

AN OBJECT-BASED APPROACH FOR FLOOD MAPPING IN VEGETATED AREAS BASED ON S-1 AND S-2 IMAGERY

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HYDRO-CLIMATE EXTREMES LAB (H-CEL) & REMOTE SENSING | SPATIAL ANALYSIS LAB (REMOSA)







CONTEXT & GOAL

Open flooding: distinguishable on SAR due to specular reflection Flooded vegetation: often complex backscattering mechanisms depending on vegetation type and structure

Pixels: most often used in image processing ⇔ Objects: allow for consideration of context and inclusion of additional input features



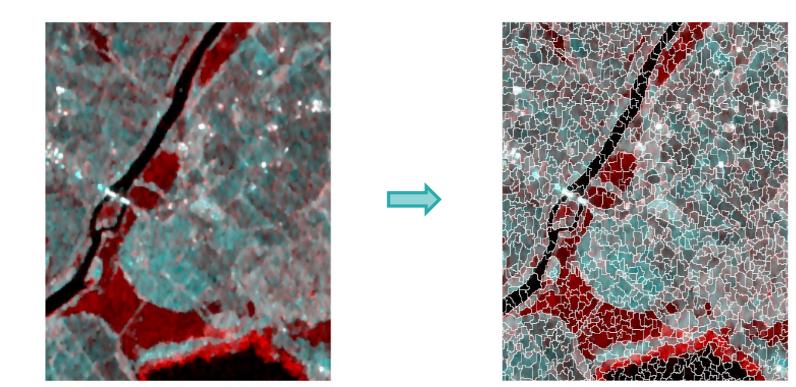
CONTEXT & GOAL

 \rightarrow Up to which extent can floods in vegetated areas be observed using freely available Sentinel imagery? What is the added value of S-2 imagery?

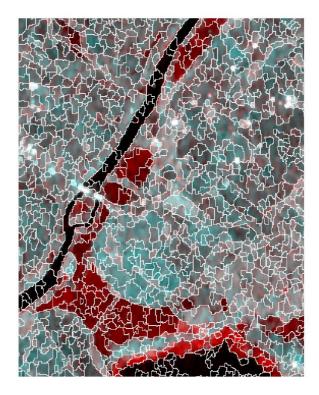




- Image segmentation 1.
 - Object prior using quickshift clustering on VV and VH bands d. flood/reference image pair
 - Object refinement based on objects' mean, st. dev. and shape b.

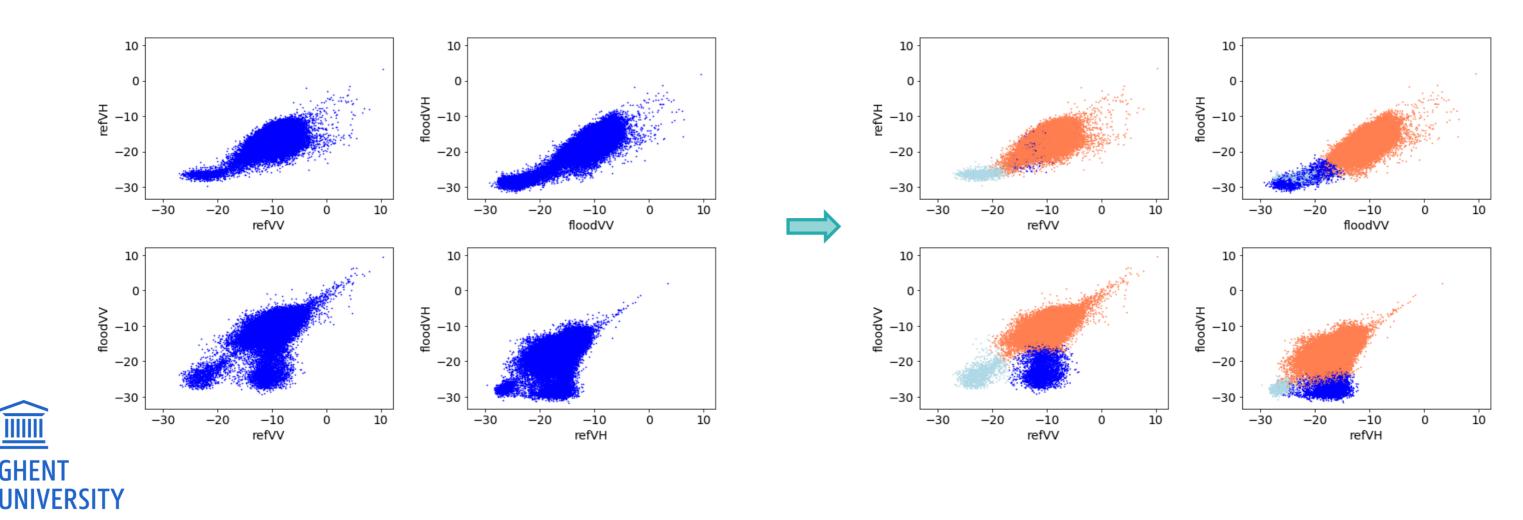






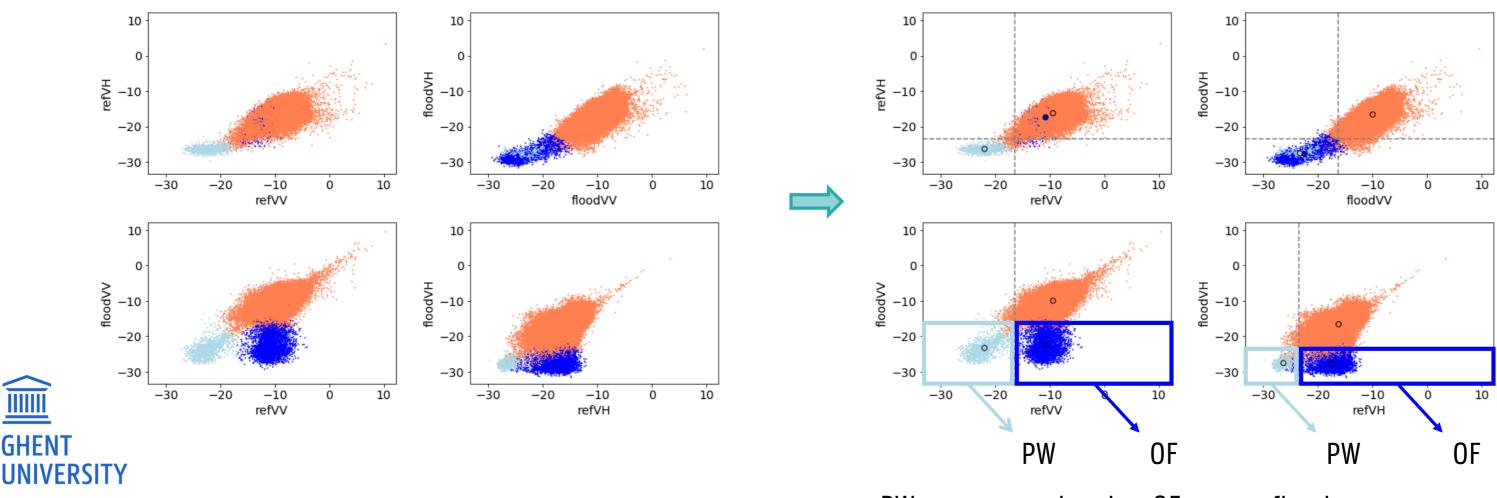
- Image segmentation 1.
- Spectral clustering on object means from SAR (and optical) features 2.

based on eigenvalues of similarity matrix



K-means clustering after dimensionality reduction

- Image segmentation 1.
- Spectral clustering on object means 2.
- Classification of clusters of visible flooding based on cluster centroids and 3. thresholds obtained from tiled thresholding



PW = permanent water; OF = open flood

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- 1. Image segmentation
- 2. Spectral clustering on object means
- 3. Classification of clusters
- 4. Post-processing [in progress]

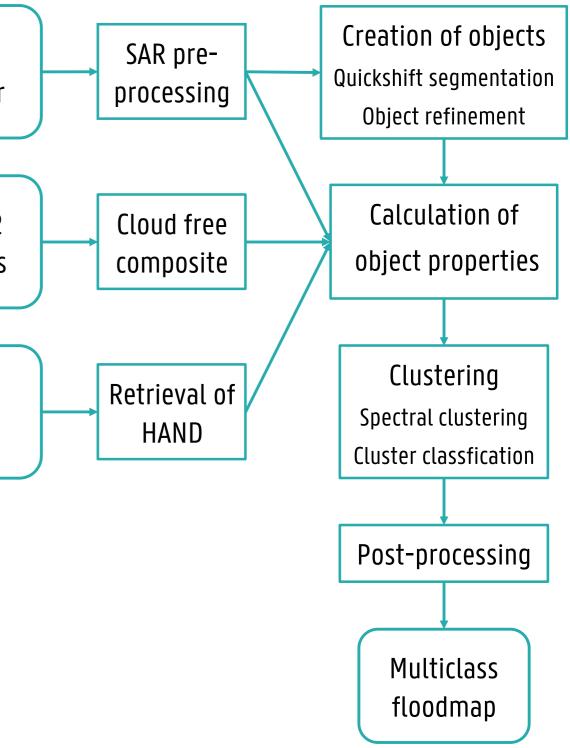
Sentinel-1 image pair

Sentinel-2 time series

Digital Elevation Model

Benchmark = pixel/segment-based thresholding:Flood \iff flood_VH < T_VH & flood_VV < T_VV</td>PW \iff floodVH < T_VH & flood_VV < T_VV & ref_VV < T_VV & ref_VH < T_VH</td>

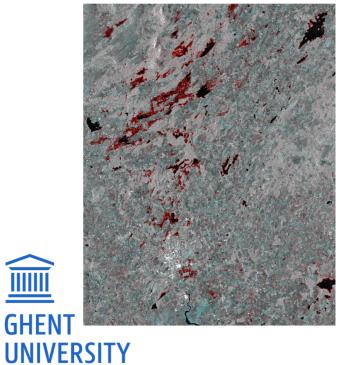




STUDY CASES

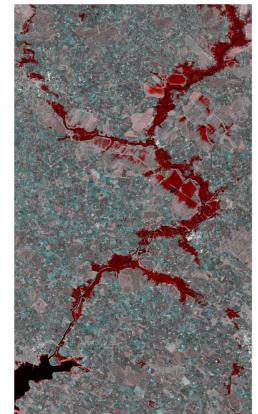
- 5 study cases (results shown for 3):
 - 3 cases comprising mainly visible flooding (cfr. Landuyt et al., 2019)
 - 2 cases comprising vegetated areas
- Annotation of truth classes: dry land (DL), permanent water (PW), open flood
 - (OF), flooded vegetation (FV, if present)





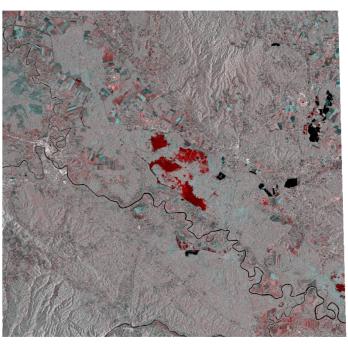
GHEN

River Shannon, 2016



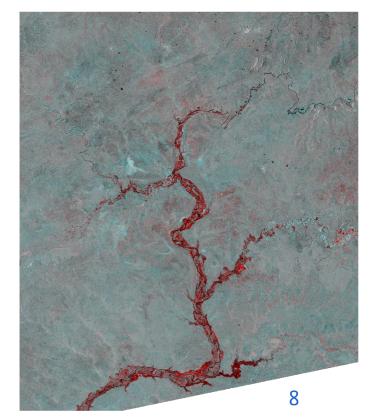
Lake Tay, 2017





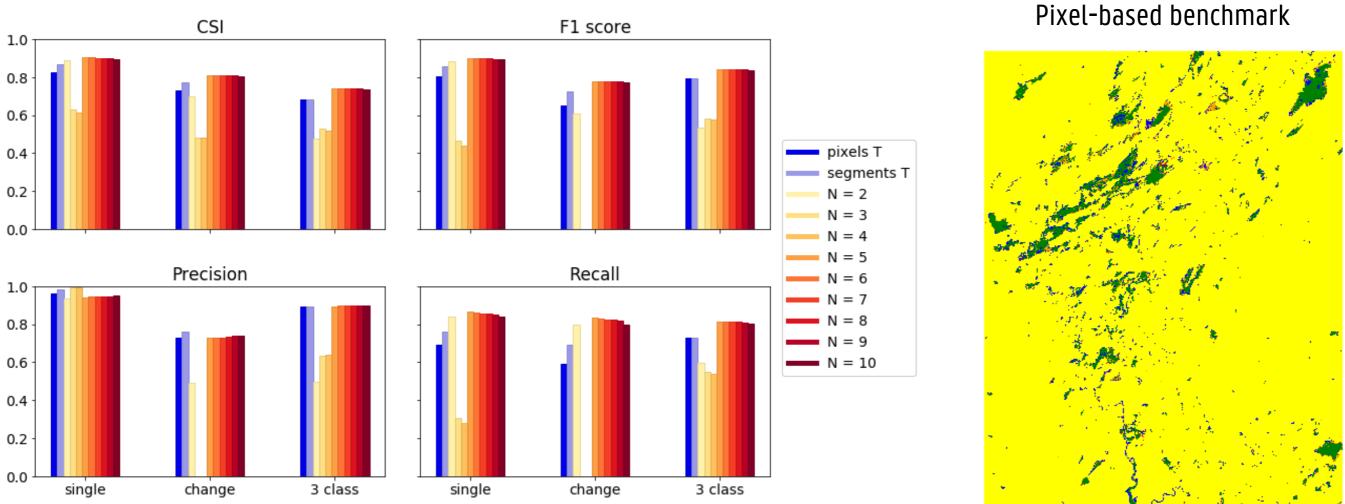
River Volta, 2018

River Sava, 2019



RESULTS – VISIBLE FLOODING

River Fergus – clustering on SAR mean features:

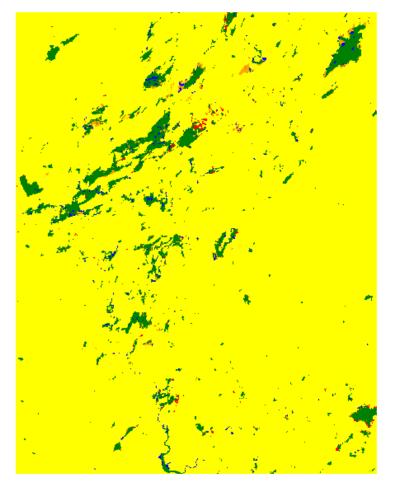




 \rightarrow Overclustering needed to fully capture the different classes \rightarrow Improved accuracies compared to benchmark

single: OF + PW vs. other change: OF vs. other 3class: OF vs. PW vs. DL

Object-based clustering, N = 8



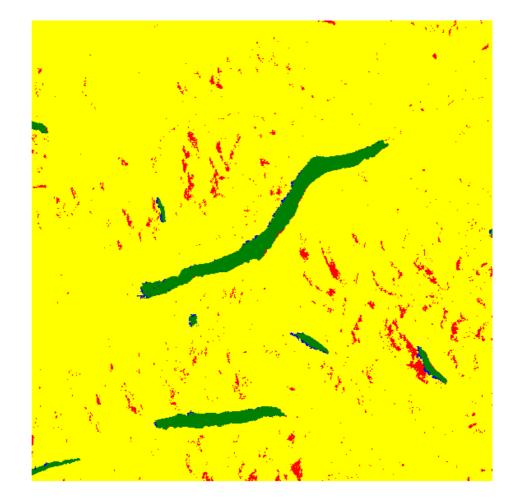
TN	
TP	
FN	
FP	
PW-flood confusion	

RESULTS – VISIBLE FLOODING

Lake Tay – clustering on SAR (+ HAND) mean features:

Pixel-based benchmark

Clustering on SAR mean features, N = 8



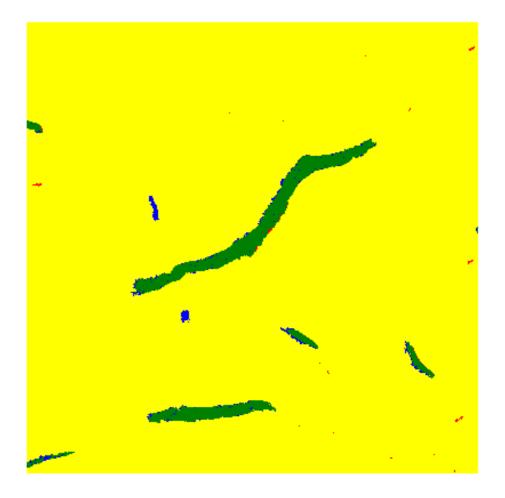


TN

FP

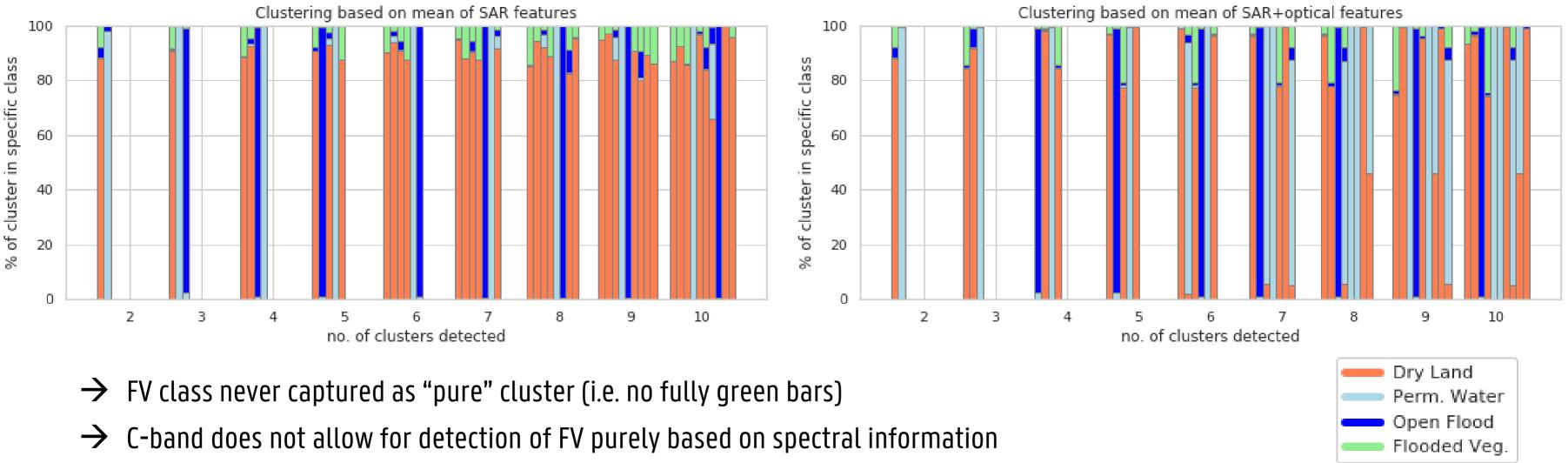
 \rightarrow Clustering-based classification limits false negatives due to wind roughening \rightarrow Inclusion of HAND in clustering space (without additional classification rules) eliminates false positives due to radar shadow

Clustering on SAR+HAND mean features, N = 8



RESULTS – FLOODED VEGETATION

River Sava - clustering on SAR (+ optical) features: Does clustering algorithm find a FV cluster?





CONCLUSIONS AND FUTURE PERSPECTIVES

- By taking into account the inherent data structure, object-based clustering approach leads to improved accuracies compared to thresholding benchmarks for visible flooding
- Unsupervised clustering is not capable of capturing flooded vegetation as a separate cluster/class
- (Potential) FV areas might be flagged by means of context-based postprocessing
- Inclusion of time series information might improve characterization of vegetation





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