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MEAN-SHIFT BASED POLARSAR IMAGE ENHANCEMENT

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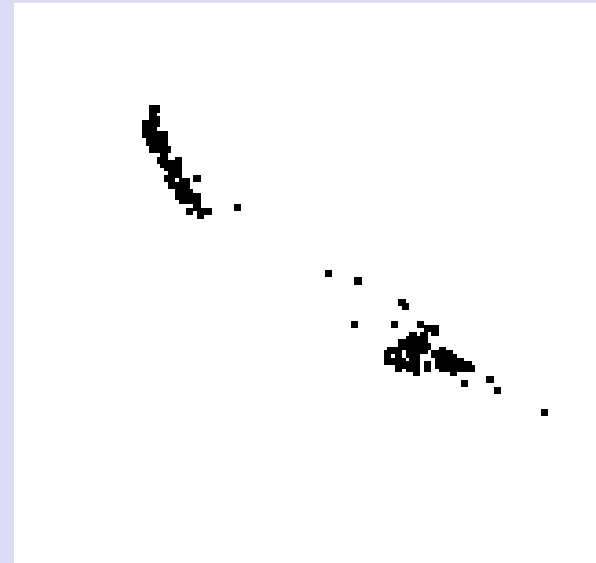
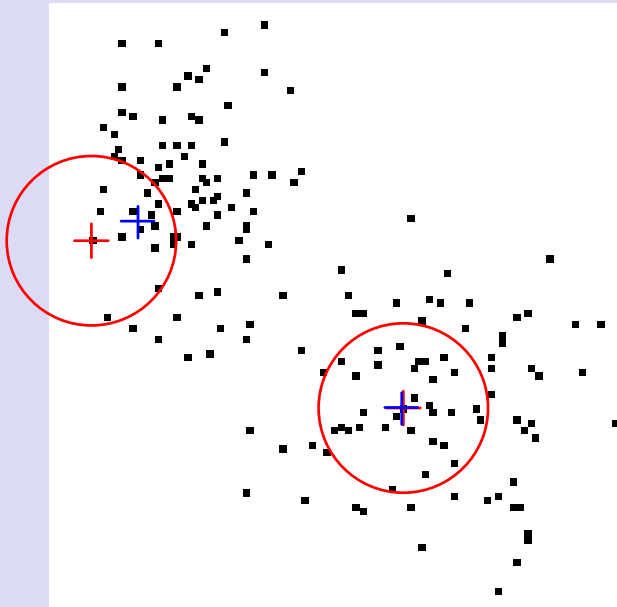
The background of the slide is a complex aerial PolSAR image. It features a grid-like pattern of fields, with a prominent purple patch in the lower right quadrant. The image is overlaid with a light blue, rounded rectangular shape containing text.

MEAN-SHIFT BASED POLSAR IMAGE ENHANCEMENT

- MeanShift Clustering
- Distance measures for PolSAR images
- Tensor spatial attribute

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- **Mean-Shift clustering move every data points toward higher probability density zones (modes)**
- **Density \rightarrow point count over a window (histogram)**
- **Direction toward higher density \rightarrow position of weighted mean (window)**



MEAN-SHIFT

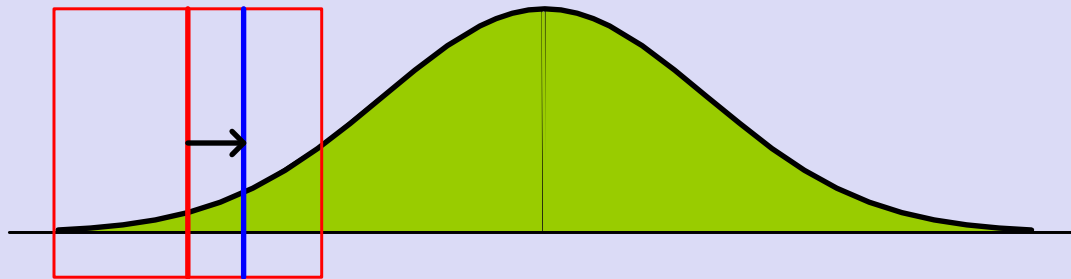
$$D_{rad} = D(Z_i, Z_j)^{1/2} / F_{rad}$$

$$D_{spatial} = \text{Distance between pixels} / F_{spatial}$$

$$\text{Weight} = \text{EXP} [- (D_{rad}^2 + D_{spatial}^2)]$$

Mean = weighted pixel mean

$$F_{shift_R} = \alpha \text{value}_R + (1-\alpha) \text{Mean}_R \quad (\text{radiometric value})$$



- Radiometric distance $D(Z_i, Z_j)$ for PolSar images
- Z_k is pixel covariance matrix
- Non textured PolSar image
- Z_k follows a complex Wishart distribution

$$p(Z_k | \Sigma) = \frac{L^{3L} |Z_k|^{L-3} \exp\{-L \operatorname{tr}(\Sigma^{-1} Z_k)\}}{\pi^3 \Gamma(L) \Gamma(L-1) \Gamma(L-2) |\Sigma|^L}$$

- Log of the likelihood ratio statistic is

$$D(Z_i, Z_j) = 2 \ln \left| \frac{1}{2} (Z_i + Z_j) \right| - \ln |Z_i| - \ln |Z_j|$$

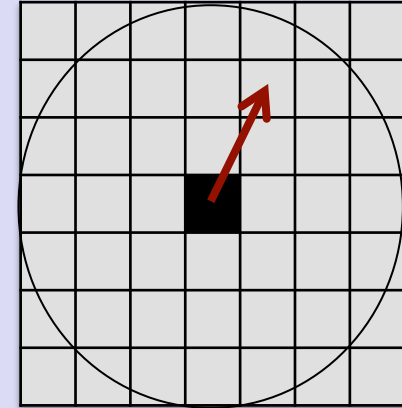
- **Distance between pixels \rightarrow Euclidian distance**

Gaussian like weight ($F_{spatial} = \sigma$)

Weight = EXP [- ($D_{rad}^2 + D_{spatial}^2$)]

Limited to a window (11x11)

$F_{hift}_R = \alpha \text{ value}_R + (1-\alpha) \text{ Mean}_R$



- **Shifting the pixel position**

$F_{hift}_p = \alpha \text{ value}_p + (1-\alpha) \text{ Mean}_p$ (pixel position)

Distances between pixels will change

- **Integrating other distances (texture, shape)**
- **Using weight to define a new attribute**

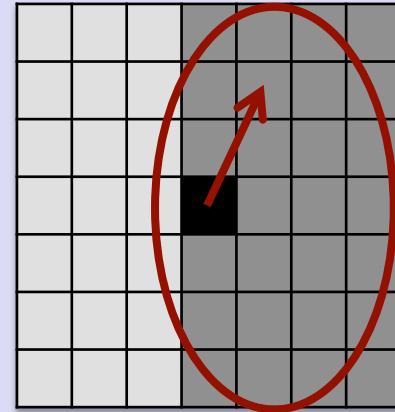
$p_i = (x_i, y_i) \rightarrow$ pixel position

$V_i =$ position covariance or tensor

$$V_i = \sum_j w_{i,j} (p_j - p_i) (p_j - p_i)^t$$

Use V_i ellipse shape (orientation, elongation)

Shape indicate edge orientation



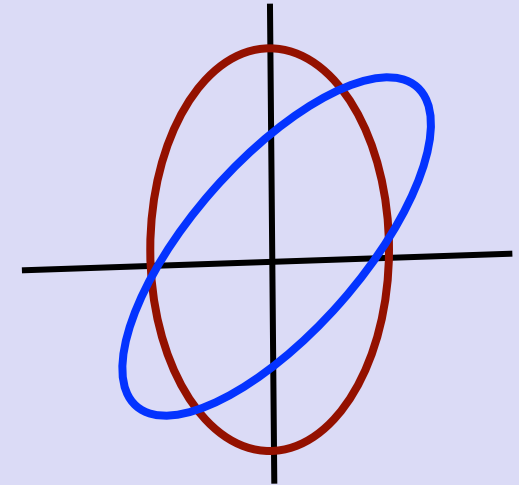
- **Using V_i in weight calculation**

Use S1 measure of Garcia to calculate the difference between V_i and V_j

(BMC Evolutionary Biology 2012, 12:222)

$$D_V = S1(V_i, V_j)^{1/2} / F_V$$

$$\text{Weight} = \text{EXP} [-(D_{rad}^2 + D_{spa}^2 + D_V^2)]$$

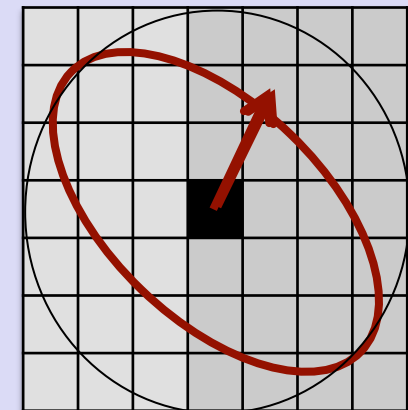


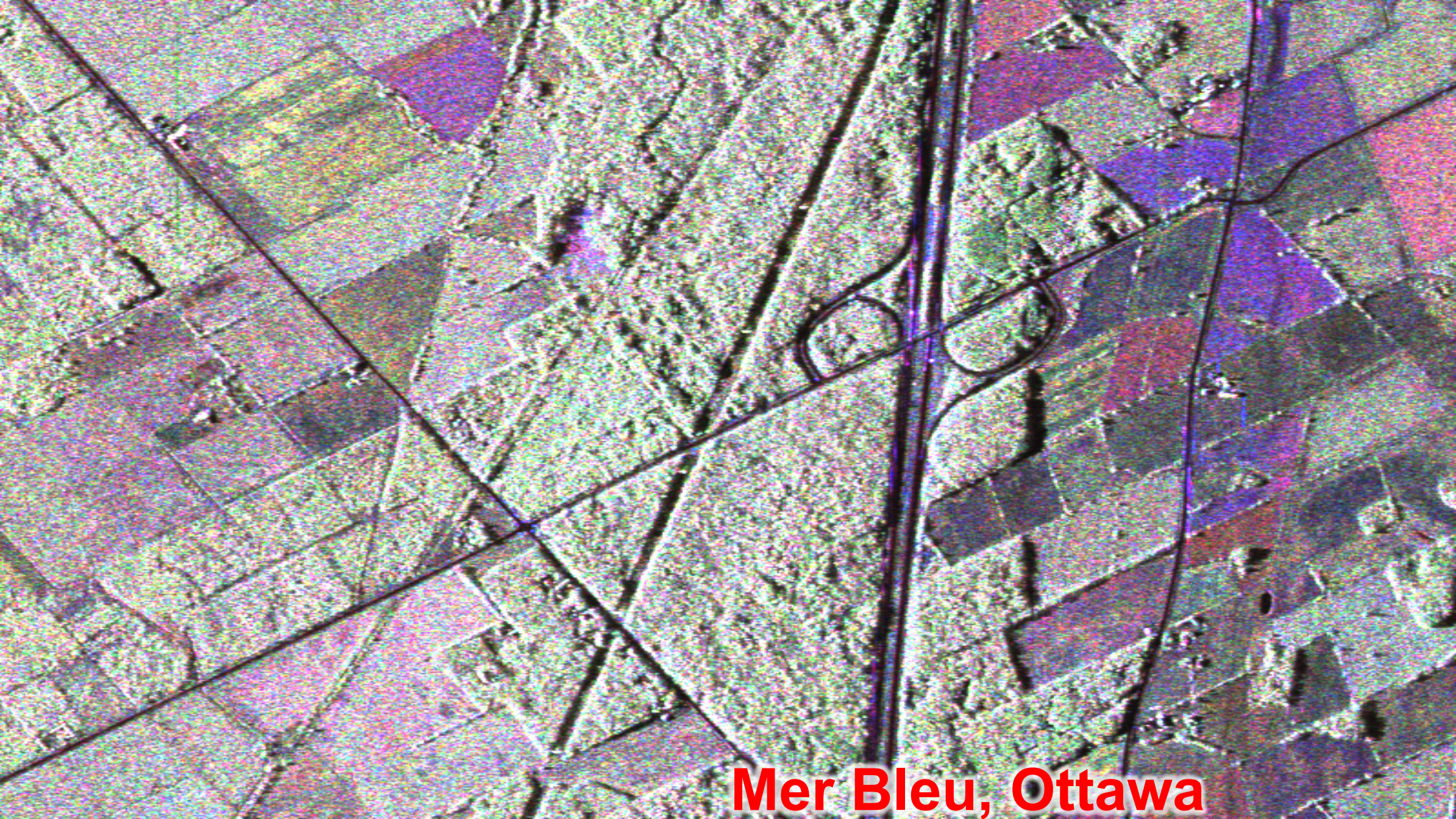
- **Shifting the value of V_i**

$$F_{\text{shift}_V} = \alpha \text{ value}_V + (1-\alpha) \text{ Mean}_V$$

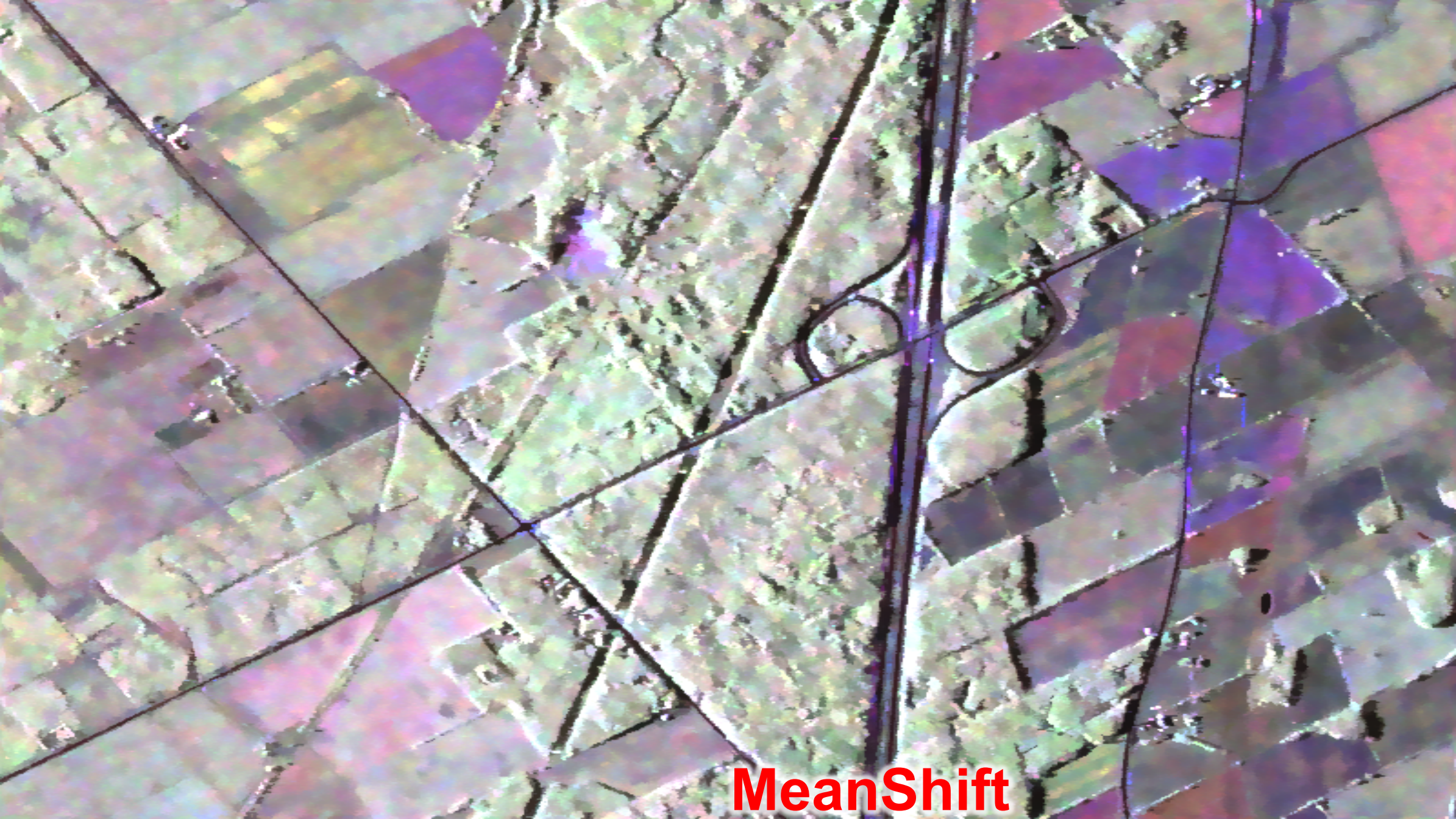
- **Mahalanobis pixel distance**

Use V_i to calculate Mahalanobis pixel distances

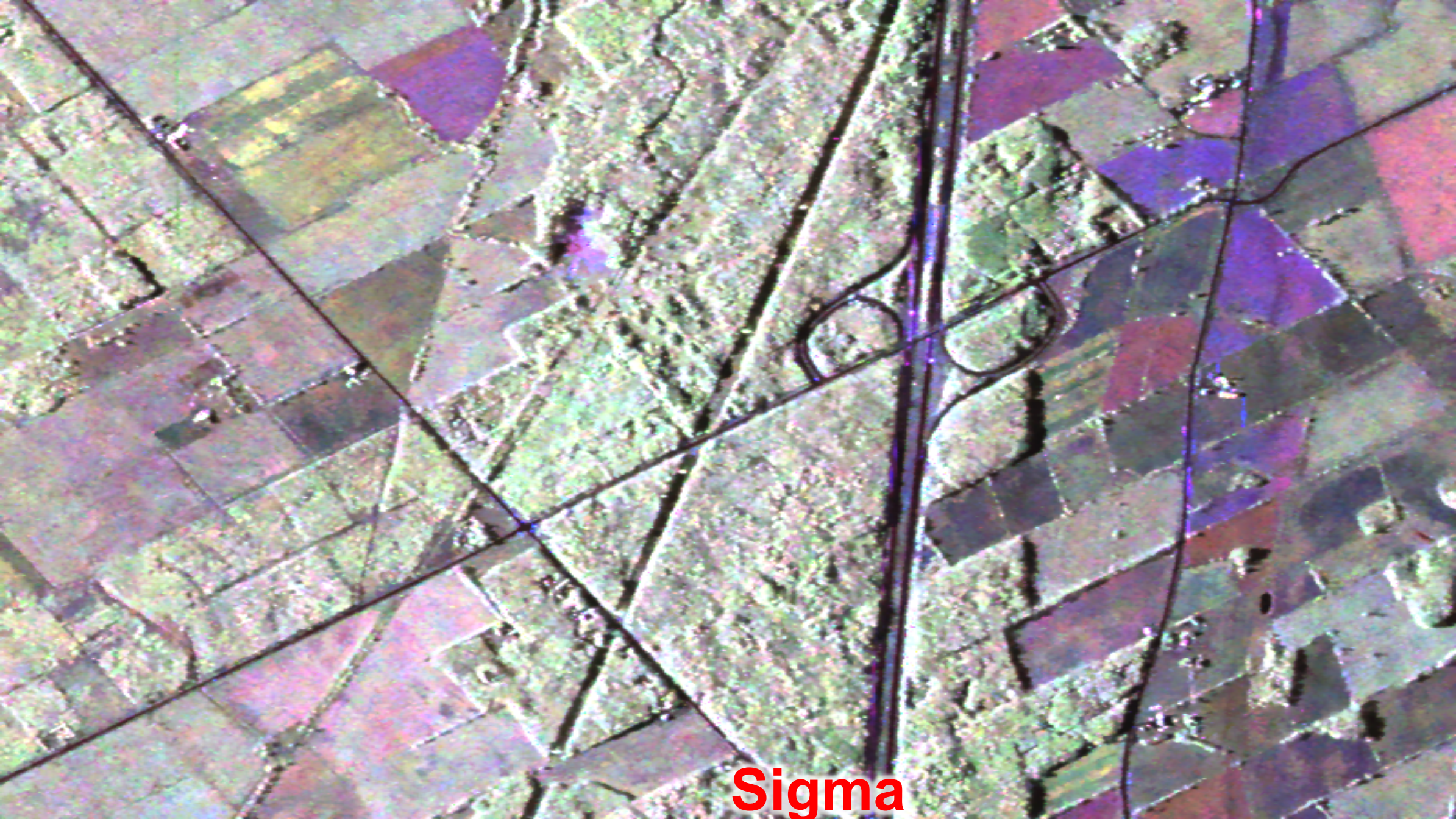




Mer Bleu, Ottawa



MeanShift



Sigma



IDAN

Oberfaffenhofen



MeanShift



Sigma



original



meanshift



sigma



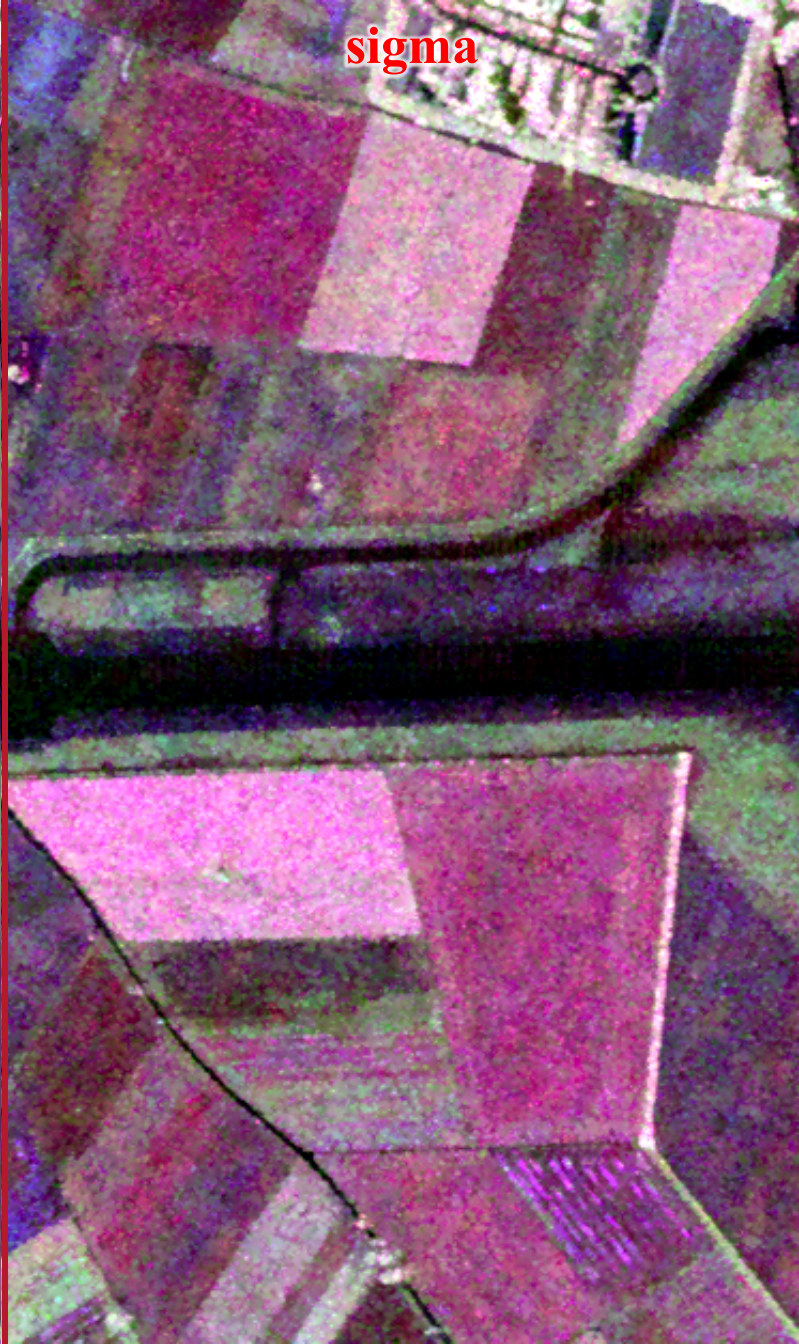
original



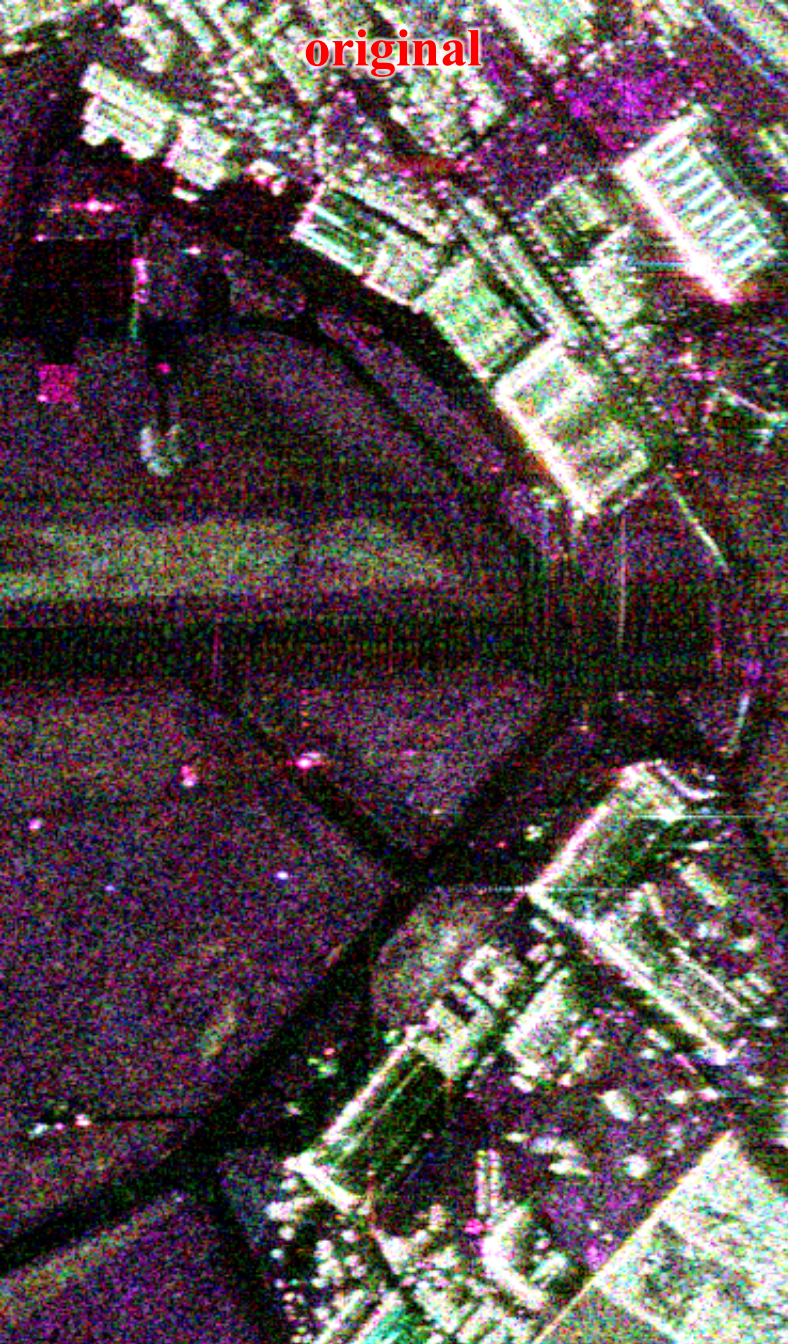
meanshift



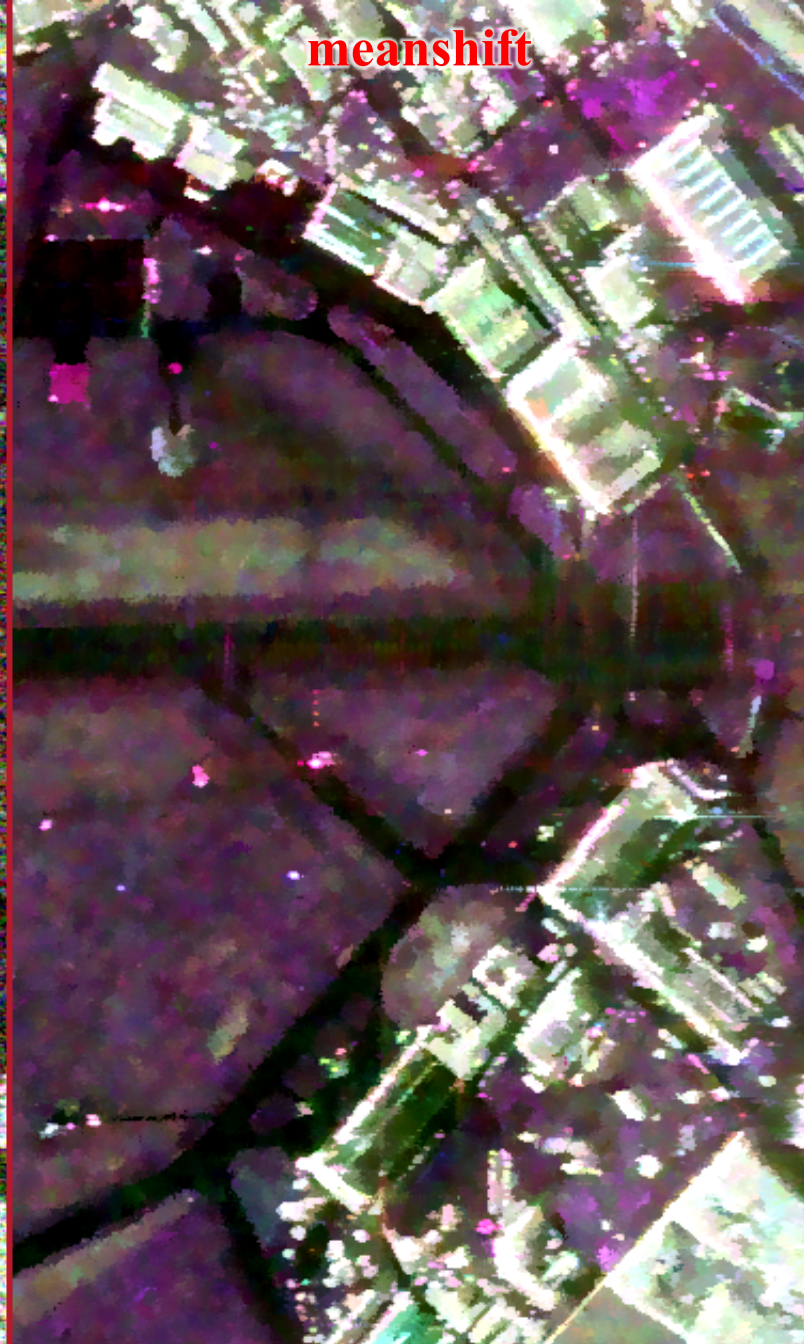
sigma



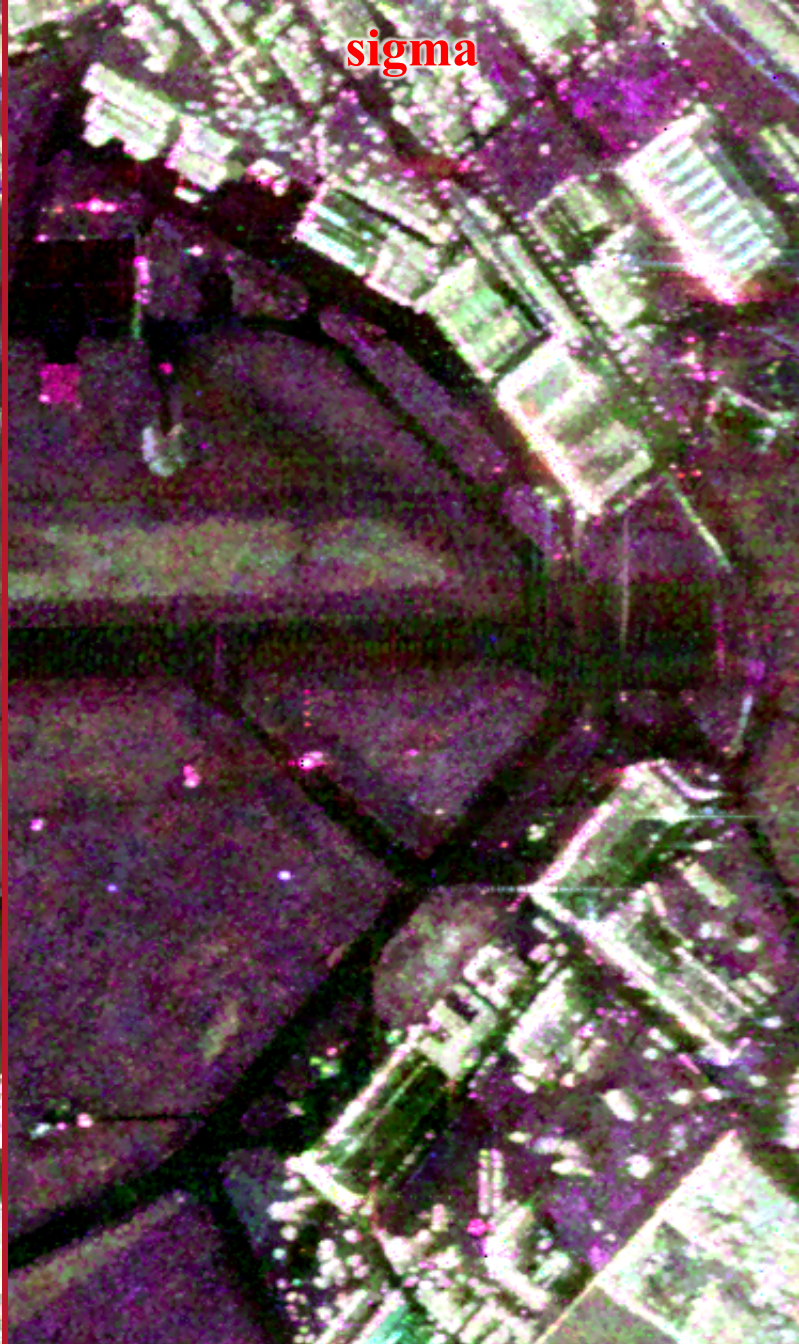
original



meanshift



sigma



original



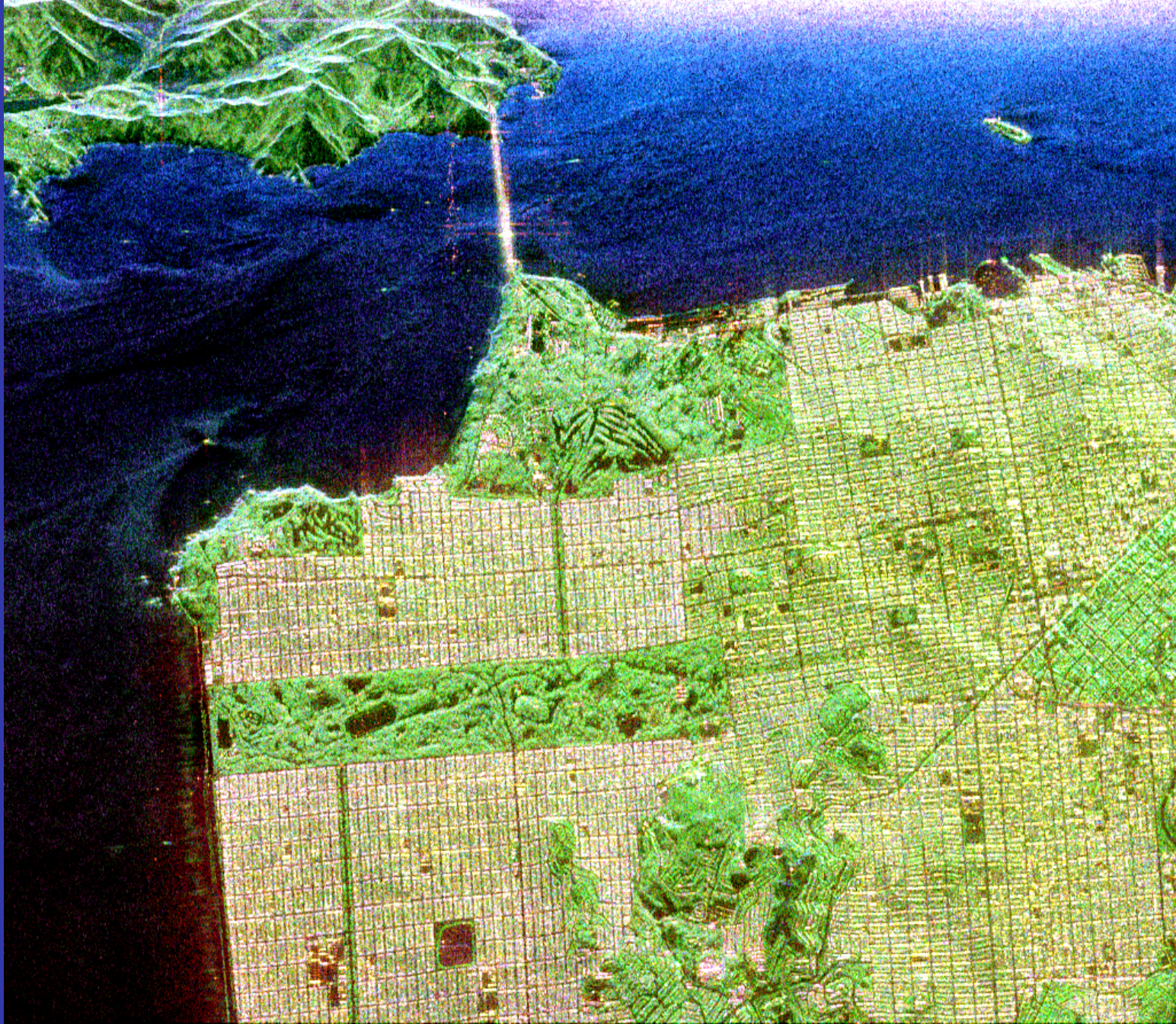
meanshift



sigma



San Francisco



MeanShift



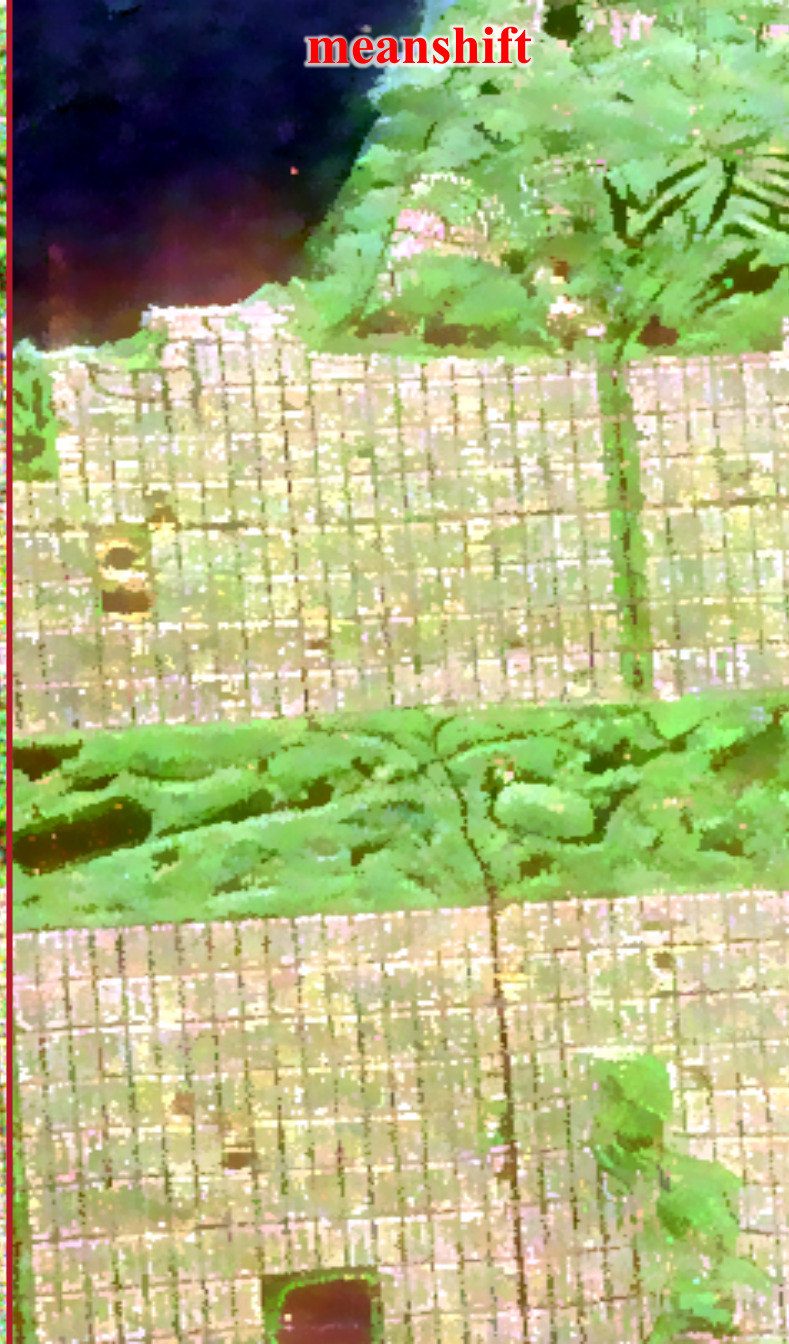
Sigma



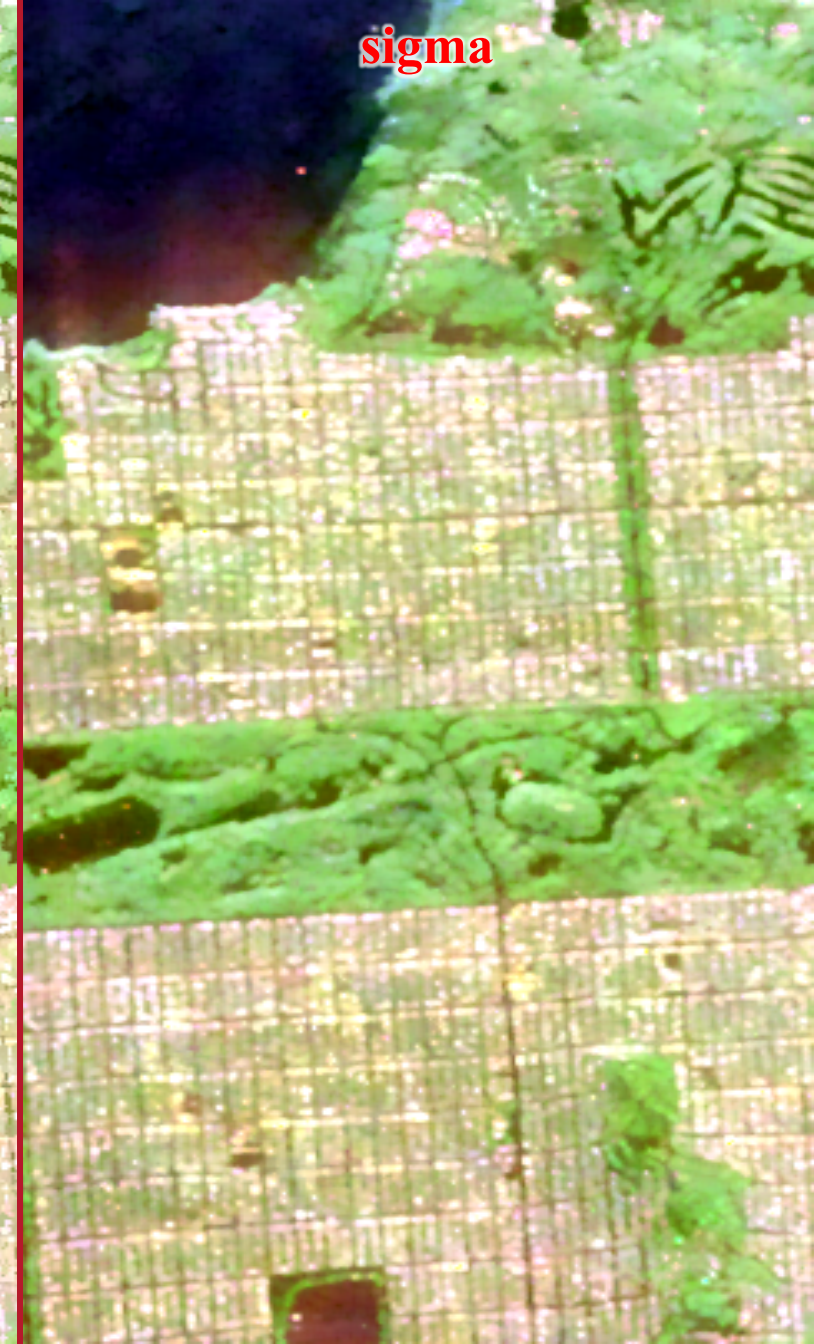
original



meanshift



sigma



CONCLUSION

- MeanShift can perform good image filtering.
- Position covariance tensor can provide a good textural attribute (ellipse orientation and elongation).
- Spatial attribute can be used in MeanShift to preserve edges.