

# Plane Extraction from Point Cloud Data

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# Planar Segmentation Algorithm

Range image *versus* point clouds

Row  $\times$  Column image representation

- Obvious neighbour relations
- Easier region growing algorithms

3D Point Clouds

- Neighbour relations in  $R^3$
- Good data structures can help with neighbour connections

Segmenting range image into planar regions:  
Use region growing algorithm

## Surface Detection Main Algorithm

```
% find surface patches
[NPts,W] = size(R);
planelist = zeros(20,4);
foundcount=0;
while notdone

    % select small local surface patch from remaining points
    [oldlist,plane] = select_patch(remaining);

    % grow patch
    stillgrowing = 1;
    while stillgrowing

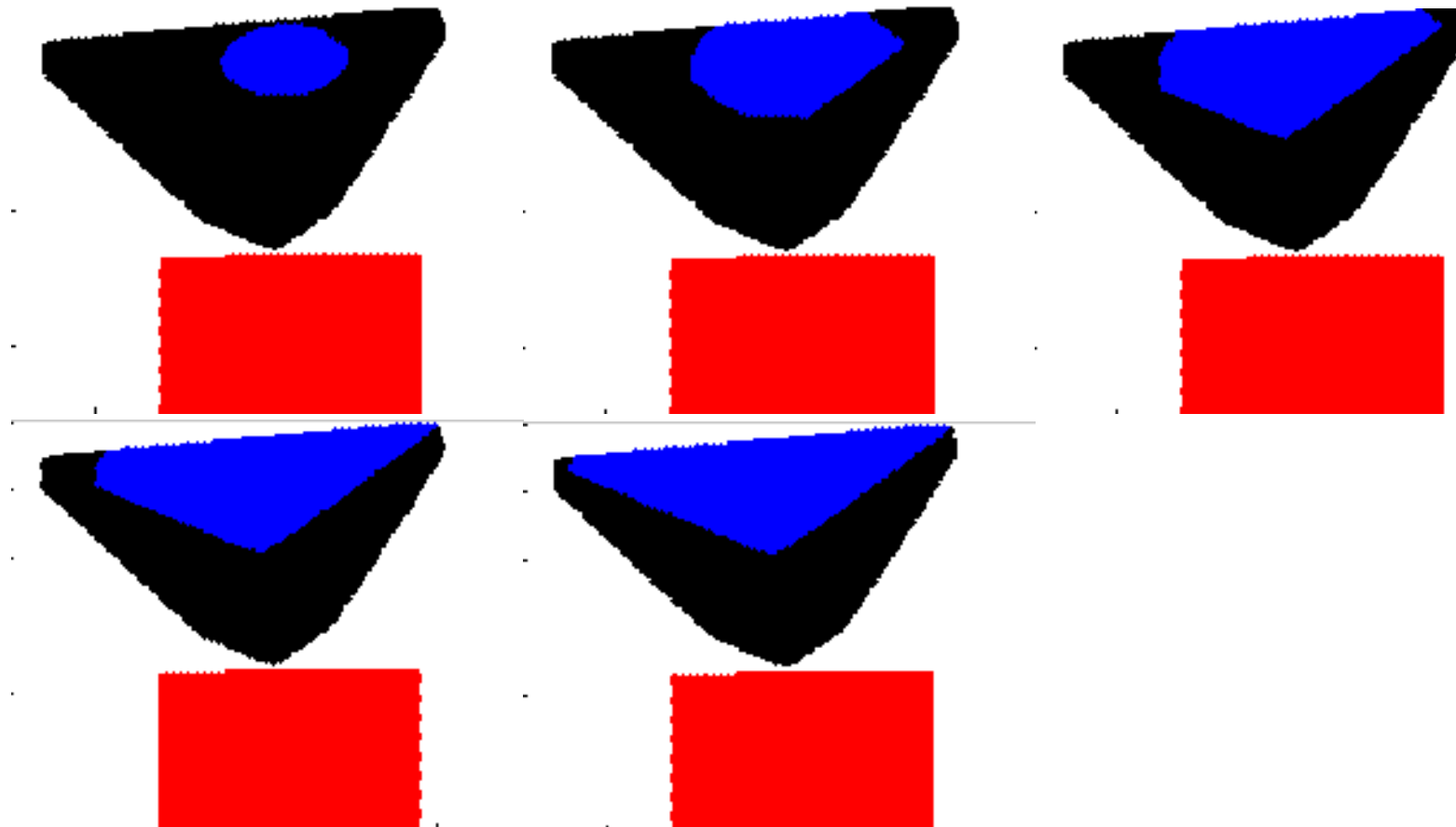
        % find neighbouring points that lie in plane
        stillgrowing = 0;
```

```
[newlist,remaining] = getallpoints(plane,oldlist,  
                                remaining,NPts);  
  
[NewL,W] = size(newlist);  
[OldL,W] = size(oldlist);  
if NewL > OldL + 50  
    % refit plane  
    [newplane,fit] = fitplane(newlist);  
    if fit > 0.04*NewL % fit going bad - stop growing  
        break  
    end  
    stillgrowing = 1;  
    foundcount = foundcount+1;  
    planelist(foundcount,:) = newplane';  
    oldlist = newlist;  
    plane = newplane;
```

# Region Growing Principles

Given a planar region formed from points  $S$  with equation  $\vec{n}'\vec{x} + d = 0$ , and a new point  $\vec{y}$ , add  $\vec{y}$  to  $S$  if:

1)  $|\vec{n}'\vec{y} + d| < \tau_p$  and 2) there is a point  $\vec{z}$  in  $S$  such that  $\|\vec{y} - \vec{z}\| < \tau_n$ .



## Plane Fitting

Given a set of datapoints  $\{\vec{x}_i\}$ , find the  $\vec{n}$  and  $d$  that best fit  $\vec{n}'\vec{x}_i + d = 0$  for all  $i$ .

Extend data:  $\vec{y}_i = [\vec{x}_i, 1]$

Extend parameters:  $\vec{p} = [\vec{n}, d]$

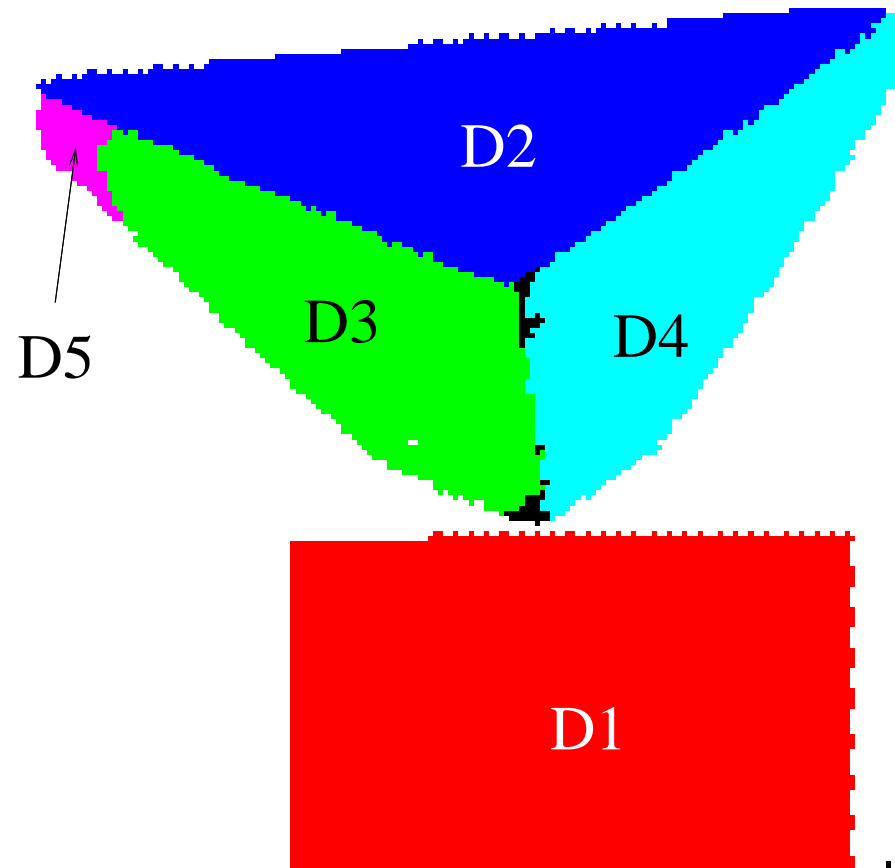
Plane equation is now:  $\vec{y}_i'\vec{p} = 0$

Least squared error:

$$\sum_i (\vec{y}_i'\vec{p})^2 = \sum_i \vec{p}'\vec{y}_i\vec{y}_i'\vec{p} = \vec{p}'(\sum_i \vec{y}_i\vec{y}_i')\vec{p} = \vec{p}'M\vec{p}$$

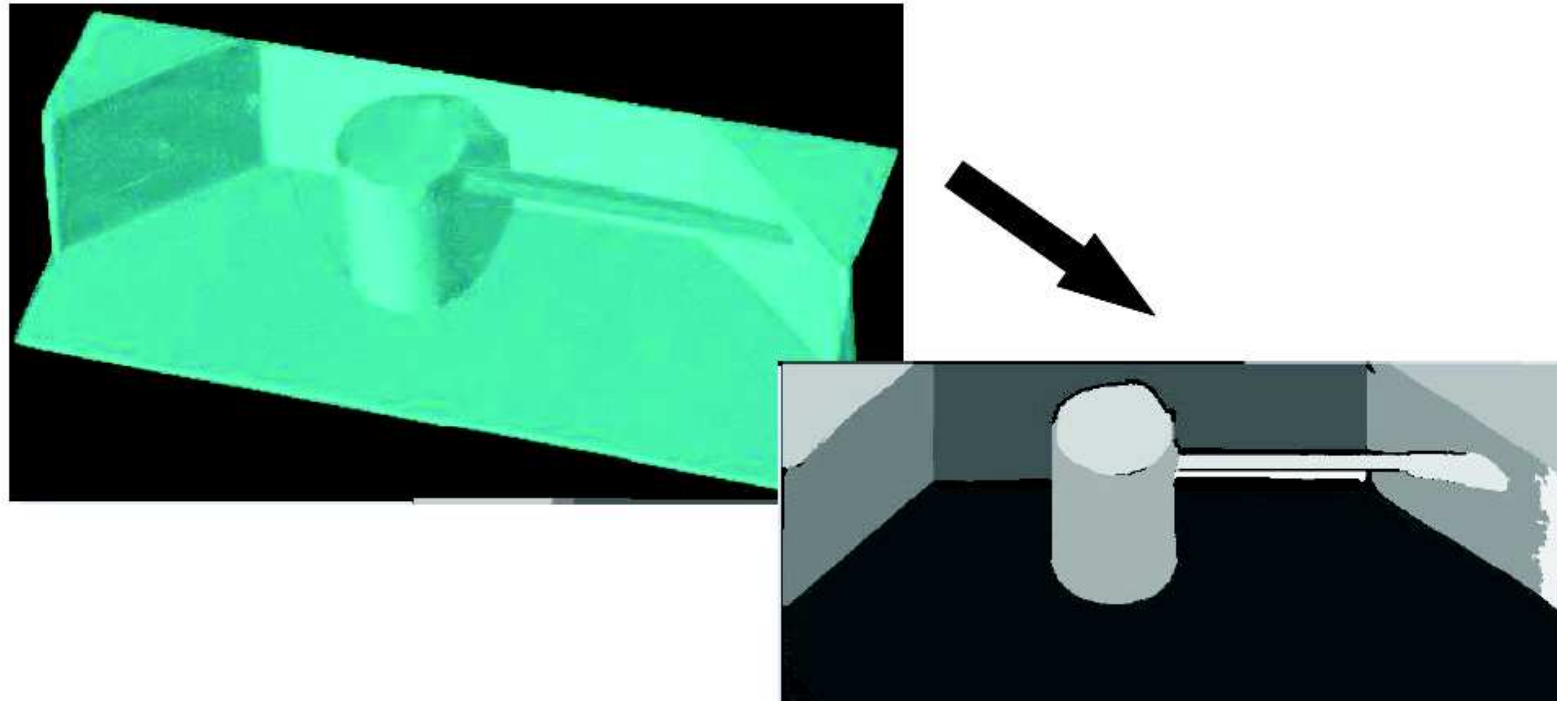
Eigenvector of smallest eigenvalue of  $M$  is desired parameter vector, provided eigenvalue is small.

## Full Segmentation



Could get 4 planes by parameter adjustment,  
but 5 means more data for matching stage

## Extensions



Extend fitting to additional surface types:  
cylinders, spheres, etc

Allows recognition of more complex objects



## What We Have Learned

- A region growing algorithm
- A least squares algorithm for plane parameter estimation
- Some idea of how well it works on relatively clean data