## BeaulieuJM.ca/publi/Bea2001a

# Utilisation of Contour Criteria in Micro-Segmentation of SAR Images 

Jean-Marie Beaulieu

Computer Science Department
Laval University
in collaboration with
Canadian Centre for Remote Sensing, Ottawa Geomatic Research Centre, Laval University

## Image Segmentation

 is the division of the image plane into regionsTwo basic questions:


1- What kind of regions do we want?

- Homogeneous regions
- Segment similarity

2- How can we obtain them?

- Algorithm design


## HIERARCHICAL SEGMENTATION BY STEP-WISE OPTIMISATION

A hierarchical segmentation begins with an initial partition $\mathrm{P}^{0}$ (with N segments) and then sequentially merges these segments.
level $\mathrm{n}+1$
level $n$
level n -1


Segment tree

## STEP-WISE OPTIMISATION

- A criterion, corresponding to a measure of segment similarity, is used to define which segments to merge.
- At each iteration, an optimization process finds the two most similar segments and merges them.
- This can be represented by a segment tree, one node per iteration, where only the two most similar segments are merged.

Sequence of segment merges.


## Segmentation by hypothesis testing

Two hypothesis
H0: segments are similar
H1: segments are different

Distributions of the statistic d under H0 and H1


Two types of errors
Type I: not merging similar segments
Type II: merging different segments

$$
\begin{aligned}
& \alpha=\operatorname{Prob}(\text { Type I errors ) } \\
& \beta=\operatorname{Prob}(\text { Type II errors ) }
\end{aligned}
$$



Select the threshold to minimise $\alpha$ or $\beta$, but not both simultaneously

## In hierarchical segmentation, type II errors

 (merging different segments) can not be corrected, while type I errors can be corrected later on.

The distribution of H 1 and $\beta$ are unknown. Reduce $\beta$ by increasing $\alpha$.

## Sequential testing:

$\alpha$ will be reduced as segment sizes increase.
$\alpha_{1+2+\ldots .} \leq$ minimum $\left(\alpha_{1}, \alpha_{2}, \ldots\right)$
$\beta_{1+2+\ldots} \geq$ maximum $\left(\beta_{1}, \beta_{2}, \ldots\right)$


## Stepwise criterion

Find and merge the segment pair ( $\mathbf{i}, \mathbf{j}$ ) that minimises $\mathrm{V}_{\mathrm{i}, \mathrm{j}}(=1-\alpha)$.


Constant value region with uniform additive noise

$$
\text { Region } \mathrm{R}_{\mathrm{k}} \propto \mathrm{~N}\left(\mathrm{~m}_{\mathrm{k}}, \sigma^{2}\right)
$$

$$
\begin{aligned}
& \mathrm{d}_{\mathrm{i}, \mathrm{j}}=\left|\mu_{\mathrm{i}}-\mu_{\mathrm{j}}\right| \\
& \mathrm{v}_{\mathrm{i}, \mathrm{j}}=\operatorname{prob}\left(\mathrm{d} \leq \mathrm{d}_{\mathrm{i}, \mathrm{j}} ; \mathrm{H} 0\right) \\
& \mathrm{v}_{\mathrm{i}, \mathrm{j}}=\int_{-\mathrm{d}_{\mathrm{i}, \mathrm{j}}}^{\mathrm{d}_{\mathrm{i}}} \frac{1}{\sqrt{2 \pi} \sigma_{\mathrm{d}}} \exp \left(\frac{-\mathrm{x}^{2}}{2 \sigma_{\mathrm{d}}^{2}}\right) \mathrm{dx} \\
& \mathrm{v}_{\mathrm{i}, \mathrm{j}}=2 \operatorname{erf}\left(\mathrm{~d}_{\mathrm{i}, \mathrm{j}} / \sigma_{\mathrm{d}}\right) \\
& \text { where } \sigma_{\mathrm{d}}^{2}=\left(1 / \mathrm{N}_{\mathrm{i}}+1 / \mathrm{N}_{\mathrm{j}}\right) \sigma^{2}
\end{aligned}
$$

## Constant value region

$$
C_{i, j}^{\text {ward }}=\frac{\mathrm{d}_{\mathrm{i}, \mathrm{j}}}{\sigma_{\mathrm{d}}}=\sqrt{\frac{\mathrm{N}_{\mathrm{i}} \mathrm{~N}_{\mathrm{j}}}{\mathrm{~N}_{\mathrm{i}}+\mathrm{N}_{\mathrm{j}}}} \frac{\left|\mu_{\mathrm{i}}-\mu_{\mathrm{j}}\right|}{\sigma}
$$

## SEGMENTATION OF SAR IMAGE

SAR IMAGE $\rightarrow$ COHERENT SIGNAL (RADAR)
$\rightarrow$ INTERFERENCE PATTERN


## MULTIPLICATIVE NOISE

$$
\mathrm{p}(\mathrm{I})=\frac{1}{\Gamma(\mathrm{~L})}\left(\frac{\mathrm{L}}{\mu}\right)^{\mathrm{L}} \mathrm{I}^{\mathrm{L}-1} \exp (-\mathrm{L} \mathrm{I} / \mu)
$$



Noise is proportional to the amplitude

## SAR criterion

Using a Gaussian approximation for large NL value, we have:

$$
\begin{gathered}
\sigma_{d}^{2}=\left(1 / N_{i}+1 / N_{j}\right) \mu_{i+j}^{2} / L \\
C_{i, j}^{s a r}=\frac{d_{i, j}}{\sigma_{d}}=\sqrt{\frac{N_{i} N_{j}}{N_{i}+N_{j}}} \frac{\left|\mu_{i}-\mu_{j}\right|}{\mu_{i+j}} \sqrt{f}
\end{gathered}
$$

The segment dispersion (difference) is divided by the segment mean

## IMPORTANT NOISE

## PROBLEM WITH THE FIRST MERGES

SHAPE CRITERIA NEEDED

## SHAPE CRITERIA

- Bonding box - perimeter Cp
-Bonding box - area $\quad \mathrm{Ca}$
-Contour length
Cl
New criteria

$$
\mathrm{C}_{\mathrm{i}, \mathrm{j}}^{\text {contour }}=\mathrm{C}_{\mathrm{i}, \mathrm{j}}^{\mathrm{sar}^{2}} \times \mathrm{Cp}^{2} \times \mathrm{Ca} \times \mathrm{Cl}
$$

## 4 regions, 4 looks, 100x100



## 10 segments, standard criterion



## 10 segments, shape criterion



## standard criterion



100 Segments


1000 Segments


2000 Segments

## Shape vs standard criterion, 1000 segments


with shape criterion

without shape criterion


## SHAPE CRITERIA

- Bonding box - perimeter Cp
-Bonding box - area $\quad \mathrm{Ca}$
-Contour length
Cl
New criteria

$$
\mathrm{C}_{\mathrm{i}, \mathrm{j}}^{\text {contour }}=\mathrm{C}_{\mathrm{i}, \mathrm{j}}^{\mathrm{sar}^{2}} \times \mathrm{Cp}^{2} \times \mathrm{Ca} \times \mathrm{Cl}
$$

Bonding box - perimeter

$$
C p=\frac{\text { perimeter of } S_{i} \cup S_{j}}{\text { perimeter of bonding box }}
$$



Bonding box - area

$$
C a=\frac{\text { area of bonding box }}{\text { area of } S_{i} \cup S_{j}}
$$



## Contour length






 ETH


## SAR image

## 1000 segments



## SAR image

## 10 K segments




Segments


Segments


1K
Segments

## CONCLUSION

- Hierarchical segmentation produces goods results
-Criterion should be adapted to the application
-The first merges should be done correctly
- Shape criteria are useful

