

# Object-Based Augmentation for Building Semantic Segmentation: Ventura and Santa Rosa Case Study

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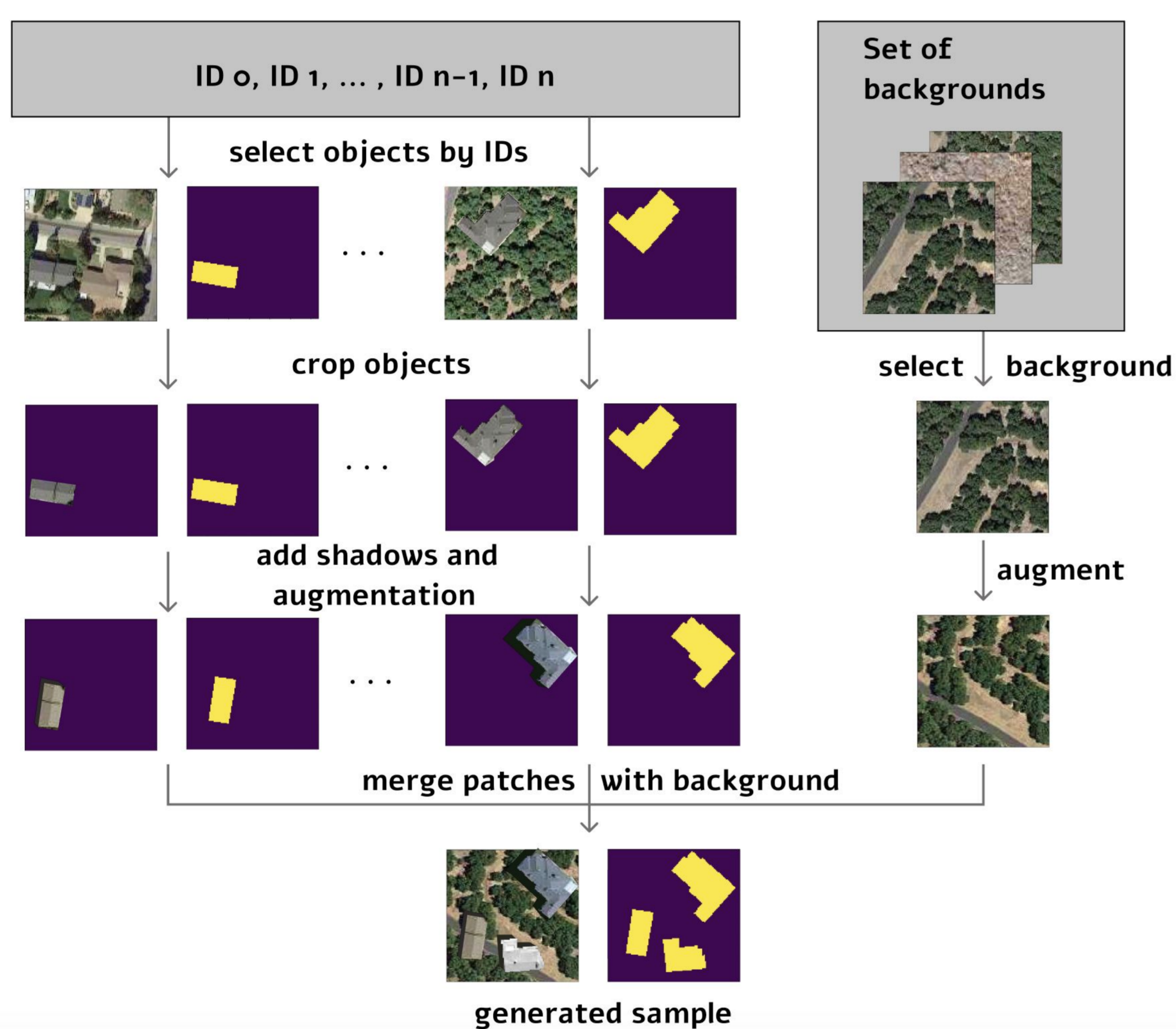
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## Background

This study focuses on the development and testing of object-based augmentation. The practical usefulness of the developed augmentation technique is shown in the remote sensing domain, being one of the most demanded in effective augmentation techniques. We propose a novel pipeline for georeferenced image augmentation that enables a significant increase in the number of training samples. The presented pipeline is called object-based augmentation (OBA) and exploits objects' segmentation masks to produce new realistic training scenes using target objects and various label-free backgrounds.

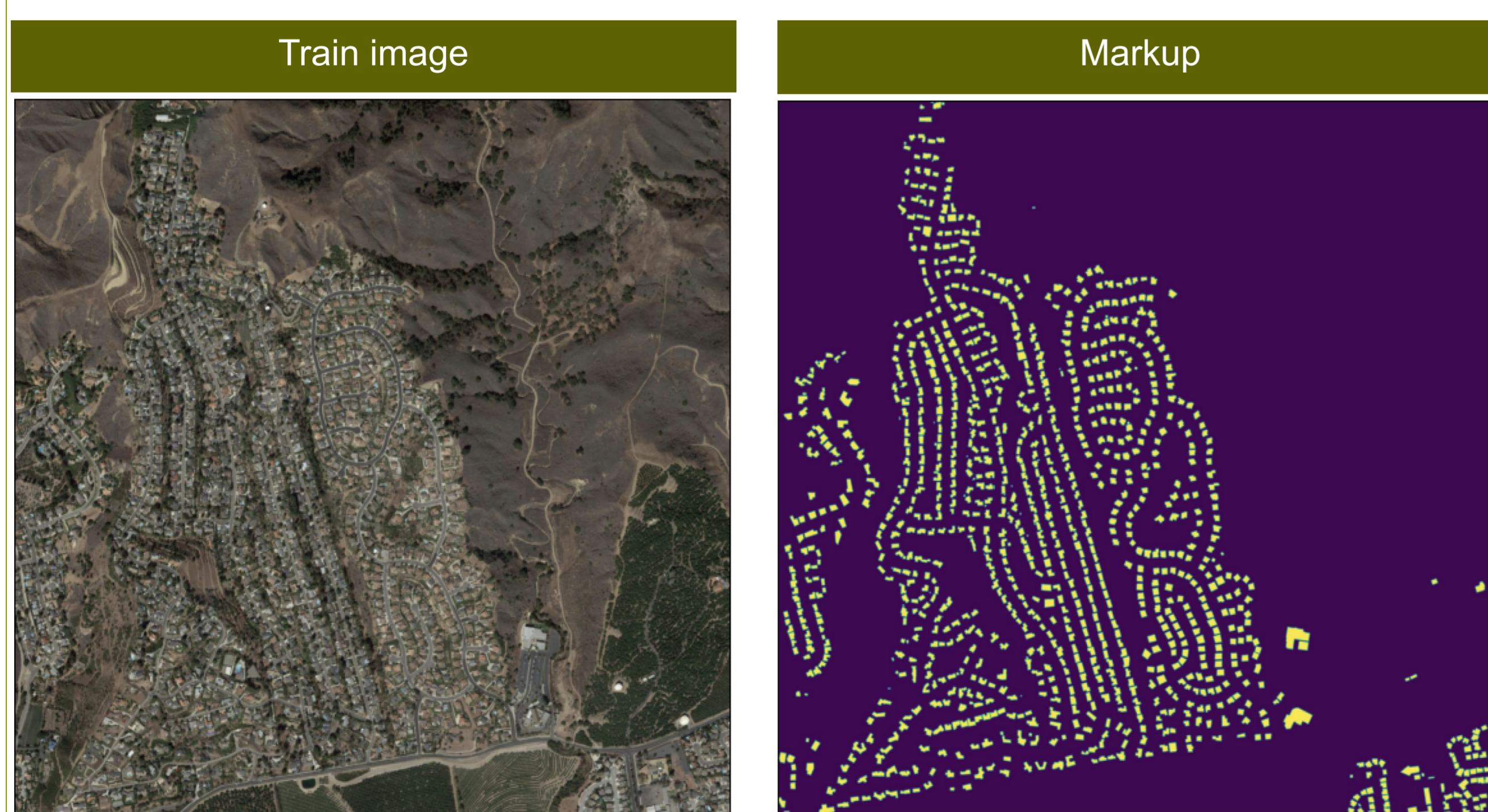
## Methods

- georeferenced images and markup
- set of backgrounds



## Data

We used high-resolution satellite images covering Santa Rosa and Ventura (California)



	Train	Validation	Test
Objects number	955	226	282
Area in hectares	390	100	93
Extra background area in hectares	2000	500	500

## Difficulties

- Small datasets
- Different environmental conditions
- Rare target objects
- Geo-spatial satellite data specificity (objects distribution, image size, shadows, etc)

## Example of training samples



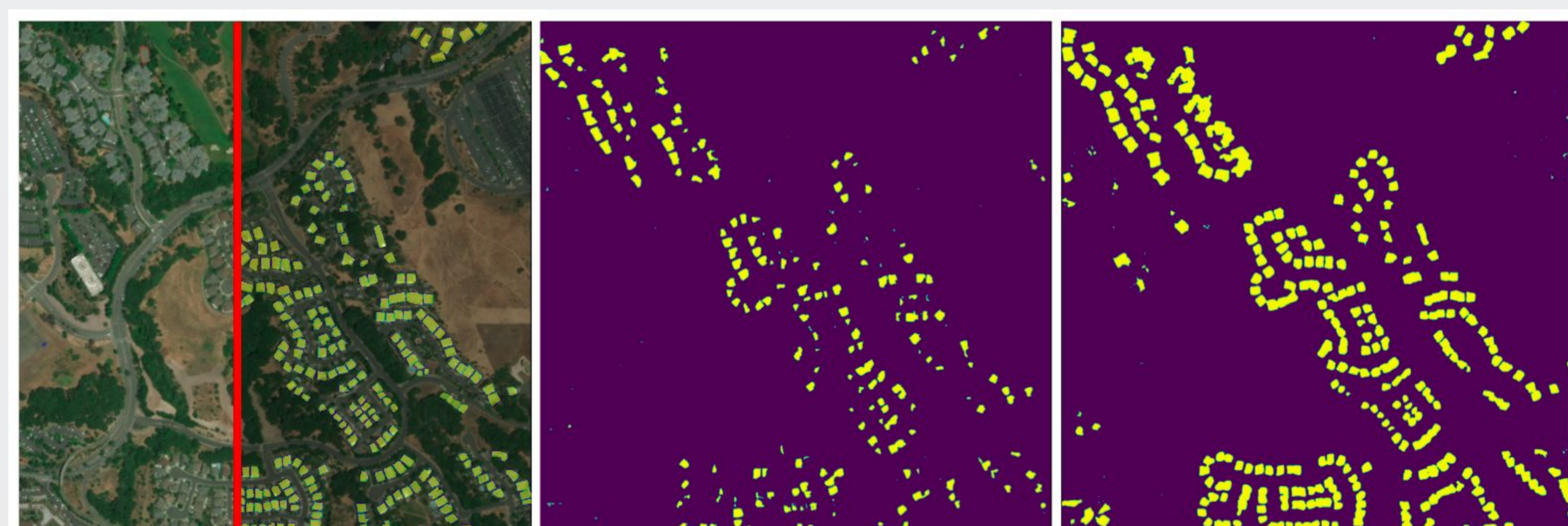
## Results

F1-score (per-pixel) is used to assess semantic segmentation quality for different augmentation setups and model's architecture

	Base augm.	Shadow	Extra background	Standard augmentation	Augmentation	F1-score
Baseline_no_augm	✓	✗	✗	No	Baseline_no_augm	0.45
Baseline	✓	✗	✗		OBA_no_augm	0.66 (+21%)
OBA_no_augm	✗	✓	✓	Yes	Baseline	0.788
OBA_no_shadow	✓	✗	✓		OBA_no_shadow	0.811 (+2.3%)
OBA_no_background	✓	✓	✗		OBA_no_background	0.81 (+2.2%)
OBA	✓	✓	✓		OBA	0.829 (+4.1%)
					OBA + optimization	0.835 (+4.7%)

Model	Baseline_no_augm			Baseline			OBA		
	Resnet18	Resnet34	Resnet50	Resnet18	Resnet34	Resnet50	Resnet18	Resnet34	Resnet50
FPN	0.325	0.367	0.186	0.741	0.762	0.784	0.802	0.813	<b>0.826</b>
U-Net	0.435	0.45	0.34	0.766	0.788	0.766	0.807	<b>0.829</b>	0.824
HRNet	Resnet101			Resnet101			Resnet101		
	0.23			0.741			<b>0.812</b>		

Model predictions (F1-score) for different architectures



Input image

Prediction without augm

Prediction with OBA

## Conclusions

- We proposed augmentation for building segmentation
- We tested different augmentation configurations
- The proposed object-based augmentation improves the performance of the remote sensing task
- The code is available

[https://github.com/LanaLana/satellite\\_object\\_augmentation](https://github.com/LanaLana/satellite_object_augmentation)