

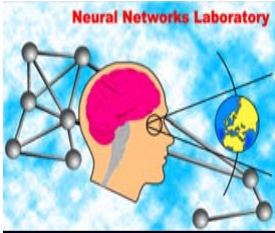
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Department Of Applied Informatics and Multimedia  
Intelligent Systems Laboratory



# Unsupervised Texture Segmentation Via Adaptive Gabor Filters

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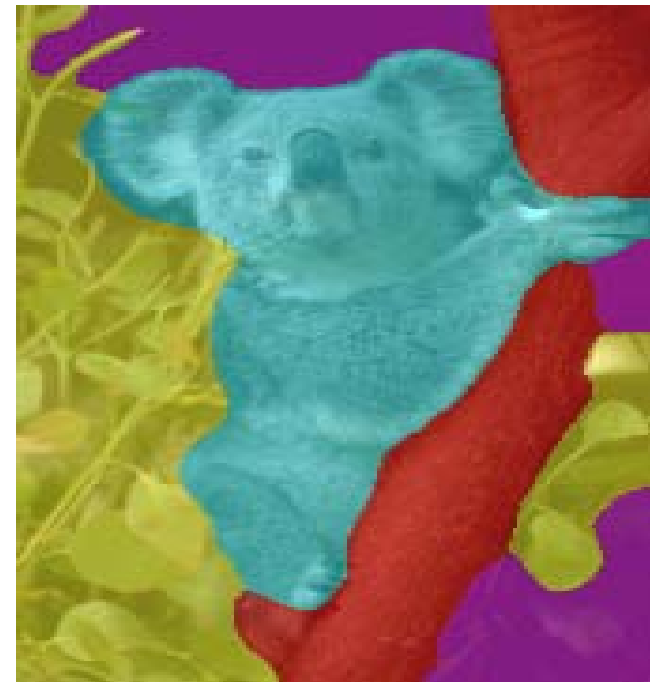


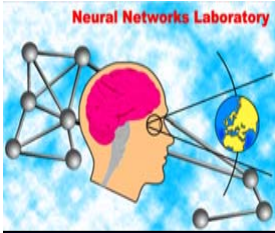
# Image Segmentation

- ❑ Partitioning an image into meaningful regions
- ❑ Necessary to reduce the amount of information
- ❑ Compact representation

*But, what is a meaningful region?*

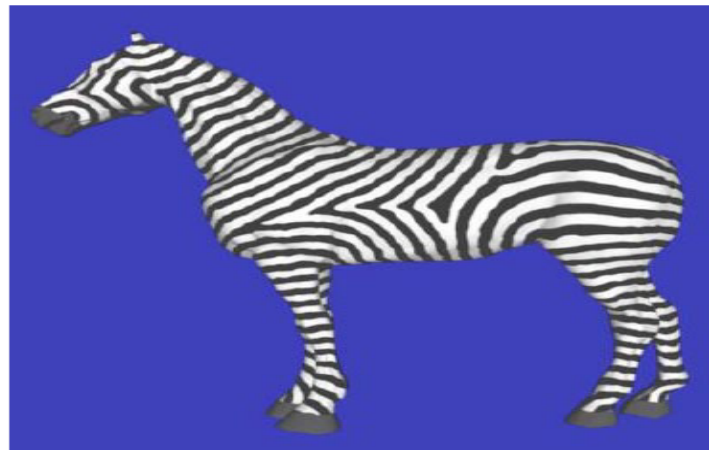
Good representation means:  
The content of regions should  
be as similar as possible  
content of different regions  
should be as dissimilar as  
possible



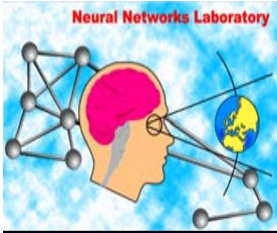


# What is texture ?

- ❑ No accurate definition.
- ❑ Often used to represent the “flavor” or “nuance” of the image.
- ❑ In our case: pixel arrangements with some kind of “structure”.

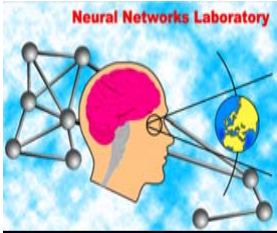


*Shapes are textured regions ?*

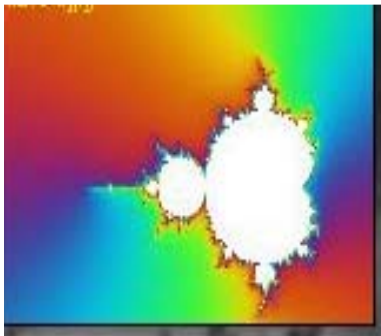
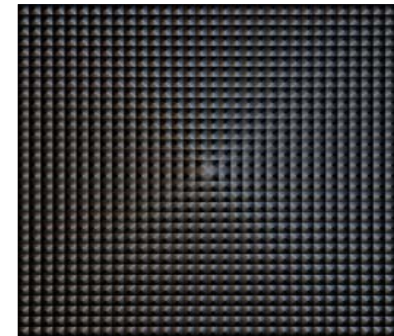
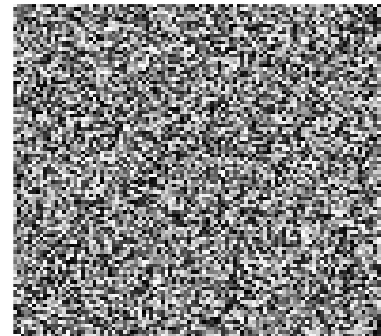
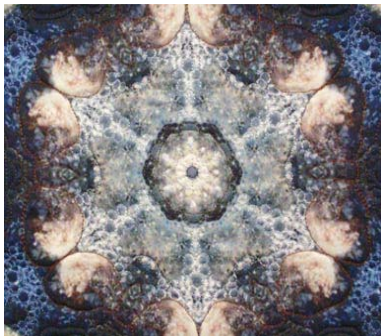


# The Problem: Texture Similarity

$$S(\text{[cobblestone texture]}, \text{[pebbles texture]}) = ?$$



# What structure?

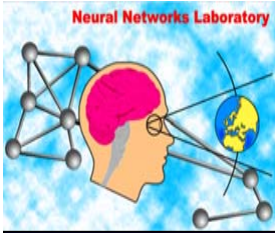


↓  
Fractal?

↓  
Both

↓  
Random

↓  
Repetition

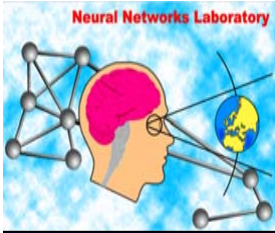


# Texture Similarity based on Response Statistics

- Collect statistics of responses over a small subimage
- Calculate distance metrics between vectors of response statistics

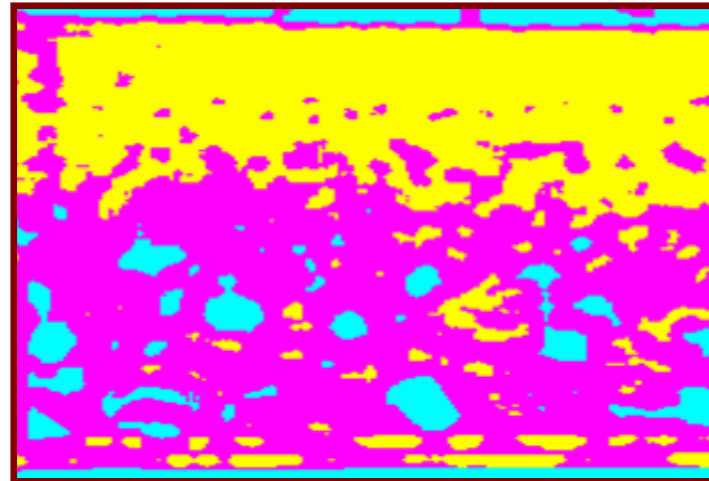
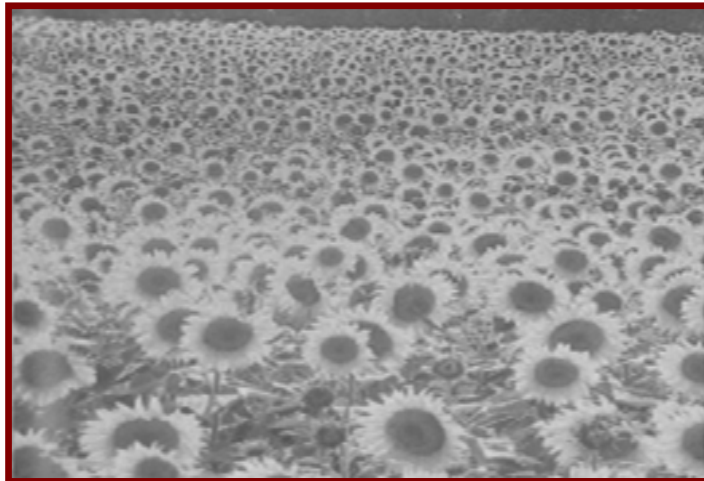
241	29	241	231	29
145	29	29	27	28
23	241	24	96	27
24	33	27	56	159
174	29	106	27	116

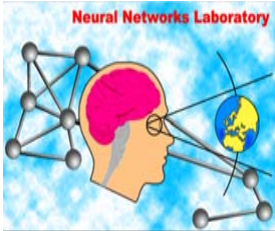
Mean: 21.4



# What is the optimal window size?

- ❑ A critical question with all the statistical texture methods.
- ❑ we don't know yet!





# Statistics of responses

Average Intensity

$$m = \sum_{i=1}^{L-1} z_i p(z_i)$$

Average Contrast

$$\sigma = \sqrt{\mu_2(z)}$$

Smoothness

$$R = 1 - 1/(1 + \sigma^2)$$

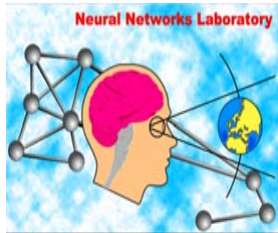
Uniformity

$$U = \sum_{i=0}^{L-1} p^2(z_i)$$

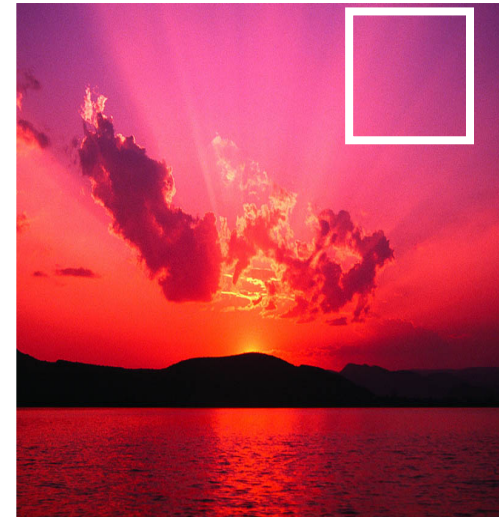
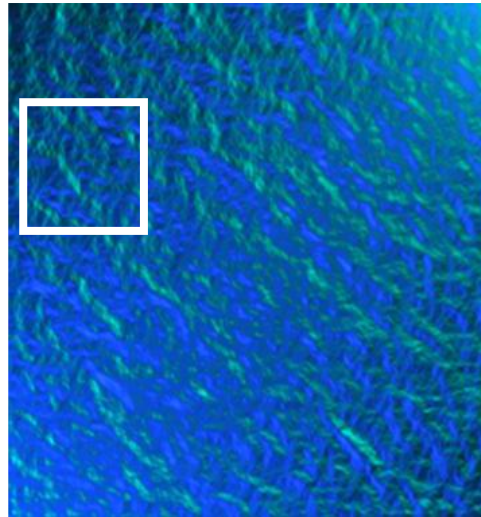
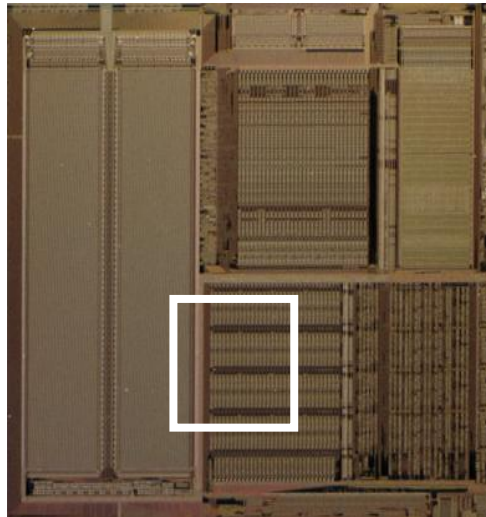
Entropy

$$e = - \sum_{i=0}^{L-1} p(z_i) \log_2 p(z_i)$$

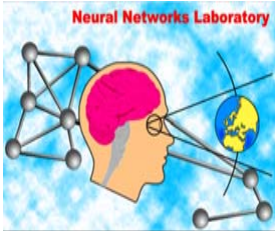




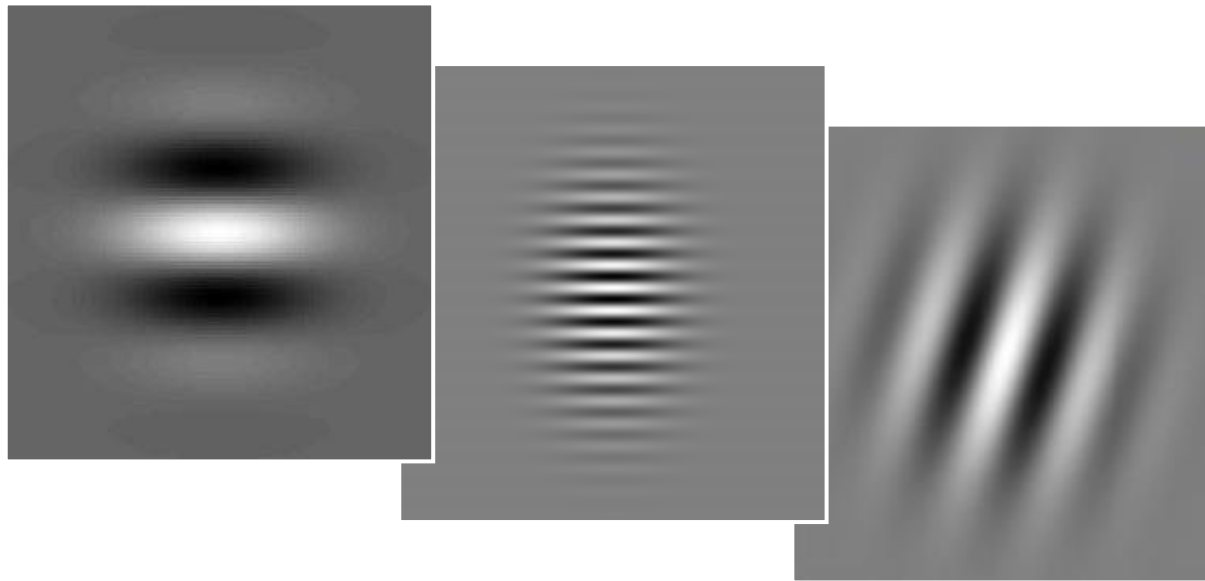
# Texture Signatures



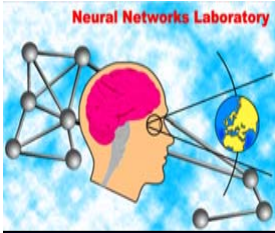
Texture	Average Intensity	Average Contrast	Smoothness	Uniformity	Entropy
Smooth	67.34	14.21	0.006	0.032	5.89
Coarse	127.04	84.18	0.055	0.007	8.64
Periodic	104.31	48.79	0.021	0.019	6.35



# Texture Representation: Filter Responses

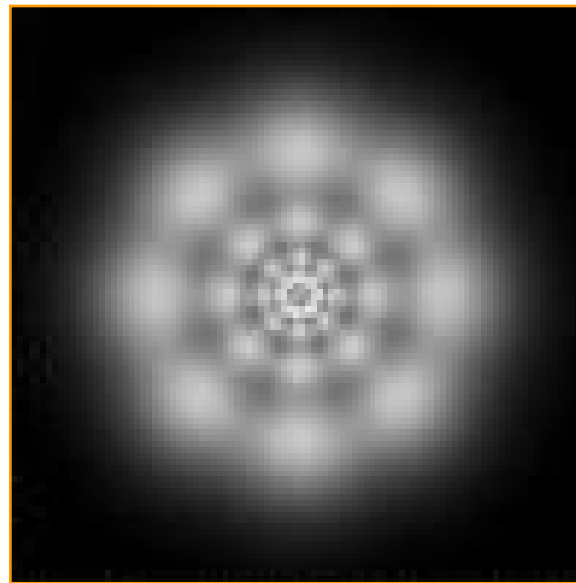


- ❑ Choose a group of Band-pass Gabor filters at different orientations, scales
- ❑ Run filters over image to get a set of *response images*  
*Each contains specific texture information*

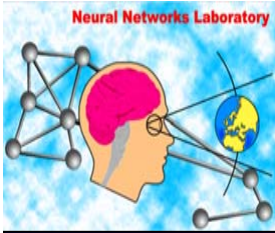


predefined tessellation of the frequency plane, consisting of overlapping filters whose centre frequencies lie on concentric circles, logarithmically spaced, centred at the origin.

This approach can lead to disadvantages since a large number of filtered images are involved and a large dimensional feature space needs to be processed

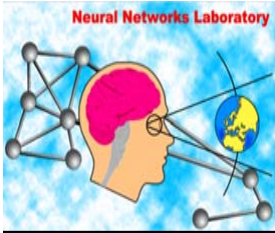


Daisy shaped filter kernels (frequency domain)

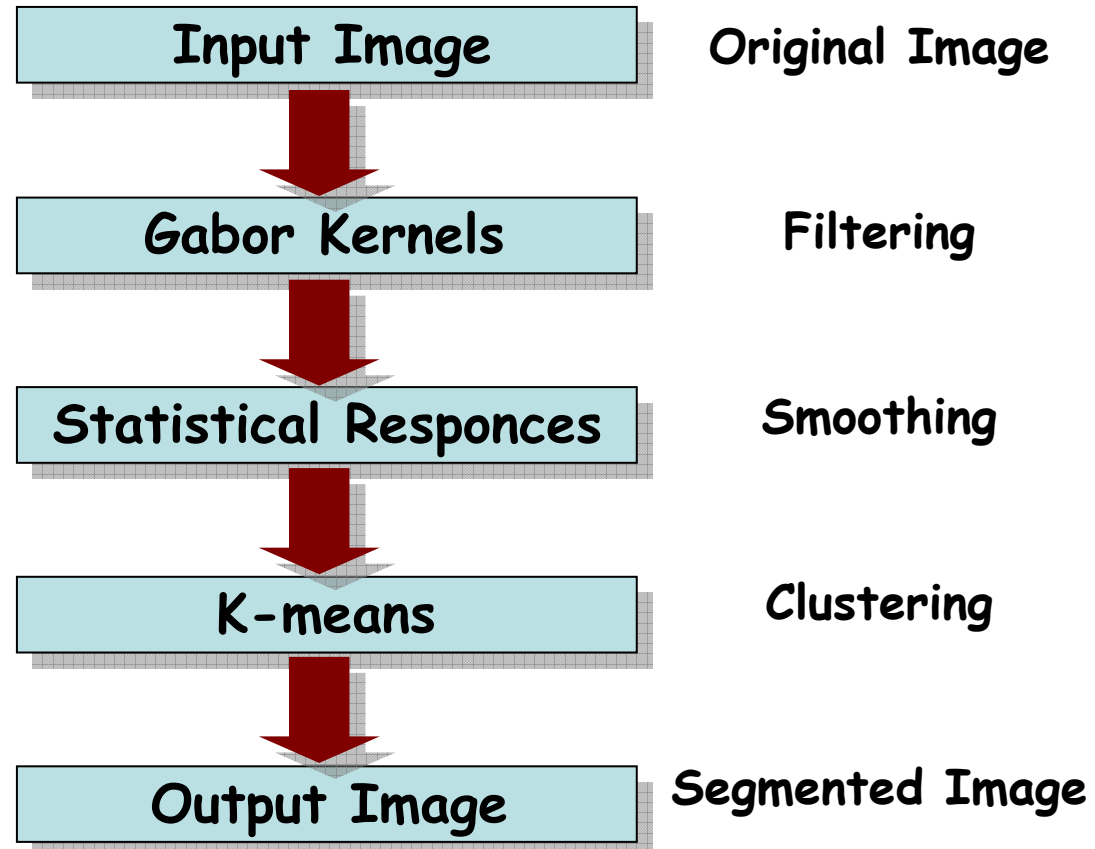


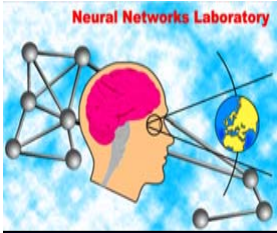
# Biological Motivation

- ❑ Later studies on human vision shows that the retina and brain have receptive fields (filters) sensitive to different spatial frequencies, at a variety of scales and translations within a region of the retina, known as the complex cells.
- ❑ Gabor analysis represents images in a way somewhat similar to complex cells

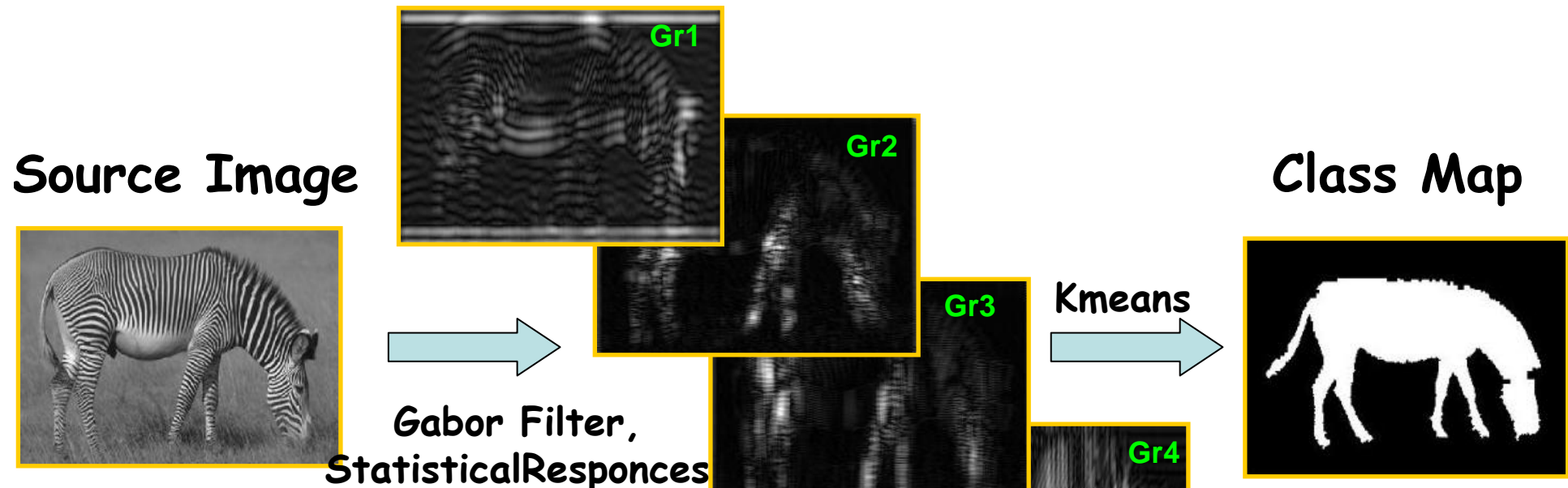


# Pixel Based Texture Classification

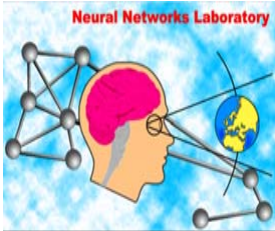




# Pixel Classification



Pixel Signatures:  $[Gr1_{ij}, GR2_{ij}, GR3_{ij}, GR4_{ij}]$



# Optimization With GAs

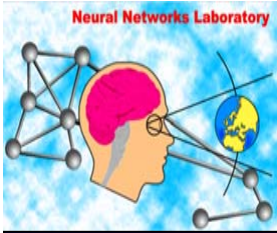
## Gabor Filter Kernel Equation

$$g(x, y) = \frac{1}{2\pi\sigma_x\sigma_y} \exp \left[ -\frac{1}{2} \left\{ \left( \frac{x}{\sigma_x} \right)^2 + \left( \frac{y}{\sigma_y} \right)^2 \right\} + 2\pi j (Ux + Vy) \right]$$

[Sx Sy Ux Vy]

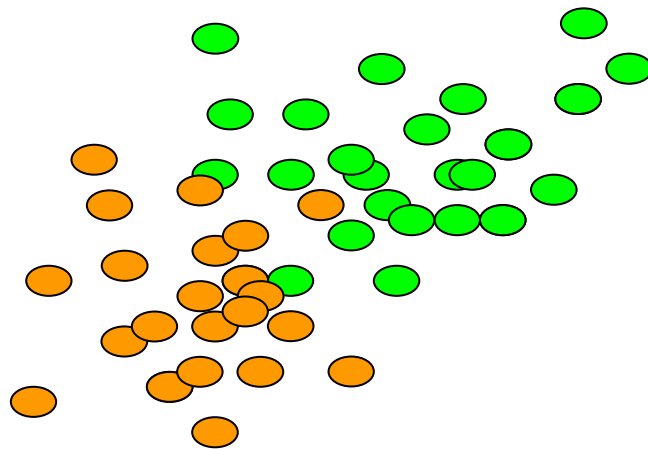
[Sx1 Sy1 Ux1 Vy1; Sx2 Sy2 Ux2 Vy2; Sx3 Sy3 Ux3 Vy3]

Single Genome corresponding to three Gabor Kernels

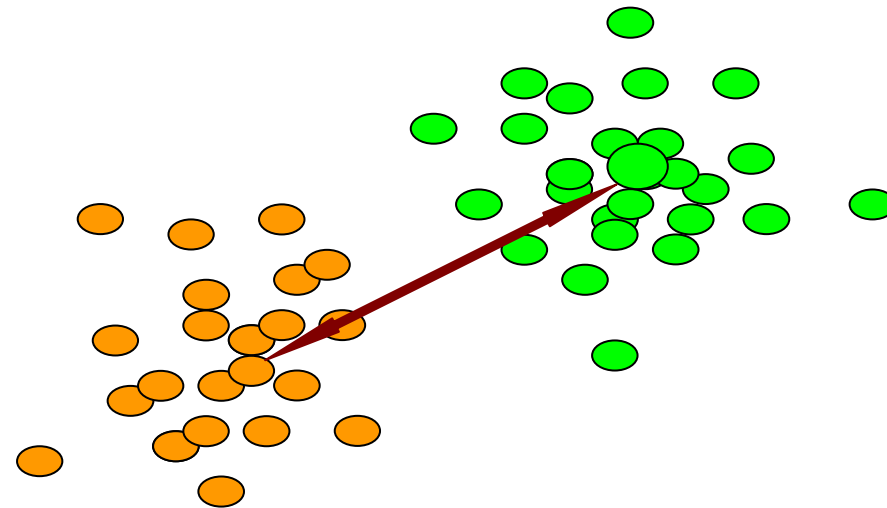


# Fitness Function

- ❑ Try to maximize the distance between clusters  
*Dissimilar content between different regions*
- ❑ Try to minimize distances between members of each cluster  
*Similar content of the same region*

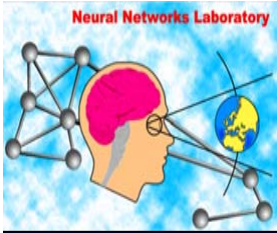


Overlapping Clusters

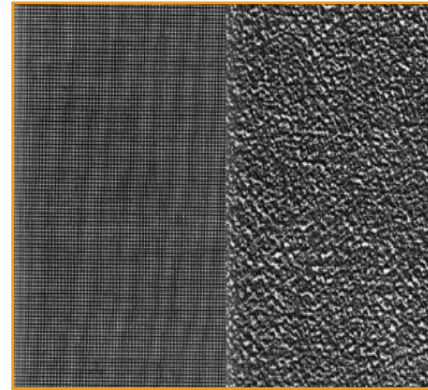


Non Overlapping Clusters





# Evolution of filter Kernels



Input Image



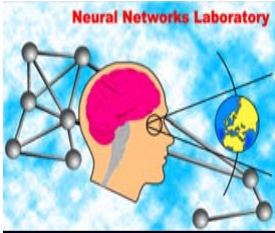
a. Initial Random Genome



b. after 20 generations



c. after 100 generations



## Conclusions

- ❑ Capturing behaviors of biological systems can lead to improved performance
- ❑ Evolutionary computation, offers solutions in difficult real world problems
- ❑ Conceptually simple procedure

## Future Work

- ❑ Optimization of statistical responses window size
- ❑ Feature selection will also be investigated,
- ❑ Dimensionality reduction of the feature vectors
- ❑ Different clustering algorithms