

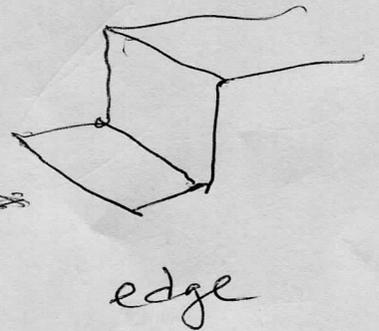
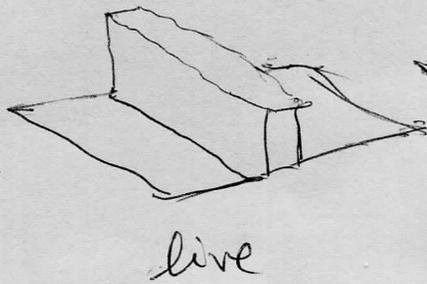
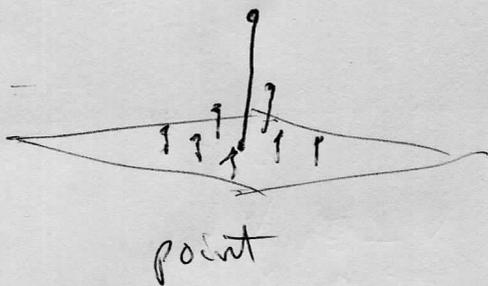
Week 9 15 Oct

segmentation: find consistent semantic regions
maybe connected or not

major approaches:

- * discontinuity : edge , split
- * similarity : region , merge

point, line, edge detection



use differential operators

point : Laplacian

line : Laplacian responds according to width of line

see CS6640 - week 9

edge : + magnitude of gradient

+ canny

Man-Hildreth

(9/2)

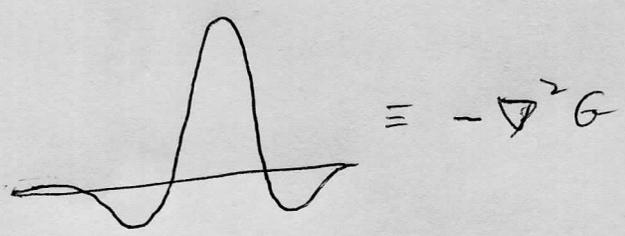
- * get 1st + 2nd deriv
- * tuned to scale

⇒ $\nabla^2 G$ Laplacian of Gaussian

$$\nabla^2 f(x,y) = \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$$

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

$$\Rightarrow \nabla^2 G(x,y) = \left(\frac{x^2+y^2-2\sigma^2}{\sigma^4}\right) e^{-\left(\frac{x^2+y^2}{2\sigma^2}\right)}$$



(1) $g(x,y) = [\nabla^2 G(x,y)] * f(x,y)$

(2) find zero-crossings

can be approximated by difference of Gaussians

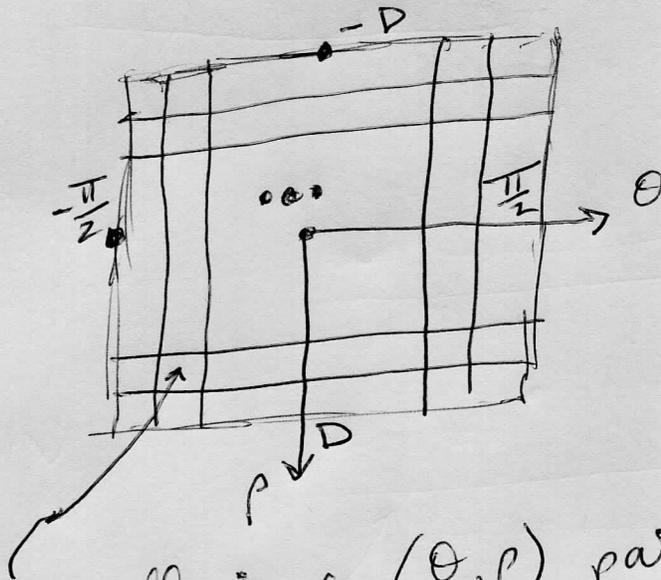
$$D_G(x,y) = \frac{1}{2\pi\sigma_1^2} e^{-\left(\frac{x^2+y^2}{2\sigma_1^2}\right)} - \frac{1}{2\pi\sigma_2^2} e^{-\left(\frac{x^2+y^2}{2\sigma_2^2}\right)}$$

$\sigma_1 > \sigma_2$

same zero crossings if $\sigma^2 = \frac{\sigma_1^2 \sigma_2^2}{\sigma_1^2 - \sigma_2^2} \ln \left(\frac{\sigma_1^2}{\sigma_2^2} \right)$

Consider: ρ - θ space
 create discrete cells

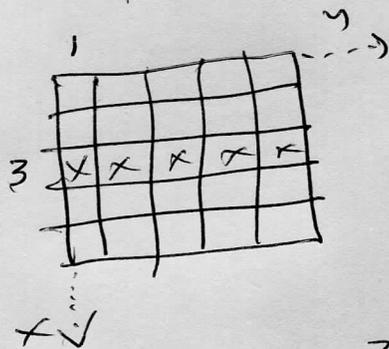
9/4



every cell is a (θ, ρ) pair

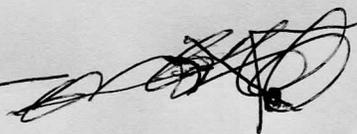
Given an (x, y) edge, every cell (θ, ρ) is incremented for which $\rho = x \cos \theta + y \sin \theta$

Consider:

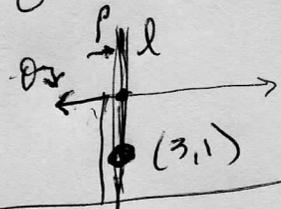


$(x, y) = (3, 1) \Rightarrow \rho = 3 \cos \theta + \sin \theta$

θ	ρ
$-\pi/2$	-1
$-\pi/4$	$\sqrt{2}$
0	3
$\pi/4$	$2\sqrt{2}$
$\pi/2$	1



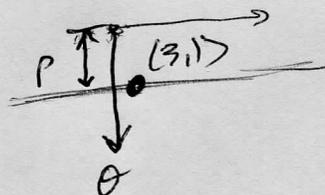
$(-\pi/2, -1)$



$(-\pi/4, \sqrt{2})$



$(0, 3)$



Thresholding

Bimodal if mostly 2 classes
find value between them

Otsu's method : maximize between class variance

```
t = graythresh(im);
iuit = imbinarize(im, t);
```

can use edge or zero-crossing locations to sample about equally from objects and background

Variable thresholding

Region Growing; + Merging + Splitting

need strong similarity measure

Clustering : kmeans

we've seen this in A2 + A3

Simple Linear Iterative Clustering (SLIC)

uses pixel location as well

```
[L, NL] = superpixel(d, 40);
```

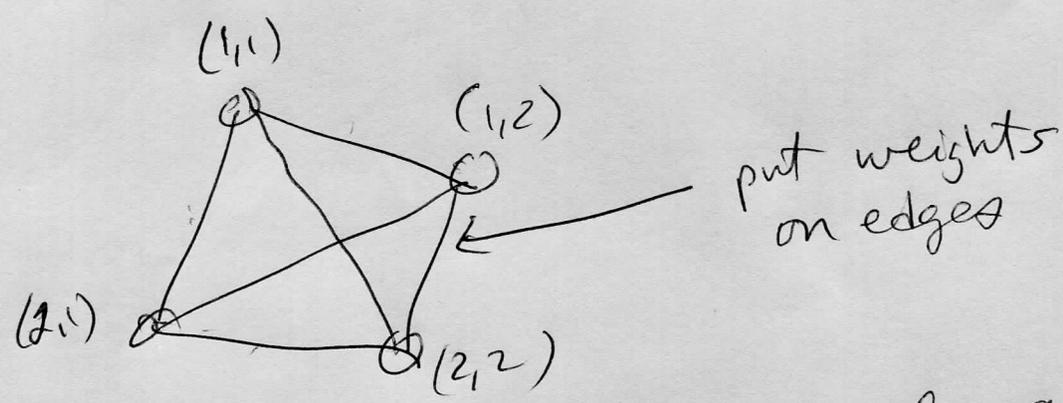
```
BW = boundarymask(L);
```

```
imshow(imoverlay(d, BW, 'red'));
```

Graphs + Graph Cut

Consider pixels as nodes in a graph
edges connect all pixels
called a K-graph (complete graph)

2x2 image \Rightarrow 4 nodes +



weights indicate how similar pixels are
(e.g., gray levels or measures based
on a window centered at pixel)

standard: $sim(pixel_1, pixel_2) = |g_1 - g_2|$

where g_1, g_2 are their gray levels

also: $sim(pixel_1, pixel_2) = \exp(-|g_1 - g_2|)$

$W \equiv$ similarity matrix

$w(p_1, p_2) = \exp(-|g_{p_1} - g_{p_2}|)$

Find eigenvalues + eigenvectors of W

Find 2 classes (use k -means)

Our object is in smaller of two classes

Watershed Variation

for every pixel, determine which neighbor has
end smallest value

if a pixel has the smallest value among its
8-neighbors, then it's a basin origin;

end label them
while some pixel has no label

check every pixel not labeled and if ~~it~~
~~has~~ its lowest value neighbor has a
value, change its value to that

end
find basin with largest spread,
& return that.