INAF Ć HaMMon Hazard Mapping and vulnerability Monitoring

Introducing HaMMon



Funded by PNRR

Backed by ICSC's spoke 3

Industrial Project

Capitalizing the acquired skills applying them to real world's issues





15 partners

Public bodies and private companies across Italy

Goals

Facing the hazardous and

extreme events **more frequent**

due to the **Climate Change**



Work Packages

0

Management



Technological infrastructure to run and deploy the applications



Post-event analysis



Seasonal forecasts and weather generator



Building features extraction from images



Vulnerability curves for earthquakes and landslides

Work Packages involving INAF

0

Management



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Vulnerability curves for earthquakes and landslides

Which are the goals?



WP2

Carry on assessment activities on the Digital Twin

WP4

Getting better estimates on the danger to which building are exposed



Which are the tasks?



WP2

- Flying drones to take pictures
- Segmenting the pictures automatically
- Doing photogrammetry
- Finalizing the Digital Twin

WP4

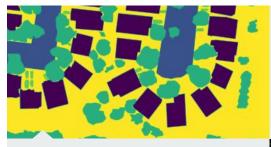
- Modelling the vulnerability curves
- Collecting aerial images
- Doing feature extraction





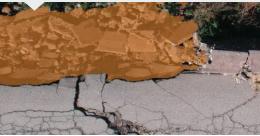


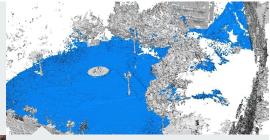
Why Segmenting?



Estimating areas







Estimating volumes (with photogrammetry)

How to segment?

Tiramisù

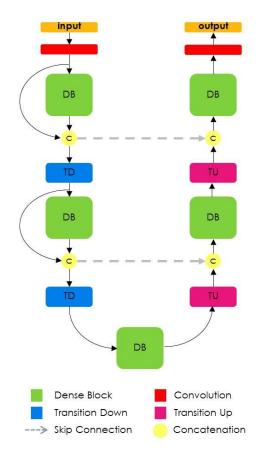
Fully Convolutional Densely Connected U-Net

How to segment?

Tiramisù

Fully Convolutional Densely Connected U-Net

- The Dense Blocks accumulate different feature maps for the input
- The Transition Down decrease size and increase feature space
- The Transition Up decrease feature space and increase size
- The Skip Connections force conditioning on the output





Experimental setup



FloodNet dataset

- Images from a drone survey made in 2017
- Presents damages left from Harvey hurricane
- Taken in Texas and Louisiana (USA)

2.343

Labelled images

10

Class of objects

Background, Building-flooded, Building-not-flooded, Road-flooded, Road-not-flooded, Water, Tree, Vehicle Pool, Grass **13GB**

of Hi-Res images

Training architecture

- The architecture is a Kubernetes cluster provided by the WP1
- The training runs on two nVidia Tesla V100 32GB
- The network storage is modular and shared

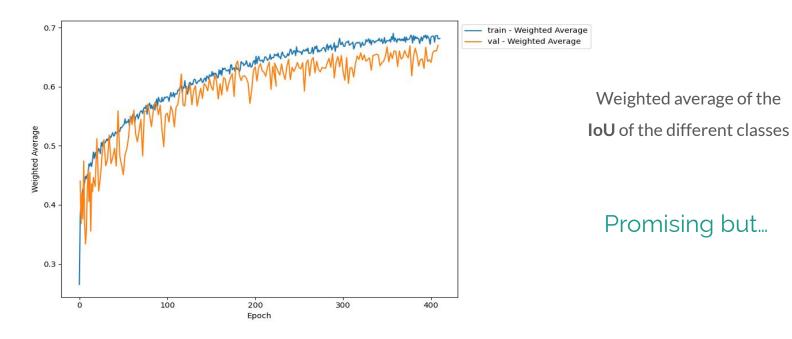


Training setup

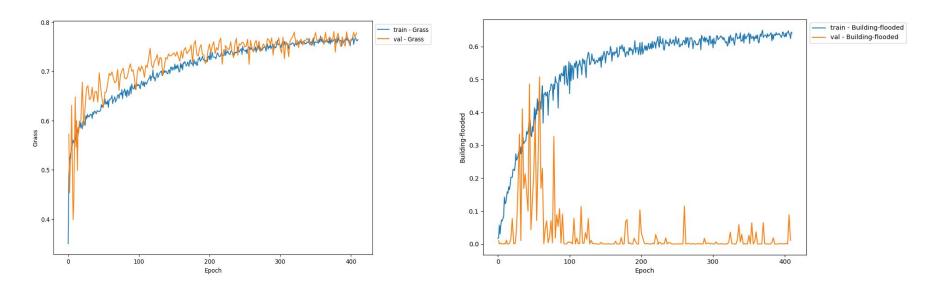
- Random crops of the dataset images at 600x600 px
- Data augmentation (flipping, scaling)
- ~ 400 epochs
- Average training time of around 100 hours



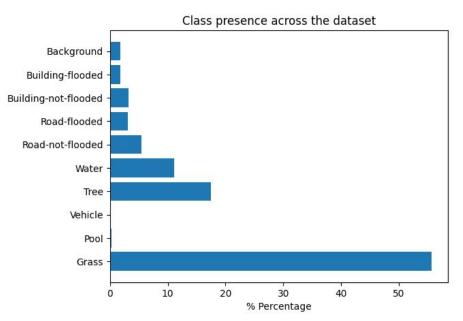
Results so far



Results so far



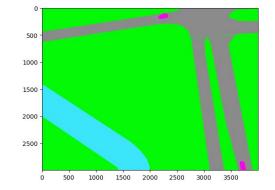
Results so far

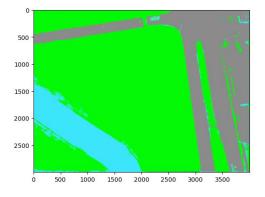


Some classes are practically missing!

Results so far







Image

Ground Truth

Prediction

What we are improving



Transfer-learning

on RescueNet dataset

- Images from a UAV made in 2018
- Presents damages left from Michael hurricane
- Taken in different USA locations

4.494

Labelled images



Class of objects

Background, Water, Building_No_Damage Building_Minor_Damage Building_Major_Damage Building_Total_Destruction Vehicle, Road-Clear, Road-Blocked, Tree, Pool 22.6GB

of Hi-Res images

Hi-Res inference

CPU/GPU

Heterogeneous computing

12MP

3000x4000 px



per image

Implementing Attention

SegFormer

Transformer encoder + MPL decoder

https://arxiv.org/abs/2105.15203

Attention U-Net

U-Net with attention decoding

https://arxiv.org/abs/1804.03999

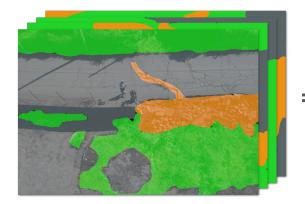
TransUNet

U-Net with attention encoding

https://arxiv.org/abs/2102.04306

Digital Twin augmentation







↑ Tiramisù







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