: determining matching feature points on the second image that correspond to the plurality of feature points.

In cooperation with reference to Figs. 9 and 10, still taking a human face as an example, assuming that a feature point A at the tip of the nose is acquired on the first image, then the objective of step 13 is to search for the position where the tip of the nose is located (i.e., the coordinates of a matching feature point A') on the second image, and then establish a correlation between a plurality of feature points and matching feature points on the first image and the second image according to this method.

With reference to Fig. 3, the specific steps of determining matching feature points in step S13 are introduced, comprising:

S131: determining, according to direction information about the first image and the second image, pre-matching feature points on the second image that correspond to the plurality of feature points.

The direction information about the first image and the second image may be determined by a plurality of means. Taking shooting a human face by a mobile phone as an example:

in one embodiment, if a camera is utilized to continuously capture the first image and the second image, then direction information about the first image and the second image may be directly determined by means of a movement sensor (such as a gyroscope, a compass, an accelerometer) built in the mobile phone.

Continuously referring to Figs. 5-7, in another embodiment, if a camera is utilized to respectively capture the first image and the second image, direction information about the first image and the second image may be determined, for example, by means of a posture measurement algorithm. Specifically, assuming that the rotation of the human face is determined based on an assumed rotation center C, by means of the existing biological anatomy knowledge, the coordinates C_X , C_Y and C_Z of the rotation center can be roughly estimated, while the distance r from the rotation center C to the center of the two eyes on the human face can be used to measure the angle α of elevation, angle β of deflection and angle γ of rotation described above. At the same time, based on the following values which can be roughly estimated: the coordinates (U_X , U_Y) on the first image and the coordinates (V_X , V_Y) on the second image of the position of the center of the two eyes of the

human face and the depth E_Z of the position of the center of the two eyes when the human face is a front face image, the calculation formulas for the angle α of elevation, angle β of deflection and angle γ of rotation can be obtained as follows:

$$\alpha = \sin^{-1}\left(\frac{V_Y - C_Y}{r}\right) - \sin^{-1}\left(\frac{U_Y - C_Y}{r}\right),$$

angle of elevation

5 where $r = \sqrt{(U_Y - C_Y)^2 + (E_Z - C_Z)^2}$;

$$\beta = \sin^{-1}\left(\frac{V_X - C_X}{r}\right) - \sin^{-1}\left(\frac{U_X - C_X}{r}\right),$$

angle of deflection

where $r = \sqrt{(U_X - C_X)^2 + (E_Z - C_Z)^2}$; and

$$\gamma = \tan^{-1}\left(\frac{V_Y}{-V_X}\right) - \tan^{-1}\left(\frac{U_Y}{-U_X}\right).$$

angle of rotation

S132: matching the second image with a pre-set second grid.

10 With reference to Fig. 10, similar to the first grid, in the embodiment of the present application, squares are used as the grid cells of the second grid. In addition, the area of the grid cells of the second grid is set as being smaller than that of the grid cells of the first grid (that is: in the first grid and the second grid of the same area, more grid cells relative to the first grid are divided in the second grid cells), so as to improve the accuracy of determining the matching feature points in the following step.

15 S133: determining centers of grid cells of the second grid where the pre-matching feature points are located as the corresponding matching feature points.

20 S14: calculating similarities between the plurality of feature points and the corresponding matching feature points and depth values of the plurality of feature points.

25 With reference to Fig. 4, the specific steps of calculating similarities between the feature points and the matching feature points in step S14 are introduced, comprising:

S141: acquiring color values of all pixels within the grid cells of the first grid

where the plurality of feature points are located.

S142: acquiring color values of all pixels within the grid cells of the second grid where the plurality of matching feature points are located.

5 The “color values” as described in step S141 and step S142 refer to color values of pixels in three channels, R, G and B.

S143: respectively calculating average color value errors between all the pixels within the grid cells of the first grid where the plurality of feature points are located and those within the grid cells of the second grid where the corresponding matching feature points are located, and determining, according to the average color value errors, the similarities between the plurality of feature points and the corresponding matching feature points. The calculation formula for the average color value error S is as follows:

$$S = \lambda \sum_i \sum_j \frac{R(F_{ij}, S_{ij}) + G(F_{ij}, S_{ij}) + B(F_{ij}, S_{ij})}{3\|F\|},$$

where i and j are coordinate values of pixels within grid cells of the first grid, F_{ij} is a feature point where a pixel is at a position of (i, j), S_{ij} is a matching feature point where a pixel is at a position of (i, j), R(F_{ij}, S_{ij}) is an absolute value of the difference between the points F_{ij} and S_{ij} on the R channel, G(F_{ij}, S_{ij}) is an absolute value of the difference between the points F_{ij} and S_{ij} on the G channel, B(F_{ij}, S_{ij}) is an absolute value of the difference between the points F_{ij} and S_{ij} on the B channel, ||F|| is the area of the grid cells of the first grid (i.e., the pixel values contained within the grid cells of the first grid), and λ is a constant, and the calculation formula is as follows:

$$\lambda = (1 + \frac{|sp.x - fp.x|}{\max X} + \frac{|sp.y - fp.y|}{\max Y}) ,$$

where maxX and maxY are the farthest distance values (in pixels) away from the corresponding feature points within one grid cell of the first grid; and fp.x and fp.y are coordinate values of the feature points, and sp.x and sp.y are coordinate values of the matching feature points.

The similarities between the feature points and the matching feature points are

determined according to the values of average color value errors obtained through calculation, and the greater the average color value error, the lower the similarity between a feature point and a matching feature point. The feature points can be screened by utilizing the similarities between the feature points and the matching feature points. Specifically:

5 first determining a discard threshold according to the similarities between the plurality of feature points and the corresponding matching feature points, and then screening out, from the plurality of feature points, feature points of which similarities with the corresponding matching feature points are less than the discard
10 threshold. When the similarity between a feature point and the corresponding matching feature point is less than the discard threshold, it is indicated that the matching feature point at this time is not a point actually corresponding to the feature point on the second image, and it is determined that the feature point matching has failed at this time, while these feature points which have failed in
15 matching do not participate in subsequent three-dimensional modeling calculation.

Continuously referring to Figs. 5-7, the depth value of a feature point is obtained by synthesizing the depths of the matching feature point at an angle of elevation, angle of deflection and angle of rotation, and performing calculation in combination with the depth values at these different angles may enable the angles to
20 have a greater influence, and improve the accuracy of the depth value calculation result. Specifically,

the depth value of a feature point at an angle of elevation is

$$Z_{\alpha} = C_z - \frac{(V_y - C_y) - (U_y - C_y) \cos \alpha}{\sin \alpha};$$

the depth value of a feature point at an angle of deflection is

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$$Z_{\beta} = C_z - \frac{(V_x - C_x) - (U_x - C_x) \cos \beta}{\sin \beta}; \text{ and}$$

the depth value of a feature point at an angle of rotation is

$$Z_{\gamma} = U_y \cos \gamma - U_x \sin \gamma; \text{ and}$$

the calculation formula of the depth value Z is obtained by synthesizing Z_{α} , Z_{β} and Z_{γ} as follows:

$$Z = Z_{\alpha} \frac{\sin\alpha}{\sin\alpha + \sin\beta + \sin\gamma} + Z_{\beta} \frac{\sin\beta}{\sin\alpha + \sin\beta + \sin\gamma} + Z_{\gamma} \frac{\sin\gamma}{\sin\alpha + \sin\beta + \sin\gamma}.$$

S15: performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points.

5 Specifically, a formula to perform weighted calculation on depth values of the feature points is as follows:

weighted depth value = depth value $Z \times$ (sum of the average color value errors of all the feature points except the feature points to be subjected to the weighted calculation / sum of the average color value errors of all the feature points).

10 S16: performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

In the embodiment of the present application, the three-dimensional modeling adopts a triangular grid, and vertices of the triangular grid are respective feature points. Of course, in other substituted embodiments, a method, e.g., polygon modeling, may also be used.

Fig. 11 is a process of three-dimensional modeling provided in another embodiment of the present application, specifically comprising the steps of:

S21: capturing images of an object in a plurality of directions.

20 S22: acquiring a plurality of feature points on one of the plurality of images.

S23: respectively determining matching feature points on the remaining images that correspond to the plurality of feature points.

S24: respectively calculating similarities between the plurality of feature points and the corresponding matching feature points on the remaining images and depth values of these matching feature points.

S25: calculating depth values of the plurality of feature points according to direction information about the plurality of images and the depth values of the matching feature points.

30 S26: performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points

and the corresponding matching feature points on the remaining images.

S27: performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

5 With reference to Fig. 12, still taking using a mobile phone to perform three-dimensional modeling on a human face as an example, a first image of the object in a reference direction and a second image, a third image and a fourth image in three directions deviating from the reference direction are captured. In addition, the first image, second image, third image and fourth image are respectively a front
10 face image, a front face elevated image, a front face left-inclined image and a front face right-inclined image of the human face. In this way, matching feature points corresponding to the feature points captured on the front face image of the human face can all be substantially found on at least one of the front face elevated image, front face left-inclined image and front face right-inclined image, and therefore the
15 calculation process for the depth value of each feature point on the front face image of the human face has synthesized the factor of angles, thereby ensuring that the depth value obtained through calculation has a higher accuracy.

 In this embodiment, the three-dimensional modeling method in the present application is described taking capturing four images as an example, but this is not a
20 limitation, and according to specific environments of application and the difference of precision requirements, the number of captured images may also be adjusted adaptively. In addition, since this embodiment does not involve improvements for other steps of the last embodiment, those steps will not be described otherwise herein, and the method and apparatus described in detail in the last embodiment
25 may both be applied in this embodiment in its entirety.

 Fig. 13 is a schematic diagram of modules of a three-dimensional modeling apparatus provided in an embodiment of the present application, specifically comprising:

 an image capture module 31 for respectively capturing a first image in a first
30 direction and a second image in a second direction of an object;

 a feature detection module 32 for acquiring a plurality of feature points on the

first image;

a matching module 33 for determining matching feature points on the second image that correspond to the plurality of feature points;

5 a calculation module 34 for calculating similarities between the plurality of feature points and the corresponding matching feature points and depth values of the plurality of feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points; and

10 a model generation module 36 for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

In the embodiment of the present application, the feature detection module 32 is specifically used for: acquiring several initial feature points on the first image, matching the first image with a pre-set first grid, and screening the several initial
15 feature points until the numbers of initial feature points retained in grid cells of the first grid are equal, and determining them as the feature points.

In the embodiment of the present application, the feature detection module 32 comprises a calculation unit 321, and the calculation unit 321 is specifically used for: calculating distances from the several initial feature points to centers of the grid
20 cells of the first grid where these points are located; and the feature detection module 32 is specifically used for determining, in the grid cells of the first grid, an initial feature point closest to the center of the grid cell where this point is located as the feature point.

In the embodiment of the present application, the feature detection module 32
25 is further used for: when a grid cell of the first grid is devoid of the initial feature points, determining the center of the grid cell as a feature point within the grid cell.

In the embodiment of the present application, the matching module 33 is specifically used for: determining, according to direction information about the first image and the second image, pre-matching feature points on the second image that
30 correspond to the plurality of feature points, matching the second image with a pre-set second grid, and determining centers of grid cells of the second grid where

the pre-matching feature points are located as the corresponding matching feature points.

In the embodiment of the present application, the calculation module 34 is specifically used for: acquiring color values of all pixels within the grid cells of the first grid where the plurality of feature points are located, acquiring color values of all pixels within the grid cells of the second grid where the plurality of matching feature points are located, and respectively calculating average color value errors between all the pixels within the grid cells of the first grid where the plurality of feature points are located and those within the grid cells of the second grid where the corresponding matching feature points are located, and determining, according to the average color value errors, the similarities between the plurality of feature points and the corresponding matching feature points.

In the embodiment of the present application, a formula for performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points is as follows:

weighted depth value = depth value $Z \times$ (sum of the average color value errors of all the feature points except the feature points to be subjected to the weighted calculation / sum of the average color value errors of all the feature points).

In the embodiment of the present application, the area of the grid cells of the second grid is smaller than the area of the grid cells of the first grid.

In the embodiment of the present application, the apparatus further comprises a screening module 35 for, prior to performing the weighted calculation on the depth values of the plurality of feature points, screening the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points.

In the embodiment of the present application, the screening module 35 is specifically used for: first determining a discard threshold according to the similarities between the plurality of feature points and the corresponding matching feature points, and then screening out, from the plurality of feature points, feature points of which similarities with the corresponding matching feature points are

greater than the discard threshold.

In the embodiment of the present application, the model generation module 36 performs three-dimensional modeling on the object by adopting a triangular grid.

Continuously referring to Fig. 13, in another embodiment of the three-dimensional modeling apparatus of the present application, the apparatus comprising:

an image capture module 31 for capturing images of an object in a plurality of directions;

a feature detection module 32 for acquiring a plurality of feature points on one of the plurality of images;

a matching module 33 for respectively determining matching feature points on the remaining images that correspond to the plurality of feature points;

a calculation module 34 for respectively calculating similarities between the plurality of feature points and the corresponding matching feature points on the remaining images and depth values of the matching feature points, calculating depth values of the plurality of feature points according to direction information about the plurality of images and the depth values of the matching feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points on the remaining images; and

a model generation module 36 for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

In the embodiment of the present application, the image capture module 31 is specifically used for: capturing a first image of the object in a reference direction and a second image, a third image and a fourth image in three directions deviating from the reference direction.

In the embodiment of the present application, the object comprises: a human face.

In the embodiment of the present application, the first image, second image, third image and fourth image are respectively a front face image, a front face

elevated image, a front face left-inclined image and a front face right-inclined image of the human face.

5 The embodiments of the present application provide a three-dimensional modeling method and apparatus. The method involves: capturing images in at least two different directions of an object, acquiring feature points on one of the images, and determining corresponding matching feature points on the images in the remaining directions, so as to calculate depth values of the various feature points; and performing weighted calculation on the depth values of the various feature points with similarities between the feature points and the matching feature points, and performing three-dimensional modeling on the object by utilizing the depth values of the various feature points after being subjected to the weighted calculation, so that the method effectively reduces the amount of calculation of three-dimensional modeling, is easily implemented on a mobile device such as a mobile phone, and has a relatively high accuracy.

15 It shall be understood by those skilled in the art that the embodiments of the present invention may be provided as methods, systems, or computer program products. Therefore, the present invention may adopt hardware only embodiments, software only embodiments, or embodiments with a combination of software and hardware. Moreover, the present invention may adopt the form of a computer program product that is implemented on one or more computer-usable storage mediums (including but not limited to a disk memory, a CD-ROM, an optical memory, and the like) that include computer-usable program code.

25 The present invention is described by referring to the flow charts and/or block diagrams of the methods, devices (systems) and computer program products according to the embodiments of the present invention. It should be understood that each flow and/or block in the flow charts and/or block diagrams and any combination of the flows and/or blocks in the flow charts and/or block diagrams may be implemented with computer program instructions. These computer program instructions may be provided to a processor of a general-purpose computer, a dedicated-purpose computer, an embedded processor or other programmable data processing devices to generate a machine, so that the instructions executed by the

processor of a computer or other programmable data processing devices generate a means for implementing the functions specified in one or more flows of the flowcharts and/or one or more blocks of the block diagrams.

These computer program instructions may also be stored in a computer-readable memory which may guide a computer or other programmable data processing devices to operate in a specific manner, so that the instructions stored in the computer-readable memory generate an article of manufacture including an instruction means which can implement the functions specified in one or more flows of the flowcharts and/or one or more blocks of the block diagrams.

These computer program instructions may also be loaded in a computer or other programmable data processing devices, so that a series of operation steps are executed by the computer or other programmable devices to realize computer-implemented processing and thus the instructions executed by the computer or other programmable devices provide steps for implementing the functions specified in one or more flows of the flow charts and/or one or more blocks of the block diagrams.

In a typical configuration, a computing device includes one or more processors (CPUs), an input/output interface, a network interface and a memory.

The memory may include a non-permanent memory, a random access memory (RAM) and/or a non-volatile memory (such as a read-only memory (ROM) or flash memory (flash RAM)) and so on in computer-readable mediums. The memory is an example of the computer-readable mediums.

The computer-readable mediums comprise permanent and non-permanent, removable and non-removable media and may realize information storage with any method or technology. The information may be a computer-readable instruction, data structure, program module or other data. The examples of computer storage mediums include but are not limited to a phase change memory (PRAM), static random access memory (SRAM), dynamic random access memory (DRAM), other types of random access memories (RAMs), read-only memory (ROM), electrically erasable programmable read-only memory (EEPROM), flash memory or other memory technologies, compact disk read-only memory (CD-ROM), digital versatile

disc (DVD) or other optical storage, magnetic cassette tape, magnetic tape and magnetic disk storage or other magnetic storage devices or any other non-transmission mediums, which can be used to store information that can be accessed by a computing device. According to the definition herein, the
5 computer-readable mediums do not include transitory computer-readable media, such as modulated data signals and carriers.

It also needs to be noted that the terms “comprising”, “including” or any other variations thereof are intended to cover non-exhaustive inclusion, so that processes, methods, merchants or devices including a series of elements not only include those
10 elements but also include other elements which are not listed expressly or include all inherent elements of such processes, methods, merchants or devices. If no more limitations are made, an element limited by “comprising a/an...” does not exclude other identical elements existing in the process, the method, the merchant, or the device which includes the element.

15 It shall be understood by those skilled in the art that the embodiments of the present application may be provided as methods, systems or computer program products. Thus, the present application may adopt full hardware embodiments, full software embodiments, or embodiments which combine software and hardware. Moreover, the present application may adopt the form of a computer program
20 product that is implemented on one or more computer-usable storage mediums (including but not limited to a disk memory, a CD-ROM, an optical memory, and the like) that include computer usable program codes.

The foregoing is merely the embodiments of the present application and is not intended to limit the present application. For those skilled in the art, various changes
25 and variations may be made to the present application. Any modifications, equivalent replacements, improvements made within the spirit and principles of the present application shall all be contained in the scope of the claims of the present application.

Claims

1. A three-dimensional modeling method, characterized in that the method comprises:

5 respectively capturing a first image in a first direction and a second image in a second direction of an object;

 acquiring a plurality of feature points on the first image;

 determining matching feature points on the second image that correspond to the plurality of feature points;

10 calculating similarities between the plurality of feature points and the corresponding matching feature points, and depth values of the plurality of feature points;

 performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points; and

15 performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

20 2. The method of claim 1, characterized in that the acquiring a plurality of feature points on the first image specifically comprises:

 acquiring several initial feature points on the first image;

 matching the first image with a pre-set first grid; and

25 screening the several initial feature points until the numbers of initial feature points retained in grid cells of the first grid are equal, and determining them as the feature points.

3. The method of claim 2, characterized in that the screening the several initial feature points until the numbers of initial feature points retained in grid cells of the first grid are equal and determining the feature points specifically comprises:

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calculating distances from the several initial feature points to centers of the grid cells of the first grid where these points are located; and

determining, in the grid cells of the first grid, an initial feature point closest to the center of the grid cell where this point is located as the feature point.

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4. The method of claim 2 or 3, characterized in that the method further comprises:

if a grid cell of the first grid is devoid of the initial feature point, then determining the center of the grid cell as a feature point within the grid cell.

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5. The method of claim 3, characterized in that the determining matching feature points on the second image that correspond to the plurality of feature points specifically comprises:

determining, according to direction information about the first image and the second image, pre-matching feature points on the second image that correspond to the plurality of feature points;

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matching the second image with a pre-set second grid; and

determining centers of grid cells of the second grid where the pre-matching feature points are located as the corresponding matching feature points.

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6. The method of claim 5, characterized in that the calculating similarities between the plurality of feature points and the corresponding matching feature points specifically comprises:

acquiring color values of all pixels within the grid cells of the first grid where the plurality of feature points are located;

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acquiring color values of all pixels within the grid cells of the second grid where the plurality of matching feature points are located; and

respectively calculating average color value errors between all the pixels within the grid cells of the first grid where the plurality of feature points are located and those within the grid cells of the second grid where the corresponding matching feature points are located, and determining, according to the average color value

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errors, the similarities between the plurality of feature points and the corresponding matching feature points.

7. The method of claim 6, characterized in that a formula for performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points is as follows:

weighted depth value = depth value $\times$ (sum of the average color value errors of all the feature points except the feature points to be subjected to the weighted calculation / sum of the average color value errors of all the feature points).

8. The method of any one of claims 5 to 7, characterized in that the area of the grid cells of the second grid is less than the area of the grid cells of the first grid.

9. The method of claim 1, characterized in that the method further comprises:

prior to performing the weighted calculation on the depth values of the plurality of feature points, screening the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points.

10. The method of claim 9, characterized in that the screening the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points specifically comprises:

determining a discard threshold according to the similarities between the plurality of feature points and the corresponding matching feature points; and

screening out, from the plurality of feature points, feature points of which similarities with the corresponding matching feature points are less than the discard threshold.

11. The method of claim 1, characterized in that the performing

three-dimensional modeling on the object adopts a triangular grid.

12. A three-dimensional modeling method, characterized in that the method comprises:

- 5 capturing images of an object in a plurality of directions;
- acquiring a plurality of feature points on one of the plurality of images;
- respectively determining matching feature points on the remaining images that correspond to the plurality of feature points;
- respectively calculating similarities between the plurality of feature points and
- 10 the corresponding matching feature points on the remaining images and depth values of the matching feature points;
- calculating depth values of the plurality of feature points according to direction information about the plurality of images and the depth values of the matching feature points;
- 15 performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points on the remaining images; and
- performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted
- 20 calculation.

13. The method according to claim 12, characterized in that the capturing images of an object in a plurality of directions specifically comprises:

- capturing a first image of the object in a reference direction and a second
- 25 image, a third image and a fourth image in three directions deviating from the reference direction.

14. The method according to claim 13, characterized in that the object comprises: a human face.

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15. The method according to claim 14, characterized in that the first image,

second image, third image and fourth image are respectively a front face image, a front face elevated image, a front face left-inclined image and a front face right-inclined image of the human face.

5 16. A three-dimensional modeling apparatus, characterized in that the apparatus comprises:

 an image capture module for respectively capturing a first image in a first direction and a second image in a second direction of an object;

 a feature detection module for acquiring a plurality of feature points on the
10 first image;

 a matching module for determining matching feature points on the second image that correspond to the plurality of feature points;

 a calculation module for calculating similarities between the plurality of feature points and the corresponding matching feature points and depth values of
15 the plurality of feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points; and

 a model generation module for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being
20 subjected to the weighted calculation.

 17. The apparatus of claim 16, characterized in that the feature detection module is specifically used for acquiring several initial feature points on the first image, matching the first image with a pre-set first grid, and screening the several
25 initial feature points until the numbers of initial feature points retained in grid cells of the first grid are equal, and determining them as the feature points.

 18. The apparatus of claim 17, characterized in that the feature detection module comprises a calculation unit which is specifically used for calculating
30 distances from the several initial feature points to centers of the grid cells of the first grid where these points are located; and

the feature detection module is specifically used for determining, in the grid cells of the first grid, an initial feature point closest to the center of the grid cell where this point is located as the feature point.

5 19. The apparatus of claim 17 or 18, characterized in that the feature detection module is further used for, when a grid cell of the first grid is devoid of the initial feature points, determining the center of the grid cell as a feature point within the grid cell.

10 20. The apparatus of claim 18, characterized in that the matching module is specifically used for determining, according to direction information about the first image and the second image, pre-matching feature points on the second image that correspond to the plurality of feature points, matching the second image with a pre-set second grid, and determining centers of grid cells of the second grid where
15 the pre-matching feature points are located as the corresponding matching feature points.

 21. The apparatus of claim 20, characterized in that the calculation module is specifically used for acquiring color values of all pixels within the grid cells of the first grid where the plurality of feature points are located, acquiring color values of
20 all pixels within the grid cells of the second grid where the plurality of matching feature points are located, and respectively calculating average color value errors between all the pixels within the grid cells of the first grid where the plurality of feature points are located and those within the grid cells of the second grid where
25 the corresponding matching feature points are located, and determining, according to the average color value errors, the similarities between the plurality of feature points and the corresponding matching feature points.

 22. The apparatus of claim 21, characterized in that a formula for performing
30 weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding

matching feature points is as follows:

weighted depth value = depth value $\times$ (sum of the average color value errors of all the feature points except the feature points to be subjected to the weighted calculation / sum of the average color value errors of all the feature points).

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23. The apparatus of any one of claims 20 to 22, characterized in that the area of the grid cells of the second grid is less than the area of the grid cells of the first grid.

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24. The apparatus of claim 16, characterized in that the apparatus further comprises a screening module for, prior to performing the weighted calculation on the depth values of the plurality of feature points, screening the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points.

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25. The apparatus of claim 24, characterized in that the screening module is specifically used for determining a discard threshold according to the similarities between the plurality of feature points and the corresponding matching feature points, and screening out, from the plurality of feature points, feature points of which similarities with the corresponding matching feature points are greater than the discard threshold.

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26. The apparatus of claim 16, characterized in that the model generation module performs three-dimensional modeling on the object by adopting a triangular grid.

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27. A three-dimensional modeling apparatus, characterized in that the apparatus comprises:

an image capture module for capturing images of an object in a plurality of directions;

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a feature detection module for acquiring a plurality of feature points on one of

the plurality of images;

a matching module for respectively determining matching feature points on the remaining images that correspond to the plurality of feature points;

5 a calculation module for respectively calculating similarities between the plurality of feature points and the corresponding matching feature points on the remaining images and depth values of the matching feature points, calculating depth values of the plurality of feature points according to direction information about the plurality of images and the depth values of the matching feature points, and performing weighted calculation on the depth values of the plurality of feature points according to the similarities between the plurality of feature points and the corresponding matching feature points on the remaining images; and

10 a model generation module for performing three-dimensional modeling on the object based on the depth values of the plurality of feature points after being subjected to the weighted calculation.

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28. The apparatus according to claim 27, characterized in that the image capture module is specifically used for capturing a first image of the object in a reference direction and a second image, a third image and a fourth image in three directions deviating from the reference direction.

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29. The apparatus according to claim 28, characterized in that the object comprises: a human face.

30. The apparatus according to claim 29, characterized in that the first image, second image, third image and fourth image are respectively a front face image, a front face elevated image, a front face left-inclined image and a front face right-inclined image of the human face.

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