Advanced Vision Assignment 3

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1 Introduction

This assignment involved analyzing 21 consecutive image frames captured with a Kinect sensor. These frames contained both RGB pixel intensity values and XYZ range values. The images showed many different angles of a cabinet containing an open book on top. Our goal was the following:

- (i) Extract the three large visible planes of the filing cabinet and compute their plane equations
- (ii) Extract the set of points belonging to the book
- (iii) Compute the translation and rotation of the three planes between frames to register all views together
- (iv) Fuse all views of the book together using the translation and rotation computed in the previous step

At last we evaluated our approach by creating a 3D fused view of the book, computing the determinant of the covariance matrix between the same point in all frames, and computed the average and standard deviation of surface normal angles which have been corrected in terms of rotation.

2 Methods

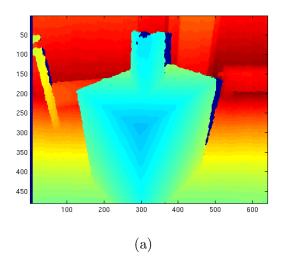
2.1 Pre-processing

Before extracting the planes, we performed the following pre-processing steps:

- We transformed the depth component of the supplied XYZ data from the Kinect sensor into a 3D point cloud representation similar to the representation used for system 6 from the lectures, to make it easier to perform region growing to extract the planes using the lecture example code
- We cleaned the data by:
 - removing all points from the data in which the Z co-ordinate (representing the depth of the point) was higher than a given threshold (we found that 1400 was an appropriate choice), as these were not relevant to the problem that we were trying to solve, since the only things we are interested are the cabinet and the book, which are in the foreground.
 - removing all points from the data in which the XYZ co-ordinates were (0,0,0), as these points were found to interfere with the plane fitting

• We normalized the data by dividing all of the X,Y, and Z co-ordinate values by 5. This was found to have a favorable effect on the speed and accuracy of the region growing algorithm

Figure 1 illustrates the effect of our transformations. The raw XYZ depth data for frame 17 (which we used as our foundation frame) is shown in the picture on the left of the figure, and the transformed data is shown on the right.



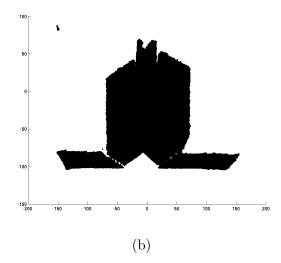


Figure 1: The raw depth image from the XYZ data (a) and the same data after the pre-processing steps had been carried out (b) for our foundation frame.

2.2 Plane Extraction

2.2.1 Patch Selection

We used region growing for extracting the points on the three planes, and managed to speed up that procedure considerably by approximating the position of the initial patches on each of the three planes. To do that we extracted the closest point in terms of Z in the range data, which was the corner of the cabinet, and then extracted three points with the following:

```
rightPoint = (cornerPoint(x) + 40, cornerPoint(y) - 20, cornerPoint(z) + 30)
leftPoint = (cornerPoint(x) - 40, cornerPoint(y) - 20, cornerPoint(z) + 30)
topPoint = (cornerPoint(x), cornerPoint(y) + 20, cornerPoint(z) + 20)
```

Here the rightPoint is our approximation for the initial patch for the right side of the cabinet, leftPoint for the left side, and topPoint for the top half (see Figure 2a).

By using these approximation points we found three points in the point cloud that were closest to the approximations, and each of them respectively turned out to be on the planes we wanted to extract. This meant that we could use them as out initial patches for region growing on each plane and not have to do an exhaustive search for points to serve as initial patches. The patches we acquired using these points were computed using a distance tolerance of 5

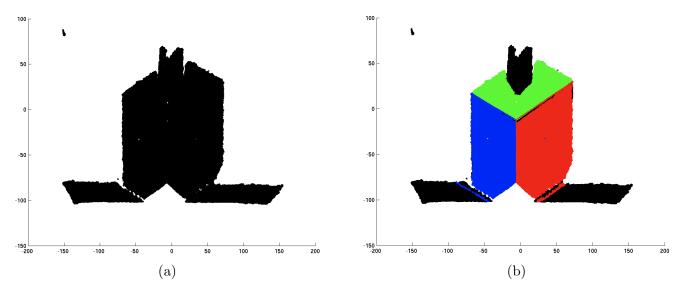


Figure 2: Reduced point cloud data with the corner point (red) and the three approximation points for the three planes (a) and region growing with initialization on the closest patch to the points (b).

2.2.2 Region Growing

In our region growing algorithm we used a plane tolerance of 1 and point distance tolerance of 150. By having a low plane tolerance we acquired more accurate regions and with such a large point distance tolerance the growing only needed about 5 iterations. When the number of new points in a region exceeded the number of points already in the region plus 50 then a plane is (re)fitted to the region. The region growing stops when the least square fitting error of that plane is bigger than 30% of the number of new points.

2.3 Book Extraction

To extract the set of points that represented the book we used couple of thresholds in order to capture the book and reduce noise. First, we used the point cloud that can be seen in Figure 2a and subtracted the points belonging to the three planes. Now to eliminate other data points, such as outside the cabinate (see Figure 2b), we subtracted all points that were to the left and right of the topmost plane, and ones that were below the corner point. Finally, to eliminate noise on the topmost plane, we extracted only the points that had the distance ≥ 1 from the top plane. The result can be seen in Figure 3.

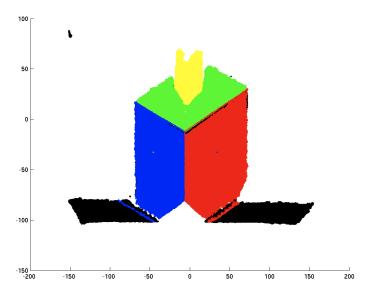


Figure 3: Book points extracted and displayed in yellow

2.4 Plane Registration

For registering views we decided to use frame number 17 as our foundation frame (see Figure ??) which we used for computing the rotation and translation between that frame and all the others.

2.5 **Book Fusion**

3 **Evaluation**

We computed the angles between the surface normals of the upwards facing plane of our foundation frame and the others. The average angle between the vectors was 0.0756 and the standard deviation was 0.0599.

4 Code

4.1 run.m

```
% Run file for Assignment 3
global model planelist planenorm facelines
load ( '... / ... / Data / data1 . mat ')
planeEq = zeros(3,4,21);
closestPoints = zeros(21,3);
bookPoints = \{\};
fig = figure(1);
for frameNo = 1:length(frame)
    clf:
    hold on
    [pointCloud planeEq planePoints closestPoints(frameNo,:)] = ...
        planeExtraction(frame(frameNo));
    bookPoints{frameNo} = bookExtraction(pointCloud, planeEq, ...
        planePoints, closestPoints(frameNo,:));
end;
% Computing surface normal angle mean and standard deviation
[angle rotationMatrices] = findRotationMatrices(planes);
angles = [angle(1:16) angle(18:21)];
angleMean = sum(angles) / length(angles);
squaredMeanDiff = (angles-mean).^2;
stdDevAngle = sqrt(sum(squaredMeanDiff)/length(angles));
4.2
     Plane Extraction
4.2.1
     planeExtraction.m
```

```
function [pointCloud planeEq planePoints closestPoints] = ...
    planeExtraction (frame)
                   % Points on each individual plane
planePoints = \{\};
                        % Points on the book
bookPoints = \{\};
closest Points = zeros(3); % Closest point in each image, i.e. the corner
xyzFrame = frame.XYZ(:,:,:);
```

```
xFrame = xyzFrame(:,:,1);
xFrame = xFrame(:);
vFrame = xyzFrame(:,:,2);
yFrame = yFrame(:);
zFrame = xyzFrame(:,:,3);
zFrame = zFrame(:);
vectorizedFrame = [xFrame yFrame zFrame];
pointCloud = vectorizedFrame(randperm(size(vectorizedFrame, 1)),:);
depthThreshold = 1400;
backgroundRowsToIgnore = find(pointCloud(:,3) > depthThreshold);
zeroRowsToIgnore = find(abs(pointCloud(:,1)) + ...
    abs(pointCloud(:,2)) + abs(pointCloud(:,3)) == 0);
index = true(1, size(pointCloud, 1));
index(backgroundRowsToIgnore') = false;
index(zeroRowsToIgnore') = false;
pointCloud = pointCloud(index,:)./5;
plot3 (pointCloud (:,1), pointCloud (:,2), pointCloud (:,3), 'k.')
closestPoints = getClosestPoint(pointCloud);
plot3 (closestPoints (1), closestPoints (2), 3000, 'r.')
[NPts,W] = size(pointCloud);
planeEq = zeros(3,4);
remaining = pointCloud;
rightPoint = [closestPoints(1)+40 \dots]
    closestPoints(2)-20 closestPoints(3)+30];
leftPoint = [closestPoints(1) - 40 \dots]
    closestPoints(2)-20 closestPoints(3)+30];
topPoint = [closestPoints(1) ...
    closestPoints(2)+20 closestPoints(3)+20];
patchPoints = [rightPoint; leftPoint; topPoint]
plot3 (rightPoint (:,1), rightPoint (:,2),3000, 'b.')
plot3 (leftPoint (:,1), leftPoint (:,2),3000, 'g.')
plot3 (topPoint (:,1), topPoint (:,2),3000, 'y.')
for i = 1 : 3
    planePatch = getClosestPointToPoint(pointCloud, patchPoints(i,:))
    % select a random small surface patch from the remaining points
    [oldlist, plane] = select_patches (remaining, planePatch);
    plot3 (oldlist (:,1), oldlist (:,2), oldlist (:,3), 'y.')
    % grow patch
    growthCycles=0;
```

```
stillgrowing = 1;
    while stillgrowing
        growthCycles = growthCycles+1
        stillgrowing = 0; %— until we find a bad fit we keep growing
         [newlist, remaining] = \dots
             getallpoints (plane, oldlist, remaining, NPts);
         [NewL,W] = size(newlist);
         [OldL,W] = size(oldlist);
        if i == 1
             plot3 (newlist (:,1), newlist (:,2), newlist (:,3), 'r.')
             save1=newlist;
         elseif i==2
             \mathbf{plot3} (newlist (:,1), newlist (:,2), newlist (:,3), 'b.')
             save2=newlist;
        else
             plot3 (newlist (:,1), newlist (:,2), newlist (:,3), 'g.')
             save3=newlist;
        end
        pause (0.01)
        if NewL > OldL + 50
             ['refitting_plane']
             % refit plane
             [newplane, fit] = fitplane(newlist);
             planeEq(i,:) = newplane';
             fitThreshold = 0.3;
             if fit > fitThreshold*NewL % bad fit - stop growing
                 break
             end
             stillgrowing = 1;
             oldlist = newlist;
             plane = newplane;
        end
    end
    [ '************ Segmentation Completed ']
    planePoints{i} = oldlist;
4.2.2 getClosestPoint.m
% gets the closest point in the image
```

end

```
function meanClosestPoint = getClosestPoint(pointCloud)
minDepth = min(pointCloud(:,3))
minDepthIndex = find(pointCloud(:,3)==minDepth)
closestPoints = pointCloud(minDepthIndex,:)
meanClosestPoint = mean(closestPoints)
```

4.2.3 getClosestPointToPoint.m

```
 \begin{tabular}{ll}  \begin
```

4.2.4 Other

Here we also used the following code from the course:

- select_patches.m Slightly modified version of select_patch.m where the exhaustive search for points is removed and instead only the three approximated plane patches are used.
- \bullet qetall points.m Tolerance variables modified.
- \bullet fitplane.m Not modified.

4.3 Book Extraction

4.4 bookExtraction.m

```
function bookPoints = \dots
       bookExtraction(pointCloud, planeEq, planePoints, closestPoints)
% subtract points in R that are on planes
allPoints = [planePoints {1}; planePoints {2}; planePoints {3}];
[, planeRowsToIgnore] = ismember(allPoints, pointCloud, 'rows');
planeRowsToIgnore(find(planeRowsToIgnore > 0));
planeIndex = true(1, size(pointCloud, 1));
planeIndex(planeRowsToIgnore') = false;
reducedPointCloud = pointCloud(planeIndex,:,:);
% subtract points too low / to left / to right of planes
minXPoint = min(planePoints \{3\}(:,1));
\max XPoint = \max(planePoints \{3\}(:,1));
bookPoints = reducedPointCloud(find( ...
    (reducedPointCloud(:,1) > minXPoint) \& ...
    (reducedPointCloud(:,1) < maxXPoint) & ...
    (reducedPointCloud(:,2) > closestPoints(2))),:);
% Get points above top plane
[abovePoints aboveIndices] = ...
    getPointsAbovePlane(bookPoints, planeEq(3,:));
```

```
bookPoints = bookPoints(aboveIndices\,,:);\\ \textbf{plot3}(bookPoints(:,1)\,,\ bookPoints(:,2)\,,bookPoints(:,3)\,,'y.');\\ \textbf{pause}(0.01);\\ \%\ saveas(fig\ ,strcat\ ('../../Images/BookExtraction3/',int2str(frameNo))\,,'png');\\ \end{cases}
```

4.5 getPointsAbovePlane.m

```
function [points indices] = getPointsAbovePlane(R, plane)
R = [R ones(size(R,1),1)];
distanceFromPlane = R*plane';
distanceTolerance = -1;
indices = find(distanceFromPlane <= distanceTolerance);
points = R(indices);</pre>
```

4.6 Plane Registration

4.7 Book Fusion