

UFMG

Contextual Description of Superpixels for Aerial Urban Scenes Classification

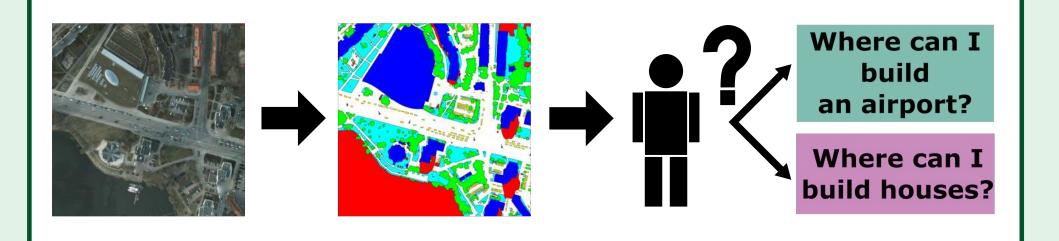
{ tiagohubner, jefersson }@dcc.ufmg.br, alexei@pucminas.br





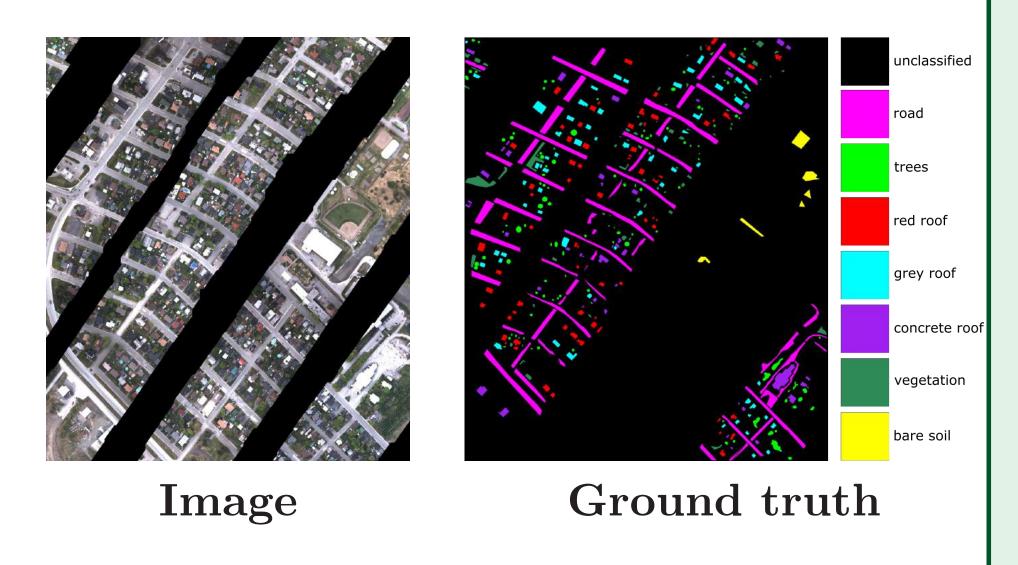
1. MOTIVATION

Thematic maps play a fundamental role in the decision-making in several areas such as urban planning, environmental monitoring and economic activities.

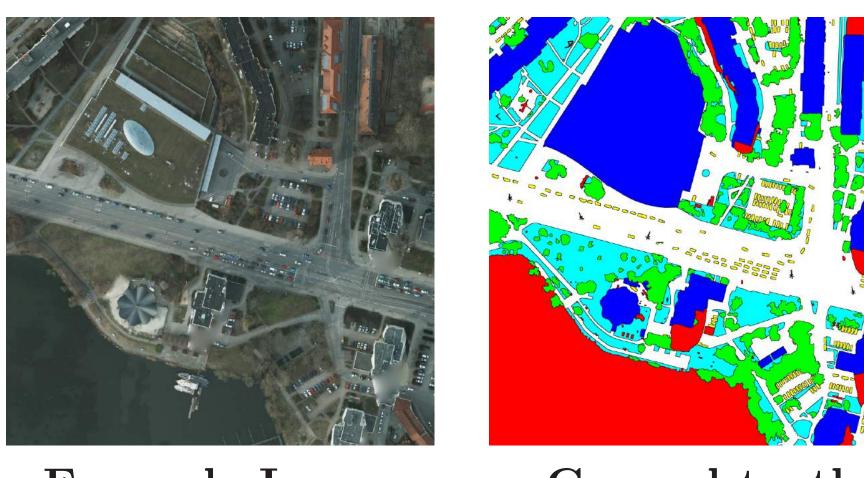


4. DATASETS

• grss_dfc_2014: 7 classes, 3769x4386 Very High Resolution (VHR) image, spatial resolution of 20 cm, taken over Québec, Canada, provides particular subset for training



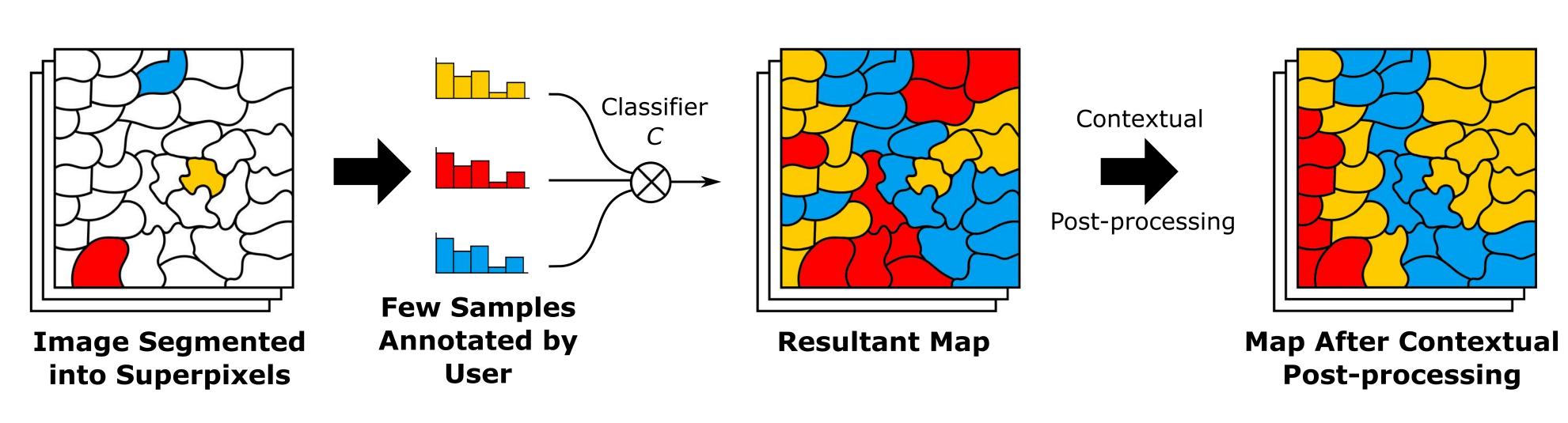
• ISPRS Postdam: 6 classes, 38 6000x6000 VHR image patches, spatial resolution of 5 cm, taken over Postdam, Germany. We selected 5 from the 38 images, once ground truth for test images is not provided



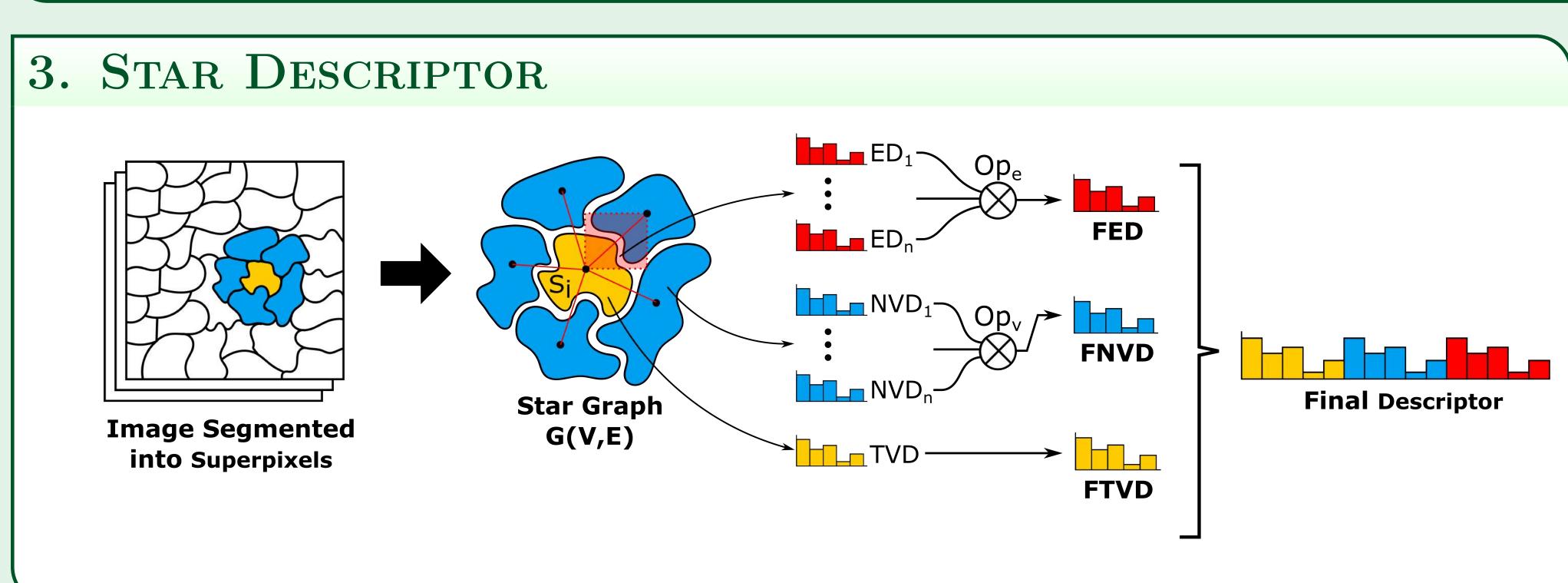
Example Image Ground truth

2. PROBLEM

widely recognized that context improves classification However, require lots of labeled data to impose traditional methods contextualconstraints That suitable for automatic thematic maps results. theover not creation.



Is there a way to encode context in the description of image elements without labeling them?



5. Experimental Setup

Segmentation: Simple Linear Iterative Clustering (SLIC)

Region Descriptors: three color (BIC, CCV, GCH) and one texture descriptor (Unser)

Edge Descriptors: two texture descriptors (LBP and Unser)

Summarization operations: sum, average and max pooling

Classifier: SVM with RBF kernel

grss_dfc_2014 protocol: specific train-ing/test subsets

ISPRS Postdam: 5-fold Cross-validation Baselines: low-level descriptors without context (NO-CTXT) and Vargas' descriptor (VAR-GAS)

7. CONCLUSIONS

- It is possible to aggregate context to automatically generate thematic maps without using lots of labeled data
- Borders might be important in urban scenarios
- Accuracy of the generated map is strongly dependent on the combination of low-level descriptors and summarization operations chosen

ACKNOWLEDGMENTS

We thank FAPEMIG, CAPES, and CNPq for the financial support.

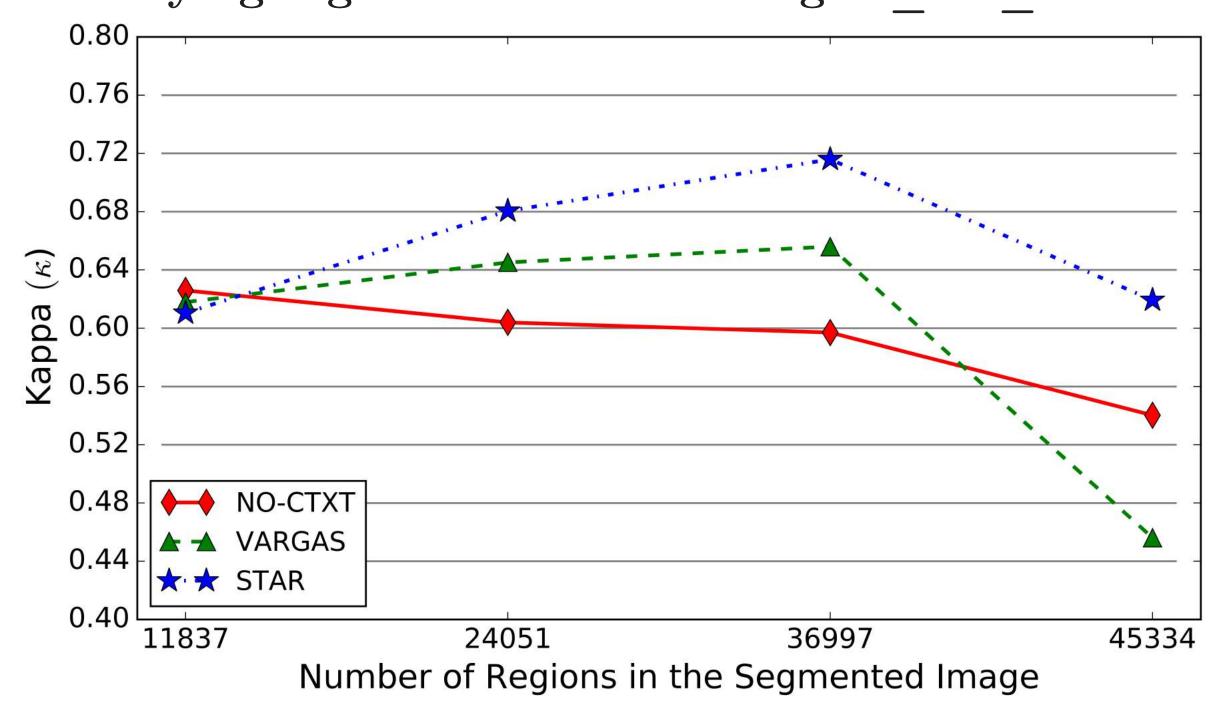
STAR

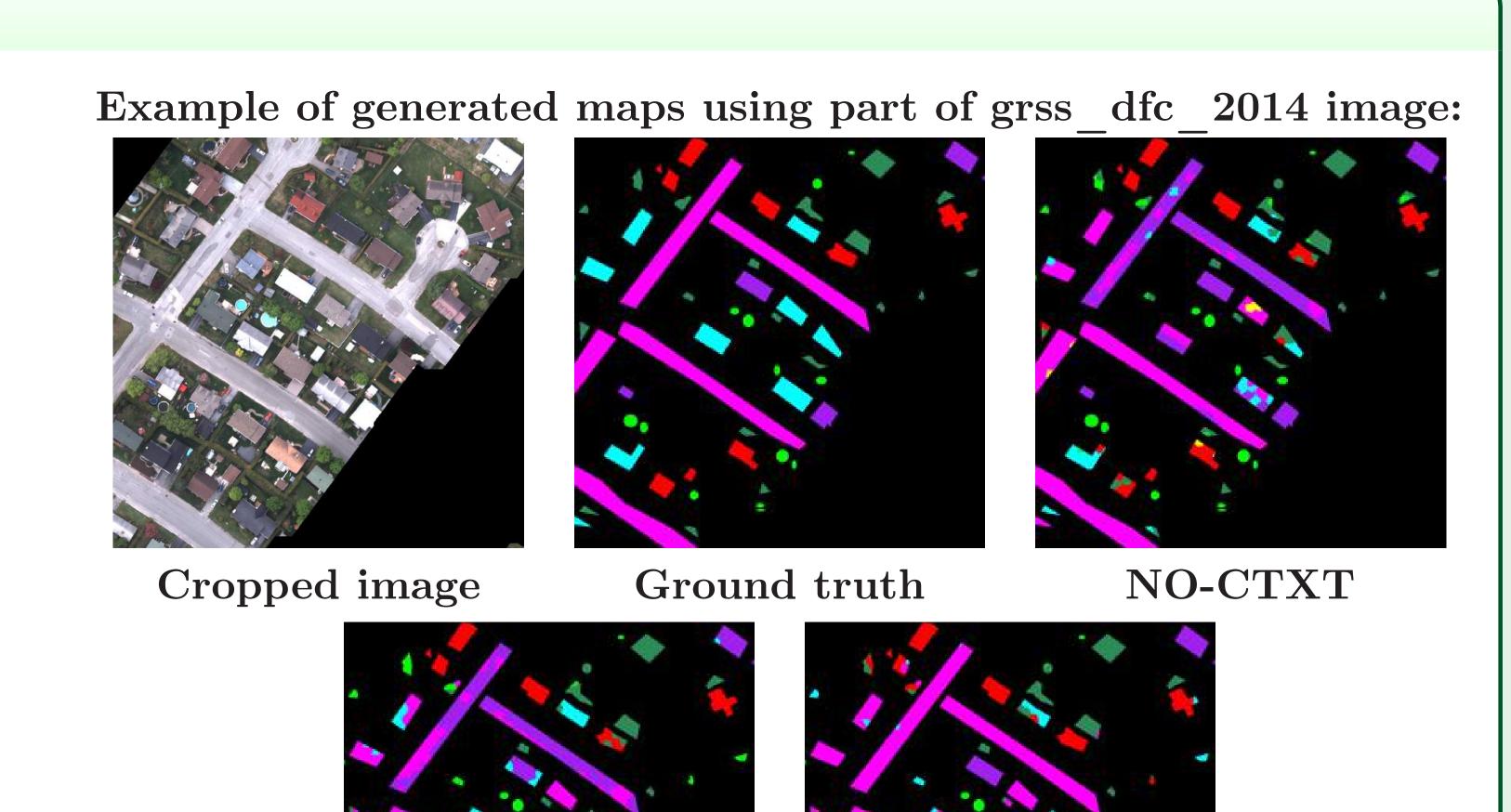
6. RESULTS

Kappa index and overall accuracy on both datasets:

	$grss_dfc_2014$		ISPRS	
Descriptors	κ	Ovr.	κ	Ovr.
NO-CTXT	0.619	0.724	0.230 ± 0.040	0.474 ± 0.065
VARGAS	0.651	0.751	$0.275 \!\pm\! 0.053$	$0.501 \!\pm\! 0.058$
\mathbf{STAR}	0.735	0.822	0.181 ± 0.049	$0.421 {\pm} 0.064$

Varying segmentation scale on grss_dfc_2014:





VARGAS