

Point Cloud Analysis Methods

Segmentation

```
% Process the data with a "Rolling Mask"
function SegmentationProcessing(obj)

    % Board string and offset
    board_state = "";
    shrink_offset = 0.002;

    % Outer Loop: Rows. Inner loop: Columns.
    for row = 0.148:-0.037:-0.111
        for column = -0.148:0.037:0.111

            % Clip data
            [tmpX, tmpY, tmpZ] = obj.ClipPointCloud(obj.X, obj.Y, obj.Z, ...
                [column+0.037-shrink_offset, column+shrink_offset],...
                [row-shrink_offset, row-0.037+shrink_offset],...
                [0.4, 0.3]);

            % Get number of data points
            d_size = size(tmpX);

            % Append to string if a piece is there or not
            if d_size(1) > 500
                board_state = board_state + "1";
            else
                board_state = board_state + "0";
            end
        end
        board_state = board_state + "/";
    end
    % DEBUG: Display final string
    disp(board_state);
end
```

Segmentation with Clustering Addon

```
%-----Clustering Addon-----%
if size(tmpX) > 0
    min_distance = 0.005;

    sizeZ = size(tmpZ);
    newZ = zeros(sizeZ(1), 1);

    p = pointCloud([tmpX, tmpY, newZ]);

    [labels, numClusters] = pcsegdist(p, min_distance);

    largest_cluster = [0 0 0];
    for cluster = 1:numClusters
        index = find(labels == cluster);
```

```

        if size(index) > size(largest_cluster)
            largest_cluster = index;
        end
    end
end

%-----END-----%

```

Clustering: Color

```

% Colors Clusters
function ColorClusters(obj, point_cloud, labels, numClusters)
    labelColorIndex = labels+1;

    % Currently Plot on figure 2. To be changed
    figure(2);
    pcshow(point_cloud.Location, labelColorIndex);

    % Change background to white. Black cluster invisible on black
    % background
    set(gca, 'color', 'w');

    % Make colormap using number of clusters
    colormap([hsv(numClusters+1);[0 0 0]]);
end

```

Clustering: Basic Approach

```

% Process data using clustering techniques
function ClusterProcessing(obj)

    % Min Distance between points for clustering
    min_distance = 0.0050; %0.0075;

    % Use clustering method
    [labels, numClusters] = pcsegdist(obj.point_cloud, min_distance);
    disp(numClusters);

    % Color clusters
    obj.ColorClusters(obj.point_cloud, labels, numClusters);
end

```

Clustering: Flattening Approach

```

% Process data using clustering techniques
function ClusterProcessing(obj)

```

```

% Min Distance between points for clustering
min_distance = 0.0050; %0.0075;

sizeZ = size(obj.Z);
newZ = zeros(sizeZ(1), 1);

p = pointCloud([obj.X, obj.Y, newZ]);
figure(3);
pcshow(p);

[labels, numClusters] = pcsegdist(p, min_distance);
disp(numClusters);

obj.ColorClusters(p, labels, numClusters);
end

```

Clustering: Iterative Approach - Active

```

% This function does the heavy work for conditioning
% 1st step: all clusters, which are in x-y too large are split with
% a kmeans clustering in two subclusters
% 2nd step: a corresponding new list of locs,
% numCluster, and all class variables is established
% 3rd step: all clusters, which are vertically too close, are fused
% in the same cluster
% 4th step: This gets repressed
function ClusterConditioning(obj, numIter)
% iterate the cluster conditioning
for clCounter = 1:numIter
    % 1st step: split too large clusters into subclusters
    labels = obj.labels;
    numClusters = obj.numClusters;
    NewNumClusters = 1; % new ordered number of total clusters
    nloc = zeros(1,3); % new ordered point location list
    nlabels = [0]; % new corresponding point list

    for i = 1:numClusters
        index = find(labels == i);
        % calculate the x-y extent of the cluster
        cl = obj.loc(index,:);
        sizeX = abs(max(cl(:,1))-min(cl(:,1)));
        sizeY = abs(max(cl(:,2))-min(cl(:,2)));

        % criteria for too large clusters for split is applied
        if ( (sizeX > (1.3*0.037)) || (sizeY > (1.3*0.037)) )
            % cluster too big, so we use kmeans clustering to
            % subdivide it
            Data2D = [cl(:,1), cl(:,2)];
            Data3D = [cl(:,1), cl(:,2), cl(:,3)];
            [idx,C] = kmeans(Data2D,2);

            % add the first new subcluster to our new data list

```

```

        new_index = find(idx == 1);
        nloc = [nloc; cl(new_index,:)];
        nlabels = [nlabels, ones(1, numel(new_index))*NewNumClusters];
        NewNumClusters = NewNumClusters + 1;
        % and add the second new subcluster
        new_index = find(idx == 2);
        nloc = [nloc; cl(new_index,:)];
        nlabels = [nlabels, ones(1, numel(new_index))*NewNumClusters];
        NewNumClusters = NewNumClusters + 1;
    else
        % the cluster checks out (i.e. is small), so we simply
        % concatenate it with the output data
        nloc = [nloc; cl];
        nlabels = [nlabels, ones(1,numel(index))*NewNumClusters];
        NewNumClusters = NewNumClusters + 1;
    end

end

% clip first elements
nloc = nloc(2:end,:);
nlabels = nlabels(2:end);

%write the data back into the class variables
obj.loc = nloc;
obj.labels = nlabels;
obj.numClusters = NewNumClusters;
% update the point cloud object
obj.ptCloud = pointCloud(obj.loc);

% 2nd step of the iteration: vertical fusion of the clusters
obj.calcClusterCenter();
obj.fuseClusters();
end

end

```

Clustering: Iterative Approach - Passive

```

% Process data using clustering techniques
function board_state = ClusterProcessing(obj, min_distance, iteration_count, ...
    slice_ratio, min_distance_increase)

% min_distance = 0.003; %0.0075;
final_data = [obj.X, obj.Y, obj.Z];
cluster_center_list = [];

for iteration = 1:iteration_count
    sizeZ = size(final_data(:,3));
    newZ = zeros(sizeZ(1), 1);

    new_pc = pointCloud([final_data(:,1), final_data(:,2), newZ]);
    %new_pc = pointCloud([final_data(:,1), final_data(:,2), final_data(:,3)]);

```

```

obj.PlotPointCloud(new_pc, iteration);
[labels, numClusters] = pcsegdist(new_pc, min_distance);
%           if numClusters ~= 0 && iteration == 5
%           obj.ColorClusters(new_pc, labels, numClusters, 2);
%           end

cluster_center_list = zeros(numClusters, 3);
new_data = [0, 0, 0];

for i = 1:numClusters
    % Find data belonging to cluster
    index = find(labels == i);
    newX = final_data(index, 1);
    newY = final_data(index, 2);
    newZ = final_data(index, 3);

    data = [newX, newY, newZ];

    [~,idx] = sort(data(:,3)); % sort just by the height column
    data = data(idx,:); % sort the whole matrix using the sort indices

    d_size = size(data);

    % Keep Portion of data
    data = data(1:round(d_size(1)/slice_ratio), :);
    new_data = [new_data; data];

    % Find center of mass of piece and insert into array
    cluster_center = mean(data);
    cluster_center_list(i,:) = cluster_center;

    % % Print Cluster Centers on a plot
    % figure(3)
    % hold on
    % plot3(cluster_center_list(i,1),
    % cluster_center_list(i,2),...
    % cluster_center_list(i,3), 'o','LineWidth',3);
    % hold off

end
new_data(1,:) = [];
final_data = new_data;
min_distance = min_distance + min_distance_increase;
end

%Map centers to grid and interpolate which squares a piece is on

board_state = "";

% Outer Loop: Rows. Inner loop: Columns.
for column = -0.148:0.037:0.111
    for row = -0.148:0.037:0.111

        [tmpX, tmpY, tmpZ] = obj.ClipPointCloud(cluster_center_list(:,1), cluster_center_list(:,2), cluster_center_list(:,3));
    end
end

```

```

        cluster_center_list(:,3), [row+0.037, row],...
        [column+0.037, column],...
        [0.4, 0.3]);

    % Get number of data points
    d_size = size(tmpX);

    % Append to string if a piece is there or not
    if d_size(1) > 0
        board_state = board_state + "1";
    else
        board_state = board_state + "0";
    end
end
board_state = board_state + "/";
end
% DEBUG: Display final string
%disp(board_state);
end

```