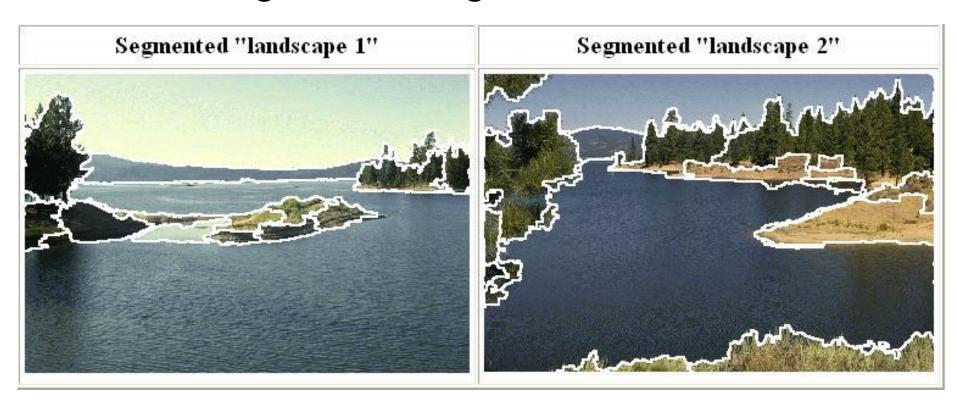
# اصول پردازش تصویر Principles of Image Processing

مصطفی کمالی تبریزی ۱۲ آبان ۱۳۹۹ جلسه سیزدهم

#### Mean-Shift

#### Mean-Shift

 An advanced and versatile technique for clustering-based segmentation



D. Comaniciu and P. Meer, <u>Mean Shift: A Robust Approach toward Feature Space</u> <u>Analysis</u>, PAMI 2002.

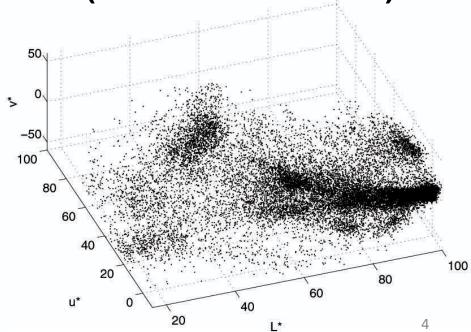
# Mean-Shift Algorithm

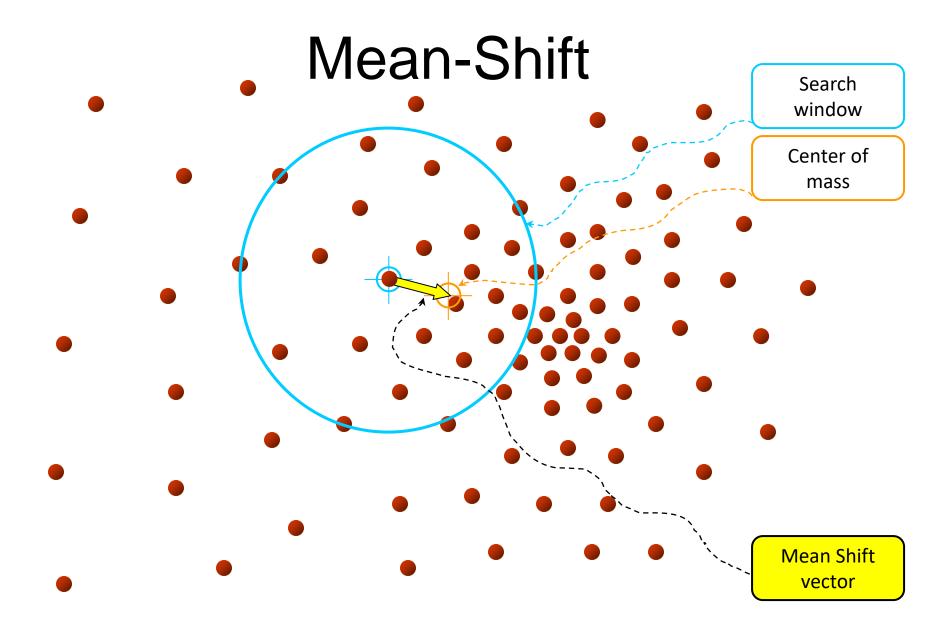
 The mean-shift algorithm seeks modes or local maxima of density in the feature space

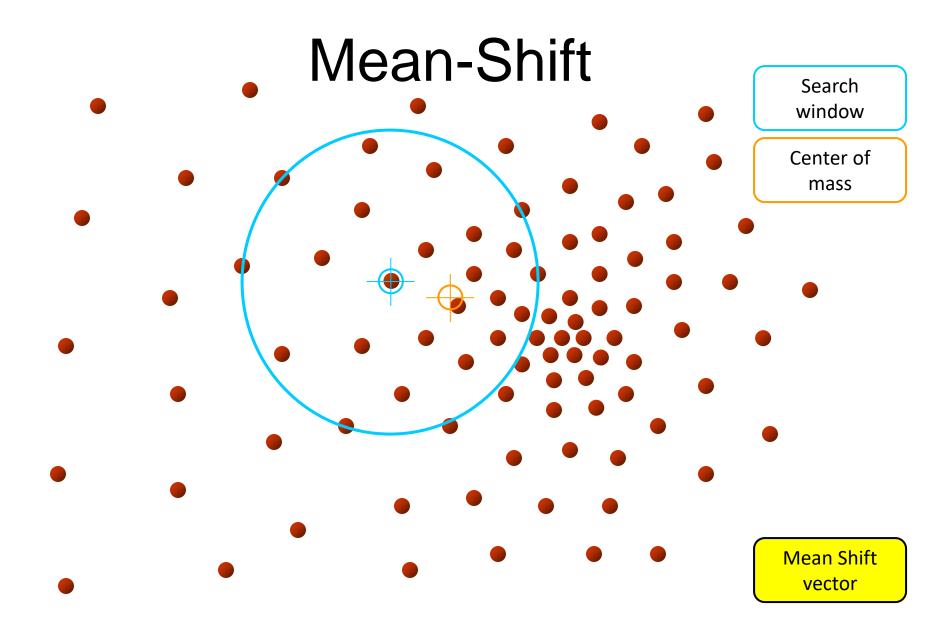
image

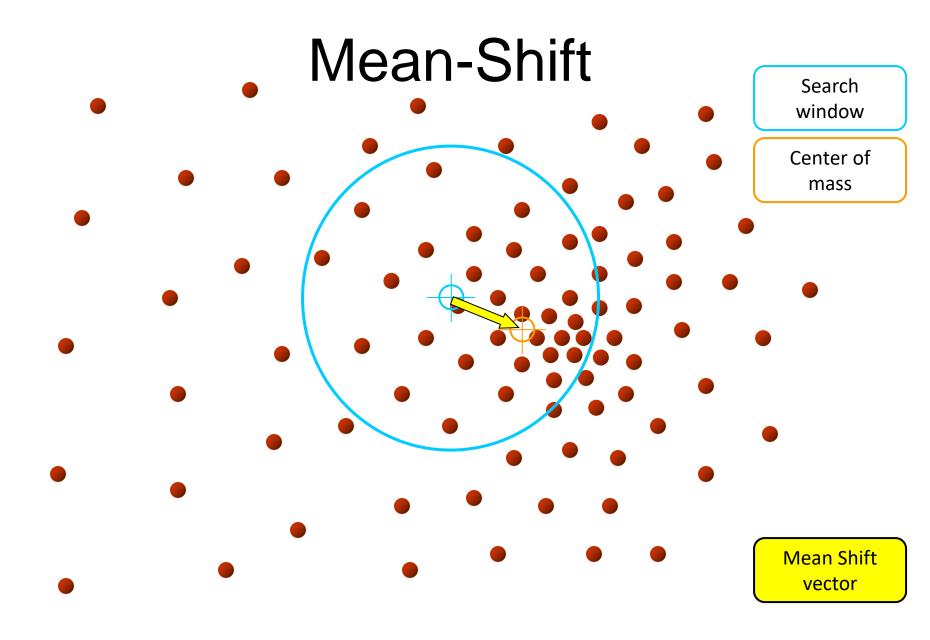


Feature space (L\*u\*v\* color values)

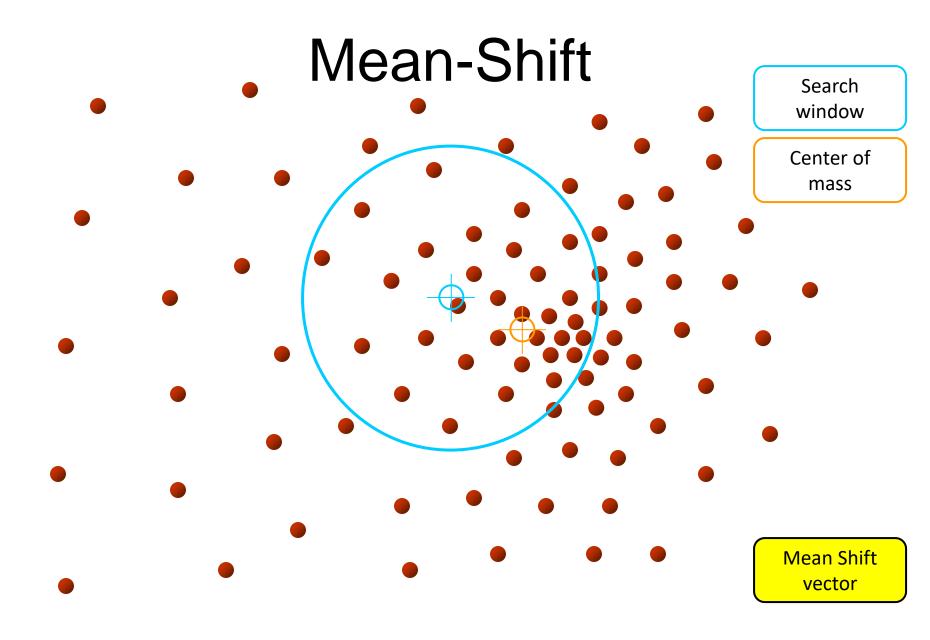


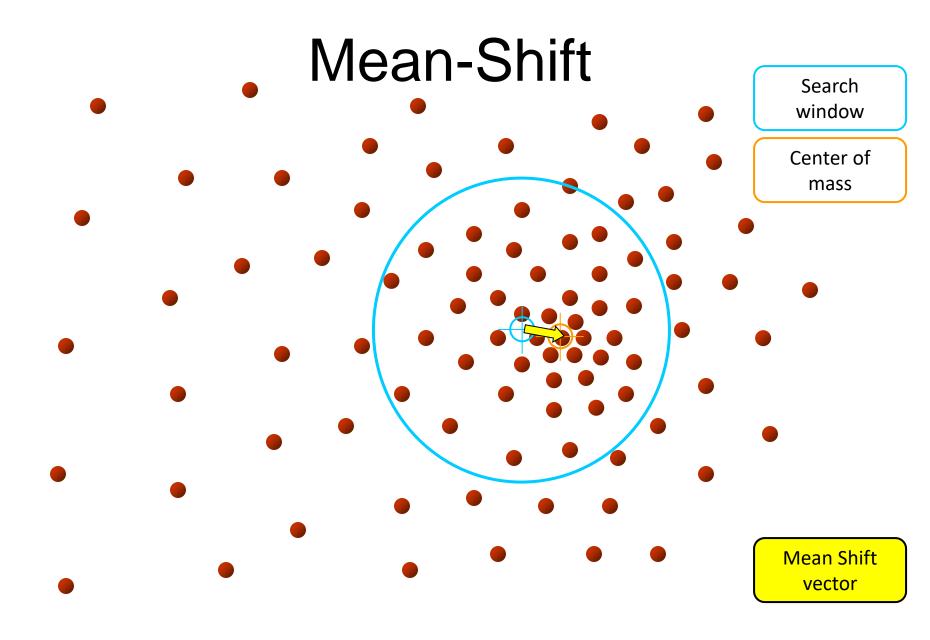


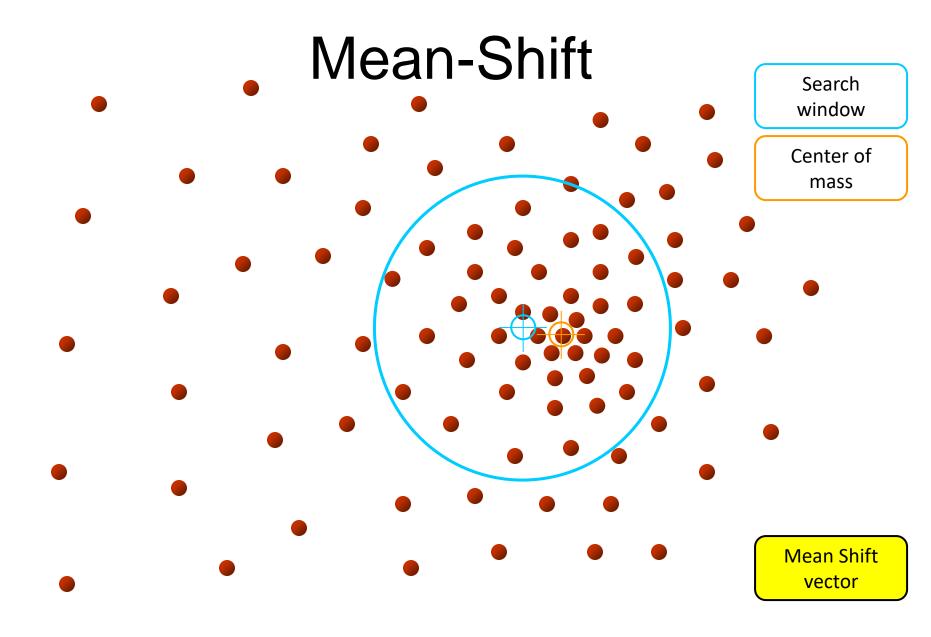




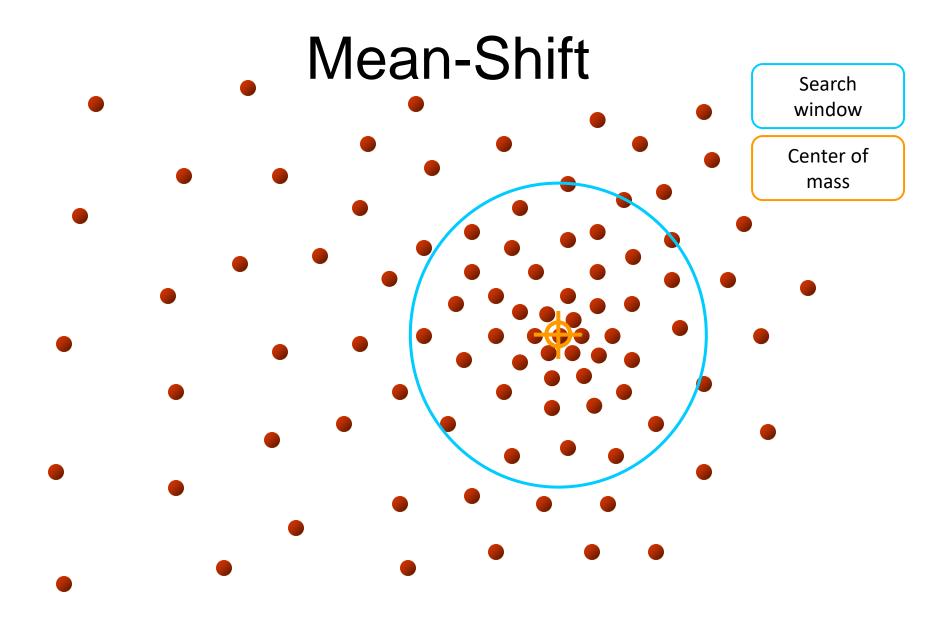
7





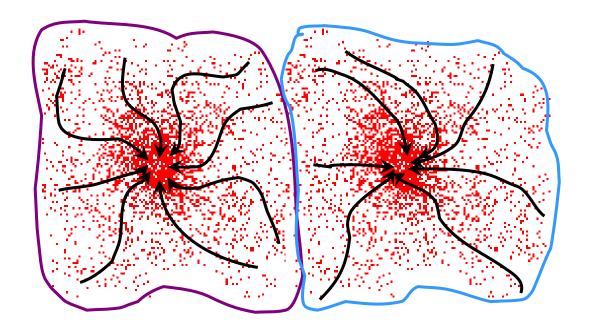


10



# Mean-Shift Clustering

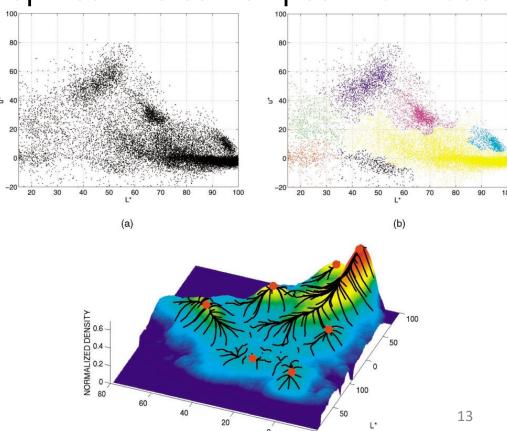
- Cluster: all data points in the attraction basin of a mode
- Attraction basin: the region for which all trajectories lead to the same mode



#### Mean-Shift Clustering/Segmentation

- Find features (color, gradients, texture, etc)
- Initialize windows at individual feature points
- Perform mean-shift for each window until convergence
- Merge windows that end up near the same "peak" or mode





# Mean-Shift Segmentation Results









### More Results

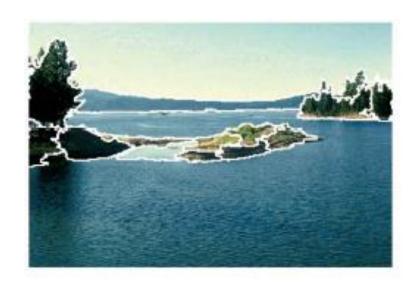








#### More Results









#### Mean-Shift Pros and Cons

#### Pros

- Does not assume spherical clusters
- Just a single parameter (window size)
- Finds variable number of modes
- Robust to outliers

#### Cons

- Output depends on window size
- Computationally expensive
- Does not scale well with dimension of feature space

#### References

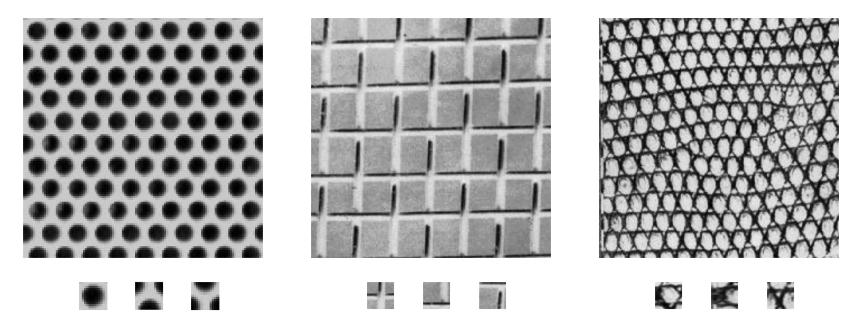
Mean-Shift
 Paper: "Mean Shift: A Robust Approach toward
 Feature Space Analysis", D. Comaniciu and P.
 Meer, PAMI 2002

#### Texture Based Segmentation

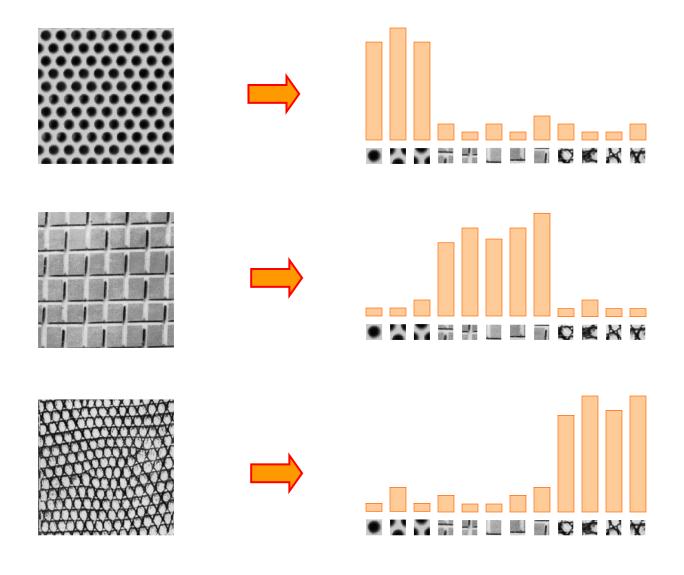
Jitendra Malik, Serge Belongie, Thomas Leung, and Jianbo Shi Contour and Texture Analysis for Image Segmentation International Journal of Computer Vision (IJCV), 2001

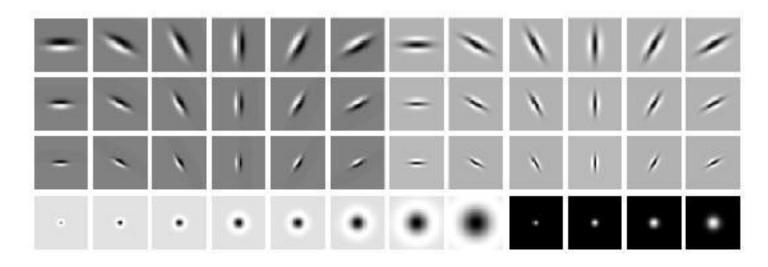
#### **Texture Recognition**

- Texture is characterized by the repetition of basic elements or textons
- For stochastic textures, it is the identity of the textons, not their spatial arrangement, that matters

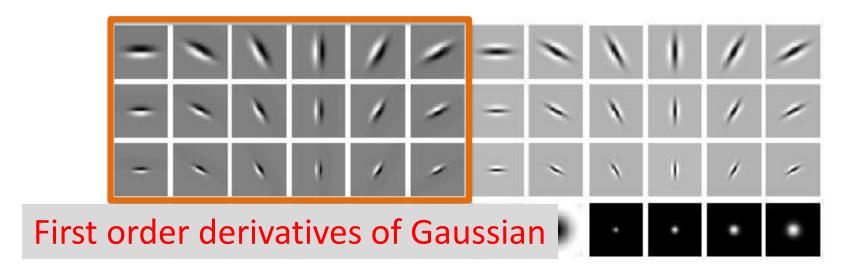


# **Texture Recognition**





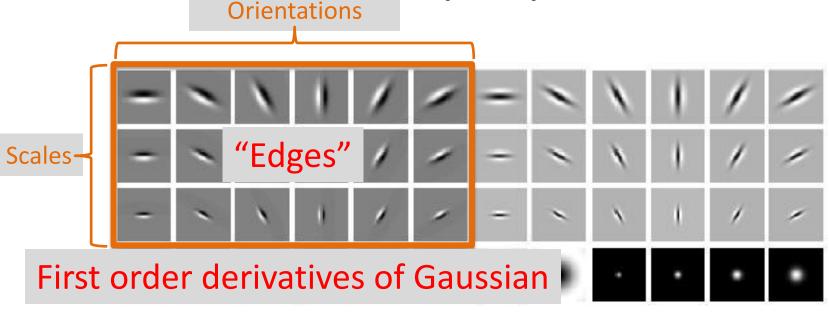
- What filters to put in the bank?
  - Typically we want a combination of scales and orientations, different types of patterns.



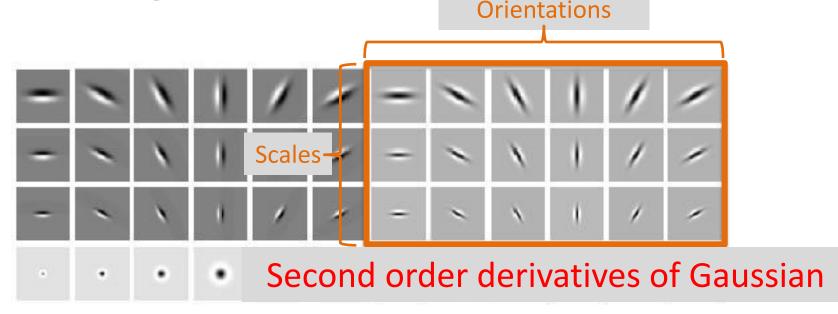
- What filters to put in the bank?
  - Typically we want a combination of scales and orientations, different types of patterns.



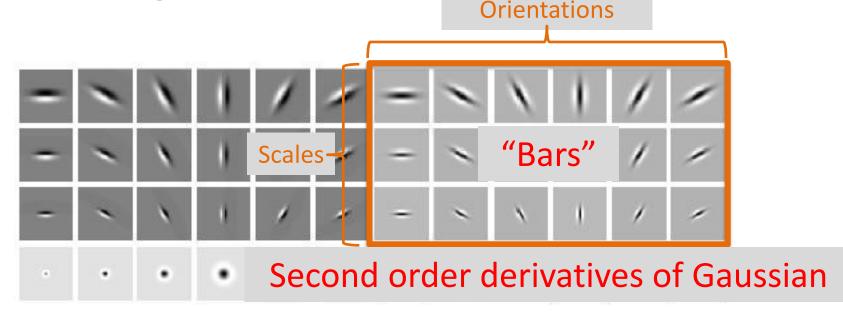
- What filters to put in the bank?
  - Typically we want a combination of scales and orientations, different types of patterns.



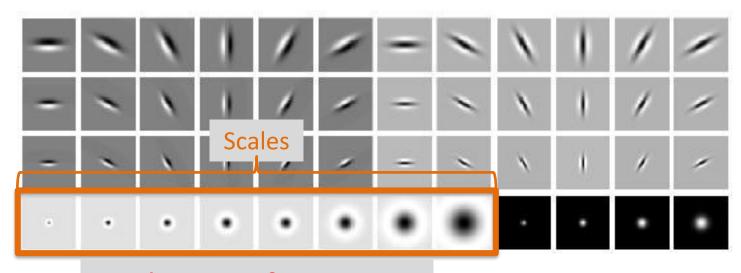
- What filters to put in the bank?
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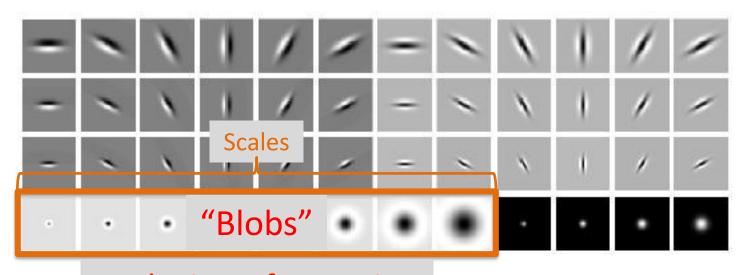
- What filters to put in the bank?
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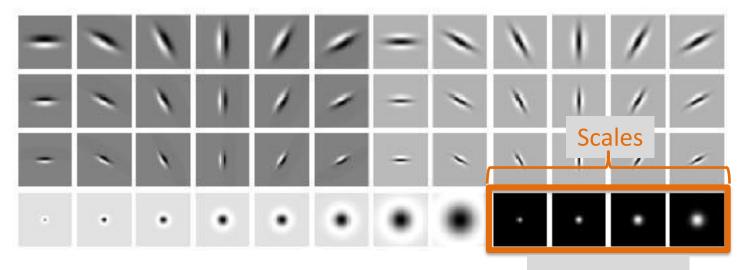
- What filters to put in the bank?
  - Typically we want a combination of scales and orientations, different types of patterns.



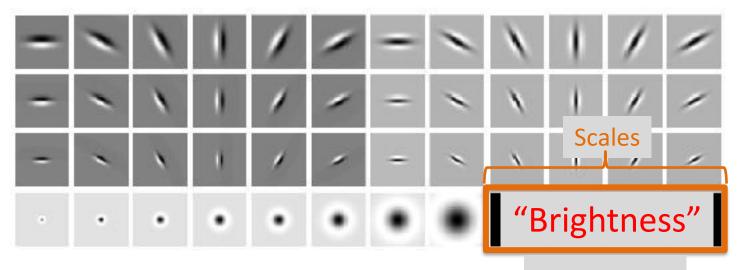
- What Laplacian of Gaussian bank?
  - Typically we want a combination of scales and orientations, different types of patterns.



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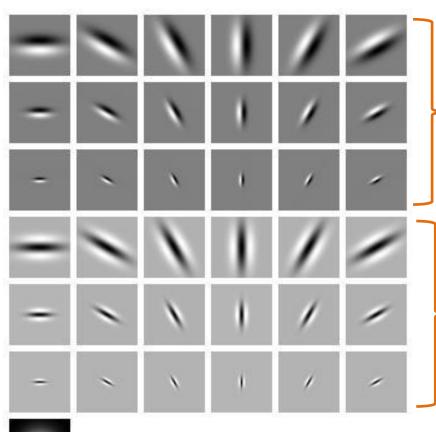


- What filters to put in the bank? Gaussian
  - Typically we want a combination of scales and orientations, different types of patterns.



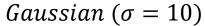
- What filters to put in the bank? Gaussian
  - Typically we want a combination of scales and orientations, different types of patterns.

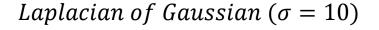
#### Root Filter Set (RFS)



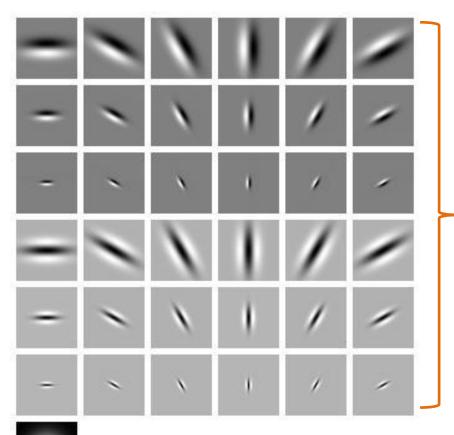
First Derivative of Gaussian (Edge) 3 scales  $(\sigma_x, \sigma_y) \in \{(1,3), (2,6), (4,12)\}$  6 orientations

Second Derivative of Gaussian (Bar)  $3 \text{ scales } (\sigma_x, \sigma_y) \in \{(1,3), (2,6), (4,12)\}$  6 orientations





## Maximum Response 8 (MR8)



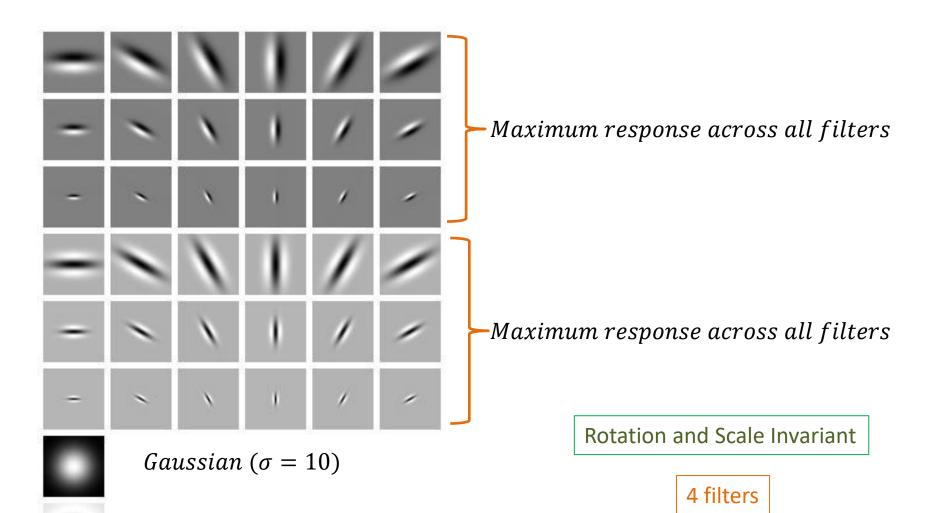
Maximum response across all 6 orientations

*Gaussian* ( $\sigma = 10$ )

*Laplacian of Gaussian* ( $\sigma = 10$ )

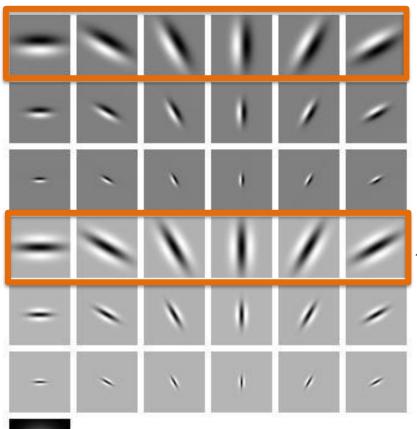
**Rotation Invariant** 

#### MRS4



*Laplacian of Gaussian* ( $\sigma = 10$ )

#### MR4



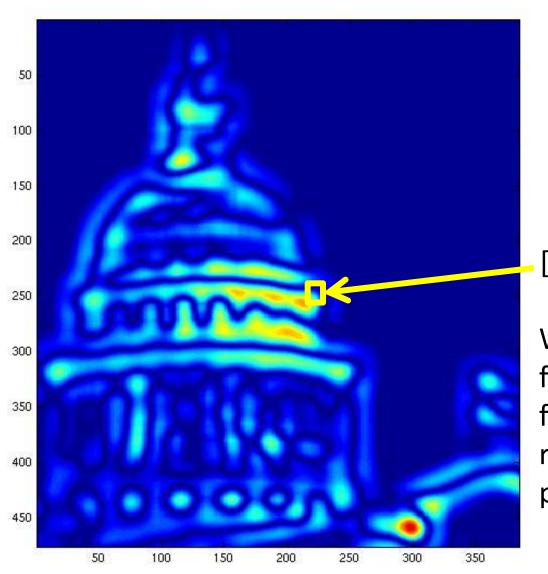
Maximum response

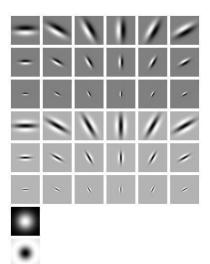
Maximum response

Gaussian ( $\sigma = 10$ )

*Laplacian of Gaussian* ( $\sigma = 10$ )

**Rotation Invariant** 





[r1, r2, ..., r38]

We can form a feature vector from the list of responses at each pixel.