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#### Abstract

Three dimensional scene analysis in an unconstrained and uncontrolled environment is the ultimate goal of computer vision. Explicit depth information about the scene is of tremendous help in segmentation and recognition of objects. Range image interpretation with a view of obtaining low-level features to guide mid-level and high-level segmentation and recognition processes is described. No assumptions about the scene are made and algorithms are applicable to any general single viewpoint range image. Low-level features like step edges and surface characteristics are extracted from the images and segmentation is performed based on individual features as well as combination of features. A high level recognition process based on superquadric fitting is described to demonstrate the usefulness of initial segmentation based on edges. A classification algorithm based on surface curvatures is used to obtain initial segmentation of the scene. Objects segmented using edge information are then classified using surface curvatures. Various applications of surface curvatures in mid and high level recognition processes are discussed. These include surface reconstruction, segmentation into convex patches and detection of smooth edges. Algorithms are run on real range images and results are discussed in detail.

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# RANGE IMAGE SEGMENTATION FOR 3-D OBJECT RECOGNITION 

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MS-CIS-88-32
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May 1988

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# UNIVERSITY OF PENNSYLVANIA THE MOORE SCHOOL OF ELECTRICAL ENGINEERING SCHOOL OF ENGINEERING AND APPLIED SCIENCE 

## RANGE IMAGE SEGMENTATION FOR 3-D OBJECT RECOGNITION

Alow Gupta<br>Philadelphia, Pennsylvania May 1988

A thesis presented to the Faculty of Engineering and Applied Science of the University of Pennsylvania in partial fulfillment of the requirements for the degree of Master of Science in Engineering for graduate work in Computer and Information Science.


## Abstract

Three dimensional scene analysis in an unconstrained and uncontrolled environment is the ultimate goal of computer vision. Explicit depth information about the scene is of tremendous help in segmentation and recognition of objects. Range image interpretation with a view of obtaining low-level features to guide mid-level and high-level segmentation and recognition processes is described. No assumptions about the scene are made and algorithms are applicable to any general single viewpoint range image. Low-level features like step edges and surface characteristics are extracted from the images and segmentation is performed based on individual features as well as combination of features. A high level recognition process based on superquadric fitting is described to demonstrate the usefulness of initial segmentation based on edges. A classification algorithm based on surface curvatures is used to obtain initial segmentation of the scene. Objects segmented using edge information are then classified using surface curvatures. Various applications of surface curvatures in mid and high level recognition processes are discussed. These include surface reconstruction, segmentation into convex patches and detection of smooth edges. Algorithms are run on real range images and results are discussed in detail.

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## Chapter 1

## Introduction

Three dimensional scene analysis in an unconstrained and uncontrolled environment is the ultimate goal of computer vision. Most of the effort in this regard has gone in extracting three dimensional information from intensity images and arriving at a meaningful and sufficiently unambiguous interpretation of the scene. However, the problem with monocular vision is the loss of 3-D information thereby making the interpretation process underconstrained. Shape from X methods have been widely studied in last two decades to extract depth of the scene using texture,shading,color,contour and motion. Depth extraction from stereo images is computationally expensive and results in sparse depth maps requiring reconstruction techniques for further interpretation. Range images on the other hand are obtained by realtime depth sensors and provide dense 3-D information of the visible surfaces.

Range images are dense depth maps measuring the distance of the physical surface from a known reference plane. Different types of ranging methods are available to obtain range information according to the application. Magnetic resonance imaging systems give true 3-D images, i.e, all the points in 3-D space are specified. Visible surfaces can be scanned by time of flight laser range finders and amplitude-modulated laser range finders. The most common and cheapest are the triangulation-based scanners. Structured lighting systems scan the scene with a laser stripe to obtain depth information of the visible surface in a
calibrated workspace. Research interest in range image processing has grown tremendously in recent years due to increasing availability of structured lighting range sensors. While these sensors can be employed in closed environment only and suffer from other drawbacks (like shadows, inability to sense highly reflective surfaces and some colors) they are useful for real-time scanning of good quality at low cost.

The range images dealt with in this work are of $z(x, y)$ type, (i.e., monge-patch surfaces) where each pixel gives the $Z$-depth at the coordinate $x$ and $y$. Since range images (or depth maps) contain explicit 3-D information about the scene it is expected that surface description and object recognition should be easier to handle with range images. However if the scene is quite complicated, then the problem cannot be solved that easily by using range images as one might think. Intensity information can be used to complement range information where ambiguity arises in interpretation, but this involves registration and correspondence problems and may even complicate the analysis.

Representation of range images is just like that of reflectance images. A two dimensional array of depth values specifying ( $\mathbf{x}, \mathrm{y}, \mathrm{z}$ ) coordinates with respect to a known coordinate frame is enough for most applications. This allows many low level intensity image processing techniques to be directly used to process range images by interpreting the pixel value as 'depth' instead of 'reflectance value'. Contrast and brightness however have to be interpreted as surfaces of varying depths.

We have addressed the problem of object and surface segmentation in this report. Segmentation is essentially goal oriented. It can be conveniently divided into two processes : initial segmentation and final segmentation. Initial segmentation process is a result of local computations done in a known neighborhood of every pixel in the image. The final segmentation process does refinement of the initial segmentation using global constraints to arrive at a global interpretation of the scene. We have not assumed any domain knowledge or limited the objects to be of certain type. Our goal is to study boundary based segmentation, surface based segmentation and integration of the two methods. It is possible to segment the scene in flat, convex and concave subparts with detailed description of individual parts using boundary and surface based techniques.

An important aspect of the object recognition problem is the robustness of the recognition approach. It is essential that algorithm be size-invariant, position invariant, orientation invariant and be able to recognize partially occluded objects. As observed by Besl and Jain [9] it is known from the results in differential geometry that Gaussian and mean curvature are visible-invariant features of a surface region in the sense that they do not change under viewpoint transformations that do not affect the visibility of that region. When a surface region is visible, its curvature measurements are invariant to changes in surface parametrization and to translations and rotations. The invariant property is important for 3-D object recognition. Since our final segmentation process will be dependent on the local computations it is necessary that the low-level features be invariant.

While these two approaches are domain independent, any high level recognition approach like model based interpretation makes use of domain specific knowledge. We use a high level volumetric approach using superquadrics described in [23] to illustrate the usefulness of initial segmentation in high level vision. Figure 1 presents the paradigm explored in this work.

It is clear that processing of range images can be divided into three major stages: low level, intermediate level and high level. After range image is acquired from the sensor, it needs to be smoothed before any useful operations can be performed on it. Though it creates localization problems, it reduces the effect of quantization which is important for surface fitting. Low level processing is data-driven with the objective of obtaining useful local features that can be used by higher processing stages. Three dimensional edges constitute important features. We have used the Laplacian of Gaussian operator of [24] to detect step edges. Smooth edges have a different significance in case of range images and are more difficult to detect. This will be discussed in chapter 3 in detail.

Computation of curvature involves computing first and second order derivatives at every pixel in the image. Based on curvature signs, initial segmentation of the scene is performed. This is further improved by region growing done with global constraints. Haralick et al [6] have described a mathematical treatment for describing the topographic primal sketch of the


Figure 1: A paradigm for Range Image Segmentation and 3-D object recognition
underlying gray tone intensity surface of a digital image. They use first and second directional derivatives to classify each picture element as one of peak,pit,ridge,ravine,saddle,flat, and hillside. Michael Brady etal [4, 5, 7] describe a study of classes of curves as a source of constraint on the surface on which they lie, and as a basis for describing it. Their approach gives a curvature primal sketch of the surface. Tracing lines of curvature in real range images is very unreliable due to the low $x-y$ resolution of the scanner and quantization and other sensing errors. Besides it is noise sensitive and computationally expensive. Besl and Jain [25, 9, 13] have done a comprehensive study of invariant surface characteristics and presented an algorithm for variable order surface fitting for image segmentation. They have summarized the field of 3-D object recognition in their excellent survey [3]

A scale-space based algorithm for extraction and representation of physical properties of a surface, using curvature properties of the surface is discussed in Fan,Medioni and Nevatia [14]. Nackman [19] has described the two dimensional critical point configuration graphs for describing the behavior of smooth functions of two variables by extracting peaks (local maxima), pits(local minima) and passes (saddle points) of a surface. Our approach is to not to go into too much detail of the surface but to label the surface as flat,convex and concave accurately. Thus local variations are ignored in favour of a more global interpretation. Yang and Kak [33] describe an algorithm to analyze the topmost object in a pile. They compute derivatives by fitting B -splines and use local curvature information to label the object as flat and curved. Their method can only handle one type of surface for the topmost object in the scene and has other problems in assuming that step edges form a closed contour, which is not true in a general range image as described in chapter 3. A new approach for surface classification using characteristic contours is proposed by Sethi and Jayaramamurthy [20]. Characteristic contours are defined as the loci of the points where the surface normals are at a constant inclination to a selected reference vector. However it requires segmented surface and normal vector at every point, which limit its usefulness to surface classification in final stage of recognition process.

Their are specific methods available to process images acquired using a light-stripe
rangefinder. Smith and Kanade [34] have done contour classification of light-stripes to produce object centered 3 -dimensional descriptions. Another method by Martin Herman [35] extracts detailed, complete descriptions of polyhedral objects from light-stripe rangefinder data.

Segmentation of scene into surface primitives is useful in many applications. Most of the techniques discussed above involve curvature determination. Hebert and Ponce [8] have used surface normals (the Extended Gaussian Images) to classify surfaces into three simple primitive surfaces: planar,cylindrical, and conic regions. Duda, Nitzan and Barrett [36] have presented an algorithm for detecting planar regions using registered range and reflectance data.

Most of the high level recognition approaches include model matching. Kuan and Drazovich [37] have represented the objects as viewpoint-independent volumetric model based on generalized cylinders. They perform feature-to-model matching based on low-level features derived from range imagery. Constructing the 3-D model of an object involves integrating data or descriptions of an object obtained from multiple views and representing this intergrated data in a coherent manner. Vemuri and Aggarwal [38] have presented an algorithm for automatic construction of models by determining the orientation of the object in the calibrated workspace and representing the object in cylindrical coordinates. Their method does not require correspondence to be established but requires registered intensity and range data of the scene while building the model. We have used superquadric models to recognize segmented objects. The classification procedure matches superquadric parameters with the parameters of the identifiable models. Since models are well defined by eleven superquadric parameters, there is no need to build models of the objects in advance.

## Chapter 2

## Acquisition and Preprocessing of Range Images

Range images obtained by different scanners differ in the format of the output. In order to apply low level techniques to the image it is necessary that the image points be quantized in Z -depth format with equal resolution factor in X and Y direction. Once converted into Z-depth format the image is smoothed.This chapter discusses some practical aspects of real range image processing which are important if any useful results are desired.

### 2.1 Range Image Acquisition

The test images used in this work were acquired by structured lighting triangulation based scanners. Figure 2 shows the ranging geometry of a typical range sensor. The trigonometry of a sensor will not be described here.

Either the laser stripe moves and scans the scene or the workspace moves under a vertical laser stripe. If the viewpoint of the sensing camera is not the same as the laser then shadows (regions with missing data) are obtained. In order to discriminate between shadows and background, (region of known depth on which object is sitting) background is assigned a nonzero depth.


Figure 2: Ranging geometry of a structured lighting scanner and Z-depth format

Contrary to the popular assumption made by researchers, it may not always be possible to represent the visible surface in Z-depth format, viewing perpendicular to the background. To be able to represent all the scanned points in Z-depth format, it is necessary to digitize the scene watching parallel to camera's line of sight. This may require rotating the scene to align the Z axis along the line of sight of camera thereby rotating the background which is no longer of constant depth. The segmentation procedure should take this into account. Also, this makes the processing viewpoint dependent. To avoid the trouble arising due to this, it is often convenient to fix the viewpoint at the cost of losing some scanned points. This problem is acute with images obtained from white scanner where $f(x, y)$ is not unique. The solution is to segment the scene from background and then rotate the scene to obtain the Z-depth image.

### 2.2 Scaling of Range Images

Sampling interval of the scanners depends on the thickness of the laser stripe, value of laser stripe increment and resolution of the camera. More often than not vertical resolution (along Y axis) is different from horizontal resolution (along X axis). Thus the sampled points are not spaced uniformly in X and Y direction. Since we apply neighborhood operators during low level processing of images, it is necessary to rescale the images uniformly in both directions. We have rescaled the Z-depth image by fitting a plane on three neighborhood points. Figure 3 illustrates the difference between unscaled and uniformly scaled images.

### 2.3 Smoothing of Range Images

Depth resolution of a range image is an important parameter in low level processing. Range scanners usually have depth resolution good enough for most applications. In fact a resolution of 0.01 inch/pixel is too fine and noise sensitive for surface fitting purposes. The problem comes in quantization of $z$ values. If entire scan depth is quantized within 8 bits (most convenient representation), effective depth resolution is drastically reduced thereby


Figure 3: Z-depth format images. left :original resolution. right : uniformly scaled
increasing the quantization error. Since surface fitting is very sensitive to quantization error, we have minimized it by following two step procedure :

1. Original depth resolution is preserved by storing the depth value unscaled in 2 bytes. This allows 64 k possible quantization levels. Scaling along Z axis is done only when needed.
2. Image is smoothed using a Gaussian operator and smoothed values are stored in floating point buffers so as not to lose any precision.

One way to reduce noise is to perform median filtering of the image. It ensures that isolated noise is reduced and edges are not smoothed.

Our approach is to study the scale-space behavior of range images. We have used Gaussian operator to smooth images. The Gaussian function in two dimensions is given by :

$$
\mathbf{G}(x, y)=\frac{1}{\sqrt{2 \pi} \sigma} e^{\frac{-\left(x^{2}+y^{2}\right)}{2 \sigma^{2}}}
$$

Since it is separable in X and Y directions, one dimensional Gaussian operator is applied separably on the image. Smoothing is controlled by the size of the operator, which is determined by $\sigma$. Gaussian operator has some nice properties that make it a unique operator for our purposes.

Yuille and Poggio [39] have proved that the Gaussian low-pass filter is the only filter with a nice-scaling behavior for linear derivative operators like the laplacian operator. It also satisfies the following conditions :

1. Filtering is shift-invariant and therefore, a convolution.

$$
\left.F \times I(x)=\int F(x-\alpha) I(\alpha) d x\right)
$$

2. The filter has no preferred scale. The filter is properly normalized at all the scales.
3. The filter recovers the whole image at sufficiently small scales.

$$
\lim _{\sigma \rightarrow 0} F(x, \sigma)=\delta(x)
$$

where $\delta(\mathrm{x})$ is the Dirac delta function.
4. The position of the center of the filter is independent of scale of the filter. Otherwise zero crossings of a step edge would change their position with change of scale.
5. The filter goes to zero as $|x| \rightarrow \infty$ and as $\sigma \rightarrow \infty$.

We have studied the behavior of increasing sigma value on edge detection, surface characterization and segmentation. As $\sigma$ value increases, window size of the Gaussian operator increases and details are lost. Figure 4 and figure 5 show the perspective plots of a range image before and after smoothing respectively.

Minor surface perturbations are smoothed easily. But the undesirable effect of uniform application of Gaussian operator is smoothing of all types of edges. Step edges (chapter 3) are smoothed to form roughly convex and concave subparts (chapter 4).This complicates edge detection, specially detection of smooth edges (concave and convex edges) that are


Figure 4: 3-D perspective plot of original image


Figure 5: 3-D perspective plot of smoothed image ( $\sigma=1$ )
further smoothed. Thin objects tend to merge into the underlying objects, making segmentation difficult. As will be discussed in chapter 3, as we go up the scale objects start merging. We have found sigma value of (window size $5 \times 5$ ) to be best suited for our experiments. In surface based segmentation technique, smoothing alters the local behavior of the surface, but makes the result more reliable, specially away from the edges. Step edges are shown as adjacent convex and concave regions. Segmentation using these effects of smoothing is discussed in chapter 4 in detail.


Figure 6: Edges in range images
allows boundaries to be read off as zero crossings of the LOG operated image. We'll discuss the significance of the LOG operator in view of range images. While step edges pose no particular problem, smooth edges are difficult to detect by local operations. In range image segmentation it is of particular interest that the a pile of objects be segmented into convex subparts. This requires detection of concave edges that will delimit the convex subparts. Mitiche and Aggarwal [28] have presented a probabilistic approach of detecting the convex and concave edges by using domain specific constraints.

Though 3-D edges are quite useful for object recognition, there are some inherent limitations in edge information that make their use limited to aiding the higher level recognition processes along with a host of invariant features. Edge classification depends on the orientation of object in 3-D space and is therefore not an invariant feature. Thus edge information cannot be the only feature used by the recognizer and it has to be used in conjunction with other features. However, as will be seen later, edge information is good enough for early segmentation of range images because the requirement of invariant features does not apply to the initial-segmentation process.

In case intensity information is available range data can be complemented by reflectance data to pick up weak 3-D edges like the step edges created by overlapping thin objects. Wong and Hayrapetian [32] used range information to segment intensity images. Gil, Mitiche and Aggarwal [27] have described experiments in combining intensity and range edges. While intensity images are certainly useful in detecting edges in the scene, they need to be registered in the same way as range images to avoid correspondence problem. This may


Figure 7: Derivatives of a cross-section of a range image
possibility of extracting smooth edges using the $\nabla^{2} \mathrm{G}$ operator.
The Gaussian distribution in one dimension is defined as :

$$
\mathbf{G}(x)=\frac{1}{\sqrt{2 \pi} \sigma} e^{\frac{-x^{2}}{2 \sigma^{2}}}
$$

The first and second derivatives are :

$$
\begin{gathered}
\mathbf{G}^{\prime}(x)=\frac{-x}{\sqrt{2 \pi} \sigma^{3}} e^{\frac{-x^{2}}{2 \sigma^{2}}} \\
\mathbf{G}^{\prime \prime}(x)=\frac{-1}{\sqrt{2 \pi} \sigma^{3}}\left(1-\frac{x^{2}}{\sigma^{2}}\right) e^{\frac{-x^{2}}{2 \sigma^{2}}}
\end{gathered}
$$

The cross-section of a range image and the profiles of first and second order derivative are shown in figure 7.

In two dimensions the LOG operator becomes :

$$
\mathbf{G}^{\prime \prime}(x, y)=\frac{-1}{\sqrt{2 \pi} \sigma^{4}}\left(2-\frac{x^{2}+y^{2}}{\sigma^{2}}\right) e^{\frac{-\left(x^{2}+y^{2}\right)}{2 \sigma^{2}}}
$$



Figure 8: 3-D edges detected in a synthetic range image. : upperleft original image. upperright thresholded step edges. lowerleft thresholded convex edges. lowerright thresholded concave edges.

It is clear that behavior of second order derivatives is unique at every type of change in surface. There are positive spikes at concave ramp edge, negative spikes at convex roof and ramp edges and zero crossings at jump edges. However there are serious practical limitations in using this response to detect concave and convex edges. The response at these edges is dependent on the convexity or concavity of the edge which is roughly the measure of angle at which the two surfaces meet. If the angle is too small and change in depth is gradual as in most situations, the response would be below or same as that due to local surface changes. See figure 8 for the step, convex and concave responses obtained in a synthetic range image having planar regions. Even in synthetic range image the responses deteriorate as image was smoothed and smooth concave and convex edges virtually disappeared.

Thresholding of zero crossings is necessary in case of range images to avoid local surface perturbations. Responses due to weak concave and convex edges would then be filtered. Zero-crossings generated by weak step edges may also lie below the thresholding value. In
range images, Value of zero-crossing has direct relationship with the magnitude of depth discontinuity. Thus selection of a threshold effectively restricts the minimum detectable depth. An object with less than acceptable height would be invisible in the edge image.

As observed in the previous chapter, it is absolutely necessary to smooth an image before attempting any local operation it. LOG operator gives following image :

$$
\mathbf{f}(x, y)=\left(\nabla^{2} \mathbf{G}\right) * \mathbf{I}(x, y)
$$

which can be written as :

$$
\mathbf{f}(x, y)=\nabla^{2}(\mathbf{G} * \mathbf{I}(x, y))
$$

Degree of smoothing depends on the value of sigma, which controls the size of the window. Larger the sigma, greater is smoothing. While this effect is interpreted in intensity images as blurring and hence reduction of details, In range images it is seen in terms of smoothing the surface at the boundaries in addition to reduction of details. This results in all types of boundaries to become smoothed and can have undesirable effects on boundary detection and surface based segmentation. We have observed that with increasing scale value, range images loose vital boundary information presenting difficulties in edge based segmentation. Empirically determined window size of 5 ( sigma value $=1$ ) is chosen for processing all the images.

The algorithm for edge detection is given below.

1. Read in the range image.
2. Convolve the image with Gaussian operator separably in $X$ and $Y$ direction.
3. Convolve the $\mathbf{G}(\mathrm{x}, \mathrm{y}){ }^{*} \mathrm{I}(\mathrm{x}, \mathrm{y})$ image with Laplacian operator:

$$
\left[\begin{array}{ccc}
0 & -1 & 0 \\
-1 & 4 & -1 \\
0 & -1 & 0
\end{array}\right]
$$

4. Label the Zero-crossing (with maximum value ) at every pixel in the $\nabla^{2} \mathbf{G}(x, y) * \mathrm{I}(x, y)$ image. Also mark the direction along which maximum crossing value is found.
5. Threshold the image at a predetermined value to label pixels as belonging to step edges.

Figures 11(b),12(b) and other figures show the magnitude of zero-crossings detected in range images.

It is observed that threshold selection is important in defining the acceptable depth, which in turn is determined by the amount of detail seen in the filtered image. Also the thresholding value is different at different scales. Threshold value varies inversely with smoothing parameter $(\sigma)$.As we go up the scale-space the need for thresholding decreases.

Next we discuss a segmentation technique based on the step edges detected by the LOG operator.

### 3.2 Segmentation of Range Images using edge information

Segmentation of objects in a range image depends on actual requirements. One should therefore define the problem of segmentation clearly in the relevant context. In order to recognize an object it is necessary to isolate the object, which is not a trivial task. In a practical environment where objects can be of any size and shape, segmentation of individual objects can be a difficult task. Objects in a range image will always be partially occluded making the problem of segmentation and recognition difficult. If the scene consists of a heap of objects and we have to recognize one object then our problem can be simplified by considering the object of immediate importance, the one on the top of the heap. But this method has only specific applications like picking parts out of a bin etc. and is not useful as a general segmentation strategy.

Another approach to a complete segmentation is to segment the picture according to the requirement. Sometimes image segmentation into different surface types may be useful
and at other times convex objects need to be segmented. Whatever the method, a segmentation process working on local information cannot always give requisite results. In fact, segmentation at lower-level of processing can at best give locally valid results which may be conflicting from global point of view. Thus a robust segmentation process has to work with higher stages of processing to yield globally valid results. This introduces the concept of feedback from higher stages so as to work in a closed loop with the goal of object recognition. Then segmentation can be considered as a part of object recognition stage.

As discussed before, 2nd order derivatives give enough information to delineate objects at the step boundaries. We will develop an algorithm for segmenting the topmost object in a heap of arbitrary objects.

First stage of segmentation process involving isolation of objects from background can be easily accomplished by thresholding the image at known background depth. The difficult part is to identify an individual object from the heap of objects. At this point it is essential to define what is meant by an 'individual object'. An object can be a complex combination of various 'primitive objects' like cube, sphere, cylinder etc.Some important questions need answering here before any further progress can be made : What are the end boundaries(edges) of the object and what are the internal boundaries of the object and how to distinguish between them?

Different types of 3-D edges are jump edges, concave roof edges, concave ramp edges, convex roof edges and convex ramp edges. Considering any of these edges as the internal or external boundary of the object is going to put restrictions on the object types that can be segmented. For example, a jump edge (Unless it is an occluding edge against the background) need not be the end boundary of the actual object. Since the segmentation process is essentially a local operation and no other knowledge is used this problem cannot be solved at this segmentation level. There are following solutions to this problem :

1. Make assumptions about the type of the objects. For example, assume the topmost object to be convex. This is a very strong assumption and requires convex internal boundary information. As noted before, this information is difficult to derive using
second order derivatives. We will see in next chapter, how surface characterization techniques can be used to approximate the presence of smooth internal boundaries of an object.
2. a priori knowledge about the scene. But this is against our approach towards a general and robust segmentation algorithm.
3. Feedback from higher stages of object recognition to eliminate need for a priori knowledge and to relax the strong assumptions made at low-level segmentation stage. Since higher levels of object recognition are global processes and may have knowledge about the domain, a closed loop segmentation procedure is bound to perform better than one having no feedback.

Thus to achieve a reliable segmentation of the initial scene, we will assume that the topmost object is delineated by jump boundaries. This may not always be true as two objects can join at convex or concave edges or one object may merge into next one due to negligible thickness of the object at the point of contact. This means that local information cannot give perfect segmentation in all the cases. In such cases we need higher level processing to figure out the right segmentation of the scene. The segmentation based on external boundary information will give only an initial estimate of segmentation. This estimate is reliable to the point that it can distinguish between objects of predetermined depth.

In the context of invariant object recognition it is important to note that step boundaries may vary with orientation of the object. Thus they are used only to segment the object and not to recognize the object. We will discuss the results of recognition and classification of the segmented object using the superquadric technique developed by Franc [23].

The block diagram of segmentation and classification process is shown in figure 9
A practical problem with using zero-crossing as step boundaries is that they do not form closed contours. The boundaries delineating the object may not be completely enclosing the object, resulting in region growing process to overflow from the object and include the neighboring object as part of the object. This drawback renders the final segmentation result very unreliable. Yang and Kak [33] use a priori knowledge about the width of the


Figure 9: Flow chart of segmentation and recognition process
object and contour tracking to extract the closed contour surrounding the object. Their method does not guarantee success in all the cases. David Heeger [26] has proposed a computational approach to gap filling. It is computationally expensive and not suited for our purpose since we want to avoid explicit contour tracing in the entire image and want the region growing method to take care of it. Peter Allen [29] has used gap filling method based on contour growing proposed by Nevatia and Babu [30]. They perform gap filling on the entire scene using the predecessor-successor graph of all connected contours.See Peter Allen's Ph.D. thesis for details. Contours are then merged based on the requirement of merging N pixel gaps. This approach is again computationally expensive. Peter Allen observes that filling of at most two pixel gaps is acceptable because of the ambiguities resulting with three or more pixel gap-filling requirement. We have implemented two pixel gap filling by constraining the region growing process near the boundaries, thus avoiding the explicit gap filling stage.

One and two pixel gap filling is accomplished by simply requiring that a pixel having a boundary pixel in its 8 -neighborhood be not grown recursively. Instead, the pixel and the boundary pixel are marked as grown. Figure 10 illustrates gap filling in one instance.

Thus we are able to avoid the contour tracing explicitly to fill gaps. Three or more pixel gaps cannot be adequately handled by gap-fillers. Some sort of post processing is necessary to further segment the segmented object in this case. One way is to trace all the boundary pixels of the segmented object and use concavity information to segment the object into parts. This approach is being implemented and will not be discussed here.

The algorithm for segmentation is given below :

1. Read in the original range image $I(x, y)$ and $\nabla^{2} G * I(x, y)$ image.
label_val $=0$
2. Segment the objects from background by thresholding at background depth (supplied by the user). In case background is not of uniform depth, a plane can be fitted to represent the background and threshold the objects from the scene.
3. Locate the $3 \times 3$ window with maximum height by averaging the pixel values in $3 x 3$
 Pixels in region.

Ungrown Pixels.

Figure 10: An example of gap filling
window at every pixel in the range image. This gives the seed region for region growing. Clearly the window lies on the topmost object. If there are more than one heap, then also only one seed region is obtained.
4. label_val = label_val +1
5. Grow the seed region recursively in all 8 directions. For gap filling procedure to work it is necessary to grow pixels in 8 -neighborhood. Let $p_{i j}$ be the pixel being grown. A pixel $\hat{p}_{i j}$ in 8 -connected neighborhood of $p_{i j}$ is not grown under one of the following conditions :
(a) depth $\left(\hat{p}_{i j}\right)<=$ background_threshold
(b) $\hat{p}_{i j}$ is already labeled.
(c) If $\hat{q}_{i j}$, any pixel in 8 -connected neighborhood of $\hat{p}_{i j}$ satisfies :
$\operatorname{laplacian}\left(\hat{q}_{i j}\right)>=$ edge_t hreshold
Then $p_{i j}$ is 1-pixel distance from an edge pixel and likely to be in a gap. Make :
$\operatorname{label}\left(\hat{p}_{i j}\right)=\operatorname{label}\left(p_{i j}\right)$ and
$\operatorname{label}\left(\hat{q}_{i j}\right)=\operatorname{label}\left(p_{i j}\right)$.
6. If number of pixels in the extracted region < acceptable size then region is invalid else it is valid.
7. If region is valid then determine supporting points of the region.
8. The region extracted in the first pass is topmost region. Subsequent regions are grown from top to bottom, left to right. If any more pixels are left to be processed then pick up any unprocessed pixel and go back to step 5 to grow region.
9. Output topmost region, all valid regions and supporting points in separate image files.

### 3.3 Segmentation Results



Figure 11: Segmentation: (a) original image from RCA. (b) Edges detected. (c) topmost object. (d) all segmented objects


Figure 12: Segmentation:(a) original image from RCA. (b) Edges detected. (c) topmost object. (d) all segmented object


Figure 13: Segmentation: (a) original image from grasp lab scanner. (b) Edges detected. (c) topmost object. (d) all segmented objects


Figure 14: Segmentation: (a) original image from RCA. (b) Edges detected at $\sigma=1$. (c) segmented parts of one object at $\sigma=1$. (d) Segmentation at $\sigma=2$, for the same threshold value.


Figure 15: Segmentation at different scales: (a) image smoothed at $\sigma=2.0$. (b) objects segmented. (c) : image smoothed at $\sigma=3.0$. (d) objects segmented

The programs are written in C and implemented on a VAX-785 running UNIX. Ikonas graphics and PM format for images are used in all programs. Images are acquired from two sources. Most of the images used as examples are from RCA range image database and remaining are scanned by Grasp lab's range scanner. All the images are digitized in Z-depth format. RCA images have better ( 12 bits/pixel ) resolution than Grasplab images ( $8 \mathrm{bits} / \mathrm{pixel}$ ). Hence more detail is seen in the former. It makes difference in detection of thin objects. Due to different Z resolution of two scanners, we have used different threshold values for the two sets of images. All the images are smoothed uniformly with Gaussian of $\sigma=1$ (window size $=5 \times 5$ ). Zero-crossings of LOG operator are thresholded to remove response due to minor surface perturbations. The threshold at a given $\sigma$ value also limits the thickness of objects that can be segmented. Threshold values are determined empirically, since histogram of zero-crossings cannot be used in determining threshold automatically as is done in intensity images. However threshold value remains same for all the images acquired from the same source. This is true for all the empirically determined parameters reported in this thesis. Background value is also known to the program and is constant given a scanner. Results of processing the images are in figures $11,12,13$ and 14 . In figure 11 all the objects are segmented correctly. The topmost object is a Cylindrical object. Figure 12 shows merging of objects because of very weak step boundary information. Figure 13 shows results of segmentation on the image obtained from grasp lab scanner. A constant offset of 100 is added to original image depth values and zero-crossings are enhanced for displaying purposes. Figure 14 exhibits different results at two scales for the same edge threshold. The scene has single object, a box with string tied around it, so that the box is divided into 4 partitions. Because of the high depth resolution of the image, edge information due to string is enough to segment the box into three parts at $\sigma=1$ (figure 14(c)). Increasing the $\sigma$ value to 2 , removes the details of the string and whole visible surface is recovered (figure 14(d)).

To study the effect of increasing sigma value on zero-crossing, one of the multiple object images is processed for $\sigma=1,2,3$. Note that objects start to merge as sigma increases, with thin objects undetectable at $\sigma=2$.(Figures 11, 15)

### 3.4 Recognition of segmented objects using Superquadrics

The surfaces extracted by the previous algorithm can be classified as one of the eight basic surface types. We will discuss this classification approach in detail in next chapter. In this section we will describe a high level recognition and classification method that classifies the segmented object into four broad categories.

We have used superquadric model recovery method implemented by Franc [23] to recognize the segmented object in a range image. Details of the procedure for superquadric fitting are discussed in Franc's Ph.D. thesis. Superquadrics are a family of parametric shapes that can be used as primitives for shape representation in computer vision [31]. Superquadrics are like lumps of clay that can be deformed and glued together into realistic looking models. However, we will consider only non-deformed superquadric models for classification of the object into one of the categories :

1. flat : Object with negligible height compared to length and width;
2. roll : A Cylindrical object.
3. box : An object with comparable height,length and width.
4. Irregular : Any object not falling in any of the above three categories.

Superquadric implicit equation is given by :

$$
\left[\left(\frac{x}{a_{1}}\right)^{\frac{2}{c_{2}}}+\left(\frac{y}{a_{2}}\right)^{\frac{2}{c_{2}}}\right]^{\frac{c_{2}}{c_{1}}}+\left[\frac{z}{a_{3}}\right]^{\frac{2}{a_{1}}}=1 .
$$

Parameters $a_{1}, a_{2}$, and $a_{3}$ define the superquadric size in $\mathrm{x}, \mathrm{y}$ and z direction respectively. $\varepsilon_{1}$ is the squareness parameter in the latitude plane and $\varepsilon_{2}$ is the squareness parameter in the longitude plane. Based on these parameter values superquadrics can model a large set of standard building blocks, like spheres, cylinders, parallelopipeds and shapes in between. Figure 16 illustrates the various types of shapes obtainable by changing two shape parameters. If both $\varepsilon_{1}$ and $\varepsilon_{2}$ are 1 , the surface defines an ellipsoid. Cylindrical shapes are obtained for $\varepsilon_{1}<1$ and $\varepsilon_{2}=1$. Parallelopipeds are obtained for both $\varepsilon_{1}$ and $\varepsilon_{2}$ are $<1$.


$$
\varepsilon_{1}=1
$$

$$
\varepsilon_{1}=1.9
$$



Figure 16: Superquadric models as function of shape parameters $\left(\varepsilon_{1}, \varepsilon_{2}\right)$ for given size paremeters $\left(a_{1}, a_{2}, a_{3}\right)$

We have restricted the model recovery procedure to fit the models with $0 \leq \varepsilon_{1}, \varepsilon_{2} \leq 1$. We will not discuss the details of model recovery here.

The Criteria used for classification are three size parameters, two shape parameters and the goodness of fit ( $G O F$ ) measure. The superquadric procedure returns a GOF measure using the following equation :

$$
G O F=\frac{1}{N} \sum_{i=1}^{N}\left[a_{1} a_{2} a_{3}\left(F\left(x, y, z ; a_{1}, a_{2}, a_{3}, \varepsilon_{1}, \varepsilon_{2}, \phi, \theta, \psi, p_{x}, p_{y}, p_{z}\right)-1\right)\right]^{2}
$$

Where $F$ is the superquadric inside-outside fuction described in Franc [23]. $\phi, \theta, \psi$ define the orientation and $p_{x}, p_{y}, p_{z}$ define position of superquadric in space.

The object given by the segmentation procedure has only the points visible to the scanner. Much of the volumetric information is lost in the Z-depth format of representation. While this is not a serious problem in case of curved objects like cylinder or segmented surfaces having volumetric information (like a tilted box viewed from above), model fitting becomes ambiguous if the visible surface is flat. If it is known that the original scene had only one object, then the supporting surface can be assumed to be the plane parallel to the known background. The problem complicates in the multiple object scenes, where it becomes impossible to assign correct depth to the segmented object. Given no prior knowledge about the surface type, we need to add points in every case to give volumetric information to the superquadric procedure. Points can be added in two ways :

1. Background is assumed to be the supporting surface of the object. Points are added on the background by backprojecting the visible surface on the background (figure 17(c). While this is desirable in case of flat surfaces, it is not right for surfaces with volumetric information.
2. Supporting points of the segmented object are used to determine the immediate supporting surface(s) of the object. Points are added vertically (figure $17(\mathrm{e})$ ) to the object. This technique is more flexible since it can handle objects not lying on the background. But it results in more points to be added in addition to assuming that


Figure 17: Horizontal and vertical addition of points.(a) object. (b) original points. (c) horizontal addition of points. (d) fit with horizontal addition. (e) vertical addition of poins. (f) fit with vertical addition.
the object is actually touching the neighboring objects, which may not be true in general.

In general it is not possible to extract correct supporting surface information from a single viewpoint. We have used horizontal addition of points in our experiments as it is faster than vertical addition and recovers the desired model.

The algorithm for model fitting,selection and classification is following :

1. Read the segmented object in Z-depth format.
2. Format conversion and point addition : Generate a list of points in 3-D space representing the object. Call it points.orig. For every point on the visible surface add
a point at the same $(x, y)$ coordinates on the background. Output the list of original and added points in points.add.
3. Superquadric fitting : Run Superquadric model fitting procedure on points.orig. Model obtained is model.orig. Run Superquadric model fitting procedure on points.add. Model obtained is model.add. Iterative superquadric fitting is stopped if one of the following conditions is met :
(a) Number of iterations $\geq 15$.
(b) Goodness of fit of $i$ th iteration $(i \leq 15)$ is $\leq$ Acceptable measure. This measure is empirically determined.
(c) If for the $j$ th $(j \geq 5)$ iteration :

$$
\sqrt{\frac{1}{5}\left[\sum_{i=j}^{j-4}\left(G O F(i)-\sum_{k=j}^{j-4} G O F(k)\right)^{2}\right]} \leq \text { Acceptable_deviation. }
$$

Condition (a) assumes that model recovery is complete by 15th iteration. Condition (b) stops the procedure if an acceptable model is obtained early in the process. Condition (c) monitors the rate convergence of fitting procedure. It terminates the fitting procedure if the GOF measures of last five iterations do not vary much. All the values used in the above three conditions are empirically determined.
4. Model selection :

THEN GOTO volume_criterion
ELSE IF $($ GOF $($ model.add $) \leq$ Acceptable_fit $)$
THEN model $=$ model.add GOTO classify.
ELSE IF $($ GOF $($ model.orig $) \leq$ Acceptable_fit $)$
THEN model $=$ model.orig GOTO classify .

ELSE
OBJECT = Irregular. Goto Done.
5. Volume_criterion : Volume can be approximated as $a_{1} \times a_{2} \times a_{3}$.
$\operatorname{IF}($ volume.add < volume.orig)
THEN model $=$ model. orig
ELSE model $=$ model.add .
6. classify : Classify model using $a_{1}, a_{2}, a_{3}$ and $\varepsilon_{1}, \varepsilon_{2}$ :
(a) IF $\left(\left(a_{3} \ll a_{1}\right) A N D\left(a_{3} \ll a_{2}\right) A N D\right.$
$\left.\left(\varepsilon_{1}<0.5\right) A N D\left(\varepsilon_{2}<0.5\right)\right)$
THEN OBJECT $=$ FLAT .
(b) ELSE IF $\left(\left(a_{1} \ll a_{3}\right)\right.$ AND $\left(a_{2} \ll a_{3}\right)$ AND
$\left.\left(\varepsilon_{1}<0.5\right) A N D\left(\varepsilon_{2}<0.5\right)\right)$
THEN OBJECT $=$ FLAT .
(c) ELSE IF ( $\left(a_{1}>T H R E S H-B O X\right) A N D\left(a_{2}>T H R E S H \_B O X\right) A N D$
$\left(a_{3}>\right.$ TH RESH_BOX) AND $\left(\varepsilon_{1}<0.5\right)$ AND $\left.\left(\varepsilon_{2}<0.5\right)\right)$
THEN OBJECT $=B O X$.
(d) ELSE IF $\left(\left(a_{1}>T H R E S H \_1_{-} R O L L\right) A N D\left(a_{2}>T H R E S H-1 \_R O L L\right) A N D\right.$
$\left(a_{3}>\right.$ THRESH_2_ROLL) AND
$\left.\left(\varepsilon_{1}<0.5\right) A N D\left(\varepsilon_{2}>0.5\right)\right)$
THEN OBJECT $=$ ROLL .
(e) ELSE OBJECT = Irregular

THRESH_BOX is the minimum acceptable dimension of the box.
THRESH_1_ROLL is the minimum acceptable width and height of the roll.
THRESH_2_ROLL is the minimum acceptable length of roll.

| OBJECT | Model | 21 | a2 | a3 | e1 | e2 | Error |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Cylinder | Orig | 16.30 | 8.65 | 75.70 | 0.24 | 0.95 | 56.21 |
|  | Add | 16.07 | 35.29 | 72.16 | 0.10 | 0.14 | 393.28 |
| Box | Orig | 3.47 | 33.6 | 45.29 | 0.10 | 0.81 | 1274.40 |
|  | Add | 38.11 | 51.14 | 39.47 | 0.10 | 0.10 | 441.16 |
| Flat | Orig | 7.46 | 46.76 | 57.72 | 0.10 | 0.52 | 319.32 |
|  | Add | 8.86 | 45.75 | 57.52 | 0.10 | 0.17 | 245.34 |
| Othas | Orig | 4.928 | 50.41 | 76.79 | 0.10 | 0.43 | 4083.61 |
|  | Add | 9.38 | 53.00 | 82.44 | 0.10 | 0.10 | 4178.83 |

Figure 18: Parameter values of the recovered models
7. Done: Output classified model with parameters. Determine Orientation and position of the model in world coordinate system.

### 3.5 Results of Superquadric Fitting and Classification

The superquadric fitting procedure and classifier were run on the objects segmented previously. The superquadric parameters for the four types of recovered objects are shown in figure 18.

Figure 19 shows model recovery on topmost object segmented in figure 14. The model selection process rejected model.orig due to large fit-error and accepted model.add. Even if model.orig had an acceptable error measure, model.add would have been selected due

RCA 21


Figure 19: Superquadric fitting and Model selection (a box) : (a) original points. (b) fitted model on original points. (c) original and added points. (d) fitted model on original and added points.
to larger volume. The acceptable error magnitude was empirically determined to be 500 . In figure 20 model.orig is selected and classified as roll because of the tremendous error difference between the two acceptable models. model.add is accepted and classified as flat in figure 22 because of volume consideration, although both the models have acceptable error measures. Finally, the film mailer in figure 23 is classified as irregular as the fit-errors of both the models is more than acceptable error measure.

The results are shown for the four classes of objects. Tapered, bent or concave objects cannot be represented by these models and hence will be classified as irregular. Franc Solina's superquadric method also allows for tapering and bending along with segmentation

RCA 19


Figure 20: Superquadric fitting and Model selection (a Cylindrical object) : (a) original points. (b) fitted model on original points. (c) original and added points. (d) fitted model on original and added points.


Figure 21: .Segmentation: upperleft : original image from grasp lab scanner ( aletter ) upperright:Edges detected. lowerleft : topmost object. lowerright : all segmented objects


Figure 22: Superquadric fitting and Model selection (a letter) (a) original points. (b) fitted model on original points. (c) original and added points. (d) fitted model on original and added points.

RCA 24


Figure 23: Superquadric fitting and Model selection (a film-mailer) : (a) original points. (b) fitted model on original points. (c) original and added points. (d) fitted model on original and added points.
of the complex objects into parts. In the next chapter we will describe a surface classification scheme that uses the output of segmentation routines described in this chapter.

## Chapter 4

## Surface Characterization and Segmentation

Surface characterization refers to the computational process of partitioning the surfaces into regions with equal characteristics. Since our ultimate goal is object recognition, classification of the surfaces by the characteristics of the surface functions is very useful. Classical differential geometry provides a complete surface description of analytic surfaces so as to obtain a complete set of surface characteristics. Surface characterization can be successfully used in intermediate and high level processing of the object recognition problem.

Important surface characteristics, that are visible-invariant are Gaussian curvature and the mean curvature. They are invariant to changes in surface parametrization and to translations and rorations of object surfaces. Guassian curvature is an intrinsic property of the surface while mean curvature is an extrinsic property of the curvature.

From differential geometry it is well known that curvature, speed, and torsion uniquely determine the shape of 3 -D surfaces. The surface characteristics of our interest are the ones with one-to-one relationship with curve shapes. The mathematics of a general surface representation scheme and calculation of Guassian and mean curvatures is described in following section.

### 4.1 Differential Geometry of Surfaces

Parametric form of equation for a regular surface $S$ with respect to a known coordinate system is :

$$
S=(x, y, z): x=x_{1}(u, v), y=x_{2}(u, v), z=x_{3}(u, v),(u, v) \in D \subseteq \mathbf{R}^{2}
$$

The surface is a locus of points in Euclidean three-space defined by the end points of the vector $\mathrm{X}(u, v)$ with $x_{i}(u, v)$ the components of the vector. These real functions are assumed to be defined over an open connected domain of a Cartesian $u, v$ plane and to have continuous second partial derivatives there. In our analysis of range images we are assuming that this condition is satisfied.

The second condition for a regular surface is automatically satisfied by Z-depth format images. It requires that the coordinate vectors $\mathbf{X}_{u}=\mathbf{X}_{1}=\frac{\partial \mathbf{X}}{\partial u}, \mathbf{X}_{v}=\mathbf{X}_{2}=\frac{\partial \mathbf{X}}{\partial v}$ are linearly independent :

$$
\frac{\partial \mathbf{X}}{\partial u} \times \frac{\partial \mathbf{X}}{\partial v}=\mathbf{X}_{1} \times \mathbf{X}_{2} \neq 0
$$

The surface in range images is given by :

$$
\mathbf{X}=\left(x_{1}, x_{2}, f\left(x_{1}, x_{2}\right)\right)
$$

and coordinate vectors become :

$$
\begin{aligned}
& \mathbf{X}_{1}=\left(1,0, \frac{\partial f}{\partial x_{1}}\right), \\
& \mathbf{X}_{2}=\left(0,1, \frac{\partial f}{\partial x_{2}}\right),
\end{aligned}
$$

These vectors are linearly independent given the first condition.
It can be shown using differential geometry techniques that first and second fundamental forms(which exist only if the surface is analytic) uniquely characterize a general smooth surface. The first fundamental form $I$ of a surface is defined as :


Figure 24: Coordinate frame at the Neighborhood of a point

$$
I(u, v, d u, d v)=d \mathbf{X} . d \mathbf{X}=\left[\begin{array}{ll}
d u & d v
\end{array}\right]\left[\begin{array}{ll}
g_{11} & g_{12} \\
g_{21} & g_{22}
\end{array}\right]\left[\begin{array}{l}
d u \\
d v
\end{array}\right]=d \mathbf{u}^{T}[g] d \mathbf{u}
$$

where [g] matrix elements are given by :

$$
g_{11}=E=\mathbf{X}_{u} \cdot \mathbf{X}_{u} g_{22}=G=\mathbf{X}_{v} \cdot \mathbf{X}_{v} g_{12}=g_{21}=F=\mathbf{X}_{u} \cdot \mathbf{X}_{v}
$$

The two tangent vectors $\mathbf{x}_{u}$ and $\mathbf{x}_{v}$ lie in the tangent plane $T(u, v)$ of the surface at the point ( $u, v$ ). [g] matrix is symmetric for an analytic surface.
figure 24 shows the coordinate frame at the Neighborhood of a point.
The first fundamental form $I(u, v, d u, d v)$ measures the small amount of movement in the parameter space ( $d u, s v$ ). The first fundamental form is invariant to surface parametrization changes and to translations and rotations in the surface. Therefore it depends on the surface itself and not on how it is embedded in the 3-D space. The metric functions $E, F, G$ determine all the intrinsic properties of the surface. In addition they define the area of a surface :

$$
A=\iint_{R} \sqrt{E G-F_{2}} d u d v
$$

The second fundamental form of the surface is given by :

$$
I I(u, v, d u, d v)=-d \mathbf{x} \cdot d \mathbf{n}=\left[\begin{array}{ll}
d u & d v
\end{array}\right]\left[\begin{array}{ll}
b_{11} & b_{12} \\
b_{21} & b_{22}
\end{array}\right]\left[\begin{array}{l}
d u \\
d v
\end{array}\right]=d \mathbf{u}^{T}[b] d \mathbf{u}
$$

Where [b] matrix elements are defined as :

$$
\begin{gathered}
b_{11}=L=\mathbf{X}_{u u \cdot \mathbf{n}} b_{22}=N=\mathbf{X}_{v v} \cdot \mathbf{n} \quad b_{12}=b_{21}=M=\mathbf{X}_{u v} \cdot \mathbf{n} \\
\mathbf{n}(u, v)=\frac{\mathbf{x}_{u} X \mathbf{x}_{v}}{\left|\mathbf{x}_{u} X \mathbf{x}_{v}\right|}=\text { unit_normal_vector }
\end{gathered}
$$

Where the double subscript denotes second partial derivatives

$$
\mathbf{x}_{u u}(u, v)=\frac{\partial^{2} \mathbf{x}}{\partial u^{2}} \quad \mathbf{x}_{v v}(u, v)=\frac{\partial^{2} \mathbf{x}}{\partial v^{2}} \quad \mathbf{x}_{u v}(u, v)=\mathbf{x}_{v u}(u, v)=\frac{\partial^{2} \mathbf{x}}{\partial u \partial v}
$$

The second fundamental form measures the correlation between the change in the normal vector $d n$ and the change in the surface position at a point ( $u, v$ ) as a function of small movement $(d u, d v)$ in the parametric space. Besl and Jain [9] have discussed the properties of first and second fundamental forms in detail. We will consider some of the important properties of Gaussian and Mean curvature in the following paragraphs.

It can be shown that the [g] matrix and the [b] matrix elements are the continuous functions with continuous second and first partial derivatives respectively and that they uniquely determine the surface type. From the [g] and [b] matrices calculated above surface shape and intrinsic surface geometry can be uniquely determined.

The Gaussian curvature function $K$ of a surface can be defined in terms of the two matrices as :

$$
K=\operatorname{det}\left(\left[\begin{array}{ll}
g_{11} & g_{12} \\
g_{21} & g_{22}
\end{array}\right]^{-1}\right) \operatorname{det}\left(\left[\begin{array}{ll}
b_{11} & b_{12} \\
b_{21} & g_{22}
\end{array}\right]\right)
$$


$H>0$ $G>0$

$M>0$ $G=0$
$M<0$
$G<0$
$M=0$
$G=0$


$$
\begin{aligned}
& M>0 \\
& G<0
\end{aligned}
$$



M: Mean
G: Gaussian

Figure 25: Basic surface types in range images (a) surface types (b) table of surface types. and the mean curvature of a surface is defined as :

$$
H=\frac{1}{2} \operatorname{tr}\left(\left[\begin{array}{ll}
g_{11} & g_{12} \\
g_{21} & g_{22}
\end{array}\right]^{-1}\right) \operatorname{det}\left(\left[\begin{array}{ll}
b_{11} & b_{12} \\
b_{21} & g_{22}
\end{array}\right]\right)
$$

The two types of curvatures are together referred to as surface curvature functions. They exhibit very important properties that enable them to be used as features for higher level of processing. For detailed discussion on the properties of surface curvature functions see Best and Jain [9]. Some of the relevant properties are summarized below :

1. Surface types can be determined by the sign of surface curvatures. They are shown in figure 25
2. Gaussian curvature exhibits isometric invariance properties.
3. Mean curvature is slightly less sensitive to noise than Gaussian curvature.
4. Gaussian curvature function of a convex surface uniquely determines the surface.
5. Mean curvature function of a graph surface taken together with the boundary curve of a graph surface uniquely determines the graph surface from which it was computed.
6. Gaussian and mean curvature are invariant to arbitrary transformations of the $(u, v)$ parameters of a surface as long as the Jacobian of the transformation is always nonzero.
7. Gaussian and mean curvatures are invariant to rotations and translations of a surface. This property enables us to obtain view-independent characteristics.
8. Gaussian curvature is an isometric invariant of a surface. Therefore it is an intrinsic surface quantity. It is independent of the the way the surface is embedded in the 3-D space.
9. Gaussian and mean curvature are local surface properties.
10. Another important property of surface curvatures is that Gaussian curvature indicates the surface shape at individual surface points. When surface is shaped like an ellipsoid in the Neighborhood of $(u, v), K(u, v)>0$. It is $<0$ for locally saddle-shaped surface and is $=0$ if the surface is flat,rdge-shaped or valley-shaped locally. Mean curvature also indicates surface shapes at individual points when considered together with the Gaussian curvature.

The above observations are very important for surface classification and have been widely studied and used in range image processing. In fact surface characteristics constitute an important part in the realization of the ultimate goal of three dimensional object recognition.

### 4.2 Computing Surface Characteristics of Range Images

Given a range image, our objective is to calculate the Gaussian and mean curvature. To compute surface curvature we need to know the estimates of the first and second partial derivatives of the depth map. Equations to get the partial derivatives can be simplified in the case of the Z-depth format range image. Parameterization takes a very simple form: $\mathbf{x}_{u}=\left[\begin{array}{lll}u & v & f(u, v)\end{array}\right]^{T}$. The $T$ superscript indicates the transpose. This gives following formulas for the surface partial derivative and the surface normal.

$$
\begin{gathered}
\mathbf{x}_{u}=\left[\begin{array}{lll}
1 & 0 & f_{u}
\end{array}\right]^{T} \quad \mathbf{x}_{v}=\left[\begin{array}{lll}
0 & 1 & f_{v}
\end{array}\right]^{T} \quad \mathbf{x}_{u u}=\left[\begin{array}{lll}
0 & 0 & f_{u u}
\end{array}\right]^{T} \\
\mathbf{x}_{v v}=\left[\begin{array}{lll}
0 & 0 & f_{v v}
\end{array}\right]^{T} \quad \mathbf{x}_{u v}=\left[\begin{array}{lll}
0 & 0 & f_{u v}
\end{array}\right]^{T} \\
\mathbf{n}
\end{gathered}=\frac{1}{\sqrt{1+f_{u}^{2}+f_{v}^{2}}}\left[\begin{array}{lll}
-f_{u} & -f_{v} & 1
\end{array}\right]^{T} .
$$

and the six fundamental form coefficients :

$$
\begin{gathered}
g_{11}=1+f_{u}^{2} \quad g_{22}=1+f_{v}^{2} \quad g_{12}=f_{u} f_{v} \\
b_{11}=\frac{f_{u u}}{\sqrt{1+f_{u}^{2}+f_{v}^{2}}} \quad b_{12}=\frac{f_{u v}}{\sqrt{1+f_{u}^{2}+f_{v}^{2}}} \quad b_{22}=\frac{f_{v v}}{\sqrt{1+f_{u}^{2}+f_{v}^{2}}}
\end{gathered}
$$

The expression for Gaussian curvature is given by :

$$
K=\frac{f_{u u} f_{v v}-f_{u v}^{2}}{\left(1+f_{u}^{2}+f_{v}^{2}\right)^{2}}
$$

And the expression for mean curvature is given by:

$$
H=\frac{f_{u u}+f_{v v}+f_{u u} f_{v}^{2}+f_{v v} f_{u}^{2}-2 f_{u} f_{v} f_{u v}}{2\left(1+f_{u}^{2}+f_{v}^{2}\right)^{3 / 2}}
$$

Thus if we are given a depth map function $f(u, v)$ that possesses first and second partial derivatives, Gaussian and mean curvature can be computed directly.

### 4.2.1 Estimation of partial derivatives of Depth Maps

Partial derivatives of the range image can be obtained by fitting a continuous differentiable function that best fits the data. There are various techniques available in mathematics that have been used by computer vision researchers to determine partial derivatives of depth maps.

## Using Discrete Orthogonal Polynomials

Besl and Jain [9] used discrete quadratic orthogonal polynomial fitting at each pixel to estimate derivatives. It is possible to control Neighborhood size for making local estimates which is important in case of actual range images.

A quadratic surface is fit at each pixel in the image, using a window convolution operator of size desired by the user.

Each point in the given window is associated with a position $(u, v)$ from the set $U X U$ where N is odd :

$$
U=\left[\frac{-(N-1)}{2}, \ldots,-1,0,1, \ldots, \frac{(N-1)}{2}\right] .
$$

The following discrete orthogonal polynomials provide the quadratic surface fit :

$$
\phi_{0}(u)=1 \quad \phi_{1}(u)=u \quad \phi_{2}(u)=\left(u^{2}-\frac{M(M+1)}{3}\right)
$$

Where $M=(n-1) / 2$. The $b_{i}(u)$ functions are normalized orthogonal polynomials :

$$
\begin{gathered}
b_{0}(u)=1 / N \quad b_{1}(u)=\frac{3 u}{M(M+1)(2 M+1)} \\
b_{2}(u)=\frac{1}{P(M)}\left(u^{2}-\frac{M(M+1)}{3}\right)
\end{gathered}
$$

Where $P(M)$ is a fifth - order polynomial in M :

$$
P(M)=\frac{8}{45} M^{5}+\frac{4}{9} M^{4}+\frac{2}{9} M^{3}-\frac{1}{9} M^{2}-\frac{1}{15} M .
$$

$b_{i}(u)$ vectors are computed according to the window size. First the surface estimate function $\hat{f}(u, v)$ is calculated :

$$
\hat{f}(u, v)=\sum_{i, j=0}^{2} a_{i j} \phi_{i}(u) \phi_{j}(v)
$$

that minimizes the mean square term :

$$
\epsilon=\sum_{(u, v) \in U^{2}}(f(u, v)-\hat{f}(u, v))^{2}
$$

Coefficients are given by :

$$
a_{i j}=\sum_{u, v \in U^{2}} f(u, v) b_{i}(u) b_{j}(v)
$$

The first and second partial derivatives can then be directly read from the $a_{i j}$ coefficients :

$$
f_{u}=a_{10} \quad f_{v}=a_{01} \quad f_{u v}=a_{11} \quad f_{u u}=2 a_{20} \quad f_{v v}=2 a_{02}
$$

After the first and second partial derivatives are determined, surface characteristics at each pixel are calculated.

## Using Difference Operators

Brady etal [4] have used $3 \times 3$ difference operators to locally compute first and second derivatives of the Gaussian smoothed surface. Neighborhood size cannot be increased in this method. The operators are :

$$
\begin{gathered}
{\left[\begin{array}{lll}
-1 & 0 & 1 \\
-1 & 0 & 1 \\
-1 & 0 & 1
\end{array}\right]\left[\begin{array}{ccc}
1 & 1 & 1 \\
0 & 0 & 0 \\
-1 & -1 & -1
\end{array}\right]\left[\begin{array}{ccc}
-1 & 0 & 1 \\
0 & 0 & 0 \\
1 & 0 & -1
\end{array}\right]} \\
\end{gathered}
$$

## Using B-Spline fitting

Yang and Kak [33] have derived $3 \times 3$ operators using B-splines for computing partial derivatives of a range map. These can be combined with Gaussian operator to increase the window size and reduce sensitivity to noise. The operators give partial derivatives at the center pixel of each operator.

$$
\begin{gathered}
\mathbf{x}_{u}: \frac{1}{12}\left[\begin{array}{ccc}
-1 & -4 & -1 \\
0 & 0 & 0 \\
1 & 4 & 1
\end{array}\right] \quad \mathbf{x}_{v}: \frac{1}{12}\left[\begin{array}{ccc}
-1 & 0 & 1 \\
-4 & 0 & 4 \\
-1 & 0 & 1
\end{array}\right] \\
\mathbf{x}_{u u}: \frac{1}{6}\left[\begin{array}{ccc}
1 & 4 & 1 \\
-2 & -8 & -2 \\
1 & 4 & 1
\end{array}\right] \quad \mathbf{x}_{v v}: \frac{1}{6}\left[\begin{array}{ccc}
1 & -2 & 1 \\
4 & -8 & 4 \\
1 & -2 & 1
\end{array}\right] \quad \mathbf{x}_{u v}: \frac{1}{4}\left[\begin{array}{ccc}
1 & 0 & -1 \\
0 & 0 & 0 \\
-1 & 0 & 1
\end{array}\right]
\end{gathered}
$$

## Least Squares Polynomial Fitting

We have used a fast least squares fitting method to derive partial derivatives in the symmetric Neighborhood of a pixel. This method allows the Neighborhood size to be controlled.

A surface fit of order $n$ can be written as :

$$
f(x, y)=\sum_{i, j=0}^{i+j \leq n} a_{i j} x^{i} y^{j}
$$

We have used second ( $n=2$ ) order fitting in the Neighborhood of every pixel to compute first and second order derivatives. Clearly, since the pixel at which derivatives are computed is at the origin, we get :

$$
\begin{gathered}
x=0 \text { and } y=0 \\
\frac{\partial f(x, y)}{\partial x}=a_{10} \frac{\partial f(x, y)}{\partial y}=a_{01} \\
\frac{\partial^{2} f(x, y)}{\partial x^{2}}=2 a_{20} \frac{\partial^{2} f(x, y)}{\partial y^{2}}=2 a_{02} \\
\frac{\partial^{2} f(x, y)}{\partial x \partial y}=\frac{\partial^{2} f(x, y)}{\partial y \partial x}=a_{11}
\end{gathered}
$$

Thus derivatives are read off directly from the coefficients. We have also used the general least squares fitting procedure for fitting polynomial on surface patches. For the purpose of


Figure 26: Effect of uniform Gaussian Smoothing
computing derivatives it is observed that we always have symmetric Neighborhood around the pixel. This fact simplifies the least squares equations. See appendix $B$ for the simplified least square fitting equations for second order bivariate approximation in a symmetric Neighborhood.

### 4.2.2 Results of Initial Segmentation

The above mentioned process is applied to actual range images and results are shown in figures $27,28,29,30,31,32,33$.

The smoothing behavior of Gaussian operator was briefly discussed in chapter 2. It is observed that step edges in range images are actually adjacent convex and concave edges. This is further amplified after smoothing the image with any size of Gaussian operator. Brady etal. [4] have restricted the Gaussian application to inside of the region. We have used Gaussian uniformly in the range image with the intention of uniformly smoothing the image for the purpose of obtaining reliable curvature estimates (see figure 26)


Figure 27: curvature estimation (a) original image. $192 \times 25612$ bits/pixel image (b) smoothed image. (c) regions. (d) error in fitting


Figure 28: curvature estimation left to right,from top; original image. $150 \times 1508$ bits/pixel image; error in fitting; segmented regions; flat regions; convex regions; concave regions.


Figure 29: Initial labeling of scene with different threshold values (a) 0.01 (b) 0.02 (c) 0.03 (d) 0.04 .


Figure 30: Labeling of scene :(a) all regions (b) convex (c) concave (d) flat.


Figure 31: Thresholded left : gauss. right : mean. black indicates zero, gray is positive and white is negative value


Figure 32: Histogram of Gaussian Curvature


Figure 33: Histogram of Mean curvature

For surface characterization purpose we use higher sigma value ( $=1.5$ ) and for post segmentation processing we work at lower level in scale. Step Edges detected at sigma $=1.0$ are used to detect region boundaries in higher level processing stage discussed in next section.

Although the response of Gaussian and Mean curvature is reliable it is necessary to threshold the values around zero. $1 \%$ of maximum was used as threshold in all examples. Figures 27 and 28 show the labeled regions and error in fitting second order polynomial at the Neighborhood of each pixel, in images with 12 bits/pixel and 8 bits/pixel respectively. The fit-error is appreciable at boundaries including smooth edges. This means that curvature estimates at the edges are not reliable. At such points curvature magnitude may not be reliable though sign of curvatures is reliable. Further results are shown only for image in figure 27.

Figure 29 shows the effect of threshold values on curvature signs. As threshold values for Gaussian and mean curvature is changed, pixel labeling may change if the curvature magnitude is not appreciable. image thresholded at $1 \%$ of maximum curvature magnitude (see figure 29(c)) has correct labelings. Further results are shown only for this threshold value. Pixels are classified as one of the eight basic types. We can classify the entire image into concave, convex, and flat regions by simply merging all neighboring pixels having similar type of surface, i.e, flat,concave or convex (see figure 30. Thresholded values of Gaussian and mean curvature are shown in figure 31. White patches indicate zero magnitude, gray indicate positive magnitude and black indicates negative curvature magnitude. It is observed that Gaussian curvature is mostly zero except for isolated patches, since the image has no spherical object. Non-zero mean curvature values are obtained at step edges and on a cylindrical object. Histogram of the magnitude of Gaussian and mean curvature (figure 32 and 33) for the entire image show appreciable mean curvature magnitude in the image and no significant Gaussian curvature. It can therefore be inferred that the scene has flat and possibly cylindrical objects.

### 4.3 Post processing of Labeled scenes

The segmentation done by labeling the individual pixels using sign of Gaussian and mean curvature is local in nature and threshold dependent. In order to interpret these labelings globally, we need to process the the labeled image with globl constraints. Besl and Jain [25] have proposed a variable order surface fitting algorithm. Surface patches are described as linear,quadric or cubic.

Our approach depends on the actual requirements. We describe two methods, ( both are preliminary ) to obtain useful segmentation given labeled image. The first method simply groups convex patches to form connected convex subparts of the scene. Second method uses the segmented objects obtained from algorithm described in chapter 3.

### 4.3.1 Obtaining Convex patches

As noted in third chapter, detection of smooth edges is difficult to extract using only local information. Curvature information at the all types of edges is easy to record. From figure 26 it is clear that edges in smoothed images can be recorded as thin convex and concave regions. In particular, convex edges are of convex cylinder type, with zero Gaussian curvature but appreciable negative mean curvature. Similarly, concave edges are of concave cylinder type, with zero Gaussian curvature but appreciable positive mean curvature. Thus all types of edges give either convex or concave cylindrical response. But the edge response is obtained over wider region due to smoothing and large window size during derivative computation. It is therefore not possible to have exact localization of patches obtained by merging convex regions.

A simple algorithm for obtaining convex patches is given below :

1. Read the labeled image.
2. Label each patch as a region.
3. Initialize the region data structure to record surface type, number of pixels, topmost pixel in the region, neighbours of the region,extremities of the region and the label


Figure 34: Convex patches


Figure 35: Convex patches
assigned to the region.
4. For the next unprocessed topmost region of the type flat or peak(convex sphere) or ridge (convex cylinder) with acceptable number of pixels do:
(a) Extend the original region to include all neighboring regions of type flat or ridge. Other types of regions are considered concave or part of other convex subpart. peak patches are not included because they will be selected as seed region.
(b) Repeat the above step to extend the region, till it is not possible to grow any more.
5. Output the convex subparts. End.

Figures 34 and 35 show convex patches obtained from labeled image obtained in figure27 and in figure 28(a) respectively. Majority of objects in figure 34 are merged into one convex patch while they are separated in figure 35

### 4.3.2 Object Surface Classification

Surfaces on the segmented objects can be classified as one of basic surface type using the initial labeling based on sign of curvatures. Yang and Kak [33] have used extended Gaussian images to identify surface type on isolated surfaces. Histogram of labels in an isolated object can give some idea about the surface and guide the surface fitting process.

The classification algorithm is as follows :

1. Read in the segmented objects image and labeled surface image.
2. For each object in the image do :
(a) Erode the object in labeled image so as to remove points within 5 pixel distance of the object boundary. This reduces the effect of smoothing and window size during curvature estimation which is mainly contributed by pixels near the boundary, and does not reflect the nature of region.


Figure 36: Classified surfaces
(b) Histogram the remaining pixel-label values.
(c) If more than $90 \%$ pixels are of one type,either flat or cylinder or sphere then the surface can be classified as such. If there are two or more peaks in the histogram, object has more than one surface type.
(d) In single surface cases fit the best fitting surface on the points. Output the description of surface.
(e) Further processing by region growing by surface fitting is necessary to smoothen the surface patches. Fit surfaces on individual patches and merge them by region growing.

This algorithm is being implemented. Initial classification process will classify the surfaces in figure 11 as 5 plane surfaces, 1 cylindrical and 1 irregular surface ( the film mailer). See figure 36.

First and second order polynomials were fitted on flat and non-flat surface patches respectively in image of figure 11. The reconstructed image is shown in figure 37 . Besl


Figure 37: Original and reconstructed Images. left: original images right: reconstructed images obtained by fitting 1st and 2nd order surfaces on patches labeled by segmentation process.
and Jain $[13,25]$ have used initial labeling to obtain seed regions in the final region growing process. They perform variable order surface fitting to approximate the scene as a collection of piece-wise continuous functions.

## Chapter 5

## Discussion

Though results of running the various algorithms described in previous chapters on images acquired from different scanners are consistent, there is scope for refinement of all the approaches. We will discuss the merits and demerits of each method and suggest improvements.

We need to study the scale-space behavior of range images in detail. This would lead to a better understanding of the scale at which range images should be handled. We have noticed that thresholding of zero-crossings makes the entire segmentation procedure dependent on the threshold value. Though we have obtained consistent results with a fixed empirically determined value for all the images obtained from a particular scanner, threshold selection is not automatic. Secondly, even with right threshold value the region may not be completely bounded by the zero-crossings (in case of overlaps by thin objects or sensor noise) To make the whole process less sensitive to threshold, following post-processing steps (region splitting ) are suggested :

1. Read in the segmented object.
2. Trace the contours around the object as it is defined now and also any other boundaries that are now lying inside the object. Except for the bounding contours, other contours may not be closed. They may simply lie within the region and actually are boundary of the real object. In such a case mark the beginning and end of the contour. If the
contour touches the closed contour then mark the point of contact as end of inner contour.
3. In all the contours mark the concavities.
4. Now split the region by connecting two contours (gap filling) or connecting two points of concavity ( gap filling or region splitting) or connecting an end point of contour with a concavity, based on predetermined gap filling distance.
5. The output is the segmented object.

The above method should be indifferent to threshold values on higher side as it splits the region consisting of more than one regions. To reduce the sensitivity to low threshold values (which will result too many small regions) some sort of merging is required. Merging is a much difficult task, so it is better to keep the threshold high and have the post-segmentation process perform the splitting, rather than initial segmentation performing splitting due to low threshold value.

Another solution to splitting is to let higher level recognition process make globally valid observations to split the region. The higher level procedure may use a priori information or may make some assumptions or apply global constraints to split the region. Franc Solina's (see [23]) superquadric procedure can split the regions into identifiable parts by performing model fitting on individual part of the object.

In chapter 4 we noticed that labeling of the scene based on curvature sign is threshold sensitive. While thresholding around zero is necessary to obtain meaningful results, it is not clear how that value can be automatically determined. Curvature determination being local, the labeling is sensitive to noise and surface texture. It is not well understood how to generate a global interpretation of such surfaces.

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## Appendix A

## 2nd order Least squares fitting in symmetric neighborhood

The approximating polynomial is written as :

$$
I(x, y)=a_{20} x^{2}+a_{11} x y+a_{02} y^{2}+a_{10} x+a_{01} y+a_{00}
$$

The square error term is :

$$
\begin{gathered}
\Sigma^{2}=\sum_{i=1}^{N}\left(Z_{i}-I_{i}\right)^{2} \\
\Sigma^{2}=\sum_{i=1}^{N}\left(Z_{i}^{2}-2 Z_{i} I_{i}+I_{i}^{2}\right)
\end{gathered}
$$

To minimize the least squares term, let $\frac{\partial \Sigma^{2}}{\partial X_{i}}=0$.
which is :

$$
\begin{gathered}
\frac{\partial \Sigma^{2}}{\partial a_{20}}=-2 Z_{i} x_{i}^{2}+2 x_{i}^{2} I_{i}=0 \\
\frac{\partial \Sigma^{2}}{\partial a_{11}}=-2 Z_{i} x_{i} y_{i}+2 x_{i} y_{i} I_{i}=0 \\
\frac{\partial \Sigma^{2}}{\partial a_{02}}=-2 Z_{i} y_{i}^{2}+2 y_{i}^{2} I_{i}=0
\end{gathered}
$$

$$
\begin{gathered}
\frac{\partial \Sigma^{2}}{\partial a_{10}}=-2 Z_{i} x_{i}+2 x_{i} I_{i}=0 \\
\frac{\partial \Sigma^{2}}{\partial a_{01}}=-2 Z_{i} y_{i}+2 y_{i} I_{i}=0 \\
\frac{\partial \Sigma^{2}}{\partial a_{00}}=-2 Z_{i}+2 I_{i}=0
\end{gathered}
$$

Writing :

$$
S_{j l k}=x_{i}^{j} y_{i}^{l} z_{i}^{k}
$$

we get :

$$
\left[\begin{array}{cccccc}
S_{400} & S_{310} & S_{220} & S_{300} & S_{210} & S_{200} \\
S_{310} & S_{220} & S_{130} & S_{210} & S_{120} & S_{110} \\
S_{220} & S_{130} & S_{040} & S_{120} & S_{030} & S_{020} \\
S_{300} & S_{210} & S_{120} & S_{200} & S_{110} & S_{100} \\
S_{210} & S_{120} & S_{030} & S_{110} & S_{020} & S_{010} \\
S_{200} & S_{110} & S_{020} & S_{100} & S_{010} & S_{000}
\end{array}\right]\left[\begin{array}{c}
a_{20} \\
a_{11} \\
a_{02} \\
a_{10} \\
a_{01} \\
a_{00}
\end{array}\right]=\left[\begin{array}{c}
S_{201} \\
S_{111} \\
S_{021} \\
S_{101} \\
S_{011} \\
S_{001}
\end{array}\right]
$$

In a symmetric neighborhood :

$$
\begin{gathered}
S_{p q 0}=0 \text { for odd } p \text { or } q \text { and } \\
S_{p q 0}=S_{q p 0}
\end{gathered}
$$

The above system of equations reduces to :

$$
\left[\begin{array}{cccccc}
S_{400} & 0 & S_{220} & 0 & 0 & S_{200} \\
0 & S_{220} & 0 & 0 & 0 & 0 \\
S_{220} & 0 & S_{040} & 0 & 0 & S_{020} \\
0 & 0 & 0 & S_{200} & 0 & 0 \\
0 & 0 & 0 & 0 & S_{020} & 0 \\
S_{200} & 0 & S_{020} & 0 & 0 & S_{000}
\end{array}\right]\left[\begin{array}{c}
a_{20} \\
a_{11} \\
a_{02} \\
a_{10} \\
a_{01} \\
a_{00}
\end{array}\right]=\left[\begin{array}{c}
S_{201} \\
S_{111} \\
S_{021} \\
S_{101} \\
S_{011} \\
S_{001}
\end{array}\right]
$$

Which can be written as :

$$
a_{10}=\frac{S_{101}}{S_{200}} a_{01}=\frac{S_{011}}{S_{200}} a_{11}=\frac{S_{111}}{S_{220}}
$$

and

$$
\left[\begin{array}{ccc}
S_{400} & S_{220} & S_{200} \\
S_{220} & S_{040} & S_{020} \\
S_{200} & S_{020} & S_{000}
\end{array}\right]\left[\begin{array}{l}
a_{20} \\
a_{02} \\
a_{00}
\end{array}\right]=\left[\begin{array}{l}
S_{201} \\
S_{021} \\
S_{001}
\end{array}\right]
$$

## Appendix B

## Source Code Listing

Listings of source programs is included in following pages.

1. space.c : Performs smoothing,median filtering, Gaussian filtering, Laplacian, marking zero-crossings and graphics display of histogram etc. in interactive manner.
2. segment.c : Segments all objects and topmost object in the scene, given original image and zero-crossings of LOG image. Also outputs supporting points of topmost object.
3. rca_calib.c : Generates a list of points for Z-depth format image. Originally written by Franc, modified to read PM files and add points horizontally and vertically.
4. classify.c : Procedure used to select and classify the superquadric models.
5. spline.c : Computes various surface characteristics of the image. Interactively displays results and histograms using quickdraw routines. Outputs labeled image and other characteristic as desired by the user.
6. grad.c : Has code for fast least squares fitting. Polynomial is fitted in the symmetric neighborhood of the pixel.
7. merge.c : Performs processing on the image labeled according to signs of Gaussian and mean curvature. It computes convex parts of the image, fits polynomials on patches using general least squares routines in solver.c.
8. solver.c : Has code for a general least squares fitting procedure.

There are other supporting programs that are vital to the algorithms. Superquadric fitting programs are developed by Franc Solina and are not listed here.
space.c Mon Apr 18 18:07:36 1988 $\quad 1$

Program for computing scale - space description of the input imag and lots of other things in interactive manner. Later will overri scale.c.

Modifications for display of histogram done on Feb 18, 1988.
March 241988 : To read/write in both PM_C and PM_S formats.
April 11988 : All buffers made floating pt. Computation of lapla and zerocrossing made floating pt.
\#include <stdio.h>
\#nclude <math.h>
\#include <local/pm.h>
\#include <ik.h>
/* \#nclude <local/qkopt.h> */
\#define BUFSIZE 256 /* Image buffer size *
\#define BUF_NUM 4 /* of buffers available for manipulation */
float result [BUFSIZE][BUFSIZE]
float buffer[BUF_NUM] [BUFSIZE][BUFSIZE]: /* 0 stores original image * int temp_buf [BUFSIZE]:
/* temporarily store a buffer
float $x[B U F S I Z E], y[B U F S I Z E] ; / *$ to store points */
int threshold;
/* Threshold for detecting zero-xings */
int booll,bool2,bool3
char *cmt;
char input_filename[50],
int sizex,sizey:
/* size of the image */
pmpic *pm1;
u_int image_format; /* stores the format of the last image read
float min();
float get_median();
float abszz();
float squarezz() :
$\operatorname{main}(\operatorname{argc}, a r g v)$
int argc:
char **argv;
i

## apace.c Mon Apr 18 18:07:36 1988 2

FILE *infile,*out; /* pointers to input image and output image fil FILE *out2;
FILE *infs, *out_pmsmooth, *output_pm
char *smooth_cmt;
char ikonas_disp[10];
char *out_cmt;
int count;
unsigned char c;
static int nhb_x[8] $=\{0,1,1,1,0,-1,-1,-1\}$;
static int nhb_y[8] $=\{1,1,0,-1,-1,-1,0,1\}$;
float nbd[8], temp;
int $1, j, k, 1, m, n, b$;
float nbr[10]:
int gsize;
float sigma; /* size and sigma of the gaussian operator */
float sum:
float gauss[60]:
float gsum;
int offset:
char input [20];
$\begin{array}{ll}\text { char input[20]; } & \text { /* offset coordinates intialized } \\ \text { int offx }=0, \text { offy }=0 ; & \text { /* name of the output file */ }\end{array}$
int b1,b2,b3;
unsigned char *pm_point;
int disp_row;
short int *pms_point; /* pointer to short integer to hand
int factor:

PMS format */
/* \#y which PM_S image pixel to b for display on IKONAS */
printf("argc : \& ${ }^{\prime}$ \n", argc);
1f (argc $!=2$ )
1
printf("usage : scale <input-image-file-pmpic>\n"): exit (0):
\}
printf("want to display on ikonas ? "):
scanf("\%s", \&ikonas_disp[0]):
printf("read $\left.\backslash n^{\prime \prime}\right)$;
/* open the ikonas display. value of env. variable is taken */
if (strcmp ("y",ikonas_disp) == 0)
if (ikopen (NULL) $==-1$ )
f
printf("can't open ikonas. exiting $\backslash n$ ") ; exit(0):

```
spaco.c
Hon Apr 18 18:07:36 1988
    }
    /* get comment line */
    cmt = pm_cmt(argc,argv);
    /* open input pm file */
    read picture (argv[1],0):
    read picture(argv[1],1);
    printf("Rows : %d Columns : %d\n",sizex,sizey)
/* procesing of the commands starts now
    various avallable commands are :
        1. gauss : convolves the 1mage with gaussian filter
        2. cross : computes zero and other types of crossings in the gi
        buffer.
        . save : saves indicated buffer in a file.
        4. disp : displays indicated buffer on the ikonas.
        5. add : adds the two buffers.result is put in buffer
        6. sub : subtracts two buffers. result is put in buffer 1
        7. buffer: selects the current buffer.
        8. offset: offset the picture on 1konas
        . original : indicated buffer gets original picture
        0.read : Reads a flle in the designated buffer
        11.hist : Computes and displays the historam using quickdraw.
        12.1kpm : saves the 1mage displayed on 1konas in the a file in
    |five active buffers are maintained to manipulate the original ima
*/
printf(">");
    while(scanf("%s",1nput) != EOF)
    {
        If((strcmp(1nput,"median") == 0) || (strcmp(1nput,"m") == 0))
            {
            * do median filtering of the image */
            = readbuffer()
            for(1=1:1<(s1zex-1):1++)
            for(j=1;j<(sizey-1);j++)
            f
                count = 0
                    for (m=1-1;m<=1+1;m++
                    for(n=j-1;n<=j+1;n++)
                    {
                            nbr[count] = buffer[b][m][n];
```


## Mon Apr 18 18:07:36 1988

## count++;

\}
esult[1][f] = get_median(nbr)
\}
for $(1=1 ; 1<($ sizex- $)$; $1++$
or $(1=1 ; 1<($ sizex -1$) ; 1++)$
for $(j=1 ; j<($ sizey -1$) ; j++)$
buffer[b][1][j] = result[1][j]:
\} /* end of median filtering */
else if((strcmp(input,"gauss") ==0) || (strcmp(input,"g")== 1
$b=$ readbuffer()
printf("sigma :") ;
scanf("\%f",\&sigma)
printf("size of window :"):
scanf("\&d", \&gsize);
/* compute the gaussian array */
gsum $=0$;
for $(1=-$ gsize/2, $j=0 ; 1<=$ gsize/2;1++, $j++$ )
1
gauss[ $f]=(1.0 /(\operatorname{sqrt}((\text { double })(2.0 * 3.1415926)) * \text { sigma }))^{*}$ gsum $+=$ gauss [ 1$]$
printf("gauss[fd] = \&f",1,gauss[f]):
\}
1f (gsum $==0$ ) gsum $=1$;
printf("\n gsum = ff $\ln$ ", gsum) ;
/* seperably convolve x-axis */
for ( $j=0 ; j<s i z e y ; j++$ )
for ( $1=$ gsize/2; $1<=($ sizex-gsize/2) ; $1++$ )
1
$\operatorname{sum}=0 ;$
for $(k=$;
for (k=-gsize/2;k<= gsize/2;k++
sum += buffer[b][1+k][j]*gauss[k+gsize/2]
\}
result [1][f] = sum/gsum:
\}

```
apace.c
    Mon Apr 18 18:07:36 1988
    /* seperably convolve y-axis */
    for(1=0;1<sizex;1++)
        for(j=gsize/2;j<=(sizey-gsize/2); ; ++ )
            l
            sum = 0;
            for(k= -gsize/2;k<=gsize/2;k++)
                    sum += result[1][j+k]*gauss[k+gsize/2];
            }
        }
        else if((strcmp("lap", input) == 0) | (strcmp("lm,input) == 0))
    {
        b readbuffer()
        /* apply laplacian operator */
    for(1=1;1<(sizex-1); 1++)
        for (j=1;j<(sizey-1); j++)
            for(j=1;j<(sizey-1); ; ++)
                result[1][y] = -4*bu
                    +buffe
                    +buffer[b][1][j-1]+buffer[b][1][j+1];
    for(1=1:1<(sizex-1);i++)
            for(j=1; f<(sizey-1); f++)
            buffer[b][1][J] = result[1][f];
else if((strcmp("cross",1nput) == 0) || (strcmp("c", input) == 0
    b = readbuffer()
    printf("step edge magnitude desired ? enter 1 if yes >"):
    canf("%d",&booll)
    printf("concave edge magnitude desired ? enter 1 if yes >")
    scanf("%d", &bool2).
    printf("convex edge magnitude desired ? enter 1 if yes >");
    canf("%d",&bool3),
    /* trace zeros */
    for(1=1;1<(sizex-1);1++)
        for(j=1; j<(s1zey-1); j++)
            {
```

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```
for(n= 0;n<8;n++
                    f
                        nbd[n] = buffer[b][1+nhb_x[n]][f+nhb_y[n]];
                            dge p(buffer[b][1][f], nbd, &temp):
                resuIt[1][f] = temp:
            }
        for(1=0;1<(sizex);1++)
            for (j=0;j<(sizey); j++)
                buffer[b][1][f] = result[1][f];
    }
else if((strcmp("offset",input) == 0) || (strcmp("off",input) =
    {
        printf("coordinates :");
        scanf("%d", &offx);
        scanf ("%d", &offy);
    }
else if(strcmp("robert",input) == 0)
    i
        b1 = readbuffer();
            for(1=0;1<(sizex-1);1++)
            for(j=0;j<(sizey -1);j++
                    result[1][f]= (float) sqrt((double)
                    (squarezz (buffer[b1][1][j] - buffer[b1][1
            for(1=0;1<(sizex-1);1++)
            for (j=0; j<(s1zey -1); j++)
            or(j=0;j<(sizey -1);j++)
        } /* robert */
else if((strcmp("subtract",input) == 0) || (strcmp("sub", input)
    &
        /* subtract two images */
            printf("b3 = b1 - b2\n");
            b1 = readbuffer()
            b2 readuuffer()
for(1=0;1<sizex;1++
            for(j=0;j<sizey;j++)
            for
            buffer[b3][1][j] = abszz (buffer[b1][1][f] - buffer[b
            }
```



```
space.c
        else if((strcmp("read", input) == 0) || (strcmp("r",input) == 0)
        i
            b = readbuffer()
            printf("enter inputfilename :");
            scanf("&s",input_filename);
            read_picture(input_filename,b):
        }
        else if((strcmp("ikpm",input) == 0) || (strcmp("ikpm",input) ==
            printf("Nothing happened\n");
        else if((strcmp("h1st", input) == 0) | (strcmp("h", input) == 0)
            { b = readbuffer();
            display_histogram(b);
            }
        else if((strcmp("row",input) == 0) || (stramp("rowscan",input)
            {
        b= readbuffer();
        printf("row : ");
        scanf("%d",&disp row)
        while(( disp_row < sizey) && ( disp_row >= 0))
            while
                            display_row_histogram(disp row,b) ;
                printf("
                scanf("%d",&disp_row)
            }
        }
        printf(">");
    }
}
/* compute edge strength and direction */
/* mask is like this :
\begin{tabular}{ccccccc}
\hline-1 & 5 & 1 & 6 & 1 & 7 & 1 \\
\hdashline 1 & 4 & 1 & \(\times\) & 1 & 0 & 1 \\
\hdashline-1 & 3 & 1 & 2 & 1 & 1 & 1
\end{tabular}
```

```
space.c Mon Apr 18 18:07:36 1988 10
*/
edge_p(1ntensity, nbd,p_es)
float intensity, nbd[8].
    { static double delta[8] = {1.0,1.0,1.0,1.0,1.0,1.0,1.0,1.0};
        static unsigned char tbl[9] = {0,1,8,7,6,5,4,3,2};
        float *p ed;
        int 1,j, ed,halfgray,maxgray;
    float res,slope;
    res = 0.0;
    ed = -1;
    halfgray = 0;
    maxgray =255;
    if(booll ==1)
        If (bo
            if(1ntensity > halfgray)
            f
            for (1=0;1<8;1++)
                { if(nbd[1]<halfgray)
                    if(nbd[1]<halfgray)
                            slope = (float)(intensity - nbd[1])/delta[1]:
                                    1f(slope >res)
                                    |
                                    res =slope;
                                    ed = opposite(1);
                                    }
                                    }
            } }
            else if(1ntensity == halfgray)
            f for (1=0.1<8.1++
            i (1=0;1<8;1++)
                j=opposite(1);
                    if((nbd[1]<halfgray) && (nbd[f]>halfgray))
                            ( slope = (float) (nbd[f]-nbd[1])/(delta[1]);
                    if(slope > res)
                                    res =slope;
                    ed =f;
```


## space.c <br> Mon Apr 18 18:07:36 198 <br> 11

## \}

## \}

$1 \quad 1$
1f(bool2 == 1)
if
if(1ntensity > halfgray)
/* if required check for $0+0$
type of crossings because they are generated by concave surfaces
( $1=0 ; 1<8 ; 1++1$
i
$j=o p p o s i t e(1)$
if((nbd[1] == halfgray) \&\& (nbd[f] >= halfgray))
(
slope $=(f l o a t)($ intensity - nbd[1])/delta[1];
if(slope > res)
res = slope;
ed = j ;
\}
\}
)
)
if (bool3 == 1)
if(intensity < halfgray)
/* if required check for $0-$ or or 0 types of crossings because they are generated by concave surfaces
for (1=0;1<8;1++)
1
f=opposite(1)
if((nbd[1] == halfgray) \&\& (nbd[j] <= halfgray))
slope $=($ float $)($ nbd [1] - intensity) /delta[1];
if(slope > res)
res = slope;
ed $=1$;

```
space.c

\section*{\}}
)
\}
\}
If (image_format \(==P M C\) ) res \(=\) res * 5; \({ }^{*} p_{\text {_es }}=\bar{m} \ln (\) res, \((f 1 \mathrm{oa} \bar{t})(\) maxgray \()):\) \({ }^{*} p_{\text {_ }}\) ed \(=\) tbl \([\mathrm{ed}+1]\);
\}
opposite(1)
int 1:
i
Int opp;
if ( \(1<4\) )
opp \(=1+4\);
else
opp \(=1-4\);
return(opp):
\}
float \(\min (1, j)\)
float 1, 3 ;
1
If (i<j) return(1): else return(f):
)
readbuffer ()
1
int \(b ;\)
printf("buffer :"):
scanf("\%d", \&b) :
if( \((\mathrm{b}<0)\) || (b > 5)) (b=readbuffer (); return (b); )
else return(b):
\}
float abszz(num)
float num;
f float 1;
1f (num < 0) \(1=-n u m ;\) else \(1=\) num;
```

epace.c
Mon Apr 18 18:07:36 1988
13
return(1);
}
float get median(nbr)
float nbr[10];
l
float nbrl[5];
sort(nbr, nbrl);
return(nbr1[4]);
}
sort(nbr,nbrl)
float nbr[10],nbr1[5];
i
Int 1,m;
float large;
int index;
for (m=0; m<5; m++)
{
large = -100
for(1=0;1<9;1++)
!
if(large < nbr[1])
{ large = nbr[1]
Index = 1;
}
|brl(m
nbr1 [m] = large;
nbr[1ndex] = 0;
}
}
read_picture(filename,b)
char filename[]:
int b;
int
Int 1,j%
ILE *infs;
unsigned char *pm point;
unsigned char *pm_poin
short int *pms_point;
nt offset;
/* open input pm file */

```
```

space.c Mon Apr 18 18:07:36 1988 14
if ((1nfs = fopen(filename,"r"))== NULL)
{ printf("file open error :%s \n",filename);
exit(0);
}
/* read inputfile into the pmpic buffer */
if((pm1 = pm_read(infs,0))== NULL)
{printf("error in reading the pmfile %s",filename);
exit(0);
}
sizex = (pml->pm_nrow); /* % of rows */
sizey = (pml->pm_ncol); /* \# of columns */
1mage_format = pml->pm_form;
1f(pml->pm_form == PM_C)
(pm_point = (unsigned char *) pml->pm_image;
for(1=0;1<s1zex;1++)
for( }\textrm{j}=0;\textrm{j}<\mathrm{ sizey; j++)
l}\mathrm{ buffer[b][1][f] = *pm_point;
pm_point++;
}
else if(pml->pm_form == PM_S)
1 pms point = (short int *)
pms_point = (short int *) pml->pm_image;
or(1=0;1<sizex;1++)
for( }j=0;j<sizey;j++
{}\mathrm{ buffer[b][1][f] = *pms_point;
pms_point++;
}
else fprintf(stderr,"Image in unrecognized format. Buffer not initi
} /* end of read picture */
display_histogram(buff)
int buff;
int
int 1.j;

```

\section*{space.c Mon Apr 18 18:07:36 1988 \\ 15}
float xmin, xmax,ymin,ymax;
int lopt:
int npts:
int max;
int value;
for ( \(1=0 ; 1<256 ; 1++\) )
\(\mathfrak{f}^{\text {for }}\)
\(x[1]=1 ;\)
\(\mathrm{y}[1]=0\);
\}
qterm (4) :
\(\max =-1000\);
for (1=0;1<sizex; \(1++\) )
for \(\left(\mathrm{j}=0 ; \mathrm{j}\right.\) <sizey; \(\jmath^{++}\)
1
value \(=\) absyy((int) (buffer[buff][1][1])):
\(y[\) value \(]=y\) [value \(]+1\);
if \((\max <y[v a l u e]) \max =y[\) value \(]\);
\}
printf(" max : \(\left.8 d^{\prime \prime}, \max \right)\);
lopt \(=0\);
npts \(=256\);
\(x m 1 n=0.0\);
\(x \max =255.0\)
\(y \min =0.0\);
printf("ymax : "):
scanf("qf", \&ymax):
qkdraw (npts, \(x, y, 10 p t, \& x m i n, ~ \& x m a x, \& y m i n, \& y m a x)\);
qdtitl(" R H "):
qxlabl("Range"):
qylabl("Values"):
qdone ():
\}
absyy (value)
int value;
1
if (value < 0) value = -value;
if(value > 255) return(255);

\section*{pace.c Mon Apr 18 18:07:36 1988 16}
else return(value);
\}
display_row_histogram(row,buff)
int row;
int buff;
1
int 1,j;
float xmin, xmax, ymin,ymax;
int lopt;
int npts;
int max;
int value;
int row_num;
for ( \(1=0 ; 1<s 12 e y ; 1++\) )
\(x[1]=1\)
\}
for ( \(1=0 ; 1<\) sizey; \(1++\) )
\[
y[1]=0 ;
\]
qterm(4) :
\(\max =-1000 ;\)
for ( \(1=0 ; j<s i z e y ; j++\) )
( value = absyy((int) buffer[buff][row][1]): \(y[f]=\) value
\}
lopt \(=0\);
npts \(=\) sizey;
\(x m i n=0.0\);
max = sizey:
\(\min =0.0\);
max \(=255\);
qkdraw(npts, x,y, iopt, \&xmin, \&xmax, \&ymin, \&ymax) :
qdtitl(" R H ") ;
qxabl("Range");
qylabl("Values") :

\section*{space.c Mon Apr 18 18:07:36 1988 17}
qdone () :
)
float squarezz (number)
float number:
return (number*number) :
) return (number*number):

\section*{ogment.C Tue Apr 1918:21:08 1988}

\section*{}

This program performs segmentation on the laplacian edge oper image and marks the topmost object in the range image.

Modified on Aug12,1987.
Modified on Jan28, 1988 to extract supporting surface information for the segmented object.
Feb 1, 1988 : Recursive call to grow regin modifled to incorporate control over the of pending recursive calls at given time. This requires maintaining a FIFO queue of open nodes.
March 24,1988 : Modified to accept PMC and PM_S format images. Ou lmages in respective format.
region label 1 is reserved for unaccpted regions. max regions allowed is determined by MAX_REGION = MAX_REGION-1 to 2.
```

include <stdio.h>
include <math.h>
include <ik.h>
Include <local/pm.h>
define MIN_VALUE 0 /* = 0 if background > shadows (non-zero backgro
define MIN ACCEPTABLE 300 else if background $=0=$ shadows then $=-1$ \%/
Udefine MIN ACCEPTABLE $300 / \star$ Minimum of pixels in a valid region
define BUF 256 /* Buffer size in one dimension */
define MAX_NUM_CALL 500 /* Maximum of pending recursive calls
at a given time. To avoid segmentation
Idefine MAX REGION 2550 /* Maximum of regions allowed */
define UNACCEPT_REGION 1 /* Label of rejected regions */
\#define PIXEL_STACK_SIZE $10000 / *$ size of the pixel stack used for
recursive-iterative region growing */
nt range_image_buffer[BUF][BUF]; $/ \star$ input image buffer */
int lap_image_būffer[BUF][BUF]; /* laplacian image buffer */
int third_image_buffer[BUF][BUF]; $/ \star$ temporary buffer for marking
int output_buffer[BUF][BUF];
int support_image_buffer[2][BUF][BUF]:
visited points */
/* segmented 1mage buffer wit labeled object */
/* supported surface */

```
int average image[BUF][BUF];
int vector [65000];
/* average \(2 \times 2\) stored */
int edge_threshold;
/* threshold for edge strengt
/* global variables to store and column values */
long int count:
int distance;
int seedrow, seedcol;
int back threshold;
int a_depth:
float squarezz () ;
char accept_region()
1kword value = 255;
char 1konas_disp[10];
pmp1c *pm1, \({ }^{\text {ppm2, *pm3; }}\)
int label;
int pixels;
int region count:
int invalid_region_count:
struct region_type 1
int number;
int valid;
int size;
int max;
int min;
\} region [MAX_REGION]:

Int row_max;
int colūmn_max;
int max:
/* maximum allowable distance the seed region */
/* seed region coordinates *
/* threshold for background *
/* average depth */
/* Run time display on Ikonas
/* Pointers to PM-picture st
/* label value to identify re
* * of pluels in the
/* of pixels in the current
/* counts only invalid region
/* Label of the region */
/* \(=0\) if not valid. \(=1\) if val
/* \# of pixels in the region
/* maximum pixel of the regio
/* minumum pixel of the regio
struct p_stack \(\{\)
(* row number of max depth pixel */
* row number of max depth pixel */
/* depth value of the max depth pixel

1nt row;
) pixel_stack[PIXEL_STACK_SIZE];
int rownum:
int colnum;
int num_call;
int stack_length;
int current_element;
int nextrow, nextcol;
int top_region_label;
int max_pixel_region
int min_pixel_region:
int gaprow, gapcol:
/*(rownum, colnum) is the next pixel * /* popped from the stack */
/* number of calls pending at a given /* =current_element if queue is empty points to the tall of the queue \(*\) /* points to head of the queue */
/* nextcol and nextcol for the seed re /* label of the topmost region */
/* maximum pixel of the region */
/* minimum pixel of the region */
/* the nbd edge pixel found, returned
```

segment.c Tue Apr 19 18:21:08 1988 3
gapfiller(), if point lies one dist
from the edge */
pmpic *read_pmpic_int();
/************************************MAIN************************************
main(argc,argv)
int argc;
char *argv[];
{
FILE *range,*laplacian,*outfile,*outfile2,*outfile3;

```

```

int row,column; /* of row and columns in in
int offset;
char *cmt;
int done,found: /* used as booleans */
unsigned char *pm_point, *pm_point2, *pm_point 3
short int *pms_point, *pms_point2,*pms_point3;
u_1nt image_format; /* stores 1mage format PM_C o
If(argc != 3)
|
printf("segment : usage : segment <range_file> <lap_edge_
exit(1):
}
cmt = pm_cmt (argc,argv):
printf("Want to display on IKONAS ? ");
scanf("\&s",\&1konas_disp(0]):
/* open the IKONAS display. value of env. variable is taken *
if(strcmp("Y",1konas_disp)"== 0)
if (1kopen (NULL) ==-1)
{
printf("Can't open IKONAS. Exiting\n"):
exit(0):
}
if((range = fopen(argv[1],"r")) == NULL)
printf("segment : Cannot open %s\n",argv[1]):
exit(0);
}

```
sogment.c Tue Apr 19 18:21:08 1988
```

if((laplacian = fopen(argv[2],"r")) == NULL)
{ printf("segment : cannot open %s\n" argy[2])
printf("segment : Cannot open %s\n",argv(2]):
exit(0):
}
If ((pm1 = pm_read(range,0)) == NULL)
{ printf("Error in reading PM file : %s\n",argv[1]);
printf("
}
If((pm2 = pm_read(laplacian,0)) == NULL)
(')
exit(0);
}
/* open output files */
strcpy (output, argv[1]);
f((outfile = fopen(output,"w"))== NULL
{ printf("segment : Cannot open output file : %s\n",out
exit(0):
}
strcpy (output,argv[1]);
strcat (output,".top"):
If((outfile3 = fopen(output,'"w)) == NULL)
{ printf("segment : Cannot open output file : \&s\n",out
exit(0);
}
strcpy(output,argv[1]);
strcat (output,".support")
strcat (output,".support");
f((outfile2 = fopen(output, "w")) == NULL)
printf("segment : Cannot open output file : %s\n",out
exit(0):
}
row = pml->pm nrow;
column = pml->pm ncol
Image_format = pm1->pm_form;

```

\section*{sogmont.c Tue Apr 19 18:21:08 1988 5}

\section*{if((row != pm2->pm_nrow) || (column != pm2->pm_ncol)}
i
printf("Input images are not of same size. Exiting \(\ln\) "); exit(0):
\}
nrow = row; /* initialize global variables nrow and ncol * ncol \(=\) column;
printf("Edge threshold :");
scanf("fd", \&edge_threshold);
printf("Background threshold :")
scanf("\&d", \&back_threshold);
printf("Max. distance from seed region :");
scanf("\%d",\&distance) ;
/* read the input image and laplacian edge operated image */
/* first read the range image */
if (pml->pm_form \(==\) PM_C
pm_point \(=\) (unsigned char *) pml->pm_1mage; for ( \(1=0 ; 1<\) row; \(1++\)
for ( \(\mathrm{j}=0\); \(\mathrm{j}<\mathrm{col}\) umn; \(j++\) )
1
range_1mage_buffer[1][f] = *(pm_point): pm_point++;

\section*{\}}
\}
i if (pm1->pm_form \(==\) PM S )
pms_point \(=(\) short int *) pml->pm_image: for (1=0;1<row; \(1++\)
for ( \(\mathrm{j}=0\); j <column; \(\mathrm{j}++\) )
range_image_buffer[1][f] = *(pms_point): pms_point++;

\section*{\}}
/* \({ }^{\prime}\) No
Now read laplacian operated 1mage */
if (pm2->pm_form \(==\) PM_C)
pm_point \(=\) (unsigned char *) pm2->pm_1mage; for(1=0;1<row; \(1++\) ) for ( \(\mathrm{j}=0\); j <column; \(j++\) )
segment.c Tue Apr 19 18:21:08 \(1988 \quad 6\)
1
lap_image_buffer[1][f] = *(pm_point):
pm_point++;
\}
else if (pm2->pm_form \(==\) PM_S)
pms_point \(=\) (short int *) pm2->pm_image; for (1=0;1<row; 1++
for ( \(j=0 ; j<c o l u m n ; j++\) )
1
lap_1mage_buffer[1][f] = *(pms_point): pms_point++;
\(\}\)
/* Initialize output and temporary buffers */
/* for(1=0;1<row; 1++) for ( \(\mathrm{j}=0\); j <column; \(\mathrm{j}++\) )
i
third image buffer \([1][f]=0\)
output_buffer[1][J]=0;
support_image_buffer[0][1][j] =0:
support_1mage_buffer[1][1][J] \(=0\);
        \}
bzero((char *) \&third_image_buffer[0][0],sizeof(1nt)*row*colu
bzero((char *) \&output_buffer[0][0],s1zeof(int)*row*column);
bzero((Char *) \&support_image buffer[0][0][0],sizeof(1nt)*row
bzero((char *) \&support_image_buffer[1][0][0],sizeof(1nt)*row
/* Open files for output and initialize buffers */ pm_addcmt (pm1, cmt)
pm2 = pm_alloc(); /* supporting points image */ pm2->pm_nrow = row;
pm2->pm_ncol \(=\) column
om2->pm_form = 1mage_format
pm2->pm_1mage \(=\left(\right.\) char \(\left.{ }^{\text {* }}\right)\) malloc (pm_psize (pm2)):
pm3= pm_alloc(): /* top most segmented object */
pm3->pm_nrow = row;
pm3->pm_ncol = column
pm3->pm_form = image_format
pm3->pm_1mage \(=(\) char \(*)\) malloc (pm_psize(pm3)):
pm1 = pm_alloc();
pml->pm_nrow = row
/* labeled image */
```

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pm1->pm ncol = column;
pml->pm form = 1mage format;
pml->pm_1mage = (char *) malloc(pm_psize(pml));
If(1mage_format == PM_S)
{
pms point = (short int *) pml->pm image;
pms point2 = (short int *) pm2->pm_1mage;
pms_point2 = (short int *) pm2->pm_1mage;
}
else /* 1mage format is PM_C */
pm point = (unsigned char *) pml->pm_image;
pm point2 = (unsigned char *) pm2->pm_image;
pm_point3 = (unsigned char *) pm3->pm_image;
/*

* initialize the buffers */
bzero(pml->pm image,pm isize(pml));
bzero(pm2->pm-1mage,pm_1size(pm2)):
bzero(pm3->pm_1mage,pm_1size(pm3)):
/*** Segmentation of the picture starts ***/
max = -1000;
if(1mage format == PM S)
f(1mage_format == PM_S) /* starting label+1; for unlabeled
label = MAX_REGION
label = MAX_REGION;
region count = 0;

```

```

/* counts only valid regions */
nextrow = nextcol = 1; /* initialize the seed region star
/********** Loop for segmenting all the objects in the rang
do
region_count++;
label-=;
1f(region count == 1) find seed region();
else
seed_region():
1f(max > 0)
{ /* call grow region for recursive region growing */

```
```

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seedrow = row max;/* seed region coordinates in globa
seedcol = column max
third image buffer[seedrow][seedcol] = 1
pixel\overline{s}=0;
num call = 1;
stack length = 0
surre-1tngth = 0;
max pixel region = -1000
Max_1xel_ren=-1000
grow_region(row_max,column_max):
while(stack_length != current_element)
l
get_pixel(): /* returns the pixel *
num call = 1;
*printf("AAAAA"):*/
if(output buffer[rownum][colnum] != label)
i
/*printf(" BBBBB "); */
grow_region(rownum,colnum);
}
}
smooth region(label)
1f(accept_region(label) == 'n')
l
region_count--;
nvalid region count++
egion[label].number = label;
egion[label] valid = 0.
region[region_count].size = pixels;
}
else
printf("label = %d ",label)
printf("Top : (%d, %d) = %d n, row max, column max,ma
printf("pixels : %d ACCEPT "'pi\overline{x}els)
rintf("pixels : %d ACCEPT ",pixels);
printf("Max : %d Min %d \n",max_pixel_region,min_
egion[label].number = label;
region[label].valid = 1;
region[label].size = pixels;
region[label].max = max pixel region:
gion[label] min = min pixel-region
egion[label].min = min pixel region
determine support(label);
}
}
else printf("\n No more valid regions \n");

```
) while((max >0)):
/* Label all the unwanted regions as UNACCEPT_REGION */ label++; /* label of the last region */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
if (third_image buffer[1][j] \(==0\) )
output buffer [1][j] = UNACCEPT REGION;
region[UNACCEPT_REGION]. valid \(=0 ; \bar{\gamma}^{*}\) unaccept region is inv
\(\max =-1000 ;\)
for \((1=\) (MAX_REGION -1) ; \(1>=\) label; \(1--1\)
if(region \([\overline{1}]\).valid \(==1\) )
if(region[1]. max > max)
i
max \(=\) region[1].max;
top_region_label \(=\) region [1]. number
max_pixel_region \(=\) region[1].max;
min_pixel_region \(=\) region[1].min;
\}
 printf("Total \# of Invalid regions : \(\boldsymbol{q d}^{\prime} \mathrm{n}^{\prime \prime}\), invalid region_cou

```

/* output the segmented edge 1mage */

```
for ( \(1=0 ; 1<\) row; \(1++\) )
1
output_buffer[1][0]=0;
output_buffer[1][column-1] \(=0\);
\}
for ( \(1=0 ; 1<\) column; \(1++\) )
1
output_buffer[0][1] = 0; output_buffer[row-1][1] \(=0\);
\}
1f(1mage_format \(==\) PM_C)
for ( \(1=0\); ; \(1<\) row ; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
1
label = output buffer[1][j];
if ( \(1<(\) column -1\()\) )
if (output_buffer[1][1] != output_buffer[1][j+1]) 1f((region\{output_buffer[i][f]].valid == 1) ||
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(region[output_buffer[1][j+1]].valid == 1)) *pm_point = 255;
if \((1<(\) row -1\())\)
If (output_buffer[1][f] \(!=\) output_buffer[1+1][j])
if((region[output buffer[1][f]].valid == 1) |
(region[output_buffer[1+1][f]].valid \(==1\) ))
*pm_point \(=255\);
if (output buffer[1][J] == top_region_label)
*pm_point \(3=\) range_1mage_buffer[1][1]; /*top */
if (region[support 1mage buffer[1][1][f]\}.valid ==1)*
if (support_image_buffer[1][1][j] \(==\) top_region_label)
*pm_point2 = support_1mage_buffer[0][1][j]; /*sup*/
pm_point++;
pm_point2++;
pmpoint3++;
\}
else /* image_format == PM_S */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(1=0\); \(1<\) column; \(j++\) )
if
label = output_buffer[1][J];
if ( \(\mathrm{j}<(\) column-1) )
If (output buffer[1][j] := output buffer[1][j+1])
if((region[output_buffer[1][j]].valid == 1) || (region[output buffer[1] [j+1]].valid ==1)) *pms_point \(=25 \overline{5}\);
If (1<(row-1))
if (output_buffer[1][f] != output_buffer[1+1][j])
if((region[output_buffer[1][j]].valid \(==1\) ) || (region[output buffer[1+1][j]].valid == 1)) *pms_point \(=25 \overline{5}\);
if (output_buffer[1][f] == top_region_label) *pms_point3 = range_image_buffer[i][j]; /*top */ if (support image buffer[1][1][j] == top region_label)
                    *pms_point2 = support_1mage_buffer[0][1][f]; /*sup*
            pms_point++;
            pms point \(2++\)
            pms point \(3++\);
        \}
pm_write (outfile,pm1)
pm_write (outfile2,pm2);
pm_write (outfile3, pm3) ;
\} /* end of main program */
```

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gapfiller(), if point lies one dist
from the edge */
pmpic *read_pmpic_int();
/************************************MAIN************************************
main(argc,argv)
int argc;
char *argv[];
{
FILE *range,*laplacian,*outfile,*outfile2,*outfile3;

```

```

int row,column; /* of row and columns in in
int offset;
char *cmt;
int done,found: /* used as booleans */
unsigned char *pm_point, *pm_point2, *pm_point 3
short int *pms_point, *pms_point2,*pms_point3;
u_1nt image_format; /* stores 1mage format PM_C o
If(argc != 3)
|
printf("segment : usage : segment <range_file> <lap_edge_
exit(1):
}
cmt = pm_cmt (argc,argv):
printf("Want to display on IKONAS ? ");
scanf("\&s",\&1konas_disp(0]):
/* open the IKONAS display. value of env. variable is taken *
if(strcmp("Y",1konas_disp)"== 0)
if (1kopen (NULL) ==-1)
{
printf("Can't open IKONAS. Exiting\n"):
exit(0):
}
if((range = fopen(argv[1],"r")) == NULL)
printf("segment : Cannot open %s\n",argv[1]):
exit(0);
}

```
sogment.c Tue Apr 19 18:21:08 1988
```

if((laplacian = fopen(argv[2],"r")) == NULL)
{ printf("segment : cannot open %s\n" argy[2])
printf("segment : Cannot open %s\n",argv(2]):
exit(0):
}
If ((pm1 = pm_read(range,0)) == NULL)
{ printf("Error in reading PM file : %s\n",argv[1]);
printf("
}
If((pm2 = pm_read(laplacian,0)) == NULL)
(')
exit(0);
}
/* open output files */
strcpy (output, argv[1]);
f((outfile = fopen(output,"w"))== NULL
{ printf("segment : Cannot open output file : %s\n",out
exit(0):
}
strcpy (output,argv[1]);
strcat (output,".top"):
If((outfile3 = fopen(output,'"w)) == NULL)
{ printf("segment : Cannot open output file : \&s\n",out
exit(0);
}
strcpy(output,argv[1]);
strcat (output,".support")
strcat (output,".support");
f((outfile2 = fopen(output, "w")) == NULL)
printf("segment : Cannot open output file : %s\n",out
exit(0):
}
row = pml->pm nrow;
column = pml->pm ncol
Image_format = pm1->pm_form;

```

\section*{sogmont.c Tue Apr 19 18:21:08 1988 5}

\section*{if((row != pm2->pm_nrow) || (column != pm2->pm_ncol)}
i
printf("Input images are not of same size. Exiting \(\ln\) "); exit(0):
\}
nrow = row; /* initialize global variables nrow and ncol * ncol \(=\) column;
printf("Edge threshold :");
scanf("fd", \&edge_threshold);
printf("Background threshold :")
scanf("\&d", \&back_threshold);
printf("Max. distance from seed region :");
scanf("\%d",\&distance) ;
/* read the input image and laplacian edge operated image */
/* first read the range image */
if (pml->pm_form \(==\) PM_C
pm_point \(=\) (unsigned char *) pml->pm_1mage; for ( \(1=0 ; 1<\) row; \(1++\)
for ( \(\mathrm{j}=0\); \(\mathrm{j}<\mathrm{col}\) umn; \(j++\) )
1
range_1mage_buffer[1][f] = *(pm_point): pm_point++;

\section*{\}}
\}
i if (pm1->pm_form \(==\) PM S )
pms_point \(=(\) short int *) pml->pm_image: for (1=0;1<row; \(1++\)
for ( \(\mathrm{j}=0\); j <column; \(\mathrm{j}++\) )
range_image_buffer[1][f] = *(pms_point): pms_point++;

\section*{\}}
/* \({ }^{\prime}\) No
Now read laplacian operated 1mage */
if (pm2->pm_form \(==\) PM_C)
pm_point \(=\) (unsigned char *) pm2->pm_1mage; for(1=0;1<row; \(1++\) ) for ( \(\mathrm{j}=0\); j <column; \(j++\) )
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1
lap_image_buffer[1][f] = *(pm_point):
pm_point++;
\}
else if (pm2->pm_form \(==\) PM_S)
pms_point \(=\) (short int *) pm2->pm_image; for (1=0;1<row; 1++
for ( \(j=0 ; j<c o l u m n ; j++\) )
1
lap_1mage_buffer[1][f] = *(pms_point): pms_point++;
\(\}\)
/* Initialize output and temporary buffers */
/* for(1=0;1<row; 1++) for ( \(\mathrm{j}=0\); j <column; \(\mathrm{j}++\) )
i
third image buffer \([1][f]=0\)
output_buffer[1][J]=0;
support_image_buffer[0][1][j] =0:
support_1mage_buffer[1][1][J] \(=0\);
        \}
bzero((char *) \&third_image_buffer[0][0],sizeof(1nt)*row*colu
bzero((char *) \&output_buffer[0][0],s1zeof(int)*row*column);
bzero((Char *) \&support_image buffer[0][0][0],sizeof(1nt)*row
bzero((char *) \&support_image_buffer[1][0][0],sizeof(1nt)*row
/* Open files for output and initialize buffers */ pm_addcmt (pm1, cmt)
pm2 = pm_alloc(); /* supporting points image */ pm2->pm_nrow = row;
pm2->pm_ncol \(=\) column
om2->pm_form = 1mage_format
pm2->pm_1mage \(=\left(\right.\) char \(\left.{ }^{\text {* }}\right)\) malloc (pm_psize (pm2)):
pm3= pm_alloc(): /* top most segmented object */
pm3->pm_nrow = row;
pm3->pm_ncol = column
pm3->pm_form = image_format
pm3->pm_1mage \(=(\) char \(*)\) malloc (pm_psize(pm3)):
pm1 = pm_alloc();
pml->pm_nrow = row
/* labeled image */
```

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pm1->pm ncol = column;
pml->pm form = 1mage format;
pml->pm_1mage = (char *) malloc(pm_psize(pml));
If(1mage_format == PM_S)
{
pms point = (short int *) pml->pm image;
pms point2 = (short int *) pm2->pm_1mage;
pms_point2 = (short int *) pm2->pm_1mage;
}
else /* 1mage format is PM_C */
pm point = (unsigned char *) pml->pm_image;
pm point2 = (unsigned char *) pm2->pm_image;
pm_point3 = (unsigned char *) pm3->pm_image;
/*

* initialize the buffers */
bzero(pml->pm image,pm isize(pml));
bzero(pm2->pm-1mage,pm_1size(pm2)):
bzero(pm3->pm_1mage,pm_1size(pm3)):
/*** Segmentation of the picture starts ***/
max = -1000;
if(1mage format == PM S)
f(1mage_format == PM_S) /* starting label+1; for unlabeled
label = MAX_REGION
label = MAX_REGION;
region count = 0;

```

```

/* counts only valid regions */
nextrow = nextcol = 1; /* initialize the seed region star
/********** Loop for segmenting all the objects in the rang
do
region_count++;
label-=;
1f(region count == 1) find seed region();
else
seed_region():
1f(max > 0)
{ /* call grow region for recursive region growing */

```
```

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seedrow = row max;/* seed region coordinates in globa
seedcol = column max
third image buffer[seedrow][seedcol] = 1
pixel\overline{s}=0;
num call = 1;
stack length = 0
surre-1tngth = 0;
max pixel region = -1000
Max_1xel_ren=-1000
grow_region(row_max,column_max):
while(stack_length != current_element)
l
get_pixel(): /* returns the pixel *
num call = 1;
*printf("AAAAA"):*/
if(output buffer[rownum][colnum] != label)
i
/*printf(" BBBBB "); */
grow_region(rownum,colnum);
}
}
smooth region(label)
1f(accept_region(label) == 'n')
l
region_count--;
nvalid region count++
egion[label].number = label;
egion[label] valid = 0.
region[region_count].size = pixels;
}
else
printf("label = %d ",label)
printf("Top : (%d, %d) = %d n, row max, column max,ma
printf("pixels : %d ACCEPT "'pi\overline{x}els)
rintf("pixels : %d ACCEPT ",pixels);
printf("Max : %d Min %d \n",max_pixel_region,min_
egion[label].number = label;
region[label].valid = 1;
region[label].size = pixels;
region[label].max = max pixel region:
gion[label] min = min pixel-region
egion[label].min = min pixel region
determine support(label);
}
}
else printf("\n No more valid regions \n");

```
) while((max >0)):
/* Label all the unwanted regions as UNACCEPT_REGION */ label++; /* label of the last region */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
if (third_image buffer[1][j] \(==0\) )
output buffer [1][j] = UNACCEPT REGION;
region[UNACCEPT_REGION]. valid \(=0 ; \bar{\gamma}^{*}\) unaccept region is inv
\(\max =-1000 ;\)
for \((1=\) (MAX_REGION -1) ; \(1>=\) label; \(1--1\)
if(region \([\overline{1}]\).valid \(==1\) )
if(region[1]. max > max)
i
max \(=\) region[1].max;
top_region_label \(=\) region [1]. number
max_pixel_region \(=\) region[1].max;
min_pixel_region \(=\) region[1].min;
\}
 printf("Total \# of Invalid regions : \(\boldsymbol{q d}^{\prime} \mathrm{n}^{\prime \prime}\), invalid region_cou

```

/* output the segmented edge 1mage */

```
for ( \(1=0 ; 1<\) row; \(1++\) )
1
output_buffer[1][0]=0;
output_buffer[1][column-1] \(=0\);
\}
for ( \(1=0 ; 1<\) column; \(1++\) )
1
output_buffer[0][1] = 0; output_buffer[row-1][1] \(=0\);
\}
1f(1mage_format \(==\) PM_C)
for ( \(1=0\); ; \(1<\) row ; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
1
label = output buffer[1][j];
if ( \(1<(\) column -1\()\) )
if (output_buffer[1][1] != output_buffer[1][j+1]) 1f((region\{output_buffer[i][f]].valid == 1) ||
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(region[output_buffer[1][j+1]].valid == 1)) *pm_point = 255;
if \((1<(\) row -1\())\)
If (output_buffer[1][f] \(!=\) output_buffer[1+1][j])
if((region[output buffer[1][f]].valid == 1) |
(region[output_buffer[1+1][f]].valid \(==1\) ))
*pm_point \(=255\);
if (output buffer[1][J] == top_region_label)
*pm_point \(3=\) range_1mage_buffer[1][1]; /*top */
if (region[support 1mage buffer[1][1][f]\}.valid ==1)*
if (support_image_buffer[1][1][j] \(==\) top_region_label)
*pm_point2 = support_1mage_buffer[0][1][j]; /*sup*/
pm_point++;
pm_point2++;
pmpoint3++;
\}
else /* image_format == PM_S */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(1=0\); \(1<\) column; \(j++\) )
if
label = output_buffer[1][J];
if ( \(\mathrm{j}<(\) column-1) )
If (output buffer[1][j] := output buffer[1][j+1])
if((region[output_buffer[1][j]].valid == 1) || (region[output buffer[1] [j+1]].valid ==1)) *pms_point \(=25 \overline{5}\);
If (1<(row-1))
if (output_buffer[1][f] != output_buffer[1+1][j])
if((region[output_buffer[1][j]].valid \(==1\) ) || (region[output buffer[1+1][j]].valid == 1)) *pms_point \(=25 \overline{5}\);
if (output_buffer[1][f] == top_region_label) *pms_point3 = range_image_buffer[i][j]; /*top */ if (support image buffer[1][1][j] == top region_label)
                    *pms_point2 = support_1mage_buffer[0][1][f]; /*sup*
            pms_point++;
            pms point \(2++\)
            pms point \(3++\);
        \}
pm_write (outfile,pm1)
pm_write (outfile2,pm2);
pm_write (outfile3, pm3) ;
\} /* end of main program */
```

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pm1->pm ncol = column;
pml->pm form = 1mage format;
pml->pm_1mage = (char *) malloc(pm_psize(pml));
If(1mage_format == PM_S)
{
pms point = (short int *) pml->pm image;
pms point2 = (short int *) pm2->pm_1mage;
pms_point2 = (short int *) pm2->pm_1mage;
}
else /* 1mage format is PM_C */
pm point = (unsigned char *) pml->pm_image;
pm point2 = (unsigned char *) pm2->pm_image;
pm_point3 = (unsigned char *) pm3->pm_image;
/*

* initialize the buffers */
bzero(pml->pm image,pm isize(pml));
bzero(pm2->pm-1mage,pm_1size(pm2)):
bzero(pm3->pm_1mage,pm_1size(pm3)):
/*** Segmentation of the picture starts ***/
max = -1000;
if(1mage format == PM S)
f(1mage_format == PM_S) /* starting label+1; for unlabeled
label = MAX_REGION
label = MAX_REGION;
region count = 0;

```

```

/* counts only valid regions */
nextrow = nextcol = 1; /* initialize the seed region star
/********** Loop for segmenting all the objects in the rang
do
region_count++;
label-=;
1f(region count == 1) find seed region();
else
seed_region():
1f(max > 0)
{ /* call grow region for recursive region growing */

```
```

Tum Apr 19 18:21:08 1988
seedrow = row max;/* seed region coordinates in globa
seedcol = column max
third image buffer[seedrow][seedcol] = 1
pixel\overline{s}=0;
num call = 1;
stack length = 0
surre-1tngth = 0;
max pixel region = -1000
Max_1xel_ren=-1000
grow_region(row_max,column_max):
while(stack_length != current_element)
l
get_pixel(): /* returns the pixel *
num call = 1;
*printf("AAAAA"):*/
if(output buffer[rownum][colnum] != label)
i
/*printf(" BBBBB "); */
grow_region(rownum,colnum);
}
}
smooth region(label)
1f(accept_region(label) == 'n')
l
region_count--;
nvalid region count++
egion[label].number = label;
egion[label] valid = 0.
region[region_count].size = pixels;
}
else
printf("label = %d ",label)
printf("Top : (%d, %d) = %d n, row max, column max,ma
printf("pixels : %d ACCEPT "'pi\overline{x}els)
rintf("pixels : %d ACCEPT ",pixels);
printf("Max : %d Min %d \n",max_pixel_region,min_
egion[label].number = label;
region[label].valid = 1;
region[label].size = pixels;
region[label].max = max pixel region:
gion[label] min = min pixel-region
egion[label].min = min pixel region
determine support(label);
}
}
else printf("\n No more valid regions \n");

```
) while((max >0)):
/* Label all the unwanted regions as UNACCEPT_REGION */ label++; /* label of the last region */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
if (third_image buffer[1][j] \(==0\) )
output buffer [1][j] = UNACCEPT REGION;
region[UNACCEPT_REGION]. valid \(=0 ; \bar{\gamma}^{*}\) unaccept region is inv
\(\max =-1000 ;\)
for \((1=\) (MAX_REGION -1) ; \(1>=\) label; \(1--1\)
if(region \([\overline{1}]\).valid \(==1\) )
if(region[1]. max > max)
i
max \(=\) region[1].max;
top_region_label \(=\) region [1]. number
max_pixel_region \(=\) region[1].max;
min_pixel_region \(=\) region[1].min;
\}
 printf("Total \# of Invalid regions : \(\boldsymbol{q d}^{\prime} \mathrm{n}^{\prime \prime}\), invalid region_cou

```

/* output the segmented edge 1mage */

```
for ( \(1=0 ; 1<\) row; \(1++\) )
1
output_buffer[1][0]=0;
output_buffer[1][column-1] \(=0\);
\}
for ( \(1=0 ; 1<\) column; \(1++\) )
1
output_buffer[0][1] = 0; output_buffer[row-1][1] \(=0\);
\}
1f(1mage_format \(==\) PM_C)
for ( \(1=0\); ; \(1<\) row ; \(1++\) )
for ( \(j=0 ; j<c o l u m n ; j++\) )
1
label = output buffer[1][j];
if ( \(1<(\) column -1\()\) )
if (output_buffer[1][1] != output_buffer[1][j+1]) 1f((region\{output_buffer[i][f]].valid == 1) ||
segmant.C Tue Apr 19 18:21:08 1988 10
(region[output_buffer[1][j+1]].valid == 1)) *pm_point = 255;
if \((1<(\) row -1\())\)
If (output_buffer[1][f] \(!=\) output_buffer[1+1][j])
if((region[output buffer[1][f]].valid == 1) |
(region[output_buffer[1+1][f]].valid \(==1\) ))
*pm_point \(=255\);
if (output buffer[1][J] == top_region_label)
*pm_point \(3=\) range_1mage_buffer[1][1]; /*top */
if (region[support 1mage buffer[1][1][f]\}.valid ==1)*
if (support_image_buffer[1][1][j] \(==\) top_region_label)
*pm_point2 = support_1mage_buffer[0][1][j]; /*sup*/
pm_point++;
pm_point2++;
pmpoint3++;
\}
else /* image_format == PM_S */
for ( \(1=0 ; 1<\) row; \(1++\) )
for ( \(1=0\); \(1<\) column; \(j++\) )
if
label = output_buffer[1][J];
if ( \(\mathrm{j}<(\) column-1) )
If (output buffer[1][j] := output buffer[1][j+1])
if((region[output_buffer[1][j]].valid == 1) || (region[output buffer[1] [j+1]].valid ==1)) *pms_point \(=25 \overline{5}\);
If (1<(row-1))
if (output_buffer[1][f] != output_buffer[1+1][j])
if((region[output_buffer[1][j]].valid \(==1\) ) || (region[output buffer[1+1][j]].valid == 1)) *pms_point \(=25 \overline{5}\);
if (output_buffer[1][f] == top_region_label) *pms_point3 = range_image_buffer[i][j]; /*top */ if (support image buffer[1][1][j] == top region_label)
                    *pms_point2 = support_1mage_buffer[0][1][f]; /*sup*
            pms_point++;
            pms point \(2++\)
            pms point \(3++\);
        \}
pm_write (outfile,pm1)
pm_write (outfile2,pm2);
pm_write (outfile3, pm3) ;
\} /* end of main program */

\section*{egment c \\ 11}
/******************************* GROW REGION ************************* /* following is the modified recursive call for the segmentation. now a pixel is examined before routine is called rather than callin the routine and then checking the pixel type. this was done to opt the stack-space which overflows in the case of large objects. even after this modification stack overflows if the object is therefore recursive region growing is not possible if the object is */

\section*{grow_region(row, col)}
int row;
int col
```

int 1.j; /* loop control variables */
int dist;
int cross_grad.

```
if (strcmp ("Y", ikonas_disp) \(==0\) )
    lwr(col, row, \&value):
output_buffer[row][col] = label;
if(max_pixel_region < range_image_buffer[row][col]) max pixel region = range image buffer[row][col]; if(min_pixel_region > range_image_buffer[row][col])
min pixel_region = range_image_buffer[row][col];

\section*{pixels++;}
/* \(\quad\) f( (row \(<(\) nrow -1\()) \& \&(\operatorname{col}<(\) ncol -1\())\) )
cross grad =sqrt ( (double)
(squarezz(range image buffer[row] [col] range-image-buffer[row+1][col+1] range - 1mage buffer[row][col+1] squarezz (range- 1mage-buffer[row+1][col])
else *
cross grad \(=0\);
if (cross grad \(!=0\) ) cross grad \(+=0\)
if (cross_grad < edge_threshold)
i
for ( \(1=\) row -1 ; \(1<\) row \(+2 ; 1++\) )
for ( \(\mathrm{f}=\mathrm{col}-1\); \(\mathrm{j}<\mathrm{col}+2 ; \mathrm{j}++\)
if((1 != row) || (1 != col))
If \(((1>0) \& \&(1<(\) nrow-1) \() \& \&(1>0) \& \&(j<(n C\)
dist \(=\) (int) (sqrt ((double) ((seedrow - i)*(s (seedcol - J)*(seedcol if (dist \(<=\) distance)
\{
if (llap image buffer[1][j] < edge thres (output buffer[1][f] != label) \&\&
(range_1mage_buffer[1][j] > back_thr 1 if 1
\[
(1, j)=0)
\]

1 third image buffer[1][1] = ;
t* check for number of pending \(c\) num call++;
if (num call \(>=\) MAX NUM CALL)
store pixel \((1, j)\);
else
grow_region(1, \()\);
\}
se /* point is actually one dista /* from an edge pixel* /* first mark the pixels vis output_bu
pixels++;
if (strcmp (" \(\overline{\mathrm{Y}}\) ", ikonas disp) \(==0\) ) (strcmp "Yvilkonas
lwr(j,,\(\&\) value); * now mark the edge pixel visi output_buffer[gaprow][gapcol] pixels++;
third image_buffer [gaprow] [gapco 1f(strcmp (" \(\overline{\mathrm{y}}\) ", 1konas_disp) \(==0\) ) (strcmp ("y", 1konas_disp) \}
else if(lap_image_buffer[1][j] >= edge_ i output buffer[1][f] = label pixels戸+;
third image buffer[1][1] \(=1\); if (strcmp ("Y",1konas_disp) \(=0\) )
lwr(J, 1, \& value): \}

\footnotetext{
\}
else /* distance exceeded */
}

\section*{sogmont. \(C\) \\ Tu* Apr 18 18:21:08 1988 \\ 13}
output_buffer[1][f] = label;
pixels++;
third_image_buffer[i][f] = 1;
if(strcmp("y",1konas_disp) \(==0\)
lwr(j,1,\&value);
* acceptable pixel */
//* cross_grad acceptable */
num_call--;
\} \(\mathrm{T}^{*}\) end of grow region */
/******************** STORE_PIXEL \& GET_PIXEL ******************************)
store_pixel (rrow, ccol)
int rrow;
int ccol:
num call--:
/* printf("FFFF "): */
pixel_stack[stack_length].row = rrow;
pixel_stack[stack_length].col = ccol;
stack_length++;
if(stack_length \(==\) PIXEL_STACK_SIZE) stack_length \(=0\);
if (stack_length \(==\) current_element)
printf("\n segment : Stack Collision While region growing \(\ln\)
)
get_pixel()
rownum = pixel stack[current_element].row;
colnum = pixel_stack[current_element].col;
if (current element \(==\) PIXEL_STACK_SIZE) current_element \(=0\);
\}
loat squarezz( \(n\) )
int \(n\);
float f;
\(f=(f l o a t)(n)\);
return (f*f) ;
/***** gap filler checks if the pixel can be considered as the part the edge. This is done by checking if the 8-connected neighbou has any edge pixel. If yes, it is considered a neighbour of edg and coordinates of edge pixel is returned in gaprow, gapcol. The pixel is not further grown. This procedure fills in 2-pixel gap without undergoing the pain of gap-filling using contour tracin which is computationally expensive. See peter allen's thesis fo performance of gap-filler (due to Nevatia \& Babu). He observes that filling of at most 2 -pixel gaps is acceptable in most case
****/
gap_filler(rrow, ccol)
int rrow, ccol:
int 1,j;
for ( \(1=\) rrow \(-1 ; 1<=(\) rrow +1\() ; 1++\) )
for \((\mathrm{j}=\mathrm{ccol}-1\); \(\mathrm{j}<=(\mathrm{ccol}+1) ; \mathrm{j}++)\)
if (lap_image_buffer[i][f] \(>=\) edge_threshold)
( /* edge \(\overline{\mathrm{p}}\) ixel in 8 -connected \(\overline{\mathrm{n}}\) bd is found */
gaprow =1;
gapcol \(=j\)
return(1):
!
/* no edge pixel in 8-connected neighbourhood is found */
return (0);
\}
smooth_region(label_value)
int label_value;
int 1,j;
/* for (i=0; i<nrow; 1++)
for ( \(\mathrm{f}=0 ; \mathrm{j}\) <ncol; \(\mathrm{j}++\) )
1
if(output_buffer[1][1] == label_value)
\}

\section*{*/ \}}
\}
\(\min (x, y)\)
int \(x, y\);
1

\section*{egment c \\ 11}
/******************************* GROW REGION ************************* /* following is the modified recursive call for the segmentation. now a pixel is examined before routine is called rather than callin the routine and then checking the pixel type. this was done to opt the stack-space which overflows in the case of large objects. even after this modification stack overflows if the object is therefore recursive region growing is not possible if the object is */

\section*{grow_region(row, col)}
int row;
int col
```

int 1.j; /* loop control variables */
int dist;
int cross_grad.

```
if (strcmp ("Y", ikonas_disp) \(==0\) )
    lwr(col, row, \&value):
output_buffer[row][col] = label;
if(max_pixel_region < range_image_buffer[row][col]) max pixel region = range image buffer[row][col]; if(min_pixel_region > range_image_buffer[row][col])
min pixel_region = range_image_buffer[row][col];

\section*{pixels++;}
/* \(\quad\) f( (row \(<(\) nrow -1\()) \& \&(\operatorname{col}<(\) ncol -1\())\) )
cross grad =sqrt ( (double)
(squarezz(range image buffer[row] [col] range-image-buffer[row+1][col+1] range - 1mage buffer[row][col+1] squarezz (range- 1mage-buffer[row+1][col])
else *
cross grad \(=0\);
if (cross grad \(!=0\) ) cross grad \(+=0\)
if (cross_grad < edge_threshold)
i
for ( \(1=\) row -1 ; \(1<\) row \(+2 ; 1++\) )
for ( \(\mathrm{f}=\mathrm{col}-1\); \(\mathrm{j}<\mathrm{col}+2 ; \mathrm{j}++\)
if((1 != row) || (1 != col))
If \(((1>0) \& \&(1<(\) nrow-1) \() \& \&(1>0) \& \&(j<(n C\)
dist \(=\) (int) (sqrt ((double) ((seedrow - i)*(s (seedcol - J)*(seedcol if (dist \(<=\) distance)
\{
if (llap image buffer[1][j] < edge thres (output buffer[1][f] != label) \&\&
(range_1mage_buffer[1][j] > back_thr 1 if 1
\[
(1, j)=0)
\]

1 third image buffer[1][1] = ;
t* check for number of pending \(c\) num call++;
if (num call \(>=\) MAX NUM CALL)
store pixel \((1, j)\);
else
grow_region(1, \()\);
\}
se /* point is actually one dista /* from an edge pixel* /* first mark the pixels vis output_bu
pixels++;
if (strcmp (" \(\overline{\mathrm{Y}}\) ", ikonas disp) \(==0\) ) (strcmp "Yvilkonas
lwr(j,,\(\&\) value); * now mark the edge pixel visi output_buffer[gaprow][gapcol] pixels++;
third image_buffer [gaprow] [gapco 1f(strcmp (" \(\overline{\mathrm{y}}\) ", 1konas_disp) \(==0\) ) (strcmp ("y", 1konas_disp) \}
else if(lap_image_buffer[1][j] >= edge_ i output buffer[1][f] = label pixels戸+;
third image buffer[1][1] \(=1\); if (strcmp ("Y",1konas_disp) \(=0\) )
lwr(J, 1, \& value): \}

\footnotetext{
\}
else /* distance exceeded */
}

\section*{sogmont. \(C\) \\ Tu* Apr 18 18:21:08 1988 \\ 13}
output_buffer[1][f] = label;
pixels++;
third_image_buffer[i][f] = 1;
if(strcmp("y",1konas_disp) \(==0\)
lwr(j,1,\&value);
* acceptable pixel */
//* cross_grad acceptable */
num_call--;
\} \(\mathrm{T}^{*}\) end of grow region */
/******************** STORE_PIXEL \& GET_PIXEL ******************************)
store_pixel (rrow, ccol)
int rrow;
int ccol:
num call--:
/* printf("FFFF "): */
pixel_stack[stack_length].row = rrow;
pixel_stack[stack_length].col = ccol;
stack_length++;
if(stack_length \(==\) PIXEL_STACK_SIZE) stack_length \(=0\);
if (stack_length \(==\) current_element)
printf("\n segment : Stack Collision While region growing \(\ln\)
)
get_pixel()
rownum = pixel stack[current_element].row;
colnum = pixel_stack[current_element].col;
if (current element \(==\) PIXEL_STACK_SIZE) current_element \(=0\);
\}
loat squarezz( \(n\) )
int \(n\);
float f;
\(f=(f l o a t)(n)\);
return (f*f) ;
/***** gap filler checks if the pixel can be considered as the part the edge. This is done by checking if the 8-connected neighbou has any edge pixel. If yes, it is considered a neighbour of edg and coordinates of edge pixel is returned in gaprow, gapcol. The pixel is not further grown. This procedure fills in 2-pixel gap without undergoing the pain of gap-filling using contour tracin which is computationally expensive. See peter allen's thesis fo performance of gap-filler (due to Nevatia \& Babu). He observes that filling of at most 2 -pixel gaps is acceptable in most case
****/
gap_filler(rrow, ccol)
int rrow, ccol:
int 1,j;
for ( \(1=\) rrow \(-1 ; 1<=(\) rrow +1\() ; 1++\) )
for \((\mathrm{j}=\mathrm{ccol}-1\); \(\mathrm{j}<=(\mathrm{ccol}+1) ; \mathrm{j}++)\)
if (lap_image_buffer[i][f] \(>=\) edge_threshold)
( /* edge \(\overline{\mathrm{p}}\) ixel in 8 -connected \(\overline{\mathrm{n}}\) bd is found */
gaprow =1;
gapcol \(=j\)
return(1):
!
/* no edge pixel in 8-connected neighbourhood is found */
return (0);
\}
smooth_region(label_value)
int label_value;
int 1,j;
/* for (i=0; i<nrow; 1++)
for ( \(\mathrm{f}=0 ; \mathrm{j}\) <ncol; \(\mathrm{j}++\) )
1
if(output_buffer[1][1] == label_value)
\}

\section*{*/ \}}
\}
\(\min (x, y)\)
int \(x, y\);
1
```

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if (x<y) return(x);
}
/****************************** ACCEPT REGION ****************************
char accept_region(lab)
int lab;
l
int 1,j,k;
if(pixels < MIN ACCEPTABLE) return('n');
else return('y')
}
/******************************** DETERMINE_SUPPORT ***********************
determine support(lab)
int lab;
int
int 1,j,k;
/* determine boundary points for all rows first */
for(1=0;1<nrow;1++)
for(j=0;j<(ncol-1);j++)
{
If((output buffer[1][j] != lab) \&\& (output buffer[1][j+1] ==
det med_support (lab, 1, j+1);
else if((output buffer[i][j]:== lab) \&\& (output_buffer[i][j+1
det_med support (lab,1,f);
}
/* determine boundary points for all columns */
for( ( = 0; f<ncol; f++)
for(1=0;1<(nrow-1);1++)
{
If((output buffer[1][f] != lab) \&\& (output buffer[i+1][f] ==
det med_support (lab,i+1,j);
else i\overline{f}((output_buffer[i][f]== lab) \&\& (output_buffer[i+1][j
det_med_support (lab,1,j);
}
}
DET MED SUPPORT ***************************
det med support(lab, row, col)
int lab;

```
int row;
int row
int
int 1,j,k;
int count \(=0\);
int num;
int wsize \(=3\);
 for \((1=0 ; 1<256 ; 1++)\) vector \([1]=0\) :
for ( \(1=(\) row-wsize \() ; 1<=(\) row+wsize \() ; 1++\) )
for \((j=(\) col-wsize \() ; j<=(\) col + wsize \() ; j++)\)
if \((1\rangle=0) \& \&\) (i<nrow) \(\& \&(j>=0) \& \&(j<\) ncol))
if (output_buffer[1][f] != lab)
count++;
vector[range_image_buffer[1][f]]++
\}
/* num \(=\) count/2; THis is for median */ num \(=1 ; \quad / \star\) This picks up the smallest depth */
/* printf("num \(=\) of count \(=\) qd", num, count) \(; * /\)
1f(num == 0)
1 support image buffer [0][row][col] = range image buffer[row][col support_image_buffer[1][row][col] = lab:

\section*{)}
else
for ( \(1=0\); num \(>0 ; 1++\) )
i num=num-vector[1]: if (vector [1] > 0) \(\mathrm{j}=1\);
support image buffer[0][row][col] \(=1\); support_image_buffer[1][row][col] = lab;
\}
)

finished()
\({ }^{1}\) int \(1,1, k\);
int finish \(=1\);
```

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for(1=1;1<nrow;1++)
for(j=1; 1<ncol; j++
if((third image buffer[1][f] == 0) \&\&
(range image buffer[1][f] > back threshold)) finish = 0;
return(f1nish);
}
/******************** FIND SEED REGION
*****************************/
find_seed_region()
{
Int 1; j,k,m,n;
/* Boolean */
int average_depth; /* average depth of the 3x3 window */
int average_
max = -1000;
max = =-1000;
for(j=2;1<ncol-2;䜣)
if((range image buffer[1][f] > back threshold)
(th1rd 1mage buffer[1][f] == 0))
{
acount = 0;
acount = 0;
average depth
ok_point = 1;
/* do it only the first time */
1f(region_count == 1)
for (m=1-1;m<1+2;m++)
for (n=1-1;n<j+2;n++)
or(n=1-1;n<j+2;n++)
(range image buffer [m][n] > back_threshold) \&\&
(range 1mage_buffer [m][n] > = 0)
{
{ acount++;
average_depth += range_image_buffer[m][n];
}
else ok_point = 0;
If(acount > 8) average image[1][f] = average_depth/acou
else average_image[1][\overline{j}]=0;
}
if (ok_point == 1)
I
If((lap_image buffer[i][f] < edge_threshold) \&\&
(range image buffer[1][f] > back threshold) \&\&
(th1rd_1mage_buffer[1][j]== 0) \&\&

```
sogment. C

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(average_1mage[1][f] > maxi)
1
max \(=\) average_image[1][1];
row_max = 1;
column_max \(=\mathrm{j}\);
\}
\}
)
\} /* end of find_seed_region */
/**************************** SEED_REGION ***********************************)
seed_region()
\((\) int \(1, j\), done, end:
int m, n;
int acount;
int cross_grad;
done \(=0\);
end \(=0\);
while \(((\) done \(==0) \& \&(\) end \(==0))\)
if((third image buffer[nextrow][nextcol] \(==0) \& \&\)
(lap_1mage_buffer[nextrow][nextcol] < edge_threshold) \&\& (range_image_buffer[nextrow](nextcol] > back_threshold))
\(i^{1}\)
acount \(=0\);
for ( \(m=\) nextrow-1; \(m\) <next row +2 ; \(m++\) )
for ( \(n=\) nextcol \(-1 ; n<n e x t c o l+2 ; n++\) )
/*
cross_grad =sqrt ((double)
(squarezz(range image buffer[m][n] -
range_1mage_buffer \([m+1](n+1])+\)
squarezz (range image buffer \([\mathrm{m}][\mathrm{n}+1]\) -
range_1mage_buffer \([m+1][n]))\) : */
cross grad \(=0\);
if((lap_image buffer[m][n] < edge_threshold) \&\& (range_1mage_buffer \([\mathrm{m}][\mathrm{n}]>\) back_threshold) \&\& (cross_grad < edge_threshold) \&\& (third_1mage_buffer \([\mathrm{m}][\mathrm{n}]==0\) )) acount+耳;
\}
if (acount > 8)
```

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{ done = 1;
max = range_image_buffer[nextrow][nextcol];
}}\mathrm{ max = rang
}
row_max = nextrow
column_max = nextcol;
nextcol++;
if(nextcol > (ncol-2))
{
nextrow++;
nextcol = 0;
if(nextrow > (nrow - 2))
end = 1;
max =-1000
}
}
}
} /* end of seed_region */
/******************************* FILL THE GAPS ***************************
fill_gaps()
int i,j,k,l;
}/*fill_gaps */
/******************************** CONTOUR TRACING *************************

```
../trial/rca_calib.c Thu May 5 17:07:25 \(1988 \quad 1\)

program for reading 2 and a \(1 / 2 \mathrm{D}\) range images
removing supporting surface,
calibrating the points,
adding vertical points
adding horizontal points,
selecting the density of grid,
outputing the points.
Jan 1988 : Modified to read PM_C format files.
Jan 1988: File reading procedure replaced to read as unsigned
Jan 1988 : Buffer declarations moved out of the program to avo run time segmentation fault

To read in supporting surface image and to add poin vertically instead of horizontally.
Changed to read rca range image files.
utput points in seperate files.
points.orig
Mar 24,1988 : Modified to read both PM C and PM S formats.
I/O : run the program as:
\%rca_calib _grid_density_ _range_image_file_ \{_support_points_f
if support image file is not specified then vertical point addi is disabled. Otherwise user is given choice to add vertical points o Input images can be either in PM_S or PM_C format. Two files are out one having just the original points : points.orig
one having all the desired added points ( background, vertical. and horizontal ) : points.add

Notes :
1. Support point image should not have depth values \(=255\). It is \(r\) for the program.
2. Backgound points are thresholded at BACK_DEPTH + TRESH.

\section*{Hnclude <math.h>}
\#include <stdio.h>
\#include <local/pm.h>
**** \#nclude "/usr/users/franc/local/pic.h" */
\#include "/usr/users/alok/trial/pic.h"
define POINTS PICDIM X * PICDIM Y
\#nclude "/usr/users/franc/local/frio.h"
../trial/rca_callb.c Thu May 5 17:07:25 1988 2
/*** Following values are Range-image scanner dependent. These corres to the SRI- test database ***/
/*\#define VERTICAL 1.531 */
**define HORIZONTAL 1.531 */
**define HEIGHT 0.245 */
/*\#define BACK_DEPTH 5.0 */
/* Following are for RCA range image scanner. measured in mm/pixel
*/
define VERTICAL 1.8796
define HORIZONTAL 1.8796
\#define HEIGHT 0.0254
*\#define VERTICAL 1.5240*/
***define HORIZONTAL 1.8796*/
*\#define HEIGHT 0.0254*/
\[
\begin{aligned}
& \text { /* X , varying rows } 0.060 \\
& \text { /* Y , varying cols } 0.074 \\
& \text { /* } \mathrm{Z} \text {, depth } 0.001
\end{aligned}
\]
/* Range 1mage scanner dependent */

/* background depth */
\#define TRESTH 0 /* original value = 5 */
define BORDER 10
/*** Big array declarations are moved out of the main program body to avoid run time memory fault. ***/
*** Important to have the image read in as unsigned char so that pix values are from 0 to 255 and not from -127 to +127 ***/
int picture[PICDIM_X][PICDIM_Y]:
int sup_points \(\left[P I C \bar{D} I M \_X\right]\left[P I C \bar{D} I M \_Y\right]\);
double point[POINTS][3], support[POINTS][3], T[4][4], newpoint[3], ve /*** additional declarations for adding hidden points ***/
double vect[3], addpoint[POINTS][3], vectpoint[3]; /* horizontal double sup_vect[3], side_add_points[POINTS][3], sup_point[3];/*vertic
pmpic *pml;
/*** Input file is in PM-format */
/*** \(\#\) of rows in input picture */
int sizey; /*** \# of columns in input picture */

\section*{../trial/rca_calib.c Thu May 5 17:07:25 1988 3}
nt argc;
char *argv[];
i
int 1, j, k, k2, k3, m, n, grid;
double x1, y1, z1, x2, y2, z2, x3, y3, z3, a, b, c;
double sqrt(), sq(), distance;
char g[10],h[10], sup [10],orig[10];
int sup_count \(=0\);
int cause;
FILE *orig_file,*add_file;
if ( \((\) argc < 3) || (argc > 4))
printf("usage: rca_calib grid_density inpic (sup_pic\} \(\backslash \mathrm{n}^{\prime \prime}\) ): else
grid \(=\) atol \((\operatorname{argv}[1])\);
read_picture (argv[2], picture);
if \((\operatorname{argc}==4)\) read picture (argv[3], sup_points);
make_matrix(T);
/* compute the supporting plane from three points
\(\mathrm{xl}=15.0\);
y1 = 5.0;
\(21=\) BACK DEPTH;
/* \(z 1=\) (double)picture[(int)x1][(1nt)y1];*/

\(\mathrm{x} 2=230.0\);
\(\mathrm{y}^{2}=5.0\);
z2 = BACK_DEPTH:
/* z2 = (double)p1cture[(int)x2][(1nt)y2];*/ fprintf(stderr,"z2 = \& \(\quad\) ", z2) ;
\(\times 3=30.0\);
\(\mathrm{y}^{3}=165.0\);
\(23=\) BACK_DEPTH:
/* z3 = (double)picture[(int)x3][(1nt)y3];*/ fprintf(stderr,"z3 \(=\) \& \(\backslash \mathrm{n}^{n}, \mathrm{z} 3\) );
\(c=\left((x 3-x 1) *\left(x 3^{*} y 2-x 2^{*} y 3\right)-(x 3-x 2)^{*}\left(x 3^{*} y 1-x 1^{*} y 3\right)\right) /(\) \(\left(x 3^{*} z 2-x 2^{\star} z 3\right)^{\star}\left(x 3^{\star} y 1-x 1^{\star} y 3\right)-\left(x 3^{*} z 1-x 1^{\star} z 3\right)^{\star}\left(x 3^{\star} y 2-x 2\right.\) );
\(b=\left(-c *\left(x 3^{*} z 1-x 1^{*} z 3\right)-(x 3-x 1)\right) /(x 3 * y 1-x 1 * y 3) ;\)

\section*{../trial/rca_calib.c Thu May 5 17:07:25 1988}
\(a=(-1-c \star z 1-b * y 1) / x 1 ;\)

fprintf(stderr, "Add hidden points horizontally ( \(\mathrm{y} / \mathrm{n}\) ) ?"); scanf("\%s", h):
if (argc \(==4)\{1 *\) supporting points file present * fprintf(stderr, "Add hidden points vertically (y/n)?"): scanf("\%s", sup):
\} else
sprintf(sup,"ヶs","no"):
\(\mathrm{k}=0\);
k2 \(=0\);
k3 \(=0\);
for ( \(j=\) BORDER; \(\}\) < sizey - BORDER; \(j=j+\operatorname{grid})\)
\{ for ( 1 = BORDER; 1 < sizex - BORDER; \(1=1+\) grid)
1
vector \([0]=j *\) HORIZONTAL;
vector \([1]=(239-1) \star\) VERTICAL;
vector \([2]=\) picture \([1][j] *\) HEIGHT
matrix_mult(vector, \(T\), newpoint):
If \(\left(h[0]==\right.\) ' \(y^{\prime}\) )
-
vect [0] \(=\) vector [0];
vect[1] = vector[1];
vect \([2]=((-1 \cdot 0-a \star 1-b \star j) / c) *\) HEIGHT
matrix_mult(vect, \(T\), vectpoint):
\}
distance \(=(a * 1+b * j+c *\) (double) \((\) picture \([1][f])+1\) sqrt(sq(a) \(+\mathbf{s q}(b)+s q(c)):\)

if (distance < -TRESH)
( point[k][0] = newpoint[0];
point \([k][1]=\) newpoint [1]
point[k][2] = newpoint[2]
if \((\mathrm{h}[0]==\) ' y '
i
( addpoint \([k][0]=\operatorname{vectpoint}[0]\) :
addpoint \([\mathrm{k}][1]=\) vectpoint[1]; addpoint \([k][2]=\) vectpoint [2];
\(\mathrm{k} \stackrel{\text { l }}{=}\)
\(\mathrm{k}=\mathrm{k}+1\);

\section*{../trial/rca_callb.c Thu Mny 5 17:07:25 1988 5} \}
        else if (distance < TRESH)
            ( support[k2][0] = newpoint [0];
            support [k2][1] \(=\) newpoint [1];
            support \([k 2][1]=\) newpoint \([1] ;\)
support \([k 2][2]=\) newpoint [2];
            \(\mathrm{k} 2=\mathrm{k} 2+1\) :
        els
        else \{\} /* points is in shadow */
    \(1\}\)
if (sup \([0]==\) ' \(y^{\prime}\) )
    for ( \(f=\) BORDER; \(f\) < sizey - BORDER; f++)
    \{ for ( 1 = BORDER; 1 < Sizex - BORDER; \(1++\) )
        ( if(sup points[1][f]>0)
            i sup count++;
            sup_count++
            for ( \(\mathrm{m}=1-\mathrm{grid} / 2 ; \mathrm{m}<=1+\mathrm{gr} 1 \mathrm{~d} / 2 ; \mathrm{m}++\) )
                            for ( \(n=j-g r i d / 2 ; n<=j+g r i d / 2 ; n++\) )
                            if((sup points \([\mathrm{m}][\mathrm{n}]==255)\) ) cause++
                    if (cause \(==0\) )
                            for (m=sup points[1][f]; \(m\) <= picture[1][f]; m=
                        for
                            sup_vect \([0]=\jmath \star\) HORIZONTAL;
                                    sup_vect \([1]=(239-1) \star\) VERTICAL;
                                    sup vect \([2]=\) m*HEIGHT
                                    matrix mult (sup vect, \(T\), sup point):
                    side add points \([k 3][0]=\) sup point [0]
                    side_add points[k3][1] = sup point [0]
                            side_add_points \([k 3][1]=\) sup_point \([1]\)
                            side_add_points[k3][2] = sup_point[2]:
                            k3++;
                    \}
                    sup points[1][f] \(=255\) :
            \}
            \}
        \}
    )
fprintf (stderr,"Remove supporting surface ( \(\mathrm{y} / \mathrm{n}\) )? "); scanf("\%s", g):
fprintf (stderr,"Add original points ? \((y / n)\) "): scanf("\%s", orlg).
```

```
../trial/rca_calib.c Thu May 5 17:07:25 1988 6
```

```
../trial/rca_calib.c Thu May 5 17:07:25 1988 6
    if(orig[0] == 'Y')
    if(orig[0] == 'Y')
    i
    i
            If((orig_file = fopen("points.orig","w")) == NULL)
            If((orig_file = fopen("points.orig","w")) == NULL)
                (fprintf(stderr,"Can't open output file : %s\n","points.or
                (fprintf(stderr,"Can't open output file : %s\n","points.or
                exit(0);}
                exit(0);}
            for(1 = 0; 1 < k; 1++)
            for(1 = 0; 1 < k; 1++)
            fprintf(orig_file,mf %f %f \n", point[1][0], point[1][1],
            fprintf(orig_file,mf %f %f \n", point[1][0], point[1][1],
        }
        }
    If((h[0]== 'Y') || (g[0]== 'Y') || (sup[0]=' ( y'))
    If((h[0]== 'Y') || (g[0]== 'Y') || (sup[0]=' ( y'))
            f((add file = fopen("points.add","w")) == NULL)
            f((add file = fopen("points.add","w")) == NULL)
            (printf("Can't open output file : %s\n","points.add");
            (printf("Can't open output file : %s\n","points.add");
                exit(0);)
                exit(0);)
    If(h[0] == 'Y') /*** hidden points to be added horizontally ?
    If(h[0] == 'Y') /*** hidden points to be added horizontally ?
        for(1 = 0; 1 < k; 1++)
        for(1 = 0; 1 < k; 1++)
            fprintf(add_file,"%f %f %f \n", addpoint[1][0], addpoint[1][
            fprintf(add_file,"%f %f %f \n", addpoint[1][0], addpoint[1][
        l
        l
    if(g[0] == 'n') /* background to be removed */
    if(g[0] == 'n') /* background to be removed */
    for(1 = 0; 1 < k2; 1++)
    for(1 = 0; 1 < k2; 1++)
        fprintf(add_file,"%f %f %f \n", support[1][0], support[1][1]
        fprintf(add_file,"%f %f %f \n", support[1][0], support[1][1]
        }
        }
    If(sup[0] == 'Y') /* vertical points to be added */
    If(sup[0] == 'Y') /* vertical points to be added */
        { for(1 = 0; 1 < k3; 1++)
        { for(1 = 0; 1 < k3; 1++)
        fprintf(add_file,"%f %f %f \n", side_add_points[i][0], side_
        fprintf(add_file,"%f %f %f \n", side_add_points[i][0], side_
        }
        }
    }
    }
1
1
*)
*)
function for multiplying matrix with a vector
function for multiplying matrix with a vector
**************************************************************************
**************************************************************************
matrix_mult(vector, matrix, result)
matrix_mult(vector, matrix, result)
double vector[3], matrix[4][4], result[3];
double vector[3], matrix[4][4], result[3];
f
f
result[0] = matrix[0][0] * vector[0] +
result[0] = matrix[0][0] * vector[0] +
matrix[0][1] * vector[1] +
matrix[0][1] * vector[1] +
                                    matrix[0][2] * vector[2] + matrix[0][3];
                                    matrix[0][2] * vector[2] + matrix[0][3];
result[1] = matrix[1][0] * vector[0] +
result[1] = matrix[1][0] * vector[0] +
matrix[1][1] * vector[1] +
matrix[1][1] * vector[1] +
matrix[1][2] * vector[2] + matrix[1][3];
```

```
matrix[1][2] * vector[2] + matrix[1][3];
```

```
```

../trial/rca_calib.c Thu May 5 17:07:25 1988 7
result[2] = matrix[2][0] * vector[0] +
matrix[2][1] * vector[1] +
matrix[2][2] * vector[2] + matrix[2][3];
}
|************************************************************************
function for making a T matrix
***************************************************************************
make matrix(TR)
double TR[4][4];
{ double rx, ry, rz, x, y, z;
x = 80.0;
y = -180.0;
z = 0;
rx = 90.0;
ry = 45.0;
rz = -15.0;
rx = rx * PI/180;
ry = ry * PI/180;
rz = rz * PI/180;
/*
homogenous transformation : Euler angles
*/
TR[0][0] = cos(rx)*\operatorname{cos(ry)*cos(rz) - sin(rx)*sin(rz);}
TR[0][1] = - cos(rx)*\operatorname{cos}(ry)*\operatorname{sin}(rz) - sin(rx)*}\operatorname{cos}(rz)
TR[0][2] = cos(rx)*sin(ry);
TR[0][3] = x;
TR[1][0] = sin(rx)*\operatorname{cos(ry)*cos(rz) + cos(rx)*sin(rz);}
TR[1][1] = - sin(rx)*\operatorname{cos(ry)*sin(rz) + cos(rx)*cos(rz);}
TR[1][2] = sin(rx)*sin(ry);
TR[1][3] = Y;
TR[2][0] = -sin(ry)*cos(rz);
TR[2][1] = sin(ry)*sin(rz);
TR[2][2] = cos(ry);
TR[2][3] = z;
TR[3][0] = 0;
TR[3][1] = 0
TR[3][2] = 0
TR[3][2] = = 1

```

\section*{double \(s q(x\)}
```

double $x$ :
(return(x*x);
\}
read picture (filename, buffer)
char filename[50]:
int buffer[PICDIM_X][PICDIM_Y];
1
int 1, j;
FILE *infs;
unsigned char *pmpoint:
short int *pms point:
/* open input pm file */
if $($ (infs $=$ fopen(filename,"r")) $==$ NULL)
printf("file open error : $\% \mathrm{~s}$ $\backslash \mathrm{n}$ ", filename): exit (0):
\}
/* read inputfile into the pmpic buffer */

```
```

f((pm1 = pm_read(infs,0)) == NULL)

```
f((pm1 = pm_read(infs,0)) == NULL)
        printf("error in reading the pmfile ts",filename);
        printf("error in reading the pmfile ts",filename);
        exit(0):
        exit(0):
    }
    }
sizex = (pml->pm_nrow); /* # of rows */
sizex = (pml->pm_nrow); /* # of rows */
sizey = (pml->pm_ncol); /* # of columns */
sizey = (pml->pm_ncol); /* # of columns */
fprintf(stderr,"rows : fd ; columns : fd\n",sizex,sizey);
fprintf(stderr,"rows : fd ; columns : fd\n",sizex,sizey);
If(pml->pm_form == PM_C)
If(pml->pm_form == PM_C)
    {
    {
        pm_point = (unsigned char *) pml->pm_image
        pm_point = (unsigned char *) pml->pm_image
            for(1=0;1<s1zex;1++)
            for(1=0;1<s1zex;1++)
                    for(j=0; f<sizey;}\++
                    for(j=0; f<sizey;}\++
                    {
                    {
                    buffer[1][f] = *(pm_point):
                    buffer[1][f] = *(pm_point):
                    pm_point++;
                    pm_point++;
                    }
                    }
    |
    |
    lse if(pm1->pm_form == PM_S)
```

    lse if(pm1->pm_form == PM_S)
    ```
\}

\section*{../trial/rca_calib.c Thu May 5 17:07:25 1988 \(\quad 9\)}
pms_point \(=\) (short int *) pml->pm_image;
for ( \(1=0 ; 1<\) sizex; \(1++\) )
for ( \(\mathrm{f}=0 ; \mathrm{j}<\) sizey; \(\mathrm{j}++\) )
(buffer[1][J] = *(pms_point); pms_point++;
fse
else
els
printf("Image file in unrecognized format.Exiting. \(\backslash n\) "); exit(0):
1
\}/* end of read plcture *!
classify.c Had Mar 9 14:00:45 1988 1
\(\qquad\)
Program to classify the superquadric model into one of the four broad categories :
flat
roli.
IPP.
Format of the input file is as output from the rec9.out program. Run as:
flassify fit_*.originalpoints fit_*.addedpoints
\(0.00<=e 1, \mathrm{e} 2<=1.00\)

\#nclude <stdio.h>
Include <math.h>
define TIM 3
/* to implement << or >> */
define flat 1
define BOX 2
define ROLL 3
define IPP 4
double a1[3],a2[3],a3[3]; double e1[3], e2[3];
double measured goodness1, goodness
** to \({ }^{\text {* }}\) to store al, a2 and a3 /* goodness of fit */ double measured_goodness2;
char dummy[100],orig[50],add[50]:
int type1,type2;
double dum;
double \(k\) box \(=10.0\).
\[
\begin{aligned}
& k 1 \_ \text {roll }=10.0 \text { : } \\
& k 2 \text { roll }=15.0 \text { : }
\end{aligned}
\]

FILE *infs,*addfile,*outfile;
main(argc, argv)
1nt argc;
char *argu[]
int \(1, f, k, 1\) int good1, good2.

\section*{lasaify.c Wed Mar 9 14:00:45 1988 2}
strcpy (orig, argv[1]);
strcpy (add, argv[2]):
if((infs \(=\) fopen (argv[1],"r")) == NULL)
printf("classify : File open error : \%s \(\mathbf{n n}^{\prime \prime}\), argv[1]); exit (0):
\}
if((addfile \(=\) fopen(argv[2],"r")) \(==\) NULL)
printf("classify : File open error : \&s\n",argv[2]); exit (0):
,
if((outfile \(=\) fopen (argv[3], "w")) \(==\) NULL)
printf("classify : File open error : \%s n", argv[3]); \(^{\prime}\) exit(0):
\}
fscanf(infs,"\%s", dummy);
fscanf(infs,"qf \%f tf", \&al[1],\&a2[1],\&a3[1]); fscanf(infs, "qf of ifn,\&al[1],\&a2[1],\&a3[1]): fscanf(infs,"qf \%f \%f",\&al[1],\&a2[1],\&a3[1]);


fscanf(infs," \%d \%d \%d", \&dum, \&dum, \&dum) : fscanf(infs,"ヶs", dummy):
fscanf(infs,"\%f", \&measured_goodness1);
printf("qf,qf,qf\n",a1[1],a2[1],a3[1]): printf("qf,qf\n",e1[1],e2[1]); printf("qf \(\backslash n^{n}\), measured_goodness1);
fscanf (addfile," \({ }^{\text {s }}\) ", dummy) ;
fscanf(addfile," \%f \%f \&f",\&a1[2], \&a2[2],\&a3[2]): fscanf(addfile,"\%f \%f \%f", \&al[2],\&a2[2],\&a3[2]); fscanf(addfile,"tf \%f \%f", \&al[2],\&a2[2],\&a3[2]);
fscanf(addfile,"qf \&f",\&el[2],\&e2[2]):


\section*{classify.c Wad Mar 9 14:00:45 1988 3}
fscanf (addfile," \%d \%d \%d", \&dum, \&dum, \&dum) :
fscanf (addfile," \(\%\) s", dummy);
fscanf (addfile,"\%f", \&measured_goodness2);
printf(*f, tf,tf\nn,a1[2],a2[2],a3[2]):
printf("\%f, \%f \(\backslash n\) ", el[2], e2[2]);
printf("\&f \(\mathrm{nn}^{\prime \prime}\),measured_goodness2)
/* Readin the Threshold values */
printf("K for Box : ")
scanf("qf", \&k box) ;
printf("K1 and K2 for Roll : ");
scanf("\%f \%f", \&k1_roll,\&k2_roll):
printf("Goodness of fit measure : "):
scanf("tf", \&goodness);
/* FIRST classify the object according to the a1,a2,a3,e1,e2 values
typel \(=\) classified(1);
type2 = classified(2);
good1 = good_enough (measured_goodness1): good2 = good_enough (measured_goodness2);
\[
\text { if }((\text { good } 1==1) \& \&(\text { good } 2==0))
\]
/* first fit is better than second fit */
classification(1,type1);
else if \(((\) good \(1==0) \& \&(\) good2 \(==1))\)
( /* second fit is better than the first fit *
classification(2,type2);
else if((goodl == 1) \&\& (good2 == 1))
( /* both the fits are acceptable */
volume_criterion():
else
( /* Both the fits are unacceptable */ classification(3,IPP);
)
\}
classified (num)
```

classify.c Wod Mar 9 14:00:45 1988
int num;
l
int 1,j,k;
1f((TIM*a3[num] < a1[num]) \&\& (TIM*a3[num] < a2[num]) \&\&
(e2[num] < 0.5)\&\& (e1 [num] < 0.5))
return(FLAT):
else if(((a1[num]*TIM < a3[num]) || (a2[num]*TIM < a3[num])) \&\&
(e1[num] < 0.5) \&\& (e2[num] < 0.5))
return (FLAT) ;
else if((a1[num] > k_box) \&\& (a2[num] > k_box) \&\& (a3[num] > k_box)
(e1[num] < 0.5) \&\& (e2[num] < 0.5)
return(BOX);
else if((a1[num] > k1_roll) \&\& (a2[num] > k1_roll) \&\& (a3[num] > k2
(e1[num] < 0.5) \&\& (e2[num] > 0.5))
return(ROLL):
else
return(IPP);
}
classification(num,type)
int num;
int type:
l
char str[50];
char string[10]:
switch(type)
case FLAT : sprintf(string,"\&s","Flat");break;
case BOX : sprintf(string,"ts","BOX");break;
case ROLL : sprintf(string,"\&s","Roll");break
case IPP : sprintf(string,"%s","Ipp");break;
}
1f((num == 1))
{
printf("Points not added in the fit.\n");
printf("Object classified as is \n",string);
fprintf(outfile,"%s", string).
sprintf(str,"%s ts %s","cp ",orig," fit.final ");
system(str);
}

```
```

classify.c Wad Mar 9 14:00:45 1988 5
else if((num == 2))
printf("Points added to obtain the fit.\n");
printf("Object classified as %%s \n",string);
fprintf(outfile,"%s",string);
sprintf(str,"%s %s %s","cp ",add," fit.final ");
system(str):
}
else
printf("Object Classified as %s \n",string);
fprintf(outfile,"%s",string);
sprintf(str,"%s %s \&s","cp ",orig," fit.final ");
system(str):
}
}
volume_criterion()
1
double vol1,vol2;
vol1 = a1[1]*a2[1]*a3[1];
vol2 = al[2]*a2[2]*a3[2];
printf("Following volume criterion \n");
if(voll <= vol2)
classification(2,type2);
else
classification(1,type1);
}
good_enough(goodn)
double goodn;
l
if(goodn < goodness)
return(1);
else
return(0);
}

```
```

../trial/rca_calib.c Thu May 5 17:07:25 1988 7
result[2] = matrix[2][0] * vector[0] +
matrix[2][1] * vector[1] +
matrix[2][2] * vector[2] + matrix[2][3];
}
|************************************************************************
function for making a T matrix
***************************************************************************
make matrix(TR)
double TR[4][4];
{ double rx, ry, rz, x, y, z;
x = 80.0;
y = -180.0;
z = 0;
rx = 90.0;
ry = 45.0;
rz = -15.0;
rx = rx * PI/180;
ry = ry * PI/180;
rz = rz * PI/180;
/*
homogenous transformation : Euler angles
*/
TR[0][0] = cos(rx)*\operatorname{cos(ry)*cos(rz) - sin(rx)*sin(rz);}
TR[0][1] = - cos(rx)*\operatorname{cos}(ry)*\operatorname{sin}(rz) - sin(rx)*}\operatorname{cos}(rz)
TR[0][2] = cos(rx)*sin(ry);
TR[0][3] = x;
TR[1][0] = sin(rx)*\operatorname{cos(ry)*cos(rz) + cos(rx)*sin(rz);}
TR[1][1] = - sin(rx)*\operatorname{cos(ry)*sin(rz) + cos(rx)*cos(rz);}
TR[1][2] = sin(rx)*sin(ry);
TR[1][3] = Y;
TR[2][0] = -sin(ry)*cos(rz);
TR[2][1] = sin(ry)*sin(rz);
TR[2][2] = cos(ry);
TR[2][3] = z;
TR[3][0] = 0;
TR[3][1] = 0
TR[3][2] = 0
TR[3][2] = = 1

```

\section*{double \(s q(x\)}
```

double $x$ :
(return(x*x);
\}
read picture (filename, buffer)
char filename[50]:
int buffer[PICDIM_X][PICDIM_Y];
1
int 1, j;
FILE *infs;
unsigned char *pmpoint:
short int *pms point:
/* open input pm file */
if $($ (infs $=$ fopen(filename,"r")) $==$ NULL)
printf("file open error : $\% \mathrm{~s}$ $\backslash \mathrm{n}$ ", filename): exit (0):
\}
/* read inputfile into the pmpic buffer */

```
```

f((pm1 = pm_read(infs,0)) == NULL)

```
f((pm1 = pm_read(infs,0)) == NULL)
        printf("error in reading the pmfile ts",filename);
        printf("error in reading the pmfile ts",filename);
        exit(0):
        exit(0):
    }
    }
sizex = (pml->pm_nrow); /* # of rows */
sizex = (pml->pm_nrow); /* # of rows */
sizey = (pml->pm_ncol); /* # of columns */
sizey = (pml->pm_ncol); /* # of columns */
fprintf(stderr,"rows : fd ; columns : fd\n",sizex,sizey);
fprintf(stderr,"rows : fd ; columns : fd\n",sizex,sizey);
If(pml->pm_form == PM_C)
If(pml->pm_form == PM_C)
    {
    {
        pm_point = (unsigned char *) pml->pm_image
        pm_point = (unsigned char *) pml->pm_image
            for(1=0;1<s1zex;1++)
            for(1=0;1<s1zex;1++)
                    for(j=0; f<sizey;}\++
                    for(j=0; f<sizey;}\++
                    {
                    {
                    buffer[1][f] = *(pm_point):
                    buffer[1][f] = *(pm_point):
                    pm_point++;
                    pm_point++;
                    }
                    }
    |
    |
    lse if(pm1->pm_form == PM_S)
```

    lse if(pm1->pm_form == PM_S)
    ```
\}

\section*{../trial/rca_calib.c Thu May 5 17:07:25 1988 \(\quad 9\)}
pms_point \(=\) (short int *) pml->pm_image;
for ( \(1=0 ; 1<\) sizex; \(1++\) )
for ( \(\mathrm{f}=0 ; \mathrm{j}<\) sizey; \(\mathrm{j}++\) )
(buffer[1][J] = *(pms_point); pms_point++;
fse
else
els
printf("Image file in unrecognized format.Exiting. \(\backslash n\) "); exit(0):
1
\}/* end of read plcture *!
classify.c Had Mar 9 14:00:45 1988 1
\(\qquad\)
Program to classify the superquadric model into one of the four broad categories :
flat
roli.
IPP.
Format of the input file is as output from the rec9.out program. Run as:
flassify fit_*.originalpoints fit_*.addedpoints
\(0.00<=e 1, \mathrm{e} 2<=1.00\)

\#nclude <stdio.h>
Include <math.h>
define TIM 3
/* to implement << or >> */
define flat 1
define BOX 2
define ROLL 3
define IPP 4
double a1[3],a2[3],a3[3]; double e1[3], e2[3];
double measured goodness1, goodness
** to \({ }^{\text {* }}\) to store al, a2 and a3 /* goodness of fit */ double measured_goodness2;
char dummy[100],orig[50],add[50]:
int type1,type2;
double dum;
double \(k\) box \(=10.0\).
\[
\begin{aligned}
& k 1 \_ \text {roll }=10.0 \text { : } \\
& k 2 \text { roll }=15.0 \text { : }
\end{aligned}
\]

FILE *infs,*addfile,*outfile;
main(argc, argv)
1nt argc;
char *argu[]
int \(1, f, k, 1\) int good1, good2.

\section*{lasaify.c Wed Mar 9 14:00:45 1988 2}
strcpy (orig, argv[1]);
strcpy (add, argv[2]):
if((infs \(=\) fopen (argv[1],"r")) == NULL)
printf("classify : File open error : \%s \(\mathbf{n n}^{\prime \prime}\), argv[1]); exit (0):
\}
if((addfile \(=\) fopen(argv[2],"r")) \(==\) NULL)
printf("classify : File open error : \&s\n",argv[2]); exit (0):
,
if((outfile \(=\) fopen (argv[3], "w")) \(==\) NULL)
printf("classify : File open error : \%s n", argv[3]); \(^{\prime}\) exit(0):
\}
fscanf(infs,"\%s", dummy);
fscanf(infs,"qf \%f tf", \&al[1],\&a2[1],\&a3[1]); fscanf(infs, "qf of ifn,\&al[1],\&a2[1],\&a3[1]): fscanf(infs,"qf \%f \%f",\&al[1],\&a2[1],\&a3[1]);


fscanf(infs," \%d \%d \%d", \&dum, \&dum, \&dum) : fscanf(infs,"ヶs", dummy):
fscanf(infs,"\%f", \&measured_goodness1);
printf("qf,qf,qf\n",a1[1],a2[1],a3[1]): printf("qf,qf\n",e1[1],e2[1]); printf("qf \(\backslash n^{n}\), measured_goodness1);
fscanf (addfile," \({ }^{\text {s }}\) ", dummy) ;
fscanf(addfile," \%f \%f \&f",\&a1[2], \&a2[2],\&a3[2]): fscanf(addfile,"\%f \%f \%f", \&al[2],\&a2[2],\&a3[2]); fscanf(addfile,"tf \%f \%f", \&al[2],\&a2[2],\&a3[2]);
fscanf(addfile,"qf \&f",\&el[2],\&e2[2]):


\section*{classify.c Wad Mar 9 14:00:45 1988 3}
fscanf (addfile," \%d \%d \%d", \&dum, \&dum, \&dum) :
fscanf (addfile," \(\%\) s", dummy);
fscanf (addfile,"\%f", \&measured_goodness2);
printf(*f, tf,tf\nn,a1[2],a2[2],a3[2]):
printf("\%f, \%f \(\backslash n\) ", el[2], e2[2]);
printf("\&f \(\mathrm{nn}^{\prime \prime}\),measured_goodness2)
/* Readin the Threshold values */
printf("K for Box : ")
scanf("qf", \&k box) ;
printf("K1 and K2 for Roll : ");
scanf("\%f \%f", \&k1_roll,\&k2_roll):
printf("Goodness of fit measure : "):
scanf("tf", \&goodness);
/* FIRST classify the object according to the a1,a2,a3,e1,e2 values
typel \(=\) classified(1);
type2 = classified(2);
good1 = good_enough (measured_goodness1): good2 = good_enough (measured_goodness2);
\[
\text { if }((\text { good } 1==1) \& \&(\text { good } 2==0))
\]
/* first fit is better than second fit */
classification(1,type1);
else if \(((\) good \(1==0) \& \&(\) good2 \(==1))\)
( /* second fit is better than the first fit *
classification(2,type2);
else if((goodl == 1) \&\& (good2 == 1))
( /* both the fits are acceptable */
volume_criterion():
else
( /* Both the fits are unacceptable */ classification(3,IPP);
)
\}
classified (num)
```

classify.c Wod Mar 9 14:00:45 1988
int num;
l
int 1,j,k;
1f((TIM*a3[num] < a1[num]) \&\& (TIM*a3[num] < a2[num]) \&\&
(e2[num] < 0.5)\&\& (e1 [num] < 0.5))
return(FLAT):
else if(((a1[num]*TIM < a3[num]) || (a2[num]*TIM < a3[num])) \&\&
(e1[num] < 0.5) \&\& (e2[num] < 0.5))
return (FLAT) ;
else if((a1[num] > k_box) \&\& (a2[num] > k_box) \&\& (a3[num] > k_box)
(e1[num] < 0.5) \&\& (e2[num] < 0.5)
return(BOX);
else if((a1[num] > k1_roll) \&\& (a2[num] > k1_roll) \&\& (a3[num] > k2
(e1[num] < 0.5) \&\& (e2[num] > 0.5))
return(ROLL):
else
return(IPP);
}
classification(num,type)
int num;
int type:
l
char str[50];
char string[10]:
switch(type)
case FLAT : sprintf(string,"\&s","Flat");break;
case BOX : sprintf(string,"ts","BOX");break;
case ROLL : sprintf(string,"\&s","Roll");break
case IPP : sprintf(string,"%s","Ipp");break;
}
1f((num == 1))
{
printf("Points not added in the fit.\n");
printf("Object classified as is \n",string);
fprintf(outfile,"%s", string).
sprintf(str,"%s ts %s","cp ",orig," fit.final ");
system(str);
}

```
```

classify.c Wad Mar 9 14:00:45 1988 5
else if((num == 2))
printf("Points added to obtain the fit.\n");
printf("Object classified as %%s \n",string);
fprintf(outfile,"%s",string);
sprintf(str,"%s %s %s","cp ",add," fit.final ");
system(str):
}
else
printf("Object Classified as %s \n",string);
fprintf(outfile,"%s",string);
sprintf(str,"%s %s \&s","cp ",orig," fit.final ");
system(str):
}
}
volume_criterion()
1
double vol1,vol2;
vol1 = a1[1]*a2[1]*a3[1];
vol2 = al[2]*a2[2]*a3[2];
printf("Following volume criterion \n");
if(voll <= vol2)
classification(2,type2);
else
classification(1,type1);
}
good_enough(goodn)
double goodn;
l
if(goodn < goodness)
return(1);
else
return(0);
}

```


Program to compute the first and second order derivatives of the \(r\) image and finally the Gaussian and Mean curvature at all the image po is given below. The program outputs the sign map of gaussian and mean curvature.

Mar 17, 1988 Interactive processing.
Can handle PM_S and PM_C images. Outputs only PM_C im
Mar 25, 1988 Display histogram of arbitrary parameter on IKONĀS using quickdraw.
\#nclude <stdio.h>
\#nclude <math.h>
\#include <local/pm.h>
\#nclude <ik.h>
Include "/usr/users/alok/advanced/spline_include.h"
\#define SCALE 1.978
\#define 2SCALE 1.500
\#define REGION_SIZE 1000
define NSYS 25
\#define NDATA 1000
struct region_type 1 int order:
float fit error
int label;
int size;
float normal vector[3];
\} regions[REGION_SIZE];
struct Image \(\{\)
float \(x u, x v, x u u, x v v, x u v\);
float fit_error;
float gauss,mean;
int label;
int th_mean;
int th gauss
float \(\bar{q}\);
float sqrtg;
* should be same as xscale and yscale as mm/pixel ;for Gus' scanner < \(1 / \star\) /* zscale in the digitized image */
apline.c
float cosph1;
| image, parm[BUFSIZE][BUFSIZE]; \(/ \star\) for parameters of pixels *
float ri[BUFSIZE][BUFSIZE]; /* range image */
int buffer[BUFSIZE][BUFSIZE];
int line[BUFSI2E];
float linef[BUFSIZE];
int of \(f x\) of \(f y\);
struct mxvali
float fit_error:
float gauss;
float mean
float q;
float sqrtg:
float cosphi;
float depth;
int label;
int th_gauss;
int th_mean;
\}maxv,minv;
float x[1024];
float y[1024];
float value;
float \(x m i n, x m a x, y m i n, y m a x ;\)
int lopt;
int npts;
/* followind parameters are returned by the least square fitting pr
double ix, 1y, ixx, 1yy, ixy, iyx, 1xxx, iyyy, ixyy, ixxy ;
double a30, a21, a12, a03, a20, a11, a02, a10, a01, a00 ;
double xu, xv, xuu, xvv, xuv;
double fit_error; /* the surface fit error */
int s_order: /* the order of the surface fitted in the nbd */
/* following are global to this file */
nt csize, rsize :
int offset1 ;
nt approx ;
int soorder; /* fitted surface order */
double lsqerr():
FILE *dumpfile;
int todump;
spline.c Wed Mar 30 11:54:33 1988 3 ".fit-error", \(/ \star\) to the input file-name to ".quad-var", /* output-file-name
". zero-mean",
"zero-mean":
". zero-gauss
.. ccaf"
. ccaf".
".sign-mean",
".sign-gauss",
".mag-pcd",
". pda"
".mgc".
".mmecritical"
.n-critical
". .region" \({ }^{\prime}\);
int mask u[5][5] = 1
\(\{-1,-6,-10,-6,-1\}\),
\(\{-2,-20,-52,-20,-2\}\)
\(\{0,0,0,0,0\}\).
\(\{2,20,52,20,2\}\)
\(\{1,6,10,6,1\}\}\) :
int mask_v[5][5] = 1
\(\{-1,-2,0,2,1\}\).
\(\{-6,-20,0,20,6\}\).
\(\{-10,-52,0,52,10\}\),
\(\{-6,-20,0,20,6\}\).
\(\{-1,-2,0,2,1\})\) :
int mask_uu[5][5] = 1
\((1,6,10,6,1)\)
\(\{0,8,32,8,0\}\),
\(\quad(-2,-28,-84,-28,-2\}\)
\(\{-2,-28,-84,-28,-2\}\),
\((0,8,32,8,0\}\)
\(\left\{\begin{array}{l}\{0,8,32,8,0\} \text {, } \\ \{1,6,10,6,1)\} \text {; }\end{array}\right.\)
int mask_vv[5][5] = 1
\(\{1,0,-\overline{2}, 0,1\}\).
\(\{1,0,-2,0,1\}\).
\(\{6,8,-28,8,6\}\),
\(\{10,32,-84,32,10\}\).
\(\{6,8,-28,8,6\}\),
\(\{1,0,-2,0,1\}\}\);
int mask uv[5][5] = 1
\(\{1,2,0,-2,-1\}\).
apline.c Wed Mar 30 11:54:33 1988
\(2,12,0,-12,-2\}\).
\(\{0,0,0,0,0\}\).
\(\{-2,-12,0,12,2\}\).
\(\{-1,-2,0,2,1\})\) :
int op_u[3][3] \(=\{\{-1,-4,-1\}\),
\(\{0,0,0\}\).
\(\{1,4,1\}\}\) :
int op_v[3][3] = 1
\(\{-1,0,1\}\),
\(\{-4,0,4\}\),
int op_uu[3][3] =
\(\{1,4,1\}\),
\(\{-2,-8,-2\}\).
\(\{1,4,1\}\}\);
int op_vv[3][3] = 1
\(\{1,-2,1\}\),
\(\{4,-8,4\}\),
\(\{1,-2,1\}\}\);
int op_uv[3][3] \(=\)
\(\{1,0,-1\}\),
\(\{0,0,0\}\).
\(\{-1,0,1\}\}\);
float weight_u \(=288.0\)
loat weight \(-v=288.0\)
float welght_uu \(=144.0\);
float welght_vv \(=144.0\);
float weight_uv =96.0;
float convo():
float convo l():
float abszz():
float div_u \(=12.0\)
float div_v = 12.0;
float div_uu= 6.0 ;
float div_vv= 6.0 :
float div_uv= 4.0 :
float max_mean \(=-1000.0\);
float max_gauss= -1000.0;

\section*{spline.c Wed Mar 30 11:54:33 1988}
float min_mean \(=1000.0\);
float min_gauss= 1000.0;

\section*{main (argc,argv)}
int argc;
char *argu[];
1
int \(1, j, k, 1, m, n\);
pmp1c *pm1,*pm2;
FILE *infile,*outfile
int row, col;
char *cmt;
unsigned char *pm_point_uchar:
short int *pm_point_short;
int option;
float gauss,mean
char temp[40]:
int offset;
pmpic *pmp[16]
unsigned char *uchar point[16];
int temp_value;
int mean_val,gauss_val;
int bool[17]:
FILE *fp[16];
float th_mean, th_gauss;
float t1,t2,t3;
float \(q\);
char \(c_{i}\)
char smooth;
int back_threshold;
int to_smooth;
int to_print files; \(\quad / \star=1\), if all outputs desired in a file */ char string[30];
float sf;
pmpic *pm;
FILE *fpl;
double scale,zscale; \(\quad / *\) scale \(=\) yscale \(=x s c a l e ; ~ z s c a l e ~ i s ~ f o r ~ d ~\) u_int image_format;
int subs;
```

1f((argc > 3) || (argc < 2))
printf("Usage : spline input_filename output_diagonistics_file\
exit(0);
}
1f((1nfile=fopen(argv[1],"r")) == NULL)

```
spline.c

\section*{Mod Mar 30 11:54:33 1988}
( printf("Can't open \%s\n", argv[1]): exit(0):
\}
todump \(=0\);
1 f (argc \(==3\)
if
```

if((dumpfile=fopen(argv[2],"w")) == NULL)
( printf("Can't open \%s\n", argv[2]);
exit (0) ;
$\stackrel{?}{\text { todu }}$
todump $=1 ;$

```
    )
/* \(1 \mathrm{f}(1\) kopen (NULL) \(==-1\) ) printf(" Can't open IKONAS \(\left.\backslash n^{\prime \prime}\right) ; * /\)
init_structure (): /* initializes the structures */
cmt \(=\) (char *) pm_cmt (argc,argv); /* get the command line */
\(\operatorname{if}((\mathrm{pml}=\mathrm{pm}\) _read \((\operatorname{lnf11e}, 0))==\operatorname{NULL})\)
( printf("Error in reading PM-Format file : \% \({ }^{\prime} \backslash n^{\prime}\), argv[1]); exit(0):
)
row \(=\) pml->pm_nrow
col = pml->pm_ncol
rsize = row;
csize = col;
of \(f x=\) offy \(=0\);
/* copy the input image into the image */
if \((\mathrm{pml}->\mathrm{pm}\) form \(=\) \(=\) PM_C) /* one byte per pixel picture */
i
mpoint_uchar \(=\) (unsigned char *) pm1->pm_1mage;
image_format \(=\) PM_C;
zscale \(=\) ZSCALE; \(/ *\) for Gus' scanner */
scale = SCALE;
printf("initializing buffer \(\\) n"):
for ( \(1=0\); \(1<\) row; \(1++\) )
for ( \(\mathrm{j}=0 ; \mathrm{j}<\mathrm{col} ; \mathrm{j}++\) )
i
ri[1][f] = ((float) *pm_point_uchar)*((float)(zscale/sca if(maxv.depth < ri[i][j]) maxv.depth = ri[1][j];

\section*{Wed Mar 30 11:54:33 1988 7}
if(minv.depth > ri[i][f]) minv.depth = ri[i][j]; pm_point_uchar++;
\}
else if(pml->pm_form \(==\) PM_S) /* short integer picture */ 1
pm point short \(=(\) short int *) pml->pmimage;
image format \(=P M S\);
zscale \(=1.00\)
printf("initializing buffer \(\ln\) ")
for ( \(1=0 ; 1<\) row \(; 1++\) )
for ( \(j=0 ; j<\mathrm{col} ; j++\)
1
ri[1][f] = ((float) *pm point short)*((float) (zscale/scal if(maxv.depth < ri[1][j]) maxv.depth = ri[1][1];
if(minv.depth > ri[1][f]) minv.depth \(=r 1[1][j]\);
pm_point_short++;
\}
\(\stackrel{\}}{\text { else }}\)
fprintf(stderr,"unrecognized PM format in image"): exit (0):
\}
/* background threshold is the background value in image as it is \(r\) and not in uniformly scaled ( in \(Z\) direction ) image */
printf(" Background Threshold : ");
scanf("\%d", \&back_threshold);
back_threshold \(=\) (back_threshold * zscale)/scale;
read_specs (): /*, read specifications */
printf("Gaussian-smooth the picture ?(1 if yes) :");
scanf("\%d", \&to_smooth);
1f(to_smooth \(==1\) ) gaussian(ri,row,col);
to print files \(=0\) :
1f(to_print_files \(==1\) )
(
printf(" Following outputs can be obtained : \(\mathrm{In}^{\prime \prime}\) ):
printf(" Indicate your choice by entering integers in a lin printf(" 1. Square root of metric determinant (edge magnitude) printf(" 2. Quadratic Variation (flatness measure image Q. \(\mathrm{nn}^{\prime \prime}\) ) printf(" 3. Zeros of the mean curvature : \(\left.H=0 \backslash n^{n}\right)\);
spline.c
printf(" 4. Zeros of the Gaussian curvature : \(k=0 \backslash n^{*}\) ): printf(" 5. Zeros of the cosine-of-the-coordinate-function cos printf(" 6. Cosine of the Coordinate angle Function :cos \(0 \backslash n^{\prime \prime}\) ) printf(" 7. Sign regions of mean curvature \(\left.\operatorname{sgn}(\mathrm{H}) \backslash \mathrm{n}^{\prime}\right)\); printf(" 8. Sign regions of Gaussian curvature, \(\operatorname{sgn}(\mathrm{K}) \mathrm{nn}^{\prime \prime}\) ) printf(" 9.Magnitude of principal curvatures difference :sqrt ( printf(" 10.Principal direction angle n \(^{\prime \prime}\) ):
printf(" 11.Magnitude of Gaussian curvature \(\left.K \backslash n^{\prime \prime}\right)\) :
printf(" 12.Magnitude of Mean curvature H\n");
printf(" 13.Non degenerate critical points image :fu=fv=0<>0\n
printf(" 14.Critical points image\n");
printf(" 15.Labeled regions n \(^{\prime \prime}\) );
printf(" 16.ALL OF THE ABOVE n' \(^{*}\) ):
\}
/* read in user requirements and set boolean corresponding to \(t\)
for ( \(1=0 ; 1<17 ; 1++\) ) bool[1] \(=0\);
\(1=0\);
while ( \(1=0\) )
scanf("\%d", \&n);
getchar():
1f \(((n>0) \& \&(n<17))\) \{bool \([n]=1\);
else \(\{1=1 ;\}\)
\(1 \stackrel{\}}{=}\)
if (bool[16] \(==1\) )
for ( \(1=0 ; 1<17 ; 1++\) ) bool[1] = 1;
printf("\n Outputs are:\n\n");*/
/* generate output filenames */
/* for ( \(1=1 ; 1<16 ; 1++\) )
if (bool[1] == 1)
if
strcpy (temp, argv(1]):
strcat (temp, name ext[1])
printf(" is \(\backslash n^{\prime \prime}\),temp):
\(\mathrm{fp}[1]=\) fopen (temp, "w")
pmp[1] = pm_alloc():
pmp [1]->pm_nrow \(=\) row;
npmp[1]->pm_ncol \(=\) col \(; ~\)
pmp[1]->pm_Image \(=(\) char *) malloc(pm psize (pmp[1])) uchar_point \([1]=\) (unsigned char *) pmp[1]->pm_image;
j*/ /*
* Now compute the derivatives and Gaussian and Mean curvature at e image point
```

zpline.c Wed Mar 30 11:54:33 1988 9
*/
printf("row : %d column %d offset1 %d\n",row,col,offset1);
printf("rsize : fd csize : td approx : fd\n",rsize,csize,approx)
compute_amatrix();
for(1=offset1;1<row-offset1;1++)
l
for(j=offset1;j<col-offset1;j++)
if(ri[1][j] > back_threshold)
I
If((approx == 1) || (approx == 2) || (approx == 3))
i
poly2(1,j):
xu = ix;
xv = 1Y;
xuu = 1xx;
xvv = 1yy;
xuv = 1xy;
else
if(approx == 5) /* B-spline fitting using Kak-Hwang mask
l
xu = convo(1, J,mask_u,weight_u,5):
xv = convo(1, 1,mask_v,weight_v,5);
xuu= convo(1,j,mask_uu,weight_uu,5);
xvv= convo(i,j,mask_vv,we1ght_vv,5);
xuv= convo(1,j,mask_uv,weight_uv,5);
} }
else /*without smoothing B-spline fitting using Kak-Hwan
{
xu = convo_l(1, j, op_u, div_u,3);
xu = convo_1(1, ,oo_u,div_u, 3)
xuu= convo_l (1, j,op_uu,div_uu,3);
xvv= convo_l (i,j,op_vv,div_vv,3);
xuv= convo_l(1,j,op_uv,div_uv,3);
}
(parm[1][f].xu) = xu;
parm[1][f].xv = xv;
parm[1][j].xuu = xuu;
parm[1][j].xvv = xvv
parm[i][j].xuv = xuv;
gauss = (xuu*xvv-xuv*xuv)/

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 10((float) (pow ((double) (1+xu*xu+xv*xv), (double) (1.5))));
if(fit_error >255) fit_error = 255;
if(fit_error > maxv.fit_error) maxv.fit_error = fit_error;
if(fit_error < minv.fit_error) minv.fit_error = fit_error:
parm[i][j].mean $=$ mean;
) /
/* End of loop to compute derivatives and curvature values */

```
```

/* open ikonas display */

```
```

/* open ikonas display */
printf("Can't open IKONAS \n");

```
```

    printf("Can't open IKONAS \n");
    ```
```

printf("ENTER >");
scanf("\%s", string);
while((strcmp("exit", string) $!=0$ ) \& (strcmp("quit", string) $!=0)$ )
\{/* interactive loop to do things starts here */
if (strcmp ("comp", string) $==0$ )
(strcmp ("comp*, string) $==0)$
(/* compute general things */
for ( $1=$ offset $1 ; 1<$ row-offset $1 ; 1++$ )
for ( $j=$ offset $1 ; j<c o l-o f f s e t 1 ; j++)$
if(ri[i][j] > back_threshold)
1
/*** sqrt g ***/
$\begin{aligned} \operatorname{parm}[1][j] . \text { sqrtg } & =(f l o a t) \operatorname{sqrt}((\text { double })(1+\operatorname{parm}[1][f] .\end{aligned}$
parm[1][j].sqrtg $=(f l o a t)$ sqrt ((double) $(1+\operatorname{parm}[1][f]$.

## pline.c <br> pline.c

```
If(gauss > maxv.gauss) maxv.gauss = gauss;
```

If(gauss > maxv.gauss) maxv.gauss = gauss;
if(mean > maxv.mean) maxv.mean = mean;
if(mean > maxv.mean) maxv.mean = mean;
if(gauss < minv.gauss) minv.gauss = gauss;
if(gauss < minv.gauss) minv.gauss = gauss;
if(mean < minv.mean) minv.mean = mean;
if(mean < minv.mean) minv.mean = mean;
if(gauss < minv.gauss) minv.gauss = gauss

```
if(gauss < minv.gauss) minv.gauss = gauss
```

if(fit_error > maxv.fit_error) maxv.fit_error = fit_error; if(fit_error < minv.fit_error) minv.fit_error = fit_error:

```
parm[1][f].fit_error = fit_error:
parm[1][j].gauss = gauss
```

\} $/ *$ End of loop to compute derivatives and curvature values */
for ( $1=$ offset $1 ; 1<$ row-offset $1 ; 1++$ )
if (parm[1][f].sqrtg > maxv.sqrtg)
maxv.sqrtg = parm[1][j].sqrtg;
if (parm[1][j].sqrtg < minv.sqrtg)
minv.sqrtg = parm[1][j].sqrtg;
/*** Q ***
parm[1][f].q $=$ (float) (parm[1][f].xuu*parm[1][f].xuu

2*parm[1][f].xuv*parm[1][f].xu parm[1][]] xvv*parm[1][]]
if (parm[1][f].q > maxv.q)
maxv.q $=$ parm[1][f].q;
maxv.q $=$ parm[1] (parm[1][f].q < minv.q)
minv.q = parm[1][f].q;
/*** zeros of cos phi ***/
parm[1][f].cosph1 = (float) (parm[1][f].xu*parm[1][f]. ((float) (sqrt ( (double) (1+parm[1][j].xu^parm[1][j].x parm[1][J].xv*parm[1][J].xv +
parm[1][j].xv*parm[1][j].xv + parm[1][J].xu*parm[1]
1f(parm[1][j].cosph1 > maxv.cosph1)
maxv.cosph1 = parm[1][f].cosph1:
if(parm[1][1].cosph1 < minv.cosph1)
minv.cosph1 $=$ parm[1][f].cosph1:

## \}

\}
else if(stramp("1kclose",string) $==0$ ) ikclose():
else if(stromp("ikopen",string) $==0$ ) 1kopen (NULL):
else if(strcmp("thresh",string) == 0)
\{
printf("threshold for Gaussian curvature > "):
scanf("\&f", \&th_gauss):
printf("threshold for Mean curvature > "):
scanf("\&f", \&th_mean):
for (1=offset1; $1<$ row-offset $1 ; 1++$ ) for ( $1=$ offset 1 ; $1<$ col-offset $1 ; j++$
if(ri[1][f] > back threshold)
1
If(fabs((double)parm[1][f].gauss) <= th gauss) parm[1][1].th_gauss $=0$; $/ *$ gaussian curv $=0 \star$ else if (parm[1][J].gauss < 0) parm[1][f].th gauss $=2 ; 1 *$ gaussian curv is -ve else parm[i][f].th gauss $=1$;/* gaussian curv is +v
if(parm[1][f].th gauss > maxv.th gauss)
maxv.th_gauss = parm[1][f].th_gauss;
if (parm[1][]].th gauss < minv.th gauss) minv.th_gauss = parm[1][J].th_gauss:
if(fabs((double)parm[1][1].mean) <= th_mean)
parm[1][f].th_mean $=0 ; \quad / \star$ mean curv $=0$ */
parm[1][j].th mean =
parm[1][f].th mean = 2; /* mean curv is -ve */
else parm[i][f].th mean $=1 ; / *$ mean curv is +ve */
if (parm[1][f].th mean > maxv.th_mean)
maxv.th mean = parm[1][f].th mean;
if (parm[1] [J].th mean < minv.th mean)
minv.th mean =-parm[1][f].th_mean;
1
for (1=offset $1 ; 1<$ row-offset $1 ; 1++$ ) for ( $j=$ offset $1 ; j<$ col-offset $1 ; j++$ )
if (r1[1][1] > back threshold)
switch (parm[1][f].th_gauss)
1
case 0 : switch (parm[1][1].th mean)
case 0 : parm[1][f].label = 1;break; /* flat */ case 1 : parm[1][J].label = 223;break;/* valley
case 2 : parm[1][f]. label = 159;break;/* ridge
\} break;
case 1 : switch(parm[1][1].th_mean)
case
case 0 : parm[1][1].label = 255;break; /* spurio case 1 : parm[1][1]. label = 63;break; /* pit */ case 2 : parm[1][f].label = 95;break; /* peak * ) break;
case 2 : switch(parm[1][f].th_mean)
\{
case 0 : parm[1][1].label = 31;break: /* minima case 1 : parm[1][f].label = 191;break;/* saddle case 2 : parm[1][f].label = 127;break;/* saddle )
if (parm[1][f].label > maxv.label) maxv.label = parm[1][1].label :
if (parm[1][f].label < minv.label)
minv.label = parm[1][J].label;

```
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13
    /* end of label */
else if(strcmp(string,"temp") == 0)
    ( /* put th eindicated parameter in the buffer */
            for(1=0;1<row;1++)
            for ( f=0; f<col; f++
            buffer[1][f] = parm[1][1].gauss:
    }
else if(strcmp(string,"disp") == 0)
    l scanf("%d",&option);
        printf("scale factor >")
        scanf("&f",&sf);
        for(1=0;1<row;1++)
            for( ( }=0;j<col;j++
            { switch(option)
                            witch(option)
                            case 1: line[f] = sf*parm[1][f].fit error;break
                                    case 2: line[f] = sf*parm[1][1].gauss: break
```



```
                                    case 3: line[f] = sf*parm[1][j].mean;
                                    break
                                    case 4: line[f] = sf*parm[1][j].label;
                                    case 5: line[j] = sf*parm[1][1].th gauss: break
                                    case 5: line[j] = sf*parm[1][j].th_gauss; break
                                    case 6: line[j] = sf*parm[1][j].th_mean; break
                                    case 7: line[f] = sf*parm[1][j].q;
                                    case 8: line[f] = sf*parm[1][j].sqrtg;
                                    case 10:linef[f]=sf*rm[1][f].cosph1; break
                                    case 10:l1nef[f] = sf*ri[1][f];
                                    break
                                    break
                                    break
                                    }
            lwr_n(offx,1+offy,&line[0],col):
        }
    }
    else if(strcmp(string,"offset") == 0)
    {
    scanf("&f %f",&offx,&offy)
    }
else if(strcmp(string,"hist") == 0)
    {/* display histogram using qkdraw */
```

spline.c

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scanf("\%d",\&option);
printf("xmax , xmin : "):
switch (option)
1
case 1: printf("qf ; \%fn,maxv.fit_error,minv.fit_error);b
case 2: printf("\%f ; \%f", maxv.gauss,minv.gauss):
case 3: printf("\%f ; \%f", maxv.mean,minv.mean):
case 4: printf("\& ; ; \%n", maxv.label,minv.label):
case 5: printf("\%d; \%d", maxv.th gauss,minv.th bauss) b
case 6: printf("fd ; fdn, maxv.th_mean,minv.th mean):

case 8: printf("\%f; \&fn,maxv.sqrtg, minv.sqrtg);
case 9: printf("\%f ; \%f", maxv.cosphi,minv.cosphi); case 10:printf("\%f ; \%f", maxv.depth,minv.depth);
pr
printf("xmin : xmax :"): scanf("ษ̊f \&f", \&xmin,\&xmax):
printf(" of points desired >"): scanf("\%d",\&npts):
for (1=0;1<npts; $1++$ )
$x[1]=x m i n+((f l o a t)(1) *(x \max -x m i n)) /(f l o a t)$ (npts) $y[1]=0$;
)
for (1=offset1; $1<$ row-offset1;1++)
for ( $j=$ offset 1 ; $f<c o l-o f f s e t 1 ; j++$ )
f
swit
case 1: value = parm[1][f].fit_error;break:
case 2: value = parm[1][1].gauss; break;
case 3: value = parm[1][f].mean; break:
case 4: value = parm[1][f].label; break;
case 5: value = parm[1][f].th_gauss; break;
case 6: value = parm[1][j].th_mean; break;
case 7: value = parm[1][J].q;
break;
case 8: value $=$ parm[1][f].sqrt
break;
case 9: value = parm[1][1].cosph1; break
case 10:value = ri[1][J];
ubs
if ((subs < 1000) \&\& (subs $>=0))^{\text {n }}$ y $[$ subs $]=y[$ subs $]+$
$1=1000$.
$y \min =1000 ;$
$y \max =-1000 ;$

| ```spline.c Ned Mar 30 11:54:33 1988 for(1=0;1<npts;1++) { if(ymin > y[1]) ymin = y[1]; if(ymax < y[1]) ymax = y[1]; } printf("ymin : %f ; ymax : %f\n",ymin,ymax); scanf("&f &f",&ymin,&ymax); 1opt = 0; qterm(4): qkdraw(npts, x, y, lopt, &xmin, &xmax, &ymin, &ymax); qdtitl(" Histogram"); qxlabl("<-- values -->"); qylabl("freq"): qdone(); } else if(strcmp(string,"row") == 0) { /* row histogram display using qkdraw */ scanf("%d", &option); printf("row :"); scanf("%d",&1): while((1 < row) && (1 >= 0)) l ymin = 1000; ymax =-1000; for( }\textrm{f}=0;\textrm{f}<\textrm{col};j++ l switch(option) { case 1: linef[f] = parm[1][f].fit_error;break; case 2: linef[j] = parm[1][j].gauss; break; case 3: linef[j] = parm[1][f].mean; break; case 4: linef[j] = parm[1][f].label; break; case 5: linef[f] = parm[1][f].th_gauss; break; case 6: linef[j] = parm[1][f].th_mean; break; case 7: linef[f] = parm[1][f].q;- break; case 8: linef[f] = parm[1][j].sqrtg; break; case 9: linef[j] = parm[1][j].cosph1; break; case 10:linef[f] = ri[1][f]; break; } if(linef[f] < ymin) ymin = linef[f]; If(linef[j] > ymax) ymax = linef[j]; } xmin = 0.0; xmax = col;``` | ```spline.c Wed Mar 30 11:54:33 1988 16 printf("ymin : %f ; ymax : %f\n",ymin,ymax); scanf("%f &f",&ymin,&ymax); 1opt = 0; npts = col; for(1=0;1<col;1++) x[1] = 1; qterm(4): qkdraw(npts,x,linef,iopt,&xmin, &xmax, &ymin, &ymax); qdtitl(" ROW "): qxlabl("col-->"); qylabl("Values"): qdone(); printf("row :"); scanf("%d",&1); } } else if(strcmp(string,"save") == 0) { /* save in a file */ printf("Save what ? "); scanf("%d", &option); printf("scale factor : "); scanf("%f",&sf); printf("outputfilename :"); scanf("%s",string); fp1 = fopen(string,"w"); pm = pm_alloc(); pm->pm_nrow = row; pm->pm_ncol = col; pm->pm_form = PM_C; pm->pm_image = (\overline{char *) malloc(pm_isize(pm));} pm_point_uchar = (unsigned char *) pm->pm_image; for(1=0;1<row;1++) for( (j=0; j<col; j++) l switch(option) { case 1: *(pm_point_uchar) = (unsigned char) sf* case 2: *(pm_point_uchar) = (unsigned char) sf* case 3: *(pm_point_uchar) = (unsigned char) sf*``` |
| :---: | :---: |

apline.c

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17
case 4: *(pm_point_uchar) = (unsigned char) sf case 5: *(pm point_uchar) = (unsigned char) sf case 6: *(pm_point_uchar) = (unsigned char) sf* case 7: *(pm point uchar) = (unsigned char) sf* case 8: *(pm point uchar) $=$ (unsigned char) sf case 9: *(pm point uchar) = (unsigned char) sf case 10:*(pm point uchar) = (unsigned char) sf f

## pm_point_uchar++;

## write (fp1,pm)

fclose(fpl):
\}
printf("ENTER > ") scanf("\%s",string)
| /* interactive processing loop ends */
/* put the outputs in respective files */
/* printf("\n Putting results in output files $\left.\backslash \mathrm{n}^{\prime \prime}\right)$;

$$
\operatorname{for}(1=1 ; 1<16 ; 1++)
$$

$$
\text { if }((1 \text { ! }=10) \& \&(\text { bool }[1]==1))
$$

if (pm_write (fp[1], pmp[1]) $==$ NULL)
printf(" Can't write output files $\backslash n \times$ ): exit(0):
\}
)
*/
/* End of main program */
float convo(indi,indj,mask,base,size)
int indi,indj:
int mask[5][5];
float base
int size;
1 int 1, j, k, l:
float sum;

```
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    float final;
    sum = 0.0;
    for (k=0,1=1ndi-2;1<=1nd1+2;1++,k++
        for (l=0,j=1ndj-2; j<=1ndj+2;j++,l++
            sum += (float)mask[k][l]*(float)r1[1][f];
    final = sum/base;
    return(final);
}
float convo_l(1ndi,1ndj,mask,base,size)
int indi,indj;
int mask[3][3]
float base;
int size;
{
    int 1,j,k,l;
    float sum;
    float final;
    sum = 0.0;
    for (k=0,1=1nd1-1;1<=1nd1+1;1++,k++)
        for (l=0,j=1ndj-1; j<=1ndj+1;j++,l++
        sum += (float)mask[k][l]*(float)ri[1][f];
    final = sum/base;
    return(final):
}
float abszz(num)
float num;
I
    float b;
    if(num >0) b=num; else b= -num;
        return(b):
    }
init_structure()
    1nt 1,j;
/* initialize the parm structure first */
```


## spline.c <br> Wod Mar 30 11:54:33 1988 <br> 19

bzero(\&parm[0][0],sizeof(1mage)*BUFSI2E*BUFSIZE);
/* initialize maxv and minv structures */
maxv.fit error $=-1000.00$;
maxv.gauss $=-1000.00$
maxv.mean $=-1000.00$;
maxv.q $=-1000.00$;
maxv.sqrtg $=-1000.00$; maxv.cosph1 $=-1000.00$;
maxv.label $=-1000$;
maxv.th gauss $=-1000$
maxv.th-mean $=-1000$;
maxv.depth $=-1000.00$.
minv.fit error $=1000.00$;
minv.gauss $=1000.00$;
minv.mean $=1000.00$
minv.q = 1000.00;
minv.sartg $=1000.00$; minv.cosphi $=1000.00$ minv. label $=1000$
minv.th gauss $=1000$; minv th mean $=1000$; minv. depeth $=1000.00$;
) /* end of init */
define RTD (180.0/3.14159)
define MAXGRAY 255
define WINDOW 5
define LFLOAT 4
define NARGS 2
\#define MAXORDER 3
define NSYS 10
/* maximum \# of coefficients possible */
include <stdio.h>
include <math.h>
include <strings.h>
Include <local/pm.h>
include "/usr/users/alok/advanced/spline_include.h"
char *malloc(), *strcpy() :
int getline(), body() ;
ouble sqrt(), atan2() ;
double lsqerr1():
double a1[200][NSYS],a2[200][NSYS],a3[300][NSYS]
/* matrices to store $A$ of $A x=b$; for $1 s t, 2 n d$ and 3 rd order fitt
extern double 1x, 1y, 1xx, 1yy, 1xy, 1yx, 1xxx, 1yyy, ixyy,
xtern double a30, a21, a12, a03, a20, a11, a02, a10, a01,
extern double fit_error;
extern int s_order:
xtern int csize, rsize
extern int offset1;
extern int approx
extern float ri[BUFSIZE][BUFSIZE]
xtern File *dumpfile:
extern int todump;
read_specs (
1
pmpic *area, *(newarea[4])
FILE *fdin, *fdout[3] ;
char buffer[BUFSIZE],
nargs,
*cmt, *cmd,
ptr, window ;
int $1, \mathrm{~J}$.
ncol, nrow,
count, nfiles, nf,
grad, mingrad, maxgrad ;

## grad.c Wod Mar 30 12:46:58 $1988 \quad 2$

float *fptr:
double $e$, theta, scalel;
printf("intensity is approximated locally by a polynomial.\n") printf("type 1 for bi-linear approximation. ${ }^{\text {n" }}$ ) ; printf("
for quadric
(n")
printf(" 3 for bi-cubic (n")
printf(" 4 for b-spline $\left.\backslash n^{\prime \prime}\right)$ :
printf(" 5 for smooth-b-spline\n");
printf("value = ") ;
scanf("\%d", \&approx)
if ((approx<1) || (approx>5))
fprintf(stderr,"value must be $1,2,3,4$ or $5 . \ n$ ")
exit (0) ;
\}
getchar();
printf("window size = ") ;
if (getline (buffer) ==0)
if (approx==1) window $=2$;
else if (approx==2) window $=3$
else if (approx==3) window $=5$
fprintf(stderr,"window size = \%d\n", window) ;
\}
else
window = atoi (buffer) ;
if ((window/2*2==window) \&\& (approx!=1)) fprintf(stderr,"window size must be an odd number. $\mathrm{In}^{\prime \prime}$ ) : exit(0) :
if $($ (approx==1) \&\& (window<2)) fprintf(stderr,"window must be $>=2 . \ln ")$ exit (0) :
1f ${ }^{\}}$
f(approx==2) \&\& (window<3)) fprintf(stderr,"window must be >= $3 . \backslash n "$ ) exit (0) :
$1{ }^{\text {f }}$
1f ((approx==3) \&\& (window<5)) fprintf(stderr, "window must be >= 5.\n") : exit (0) :
$)^{\}}$
offsetl $=$ window/2


$$
\begin{aligned}
& \begin{array}{l}
\mathrm{s} 220=\mathrm{s} 220+\mathrm{x} 2{ }^{*} \mathrm{y} 2 ; \\
\mathrm{s} 600=\mathrm{s} 600+\mathrm{x} 4 \text { * } \mathrm{x} 2 ;
\end{array} \\
& \begin{array}{l}
s 600=s 600+x 4 \star x^{2} ; \\
s 420=s 420+x 4 * y^{2} ;
\end{array} \\
& \text { s001 = s001 + z ; } \\
& \text { s011 }=\operatorname{sol1}+\mathrm{y} \text { * } \mathrm{z} \\
& \mathrm{~s} 021=\mathrm{s} 021+\mathrm{y} 2 \text { * } \mathrm{z} \text {; } \\
& \text { s031 }=5031+y^{2} * y * z \text {; } \\
& \text { s101 }=\text { s101 }+x \text { * } z \text {; } \\
& \text { s111 }=\operatorname{sil1}+x{ }^{*} y^{*} z \text {; } \\
& s 121=s 121+x * y 2{ }^{*} z \text {; } \\
& \text { s201 = s201 + x2 * z ; } \\
& \begin{array}{l}
\mathrm{s} 211=\mathrm{s} 211+\mathrm{x} 2 \star \mathrm{y} \text { * } \mathrm{z} \text {; } \\
\mathrm{s} 301=\mathrm{s} 301+\mathrm{x} 2 \star \mathrm{x} \text { * } \mathrm{z} ;
\end{array} \\
& \text { ) }
\end{aligned}
$$

if (approx==1) (
coeffl[0] = al0 = s101 / s200 :
coeffl[1] $=\mathbf{a 0 1}=$ s011 / s200
coeff1[2] $=\mathbf{a 0 0}=\mathbf{s 0 0 1} / \mathbf{s 0 0 0}$
$a 30=a 21=a 12=a 03=0.0$
$a 20=a 11=a 02=0.0$
fit_error $=$ error1 = lsqerrl(al,coeff1,b_mat,k,3, \&p_averb); /* c
f
else if (approx==2)
coeff2[3] = al0 = s101/s200 ;
coeff2[4] = a01 = s011 / s200
coeff2[5] = a00 = s001 / s000
$\operatorname{det} 2=\operatorname{determ}(\mathrm{s} 400, \mathrm{~s} 220, \mathrm{~s} 200, \mathrm{~s} 400, \mathrm{~s} 200, \mathrm{~s} 000)$;
inverse(s400, s220, s200, s400, s200, s000, det2, inv2 ) ;
coeff2[0] = a20 = inv2[1] * s201 + inv2[2] * s021 + inv2[3] * s0 coeff2[1] = a02 = inv2[4] * s201 + inv2[5] * s021 + inv2[6] * s0 coeff2[5] = a00 = inv2[7] * s201 + inv2[8] * s021 + inv2[9] * s0 $\mathrm{a} 30=\mathrm{a} 21=\mathrm{a} 12=\mathrm{a} 03=0.0$;
coeff2[2] $=$ all = s111 / s220
fit_error $=$ error2 $=$ lsqerr1 (a2, coeff2,b_mat,k,6,\&p_averb);
)
else if (approx==3) \{
$\operatorname{det} 1=\operatorname{determ}(\mathrm{s} 600, \mathrm{~s} 420, ~ s 400, ~ s 420, ~ s 220, ~ s 200)$;
det2 $=$ determ( s400, s220, s200, s400, s200, s000) ;
inverse(s600, s420, s400, s420, s220, s200, det1, inv1 ) : inverse(s400, s220, s200, s400, s200, s000, det2, inv2 ) ;
coeff $3[1]=a 03=\operatorname{lnv1}[1] * \operatorname{s031}+\operatorname{inv1[2]} * \operatorname{s211}+\operatorname{inv1[3]}$ * s0 coeff $3[2]=\mathrm{a} 21=$ invl[4] * s031 + inv1[5] * s211 + inv1[6] * s0
$\operatorname{coeff} 3[8]=a 01=$ inv1[7] *s031 + inv1[8] * s211 + inv1[9] * s0
grad.c Wod Mar 30 12: 16:58 1988
coeff $3[0]=a 30=1 n v 1[1] * s 301+\operatorname{lnv} 1[2] *$ s121 + inv1[3] * s1 coeff $3[3]=a 12=$ inv1[4] * s301 + inv1[5] * s121 + inv1[6] * s coeff $3[7]=$ al0 $=$ inv1[7] * s301 + inv1[8] * s121 + inv1[9] * s1

 coeff3[9] $=\mathrm{a} 00=$ inv2[7] * s201 + inv2[8] * s021 + inv2[9] * s0
coeff $3[6]=$ all $=$ s111 / s220 ;
f1t_error $=$ error3 = lsqerr1 (a3, coeff3,b_mat,k,10, \&p_averb);
\}
/* the program computes the least square fitting error by calling $t$ sqerri() routine in the source file solver.c. the calling parame a: the $n^{\star}$ m matrix.
$x: m^{\star} 1$ matrix having the values of coefficients.
$b:$ observed values at the $n$ points.
p_averb : value returned = average value at $n$ points.
integers.
called as : double lsqerrl(a, $x, b, n, m, p$ averb) double a[][nsys], x[],b[],*p_averb; int m, n ;
returns lsq error (double). */
/* $1 f($ (errorl <= error2) \&\& (error1 <= error3))
1
s_order = 1;
fit_error = error1;
$\mathrm{a} 01=\operatorname{coeff1[1];}$
al0 $=\operatorname{coeff1[0];~}$
a00 $=\operatorname{coeffl[2];~}$
$a 30=a 21=a 12=a 03=0.0$;
$a 20=a 11=a 02=0.0$;
)
else if(error2 <= error3)
1
s_order $=2$;
fīt_error = error2;
$\mathrm{a} 00=\operatorname{coeff} 2[5]$;
a01 = coeff2[4];
a10 $=$ coeff2[3];
al1 $=$ coeff2[2];
$\mathrm{a} 02=\operatorname{coeff} 2[1]$;
a20 $=$ coeff2[0];

```
grad.c Wod Mar 30 12:46:58 1988 7
    else
        s_order = 3;
            fit_error = error3;
        }
    */
```

    1f(todump \(==1\) ) printf(dumpfile, "x \(=\) id \(y=\) of error \(=\) of \(\backslash n "\), row, \(c\)
    /* if (todump $==1$ ) printf(dumpfile," $x=$ \%d $y=$ ord errorl $=$ \%f error2
$1 \mathrm{x}=\mathrm{a} 10$;
$1 \mathrm{y}=\mathrm{a} 01$ :
$1 \mathrm{xx}=2.0$ * a20 :
$1 y y=2.0$ * a02
$1 \times y=a 11$;
$1 x y=a 11$
$1 y x=a 11$
${ }_{1 \times x x}=6.0^{\circ}$ * a 30 ;
1yyy $=6.0 \star$ a03:
IXYY $=6.0 \star$ a03
1XYY $=2.0 \star \mathrm{a} 12$ :
1xyy $=2.0 \star$ a12 $;$
ixxy $=2.0 \star$ a21 $;$

return(0):
) /* Main */
/*
compute determinant of symmetric $3 \times 3$ matrix.
double determ( b11, b12, b13, b22, b23, b33 ) double b11, b12, b13, b22, b23, b33;
f
double templ, temp2 :
temp1 = b11 * b22 * b33 + b12 * b23 * b13 + b12 * b23 * b13 ;
temp2 = b13 * b22 * b13 + b12 * b12 * b33 + b23 * b23 * b11 ;
return ( temp1 - temp2) :
\}
/* get inverse of symmetric matrix. */
inverse( b11, b12, b13, b22, b23, b33, det, inv ) double b11, b12, b13, b22, b23, b33;
double det
double inv[] :

```
inv[0] \(=0.0\) :
inv[1] \(=(\mathrm{b} 22 \times \mathrm{b} 33-\mathrm{b} 23 * \mathrm{~b} 23) /\) det ;
inv[2] \(=-(\mathrm{b} 12 \star\) b33 - b13 * b23) / det :
inv[3] \(=\) (b12 * b23-b22 * b13) / det :
inv[4] \(=\) inv[2]:
inv[5] \(=\) (b11* b33-b13* b13) / det ;
\(\operatorname{lnv}[6]=-(b 11 \star\) b23-b12 * b13) / det ;
\(\operatorname{lnv}[7]=-\operatorname{inv}[3]\);
inv \(\operatorname{lnv}[8]=\) inv \([6]\)
inv[9] \(=\) (b11 * b22-b12 * b12) / det ;
return(0) :
\({ }_{\}}^{\text {ret }}\)
```

int body (size, x)
int size, $x$;
if $(x<0)$ return $(0)$ :
else if ( $x>=s i z e$ )
return(size-1) ;
else
return(x) ;
\}
getline(s)
char $s[]$;
cha
int c, 1 :
$1=0$;
while( (c=getchar()) != $\backslash \mathrm{n}^{\prime}$ \&\& $\mathrm{c}!={ }^{\prime} \backslash 0^{\prime}$ )
$\mathrm{s}[1++]=\mathrm{c}$;
$s[1]=10^{\circ}$;
return(1) :
\}
double b11, b12, b13, b22, b23, b33

## grad.c Wed Mar 30 12:46:58 $1988 \quad 9$

double lsqerrl(A, $x, b, m, n, p$ averb)
double $A[][N S Y S], x[], b[]$,
*p_averb ;
int $\quad \mathrm{m}, \mathrm{n}$;
int
double sum, errsum,bsum ;
errsum=bsum=0.0;
for $(1=0 ; 1<m ; 1++) \quad\{$
sum $=0.0$;
for ( $j=0 ; j<n ; j++$ )
sum $=\operatorname{sum}+A[1][j] * x[j]$;
sum $=$ sum - b[1].
errsum $=$ errsum + fabs (sum)
bsum = bsum + fabs (b[1]) ;
*p_averb = bsum / (double) m ; return(errsum);
\} /* lsqerr */

## merge.c Hed Apr 20 21:59:01 1988 1

P* Program for further segmentation of the scene obtained after label using the sign of mean and gaussian curvatures.

The program calls routines in solver.c to fit the surface on specif data.

March 30, 1988
April 4, 1988 : Initialize neighbour structure. Regions smaller th size 6 are merged with the best fitting neighbouri size 6 are merged with the best fiting neighbour in the form of labeled convex/concave subparts of scene seperated by convex/concave/jump edges.
*/
\#include <stdio.h>
\#include <math.h>
\#include <local/pm.h>
\#include <ik.h>
\#include </usr/users/alok/advanced/spline include.h>
\#define REGION SIZE 1000
\#define REGION_SIZE 100
\#define NSYS 20
\#define MIN_REGION_SIZE 50
\#define PIXEL_STACK_SIZE 8000
\#define MAX_NUM CALL 400
\#define ACCEPT ERROR 2.00
\#define THRESH ERROR 0.20
\#define flat 1
\#define PEAK 95
\#define RIDGE 159
\#define MIN 31
\#define SADRID 127
/* so many regions in the */
/* so many points in a region */
/* so many variables at the most *
/* so many pixels in an acceptable seed region */

* size of pixel-stack used for rec -iterative region growing */
* Maximum \# of pending recursive c at a given time. */
/* value of acceptable error */
/* threshold error that is accepta while region growing */
/* spherical */
/* spherical */
/* minimal */
/* Saddle ridge */
float buffer[BUFSIZE][BUFSIZE]: int label[BUFSIZE][BUFSIZE]; short int lap[BUFSIZE][BUFSIZE short int lap [BUFSIZE][BUFSIZE]; char attempt[BUFSIZE][BUFSIZE];
/* stores input image */
/* stores labeled 1mage */ /* stores laplacian image * /* reconstructed image */ /* used to keep track of pixels region grow() */
marge. 0
Wed Apr 20 21:59:01 1988
double lsqerr():
int congrowlab;
ikword ikvalue = 255;
int row, col;
nt row, col.
int seedrow, seedcol
nt seedrow, seedcol
nt num points
double avect [NPOINTS] [NSYS]:
ouble bvect [NPOINTS]
double result[NSYS]:
int surf type;
int surf_type;
int cmin,
int $t r$;
nt edge threshold
int to_merge;
struct region_type \{
int valid;
int label;
int order:
double fit error
int surf_type:
int size;
int done;
int center[2];
float coeffs[6].
int neighbour[500]
int num neigh:
int rmin, rmax, cmin, cmax
int dmax;
regions[REGION_SIZE], reg;
Int dmax;
int con_label[REGION_SIZE];
int colōr[REGION SIZE];
int convex label;
float error_in_fit();
int to disp;
char ikonas_disp[2];
/* for stack management during final region growing */
struct p_stack \{
/* stores the open pixels in circular
int row;
int col;
\} pixel_stack[PIXEL_STACK_SIZE];

| int rownum; | /*(rownum, colnum) is the next pixel * |
| :---: | :---: |
| int colnum; | /* popped from the stack */ |
| int num_call; | /* number of calls pending at a given |
| int stack_length; | /* =current_element if queue is empty points to the tall of the queue */ |
| int current_element; | /* points to head of the queue */ |

main (argc,argv)
int argc;
char *argv[]:

```
int 1,j,k,l,m,n
/* local variables */
```

pmpic *pm1,*pm2,*pm3; /* pmpic pointers to range and labeled im double scale, zscale;
/* scale of the input image */ int image_format;
/* stores input image format */
unsigned char *uchar_point;
float *float point; /* original image is scaled and smoothed short int *short_point;
FILE *imagefile, *labelfile;
FILE *loc_file,*rec_file,*outputfile,*outfile,*lab_file,*lapfile,*c int want;
double $x, y$, averb;
int valid $=0$, invalid $=0$;
int imageformat;
int region_num;
int pixel_val;
float fit_error,min_error;
int found;
int lab;
int surface_fit; $/ *=1$; if surface fit is to be done
if(argc $!=4$ )
fprintf(stderr,"Usage : merge \{scaled \& smoothed input PM_F ima exit (0);
\}
printf("Want to display region growing on IKONAS. y if yes > "); scanf("8s",\&1konas_disp[0]);

## arge.

printf("Edge_threshold : ") scanf("\%d", \&edge_threshold);

```
1f((1magefile = fopen(argv[1],"r")) == NULL)
```

    1
        fprintf(stderr,"Cannot open input range image file : \%s 1 n ", arg
        exit(0):
    \}
    if((labelfile $=$ fopen(argv[2],"r")) $==$ NULL $)$
1
fprintf(stderr, "Cannot open image label file : is $\ln ", \operatorname{argv}[2])$;
exit(0);
\}
1f((lapfile $=$ fopen(argv[3],"r")) == NULL)
1f
fprintf(stderr,"Cannot open laplacian operated file : \%s $\ln ", a r$
exit (0):
\}
outputfile $=$ fopen("log", "w");
1f $($ (pml $=$ pm_read $(1$ magefile, 0$))==$ NULL $)$
,
fprintf(stderr, "Error in reading PM-format range image file: \%
exit (0):
\}
if $((\mathrm{pm} 2=$ pm_read(labelfile, 0$))==$ NULL $)$
1
fprintf(stderr,"Error in reading PM-format range image file : \%
exit(0):
\}
if $($ (pm3 $=$ pm_read(lapfile, 0$))==$ NULL $)$
if
fprintf(stderr, "Error in reading PM-format range image file : \%
exit(0);
\}
row $=$ pml->pm nrow
col = pml->pm_ncol:

1f((row != pm2->pm_nrow) || (col != pm2->pm_ncol) || (row != pm3->p || (col != pm3->pm_ncol))

```
marge.c Wad Apr 20 21:59:01 1988 5
    exit(0);
    }
    1f(pm2->pm_form != PM_C)
        { fprintf(stderr,"Label image is not in PM_C format \n");
        exit(0):
    }
if(pml->pm_form == PM_F){
    float_point = (floa\overline{t *) pml->pm_image:}
    1mageformat =PM_F;
    scale = 74.20; /* for RCA images
                            */
    zscale = 1.00
    for(1=0;1<row;1++)
        for(j=0;j<col;j++
            (buffer[1][j] = *(float_point):float_point++;)
}
else
    If(pml->pm_form == PM_S){
        1mageformat = PM_S;
        short_point = (short int *) pml->pm_1mage;
        scale = 74.20; /* for RCA 1mages *
        zscale = 1.00; /* for RCA images *
        for(1=0;1<row;1++)
            for(j=0;j<col:j++
            {buffer[1][j] = ((float)*(short_point))*zscale/scale;short_
    }
    else
        1f(pm1->pm_form == PM_C) {
            imageformat = PM_C;
            uchar_point = (unsigned char *) pml->pm_image;
            zscale = 1.5; /* for Gus' Images */
            scale = 1.978; /* for Gus' 1mages */
            for(1=0;1<row;1++
            for (j=0; j<col; j++
            {buffer[1][j] = ((float)*(uchar_point))*zscale/scale;uchar
        }
            else {
            printf("unrecognized format in the input image \n");
            exit(0):
        }
    uchar_point = (unsigned char *) pm2->pm_image
    for(1=0;1<row; 1++)
    for(j=0;j<col;j++
        label[1][j]= = - (*(uchar_point++));
    short_point = (short int *) pm3->pm_image:
short_point \(=\) (short int *) pm3->pm_image
```

```
merge.c Wod Apr 20 21:59:01 1988 6
    for(1=0;1<row;1++)
    for( }j=0;j<col;j++
        lap[1][J] = *(short_point++);
If(pml->pm_form != PM_F)
    1f
        printf("want to smooth ? (1/0) >"):
        scanf("%d",&want);
        if(want == 1) gaussian(buffer,row,col);
    }
```

zero ((char *) \&rec image [0][0], sizeof(short int) *BUFSIZE*BUFSIZE); bzero ((char *)\&regions[0],sizeof(reg)*REGION_SIZE);
/* initialize the structure ;
fit a second order polynimial on every patch */
fprintf(stderr,"starting surface fitting on individual regions $\backslash n "$ ) seedr $=0$;
seedc
seedc $=0$;
region_label $=0$;
surface_fit =1; /* no surface fitting to be done */
ikvalue $=100$;
to_disp $=0$;
while (next_seed () $==0$ ) /* while there are seed regions */
if
surf_type = label[seedrow][seedcol]: /* label of the region */ region-label++;
num_points $=0$
$\mathrm{cmin}=1000$;
rmin $=1000$;
$c_{\max }=-1000$
$\begin{aligned} & \operatorname{rmax}=-1000 \\ & d m a x \\ &=-1000\end{aligned}$
$\operatorname{dmax}=-1000$
label[seedrow] [seedcol] = region_label;
grow_seed(seedrow, seedcol, surf_type, region_label);
if (surface_fit $==1$ )
if (num_points < MIN_REGION_SIZE) /* region is invalid if \# of 1
regions[region_label].valid $=0$;
invalid++;
$\stackrel{\}}{\text { else }}$
else

## morgo.c <br> Wod Apr 20 21:59:01 1988 7

if $(10$ - surf_type $)==$ FLAT $)$
1
lsquare (avect, bvect, result, num_points, 3)
regions[region_label).fit_error = lsqerr(avect, result,bve ?
else else

Isquare (avect, bvect, result, num_points, 6) ;
regions[region_label].fit_error = lsqerr (avect, result, bve $\stackrel{\text { \} }}{\text { regi }}$
regions[region_label].coeffs[0] = result[0];/*a00,a01,a10, a02 regions[region_label].coeffs[1] = result[1];
regions[region_label].coeffs[2] = result[2]; if ( $(0$ - surf_type) != FLAT)
1
regions[region_label].coeffs[3] = result[3];
regions[region_label].coeffs[4] = result[4];
regions[region_label].coeffs[5] = result[5];
regions[region_label].order $=2$
$\stackrel{\}}{\text { else }}$
else
regions[region_label].order = 1
\}
regions[region_label].valid = 1;
\}
regions[region_label].label = region_label;
regions[region_label].size = num_points;
regions[region_label].surf_type $=$-surf_type;
regions[region_label].center[0] $=$ seedrow;
regions[region_label].center[1] $=$ seedcol
regions[region_label]. num_neigh $=0$;
regions[region_label]. $\mathrm{cmin}=\mathrm{cmin}$;
regions[region_label].cmax $=$ cmax;
regions[region_label].rmin $=r m i n$;
regions[region_label]. rmax $=$ rmax;
regions [region_label]. done $=0$;
regions[region_label].dmax = dmax;
con_label[region_label] $=0$; /* not assigned to any convex re
fprintf(outputfile," region : \%d points : \%d surf_type \%d (row,col)= if (num_points >= MIN_REGION_SIZE) fprintf(stderr," region : od

## Wad Apr 20 21:59:01 1988

\} /* while loop to fit surfaces */
printf("\# of regions found : \%d invalid : \%d valid $\% d$ n", region_lab
fclose (outputfile); fclose (outputfile);
$\mathrm{tr}=$ region_label; /* total \# of regions */
fprintf(stderr, "Marking neighbours $\backslash n$ ");
/* initialize neighbours */
for ( $1=0 ; 1<$ row $; 1++$ )
for ( $j=0 ; j<\operatorname{col} ; j++$ )
if (label[1][f]>0)
i
if (label[i][f] ! = label[1+1][f])
$i$
make_neighbour (label[1][f], label[1+1][f]);
make_neighbour(label[1+1][j], label[1][f]);
if ?
if (label[1][f] != label[1][f+1])
i
make_neighbour (label[1][j], label[1][j+1]); make_neighbour(label[i][j+1], label[1][f]);
\}
\}
/*------------------ SURFACE FITTING / REGION GROWING */
f(surface_fit == 1)
(/* TO BE DONE ONLY IF SURFACE FITTING BASED REGION GROWING DESIRED fprintf(stderr, "Merging invalid regions into nearest best fitting r /* merge invalid regions with neighbouring valid region with least to_merge $=0$;
if (to_merge $==1$ )
for ( $1=1 ; 1<=t r ; 1++$ )
( /* for all regions do */
if (regions 11 . valid $==0$ )
( /* if the region is invalid then do */
/* returns the list of neighbours */
min_error $=1000.00$
found $=0$;
for ( $j=0 ; j<r e g i o n s[1] . n u m \quad n e i g h ; j++$
( /* find error in fitting the polynomial of neigh[j] on */
if(regions[regions[1].neighbour[f]].valid == 1)
f found $=1$;
fit_error = error_in_fit(regions[i].label, regions[i if (min_error > fit_error) \{

```
marge.c
    Wod Apr 20 21:59:01 1988 S
                        region_num = regions[i].neighbour[f];
                    min_error = fit_error;
                    }
        if(found == 1)
            {/* a suitable region has been found
                merge first region into second one */
                merge_region(regions[i].label,region_num):
            )
        else {/* no suitable valid region was found in the neighbo
                    Do nothing for the moment.*/
                fprintf(stderr,"No suitable valid region found in the nei
        }
        }
    } /* end of merging invalid regions */
    /* Main Segmentation loop starts. Bigger seed regions are merged wi
        selected set of points if the RMS error is acceptable, thatis lo
        she acceptable error of the whole region. */
/*
    fprintf(stderr,"Main segmentation starts \n");
    curr = 1;
    if(strcmp("Y",1konas_disp) == 0)
        { if(1kopen (NULL) == -1)
            fprintf(stderr,"Can't open IKONAS \n");
            to_disp = 1;
        }
    else to_disp = 0
    1kvalue = 200;
    while((lab = pick_up_seed()) != -1) */ /* returns the next seed r
/* while
    fprintf(stderr,"Growing for od :\n",lab);
        /* curr++;*
            num_call = 1;
        grow_region(regions[lab].center[0],regions[lab].center[1],lab);
        while(stack_length != current_element)
            |
                /* returns the pixel */
/*
                    get_pixel();*/ /* returns the pixel */
                    grow_region(rownum, colnum, lab)
            }
    * }
*/
```


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\} /* END OF THE REGION TO BE DONE ONLY IF SURFACE FITTING BASED REGIO GROWING IS DESIRED *

/* output the reconstructed and error image */
for ( $1=1 ; 1<=($ row -1$) ; 1++$ )
for $(j=1 ; j<=(\operatorname{col}-1) ; j++$
if(label[i] [ f$] \gg 0$
\{ /* for all pixels do */
/* compute local error at the pixel */
$x=1$ - regions [label[1][j]]. center[0];
$y=j-r e g i o n s[l a b e l[1][j]] . c e n t e r[1] ~$
$y=j$ - regions[label[1][f]].center[1];
rec_image [1][f] = $\mathbf{x}^{\star} \mathbf{x}^{\star}$ regions [label[1][f]].coeffs [5] +
$x^{\star} y^{\star}$ regions[label[1][j]].coeffs [4] +
$y^{\star} y^{\star}$ regions [label[1][1]].coeffs[3] +
$x^{\star}$ regions [label[1][j]].coeffs[2] +
$Y^{\star}$ regions[label[1][1]].coeffs[1] +
1*regions[label[1][j]].coeffs[0];
\}
/* Initialize pm buffers */
printf("Saving images. Stored as PM_S images $\ \mathrm{n}$ ");
rec_file = fopen("recimage","w");
oc_file $=$ fopen ("errorimage","w")
lab_file $=$ fopen("labelimage", "w");
$\mathrm{pml}=\mathrm{pm}$ alloc() ;
pml->pm_nrow = row
pml->pm_ncol = col;
ml->pm_form $=$ PM S
pml->pm_image $=(\overline{c h a r} *)$ malloc (pm_isize(pm1));
bzero(pm1->pm_image,pm_isize(pml));
short_point $=($ short int $*)$ pml->pm_1mage;
for ( $1=0 ; 1<$ row $; 1++$ )
for ( $j=0 ; j<\mathrm{col} ; j++$ )
\{ *short point $=$ (short int) fabs((double) (rec_image[1][f]*scale short point++;
)
pmwrite (rec file, pml)
fclose(rec_file)

## marga.c <br> Hed Apr 20 21:59:01 1988

bzero(pm1->pm_image,pm_1size(pm1)):
short_point $=$ - (short int $*)$ pml->pm_image
for ( $1=0 ; 1<$ row ; $1++$ )
for ( $\mathrm{f}=0 ; \mathrm{j}<\mathrm{col} ; \mathrm{j}++$
\{ *short_point $=$ (short int ) fabs((double) (rec_image[1][j] short_point++;
\}
pm_write(loc_file,pml) ;
fclose (loc_fyle):
/* output new label image */
$\mathrm{pml}->\mathrm{pm}$ form $=$ PM_C
pml $\rightarrow$ pm_image $=$ (char $*)$ malloc(pm_isize (pm1));
bzero(pm1->pm_image,pm_isize(pmi))
uchar_point = (unsigned char ${ }^{\star}$ ) pml->pm_image
for ( $1=0 ; 1<$ row ; $1++$ )
for ( $j=0 ; j<\operatorname{col} ; j++$ )
( *uchar_point = regions[label[i][f]].surf_type; uchar_point++
frit
pm_write (lab_file, pmi);
fclose(lab_file);

1f(1mageformat != PM_F)
1
pm1 = pm_alloc () ;
pm1->pm_nrow = row;
$\mathrm{pml} \rightarrow \mathrm{pm}$ nncol $=\mathrm{col}$;
$\mathrm{pml} \rightarrow \mathrm{pm}$ form $=\mathrm{PM} \mathrm{F}$
pm1 $\rightarrow$ pm_1mage $=(\overline{\text { char }}$ *) malloc (pm_isize (pml)):
bzero(pml->pm_image,pm_1size(pml)) ;
float point $=-\left(\right.$ float $\left.{ }^{\dagger}\right)$ pml->pm_image:
for ( $1=0 ; 1$ <row; $1++$ )
for ( $j=0 ; j<c o l ; j++$
1
*float_point $=$ buffer[1][f];
float_point++;
\}
outfile $=$ fopen("1mage","w")
pm_write (outfile,pm1)
fclose (outfile):
\}
*---------------- CONVEX PATCHES GROWING $\qquad$

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/* mark all the regions as undone */
for ( $1=1 ; 1<=t r ; 1++$ )

* label convex subparts in the image */ fprintf(stderr, "Label convex subparts in the image $\backslash n "$ );
convex label $=0$
convex labe $=0$
while((lab = get_next_region()) != -1)
f
pixel_val = pixel_val +2
convex_label++;
color[convex_label] = pixel_val;
regions [lab].done $=1$;
congrowlab = lab,
extend_region(lab) ;
\}
/* output the convex/concave image */ con_file $=$ fopen ("confile", "w");
fprintf(stderr,"\# of convex patches found : \%d", convex_label);
$\mathrm{pm1}=\mathrm{pm}$ alloc():
ml->pm_nrow = row
$\mathrm{pm1} 1>\mathrm{pm}$ _ncol $=\mathrm{col}$;
pml->pm form $=$ PM C
pm1->pm_1mage $=(\overline{c h a r} *)$ malloc (pm_isize(pm1));
bzero(pml->pm_1mage,pm_isize(pm1));
uchar_point $=$ (unsigned char *) pml $\rightarrow$ pm_image; con_label $[0]=0$
for ( $1=0 ; 1<$ row $; 1++$ )
for $(j=0 ; j<\mathrm{col} ; j++)$
1
if (con_label[label[1][j]] $!=0$
*uchar point = color[con_label[label[1][f]]];
if()con_label[label[1][j]] != con_label[label[i][j+1]]) ||
(con_label[label[1][j]] != con_label[label[1+1][j]]))
*uchar_point $=255$;
uchar_point++;
\}
pm_write (con_file, pml)
flose (con_fīle):


## ange.c

```
}/* end of main */
```


## /********************************NEXT_SEED*********************************)

## next_seed(

1
int 1,j,k,l;
while((buffer[seedr][seedc] == 0) || (label[seedr][seedc] >= 0)) \{ seedc++;
if (seedc $==($ col-1))

$$
\text { seedc }=0 \text {; }
$$

$$
\text { if }(\text { seedr }==(\text { row-1) }) \text { return }(-1) \text {; }
$$

\}
seedrow = seedr;
seedcol = seedc;
return(0):
\}
/********************************GROW_SEED**********************************)
grow seed(srow, scol, stype, slabel)
int srow;
int scol;
int stype;
int slabel
1
int 1,j;
double $x, y$ :
if(to_disp == 1) lwr(scol,srow,\&ikvalue);
If(rmin > srow) rmin = srow
if(rmax < srow) rmax = srow
if(cmin > scool) cmin = scol
if(cmax < scol) cmax = scol;
if(dmax < buffer[srow][scol]) dmax = buffer[srow][scol]:
num_points++;
$\mathrm{x}=$ srow - seedrow
$y=$ scol - seedcol;
avect[num_points][5] $=x^{\star} x$;

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avect [num_points] [4] $=x^{\star} y$;
avect[num_points] [3] $=\mathrm{y}^{\star} \mathrm{y}$;
avect[num_points][2] $=x$;
avect [num_points] [1] = y ;
avect [num_points] $[0]=1$;
bvect[num points] = buffer[srow][scol]

```
for ( \(1=\) srow- \(1 ; 1<=\) srow \(+1 ; 1++\) )
            for ( \(j=s \mathrm{col}-1 ; j<=\mathrm{scol}+1 ; j++\) )
            if \(((1>0) \& \&(1<(\) row -1\()) \& \&(j>0) \& \&(j<(\operatorname{col}-1))\)
            if (label[1][j] == stype)
                1
                    label[1][f] = slabel;
                        grow_seed(1,f,stype,slabel);
                \}
```

\} /* grow_seed */
************************** MAKE_NEIGHBOUR ***********************************)
Makes label2 neighbour of labell.
*/
make_neighbour (labell, label2)
int labell;
int label2;
1
int 1,j,k,l;
int done;
done $=0$;
$\mathrm{k}=0$;
if(regions [labell]. num_neigh $==0$ )
1
regions [labell]. num_neigh = 1; regions[label1].neighbour[0] $=$ label2; $\stackrel{\text { f }}{\stackrel{r}{2}}$
while (done $==0$ )
whil
if ( $k==$ regions[labeli].num_neigh)
(/* neighbour not yet marked in the structure */
regions[label1].num_neigh = regions[labell].num_neigh + 1; regions[labell].neighbour [k] = label2;
done = 1;
\}
else
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if(regions[labell].neighbour $[k]==$ label2)
\{/* nothing to be done */
done $=1 ;$
\}
else $k++$;
/*
\} /* end of make neighbour */
 computes fit_error of fitting labell points using label2 coefficien */
float error_in_fit(labell, label2)
int labell:
int label2;
1
int 1,j,k,1;
double error;
double averb;
float $x, y$;
$\mathrm{k}=0$;
for ( $1=$ regions[label1].rmin; $1<=$ regions [labell].rmax; $1++$ ) for ( $j=r e g i o n s[l a b e l 1] . c m i n ; j<=r e g i o n s[l a b e l 1] . c m a x ; j++)$

If (label $[1][\mathrm{j}]==$ labeli)
l
$x=1$ - regions[label2]. center[0];
$y=j-$ regions [label2].center[1];
avect $[k][5]=x^{*} x$;
avect $[k][4]=x^{*} y$;
avect[k][3] = $y^{*} y$;
avect [k][2] $=\mathbf{x}$;
$\begin{aligned} \text { avect }[k][1] & =y ; \\ \text { avect }[k][0] & =1 ;\end{aligned}$
bvect[k] = buffer[1][f];
k++;
\}
result[0] $=$ regions[label2].coeffs[0];
result[1] = regions[label2].coeffs[1];
result [2] $=$ regions[label2].coeffs[2];
result[3] = regions[label2].coeffs[3];
result [4] $=$ regions[label2].coeffs[4];
result[5] = regions[label2].coeffs[5];
error = lsqerr(avect, result, bvect, $k, 6$, \&averb);
return((float)error);
\} /* error in fit */

```
/********************************* MERGE REGION ***************************
```

merge first region into the second one.
*/
merge_region(label1, label2)
int labell
int label2
i
int 1,j,k,1;
$k=0$; $/ *$ counting \# of points */
for ( $1=$ regions [labeli]. rmin; $1<=$ regions[label1]. rmax; $1++$ )
for ( $j=$ regions [labeli].cmin; $j<=$ regions[labeli].cmax; $j++$ )
if (label[1][f] == label1)
if
label[1][f] = label2;
k++;
${ }^{1}$
regions[label2].size $=$ regions[label2].size $+k$;
\} /* end of merging regions. that was easy */

##  criterion is met and the child is of right type <br> cr $\mathrm{*} /$

erode (parent, child)
int parent:
int child;
1
int 1,j,k,l;
int lab;
curr++;
switch(regions[parent].surf_type)
i
case 1 : /* parent region is flat. */
1
/* for all types of neighbours grow the region */ attempt [regions [parent]. center[0]][regions[parent]. center[1]] grow_region(regions [parent].center[0], regions [parent].center [ \} breā̄;
case 91 : /* peak; sphere; convex */
case 159 :/* ridge, cylinder, convex */
case 31 : /* minimal, */
case 127 :/* saddle ridge */
i
switch(regions[child].surf_type)

```
    {
        case 1 : /* flat */
        case 63: /* pit , sphere,concave */
        case 223:/* valley, cylinder , concave */
        case 191: /* saddle valley, concave */
        attempt [regions[parent] . center[0]][regions [parent].center[1]]
        grow_region(regions [parent].center[0], regions[parent].center[
            break:
        }
    }break;
    case 63: /* pit , sphere,concave *
    case 223:/* valley, cylinder, concave */
    case 191: /* saddle valley, concave *
    {
        switch(regions[child].surf_type
            l
            case 1 : /* flat *
            case 91 : /* peak; sphere; convex */
            case 91 : /** peak; sphere; convex */ 
            case 31 : /* minimal, */
            case 127 :/* saddle ridge */
        ttempt [regions[parent].center[0]][regions[parent].center[1]]
        grow region(regions[parent].center[0], regions [parent].center[
            break
        }
    }
    }/* end of switch */
} /* end of erode */
/***********************************GROW REGION******************************
&/
grow_region(nrow, ncol, parent
int nrow;
int ncol;
lnt parent;
{
int 1,j:
    float x,y;
    float value;
    if(to_disp == 1) lwr(ncol,nrow,&1kvalue);
    if(label[nrow][ncol] != parent)
```

marge. $a$
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18
( /* pixel to work on */
$x$ = nrow - regions [parent]. center [0];
$y=$ ncol - regions [parent]. center[1];
value $=x^{\star} x^{\star}$ regions [parent].coeffs [5] +
$x^{\star} y^{\star}$ regions[parent].coeffs[4] +
$y^{*} y^{*}$ regions [parent]. Coeffs [3]
$x^{\star}$ regions [parent].coeffs[2]
${ }^{\star}$ regions [parent]. coeffs[1]
*regions [parent] .coeffs[0]
1f(fabs((double) (value - buffer[nrow][ncol])) <=
(fabs((double)(regions [parent].fit_error)) + THRESH_ERROR))
/* acceptable pixel */
label[nrow] [ncol] = parent
regions[parent].size $=$ regions [parent].size +1 ;
regions[label[nrow][ncol]].size = regions[label[nrow][ncol]
for $(1=($ nrow -1$) ; 1<=($ nrow +1$) ; 1++$ )
for $(j=($ ncol -1$) ; j<=($ ncol +1$) ; j++)$
if $((1>=0) \& \&(1<r o w) \& \&(j>=0) \& \&(j<c o l))$
if((1 == nrow) || (j== ncol)
1
if((lap [1][f] < edge_threshold) \&\&
(attempt[1][j] !=-curr))
f
attempt[1][1] = curr:
/* check for the \# of pending calls */
num_call++;
if (num_call >= MAX_NUM_CALL)
store_pixel $(1, j)$;
else
\}
for $(1=($ nrow -1$) ; 1<=($ nrow +1$) ; 1++)$
for $(j=($ ncol -1$) ; j<=($ ncol +1$) ; j++$
if $((1>=0) \& \&(1<$ row $) \& \&(j>=0) \& \&(j<c o l))$ if((1 = = nrow) || $(j==$ ncol))
1
if((lap[i][f] < edge_threshold) \&\&
(attempt[1][f] !=-curr))
1
attempt[1][j] = curr
num_call++;
if (num_cail >= MAX_NUM_CALL)
morgo.c Hod Apr 20 21:59:01 1988 19

$$
\text { store_pixel }(1, j) \text {; }
$$

else
\}
grow_region(1,f,parent) ;
\}
\}
\}/* region_grow */
/******************** STORE PIXEL \& GET PIXEL ******************************)

```
store_pixel(rrow,ccol)
```

int rrow;
int ccol;
\} num call--:
pixel_stack[stack_length]. row = rrow;
pixel_stack[stack_length].col = ccol;
stack_length++;
if(stack length $==$ PIXEL_STACK_SIZE) stack_length $=0$;
if(stack_length $==$ current_element)
prīntf("\n segment : ${ }^{-}$Stack Collision While region growing $\backslash n$
)
get_pixel()
i
rownum = pixel_stack[current_element].row,
colnum = pixel_stack[current_element].col;
current_element++
if (current_element == PIXEL_STACK_SIZE) current_element = 0;
\}
號
returns the seed label.
*/
pick_up_seed ()
int $1, f, k, 1$;
/* first look for flat,spherical, cylidericalregions */

## merge.c Wod Apr 20 21:59:01 $1988 \quad 20$

for ( $1=1 ; 1<=\operatorname{tr} ; 1++$ )
1
if(((regions[1].surf_type == FLAT) /*|| (regions[1].surf_type = (regions[1].surf_type == RIDGE)*/) \&\&
(regions[1].done $==0$ ) \&\&
(regions[1].fit_error <= ACCEPT_ERROR) \&\&
(regions[1].size > MIN_REGION_SIZE))
1
.done $=1 ;$ return(1);
)
$\stackrel{3}{\text { Now }}$

* Now look for any type of acceptable region */

$$
\operatorname{for}(1=1 ; 1<=\operatorname{tr} ; 1++)
$$

1f((regions[i].done $==0) \& \&$ (regions[1].size > MIN_REGION_SIZE (regions[1].fit_error <= ACCEPT_ERROR)*/)
regions[1].done $=$ return(1):
\}
\}
/* No suitable seed region is available */ return(-1);
\} /* pickupseed */
*******GET NEXT REGION* returns the label of next region to be grown as convex region */
get_next_region()
int 1,j,k,l;
int max;
int tlabel:
$\max =-1000 ;$
for ( $1=1 ; 1<=\operatorname{tr} ; 1++$ )
if((regions[1].done $==0) \& \&($ (regions[1].surf_type $==$ FLAT $) \quad|\mid$ (regions[1].surf-type $==$ PEAK) || (regions[i].surf_type $==$ RIDGE))
(regions[i].size >= MIN_REGION_SIZE) \&\&
(regions[i].dmax > max))
1
$\max =$ regions[1].dmax;

```
morge.c Mod Apr 20 21:59:01 1988 21
            tlabel = 1;
        }
        }
    1f(max != -1000) return(tlabel);
    else return(-1)
    }
/***********************************EXTEND REGION*****************************
used to extend a convex region recursively at the region level.
*/
extend region(lab)
    int la\overline{b}
    int
        int 1,j:
        int n;
        con_label[lab] = convex_label;
        for(i=0;1<regions[lab].num_neigh;1++)
        l
            n = regions[lab] neighbour[i]
            if((regions[n].done == 0) &&
            /*(regions[congrowlab].dmax > regions[n].dmax) &&*/
                    (regions[n].surf type == FLAT) )
                    ((regions[n].surf type == FLAT)
                    *(regions[n].surf_type == PEAK)
                    regions[n].sur_type == RIDGE)/*
                    (regions[n].surf type == SADRID)
                    (regions[n].surf_type == MIN)*/))
            1
            if(regions[n].size >= MIN REGION SIZE)
            { regions[n].done = 1
                regions[n].done =
                &
            else
                regions[n].done = 1,
                con_label[n] = convex_label;
                }
            }
        }
}
```

```
solvor.c Tue Apr 5 11:52:21 1988 1
/*
    Used for calculating the least square fit-error in the grad.c program.
Caution, Warning, Danger:
1) Include the following in the main program.
\#define NSYS <n> \#define NDATA <m>
2) If you want to compute least-squares error, you MUST have the following in the calling program: double lsqerr():
It returns \(E=|A x-b| * 100\)
```

Parameters passed to the least square program are: A[][], x[], b[], m, n where
$A \quad x=b$.

To check the sigularity, small constant 'epsilon' is used. current value is set to 0.00001
This is 0.K. for most application, but depending on your application, you may want to change the value.

## */

\#include <stdio.h>
\#include <math.h>
\#include "/usr/users/alok/advanced/spline_include.h"
\#define NSYS 20
\#define NDATA 10000
int prnt $=1$
double epsilon $=0.00001$
/* = -1 for printing else no printing */
solver.c Tue Apr 5 11:52:21 $1988 \quad 2$

```
/*-------------------------------Lsquar
int lsquare (A, b, x, m, n)
double A[][NSYS], b[], x[];
int m, n ;
double ATA[NSYS][NSYS], ATb[NSYS]:
    int order ;
    if (m<n) {
        1f (prnt==-1) fprintf(stderr,"insufficient # of points.\n") ;
        return(1) :
    }
    product (A, ATA, m, n)
    mult(A, b, ATb, m, n) ;
    if((order=solver(ATA,ATb,x,n)) < n) {
        return(order).
    }
    return(0) ;
    } /* Lsquare *
```


int lsqsol (A, b, x, m, n, n1)
double $A[][N S Y S], b[], x[]$
int
$\mathrm{m}, \mathrm{n}, \mathrm{n} 1$;
1
double ATA[NSYS][NSYS], ATb[NSYS]:
int order, 1;
if ( $m<n$ ) \{
if (prnt==-1) fprintf(stderr,"insufficient \# of points. ${ }^{(n ")}$; return(1) ;
\}
product ( $A, A T A, m, n$ )
mult (A, b, ATb, m, n)
for $(1=0 ; 1<n$; $1++) x[1]=0.0$.
order $=\operatorname{solver}(A T A, A T b, x, n)$;
if (order<n) \{
if (order<n1) return(order) ;

```
solver.c Tue Apr 5 11:52:21 1988
    product (A, ATA,m,nl)
        mult (A,b,ATb,m,nl).
        for (i=0; 1<n; 1++) x[1] = 0.0;
        order = solver(ATA,ATb, x,nl);
        return(order) ;
    }
    n(0) :
} /* Lsquare */
```


double $A[][$ NSYS], $x[], b[]$,
int ${ }^{*}{ }_{m}{ }^{\text {paverb }}$
m, n
int
1, j ;
double sum, errsum, bsum ;
errsum=bsum=0.0
for $(1=0 ; 1<m ; i++) \quad 1$
sum $=0.0$;
for ( $j=0 ; j<n ; j++$ )
sum $=\operatorname{sum}+A[1][f] * x[j]$;
sum $=$ sum -b[1];
errsum = errsum + fabs(sum);
bsum = bsum + fabs(b[1]) ;
\}
*p_averb $=$ bsum $/($ double $) \mathrm{m}$;
return(errsum/m);
) /* lsqerr */

double $A[][N S Y S], x[], b[]$;

## solvar.c Tue Apr 5 11:52:21 1988 5

for (1=0; 1<n; 1++)
for $(j=0 ; j<n ; j++)$ AS[1][j] =A[1][j] ;
AS[1][n] $=\mathrm{b}[1]$;
$\mathrm{x}[1]=0.0$
${ }_{3}[1$
if ( $k=$ gauss (AS, $p, n$ ) ) $=0$ )
if(prnt==-1) printf("Singular Matrix. order = \%d. $\backslash \mathrm{n} ", \mathrm{k}$ ) ;
\}
k $=\mathrm{n}$;
backsub(AS, p, x, n, k)
return (k) ;
/* Solver */
/* Gaussian elimination with full pivoting.
/*
int gauss ( $\mathrm{A}, \mathrm{p}, \mathrm{n}$ )
double A[][NSYS] ;
int p[]$, \mathrm{n}$ :

1 int 1, j, k, row, col
int itemp;
int itemp ;
double pivot, ratio, dtemp :
ouble epsilon
epsilon $=0.0000001$;
/* Initialize permutation vector. *
for (1=0; $1<n$; $1++$ )
$\mathrm{p}[1]=1$;
p
for ( $k=0$; $k<n-1$; $k++$ ) \{
/* Find the next pivot element. *
ivot $=A[k][k]$
row $=$ col $=\mathrm{k}$.
for ( $1=k$; $1<n$; $1++$ )
for ( $j=k$; $j<n$; $j++$ )

## solver.c Tue Apr 5 11:52:21 1988

if(fabs (pivot)<fabs(A[1][j])) \{

```
        pivot = A[1][j] ;
```

        row = 1 ;
        col \(=\mathrm{j}\);
    ,
    if(fabs(pivot) <epsilon) return(k) ;
    /* Exchange row *
    if (k != row)
        for ( \(1=0 ; 1<=n ; 1++\) )
            dtemp \(=A[k][1]\);
            \(\mathrm{A}[\mathrm{k}][1]=\mathrm{A}[\) row \(][1]\)
            \(A[\) row \(][1]=\) dtemp ;
        \(\}^{A}{ }^{A}\)
        \}
    /* Exchange column */
    if ( k ! = col)
        itemp \(=\mathrm{p}[\mathrm{col}]\);
        \(\mathrm{p}[\mathrm{col}]=\mathrm{p}[\mathrm{k}]\);
        \(\mathrm{p}[\mathrm{k}]=\) itemp ;
        for ( \(1=0\); \(1<n\); \(1++\) )
            dtemp \(=A[1][k]\);
            \(A[1][k]=A[1][\mathrm{col}]\)
            \(\mathrm{A}[1][\mathrm{col}]=\) dtemp :
        1
    /* Elimination. *
    for \((1=k+1\); \(1<n\); \(1++\) )
        ratio \(=A[i][k] /\) pivot
            \(\mathrm{A}[1][\mathrm{k}]=0.0\);
            for \((j=k+1\); \(j<=n\); \(j++)\)
            A[1][j] =A[1][f]- ratio * A[k][f] -
        ,
    \}
    if (fabs(A[n-1][n-1])<epsilon) return(n-1) ;
return (0) ;
return (0) ;

```
solver.c Tue Apr 5 11:52:21 1988 7
/*-------------------------------Backsub
/* Back substitution.
*/
backsub (A, p, soln, N, n)
double A[][NSYS], soln[] ;
int p[],N, n .
1
    int 1, j. k ;
    double sum, sol[NSYS] :
    for (k=n-1; k>=0; k--) {
    sum = 0.0;
        for (j=k+1; j<n; j++)
        sum = sum + A[k][j] * sol[f] ;
        sol[k] = (A[k][N] - sum)/A[k][k]
        }
    for (k=0; k<n; k++) {
        1 = p[k] ;
        soln[i] = sol[k] ;
        }
    return ;
    /* Backsub */
/*--------------------------------------
product (B, C, m, n)
double B[][NSYS], C[][NSYS] :
int
    B[] n;
{ int i, j, k
    lnt i, J, k
```

```
solver.c Tue Apr 5 11:52:21 1988
    for (1=0; 1<n; 1++)
    for (1=0; 1<n; 1++)
    for ( }j=0;j<n; j++) 
    for(k=0; k<
        sum = sum + B[k][i]*B[k][f];
        C[1][f] = sum 
    }
return ;
    } /* Product */
```


int $m, n$ :
int
int 1, J
double sum:
for ( $1=0$; $1<n$; $1++$ ) $\{$
sum $=0.0$;
for ( $j=0$; $j<m ; j++$ )
for $(j=0 ; ~ f<m ; ~$
sum $=\operatorname{sum}+A[f][1] \star b[f]$;
sum $=$ sum
$c[1]=$ sum ;
$\begin{gathered}c[1] \\ \}\end{gathered}$
return :
\} /* Mult */

## solvar.c Tue Apr 5 11:52:21 1988 5

for (1=0; 1<n; 1++)
for $(j=0 ; j<n ; j++)$ AS[1][j] =A[1][j] ;
AS[1][n] $=\mathrm{b}[1]$;
$\mathrm{x}[1]=0.0$
${ }_{3}[1$
if ( $k=$ gauss (AS, $p, n$ ) ) $=0$ )
if(prnt==-1) printf("Singular Matrix. order = \%d. $\backslash \mathrm{n} ", \mathrm{k}$ ) ;
\}
k $=\mathrm{n}$;
backsub(AS, p, x, n, k)
return (k) ;
/* Solver */
/* Gaussian elimination with full pivoting.
/*
int gauss ( $\mathrm{A}, \mathrm{p}, \mathrm{n}$ )
double A[][NSYS] ;
int p[]$, \mathrm{n}$ :

1 int 1, j, k, row, col
int itemp;
int itemp ;
double pivot, ratio, dtemp :
ouble epsilon
epsilon $=0.0000001$;
/* Initialize permutation vector. *
for (1=0; $1<n$; $1++$ )
$\mathrm{p}[1]=1$;
p
for ( $k=0$; $k<n-1$; $k++$ ) \{
/* Find the next pivot element. *
ivot $=A[k][k]$
row $=$ col $=\mathrm{k}$.
for ( $1=k$; $1<n$; $1++$ )
for ( $j=k$; $j<n$; $j++$ )

## solver.c Tue Apr 5 11:52:21 1988

if(fabs (pivot)<fabs(A[1][j])) \{

```
        pivot = A[1][j] ;
```

        row = 1 ;
        col \(=\mathrm{j}\);
    ,
    if(fabs(pivot) <epsilon) return(k) ;
    /* Exchange row *
    if (k != row)
        for ( \(1=0 ; 1<=n ; 1++\) )
            dtemp \(=A[k][1]\);
            \(\mathrm{A}[\mathrm{k}][1]=\mathrm{A}[\) row \(][1]\)
            \(A[\) row \(][1]=\) dtemp ;
        \(\}^{A}{ }^{A}\)
        \}
    /* Exchange column */
    if ( k ! = col)
        itemp \(=\mathrm{p}[\mathrm{col}]\);
        \(\mathrm{p}[\mathrm{col}]=\mathrm{p}[\mathrm{k}]\);
        \(\mathrm{p}[\mathrm{k}]=\) itemp ;
        for ( \(1=0\); \(1<n\); \(1++\) )
            dtemp \(=A[1][k]\);
            \(A[1][k]=A[1][\mathrm{col}]\)
            \(\mathrm{A}[1][\mathrm{col}]=\) dtemp :
        1
    /* Elimination. *
    for \((1=k+1\); \(1<n\); \(1++\) )
        ratio \(=A[i][k] /\) pivot
            \(\mathrm{A}[1][\mathrm{k}]=0.0\);
            for \((j=k+1\); \(j<=n\); \(j++)\)
            A[1][j] =A[1][f]- ratio * A[k][f] -
        ,
    \}
    if (fabs(A[n-1][n-1])<epsilon) return(n-1) ;
return (0) ;
return (0) ;

```
solver.c Tue Apr 5 11:52:21 1988 7
/*-------------------------------Backsub
/* Back substitution.
*/
backsub (A, p, soln, N, n)
double A[][NSYS], soln[] ;
int p[],N, n .
1
    int 1, j. k ;
    double sum, sol[NSYS] :
    for (k=n-1; k>=0; k--) {
    sum = 0.0;
        for (j=k+1; j<n; j++)
        sum = sum + A[k][j] * sol[f] ;
        sol[k] = (A[k][N] - sum)/A[k][k]
        }
    for (k=0; k<n; k++) {
        1 = p[k] ;
        soln[i] = sol[k] ;
        }
    return ;
    /* Backsub */
/*--------------------------------------
product (B, C, m, n)
double B[][NSYS], C[][NSYS] :
int
    B[] n;
{ int i, j, k
    lnt i, J, k
```

```
solver.c Tue Apr 5 11:52:21 1988
    for (1=0; 1<n; 1++)
    for (1=0; 1<n; 1++)
    for ( }j=0;j<n; j++) 
    for(k=0; k<
        sum = sum + B[k][i]*B[k][f];
        C[1][f] = sum 
    }
return ;
    } /* Product */
```


int $m, n$ :
int
int 1, J
double sum:
for ( $1=0$; $1<n$; $1++$ ) $\{$
sum $=0.0$;
for ( $j=0$; $j<m ; j++$ )
for $(j=0 ; ~ f<m ; ~$
sum $=\operatorname{sum}+A[f][1] \star b[f]$;
sum $=$ sum
$c[1]=$ sum ;
$\begin{gathered}c[1] \\ \}\end{gathered}$
return :
\} /* Mult */

```
m01ver.c Tue Apr 5 11:52:21 1988 S
    for (1=0; 1<n; 1++) (
        for(j=0; j<n; j++) AS[1][j] = A[1][j] :
        for(j=0; j<n; j++)
        x[1] = 0.0 :
        }[1
    If ((k=gauss(AS,p,n))!=0) {
        if(prnt==-1) printf("Singular Matrix. Order = (d.\n", k):
    )
    k n n;
    backsub(AS, p, x, n,k) ;
    return(k) ;
    } /* Solver */
```


1nt gauss ( $A, p, n)$
double A[][NSYS]:
int pll, $n$ :
1
int 1, j, $k$, row, col :
int
int itemp:
double pivot, ratio, dtemp ;
double epsilion ;
epsilon $=0.0000001$ :
/* Initialize permutation vector. */
for $(1=0 ; 1<n ; 1++) \quad 1$
$\mathrm{p}[1]=1$;
p
for ( $k=0 ; k<n-1 ; k++1 \quad 1$
/* Find the next pivot element. */
pivot =A[k] (k):
row = col =k;
for ( $1=k$; $1<n ; 1++$ )
for ( $j=k$; $j<n$; $j++$ ) (

```
salvar.e Tue Apre 5 11:52:21 1988 G
    if(fabs(pivot)<fabs(A[1][f])) {
            plvot = A[1][j];
            row = 1;
            col = 1 ;
        }
    if!
    if(fabs(pivot)<epsilon) return(k) :
```

    /* Exchange row */
    if (k ! = row) (
        for ( \(1=0\) : \(1<=n ; 1++\) ) 1
            dtemp \(=A(k][1]\);
            \(A(k][1]=A(\) row \(][1]\)
            \(A[\) row \(][1]=\) dtemp :
        \(1^{1}\)
        1
    /* Exchange column */
    if (k ! = col) (
        1temp \(=p[\) col \(]\)
        \(p[c o l]=p[k]\)
        \(p(k)=\) itemp ;
        for ( \(1=0\); \(1<n ; 1++\) )
            dtemp \(=A[1][k]\);
            \(A[1](\mathrm{k}]=\mathrm{A}[1][\) col \(]\);
            A[1][col] = dtemp :
            )
        1
    /* Elimination. */
for $(1=k+1 ; 1<n ; 1++) \quad 1$
ratio $=A[1][k] /$ pivot :
$A(1][k]=0.0$ :
for $(j=k+1$; $j<=n$; $j++$ )
or $(j=k+1: f<=n ; j++)$
$A[1][j]=A[1][j]-$ ratio * $A[k][j]$;
1
1
if (fabs(A[n-1][n-1])<epsilon) return(n-1):
return ( 0 ) :
1//* Gauss */

```
m01ver.c Tue Apr 5 11:52:21 1988 S
    for (1=0; 1<n; 1++) (
        for(j=0; j<n; j++) AS[1][j] = A[1][j] :
        for(j=0; j<n; j++)
        x[1] = 0.0 :
        }[1
    If ((k=gauss(AS,p,n))!=0) {
        if(prnt==-1) printf("Singular Matrix. Order = (d.\n", k):
    )
    k n n;
    backsub(AS, p, x, n,k) ;
    return(k) ;
    } /* Solver */
```


1nt gauss ( $A, p, n)$
double A[][NSYS]:
int pll, $n$ :
1
int 1, j, $k$, row, col :
int
int itemp:
double pivot, ratio, dtemp ;
double epsilion ;
epsilon $=0.0000001$ :
/* Initialize permutation vector. */
for $(1=0 ; 1<n ; 1++) \quad 1$
$\mathrm{p}[1]=1$;
p
for ( $k=0 ; k<n-1 ; k++1 \quad 1$
/* Find the next pivot element. */
pivot =A[k] (k):
row = col =k;
for ( $1=k$; $1<n ; 1++$ )
for ( $j=k$; $j<n$; $j++$ ) (

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    if(fabs(pivot)<fabs(A[1][f])) {
            plvot = A[1][j];
            row = 1;
            col = 1 ;
        }
    if!
    if(fabs(pivot)<epsilon) return(k) :
```

    /* Exchange row */
    if (k ! = row) (
        for ( \(1=0\) : \(1<=n ; 1++\) ) 1
            dtemp \(=A(k][1]\);
            \(A(k][1]=A(\) row \(][1]\)
            \(A[\) row \(][1]=\) dtemp :
        \(1^{1}\)
        1
    /* Exchange column */
    if (k ! = col) (
        1temp \(=p[\) col \(]\)
        \(p[c o l]=p[k]\)
        \(p(k)=\) itemp ;
        for ( \(1=0\); \(1<n ; 1++\) )
            dtemp \(=A[1][k]\);
            \(A[1](\mathrm{k}]=\mathrm{A}[1][\) col \(]\);
            A[1][col] = dtemp :
            )
        1
    /* Elimination. */
for $(1=k+1 ; 1<n ; 1++) \quad 1$
ratio $=A[1][k] /$ pivot :
$A(1][k]=0.0$ :
for $(j=k+1$; $j<=n$; $j++$ )
or $(j=k+1: f<=n ; j++)$
$A[1][j]=A[1][j]-$ ratio * $A[k][j]$;
1
1
if (fabs(A[n-1][n-1])<epsilon) return(n-1):
return ( 0 ) :
1//* Gauss */


