

Monitoring Water Levels with Satellite Imagery

Data Science Retreat - Batch 22
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Agenda

- Motivation
- Data collection
- Labeling workflow
- Metrics
- Model exploration
- Model refinement
- Evaluation and testing
- Demo
- Question and Answers



<https://www.nationalgeographic.com/magazine/2018/03/drying-lakes-climate-change-global-warming-drought/>

Water bodies of interest

- Lake Poopo, Bolivia
- Lake Urmia, Iran
- Lake Mojave, USA
- Aral sea, Kazakhstan
- Lake Copais, Greece
- Lake Ramganga, India
- Lake Qinghai, China
- Salton sea, USA
- Elephant Butte Reservoir, USA
- Salton sea, USA
- Lake Faguibine, Mali
- Lake Mono, USA
- Lake Walker, USA
- Lake Balaton, Hungary
- Lake Koroneia, Greece
- Lake Salda, Turkey
- Lake Burdur, Turkey
- Lake Mendocino, USA

Data

1059 images

700 Resisc45



182 Sentinel-2 hub service



177 courtesy from Olga



Train set

Test set

U-NET

Encoding Path

convolution + max
pooling operations

Filters + cropping

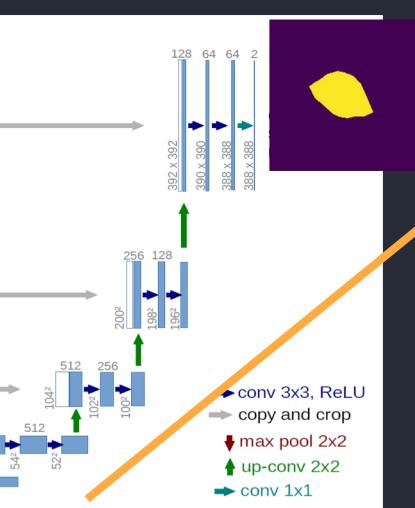
↓ resolution, ↑ context

Higher level features
LAKES!



Decoding Path

upsampling

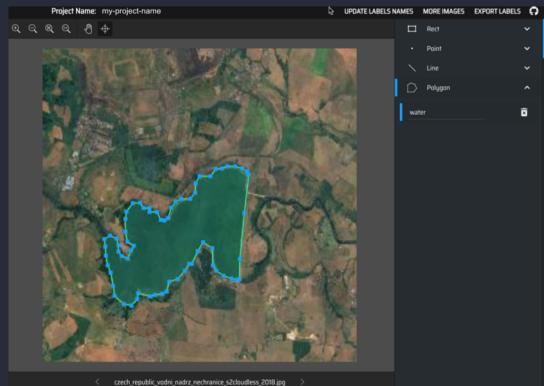


Labeling

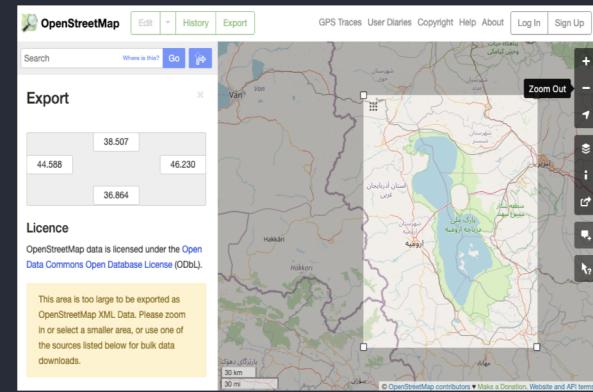
Training set



Make Sense Tool

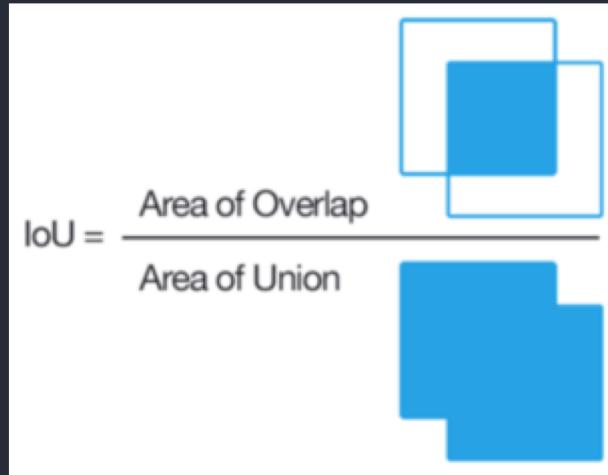


Test set

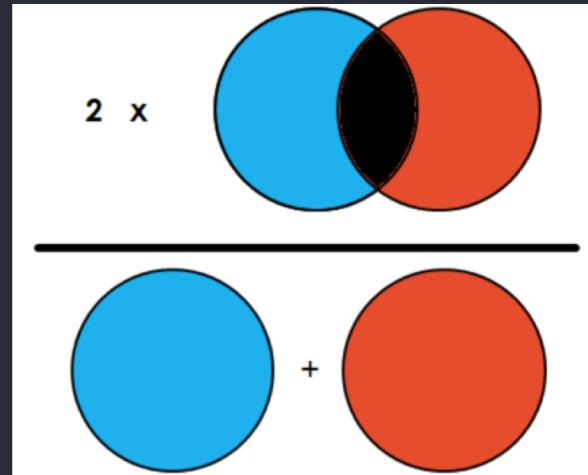


Metrics

Jaccard Index (IoU)

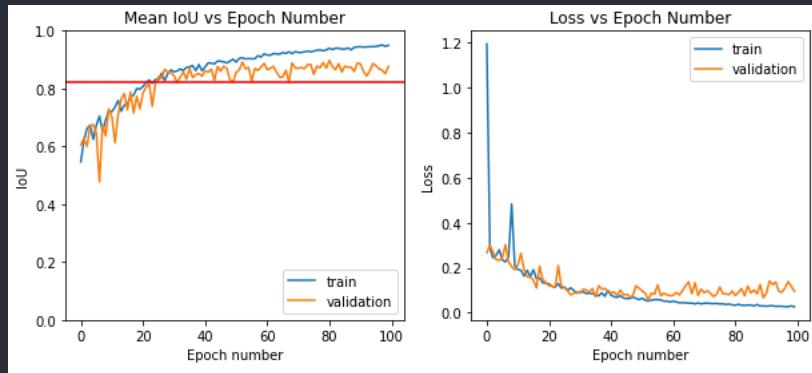


Dice Coefficient (F1 Score)



Source: <https://towardsdatascience.com/metrics-to-evaluate-your-semantic-segmentation-model-6bcb99639aa2>

Model training without image augmentation



Training : 489

Validation: 140

Test : 71

Loss: 0.09

IoU : 0.89

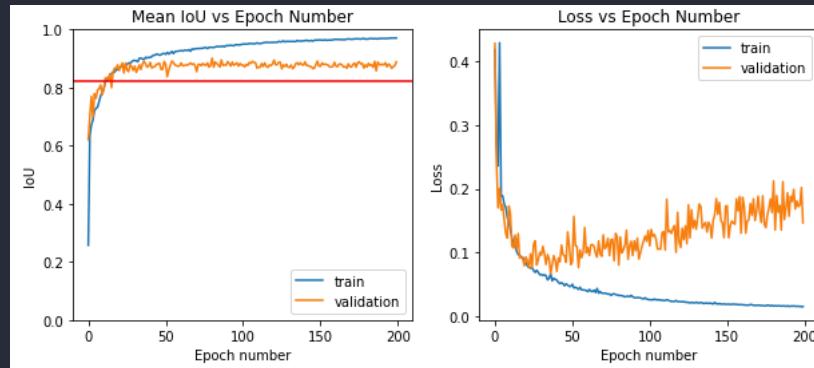
Loss function: binary cross entropy.

Image Augmentation

The following techniques have been applied during training:

- Height shift up to 30%
- Horizontal flip
- Rotation up to 45 degrees
- No shear
- Vertical flip
- Width shift up to 30%
- Zoom between 75% and 125%

Model training with image augmentation



Training : 979

Validation: 280

Test : 141

Loss: 0.18

IoU : 0.86

Loss function: binary cross entropy.

Model Refinement

The following strategies have been studied:

1. Dropout, early stopping and adaptive learning rates
2. Use a bigger rocket
3. Batch regularization
4. Residual connections
5. Dealing with imbalanced classes using Dice score
6. Different optimization algorithms
7. Refining labels using Conditional Random Fields
8. Ensemble predictions

Conditional Random Fields



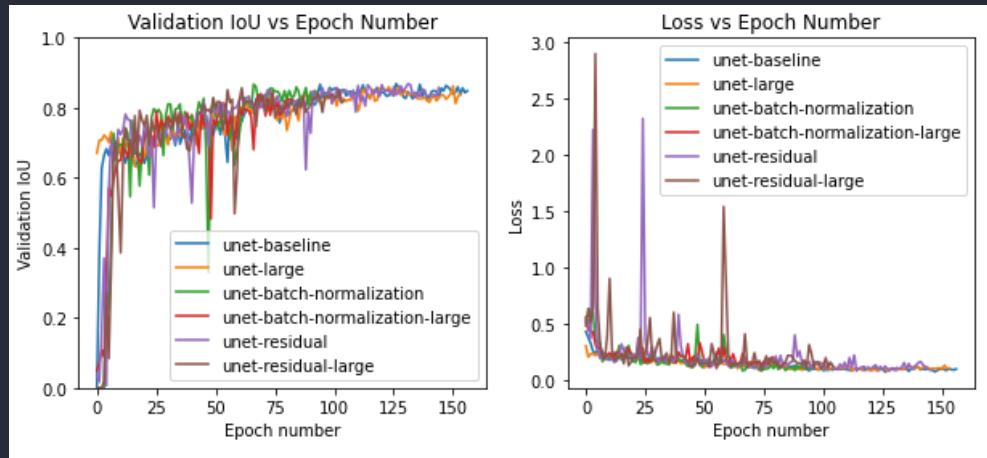
Models performance I

| | Validation results | | | | |
|---------------------|--------------------|---------------|--------------|------------------|-----------|
| | Loss | Jaccard index | Jaccard loss | Dice coefficient | Dice loss |
| unet-baseline | 0.09 | 0.87 | 0.13 | 0.90 | 0.10 |
| unet-large | 0.08 | 0.89 | 0.11 | 0.91 | 0.09 |
| unet-bn | 0.15 | 0.83 | 0.17 | 0.88 | 0.11 |
| unet-bn-large | 0.11 | 0.85 | 0.15 | 0.90 | 0.10 |
| unet-residual | 0.22 | 0.80 | 0.20 | 0.86 | 0.14 |
| unet-residual-large | 0.12 | 0.85 | 0.15 | 0.89 | 0.11 |

Models performance II

| | Test results | | | | |
|---------------------|--------------|---------------|--------------|------------------|-----------|
| | Loss | Jaccard index | Jaccard loss | Dice coefficient | Dice loss |
| unet-baseline | 0.29 | 0.63 | 0.37 | 0.73 | 0.27 |
| unet-large | 0.82 | 0.55 | 0.45 | 0.69 | 0.31 |
| unet-bn | 0.27 | 0.65 | 0.35 | 0.75 | 0.25 |
| unet-bn-large | 0.28 | 0.63 | 0.37 | 0.73 | 0.27 |
| unet-residual | 0.54 | 0.50 | 0.50 | 0.64 | 0.36 |
| unet-residual-large | 0.29 | 0.60 | 0.40 | 0.72 | 0.28 |

Models performance III

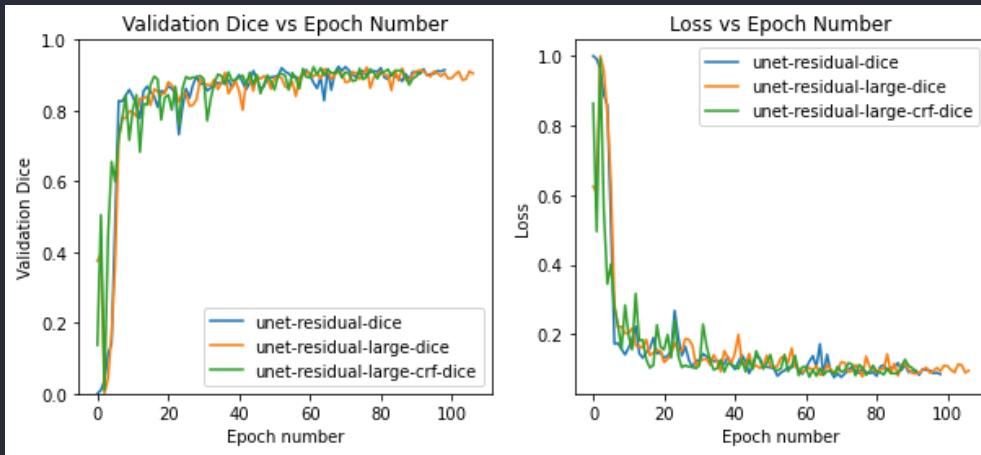


Loss function: binary cross entropy.

Models performance IV

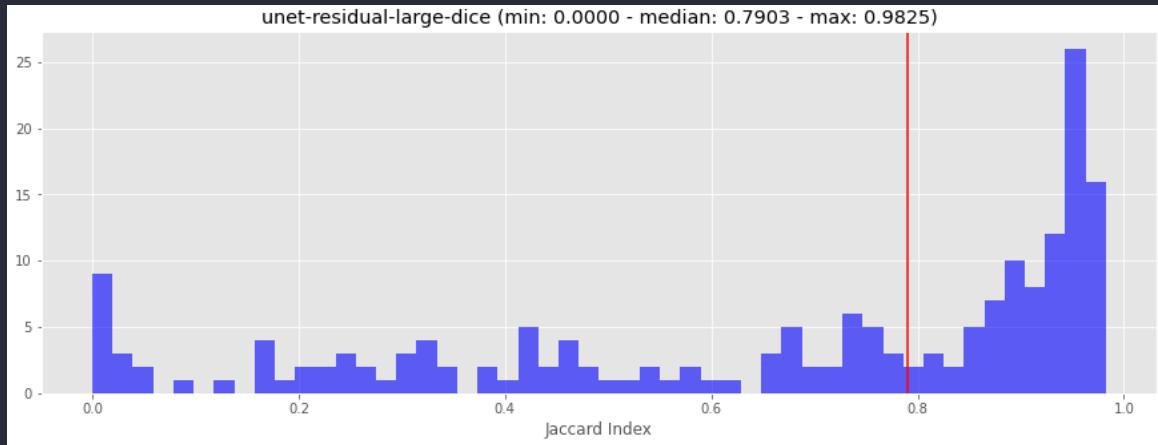
| | Validation results | | Test results | |
|------------------------------|--------------------|-----------|------------------|-----------|
| | Dice coefficient | Dice loss | Dice coefficient | Dice loss |
| unet-residual-dice | 0.92 | 0.08 | 0.73 | 0.27 |
| unet-residual-large-dice | 0.90 | 0.10 | 0.67 | 0.33 |
| unet-residual-large-crf-dice | 0.91 | 0.09 | 0.79 | 0.21 |

Models performance V



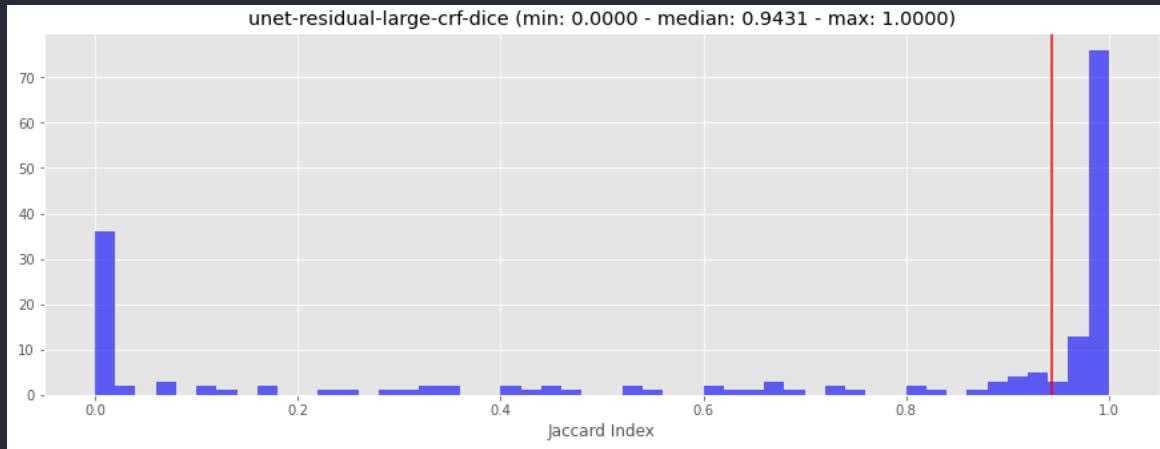
Loss function: dice coefficient loss.

Test results I - U-net residual large model



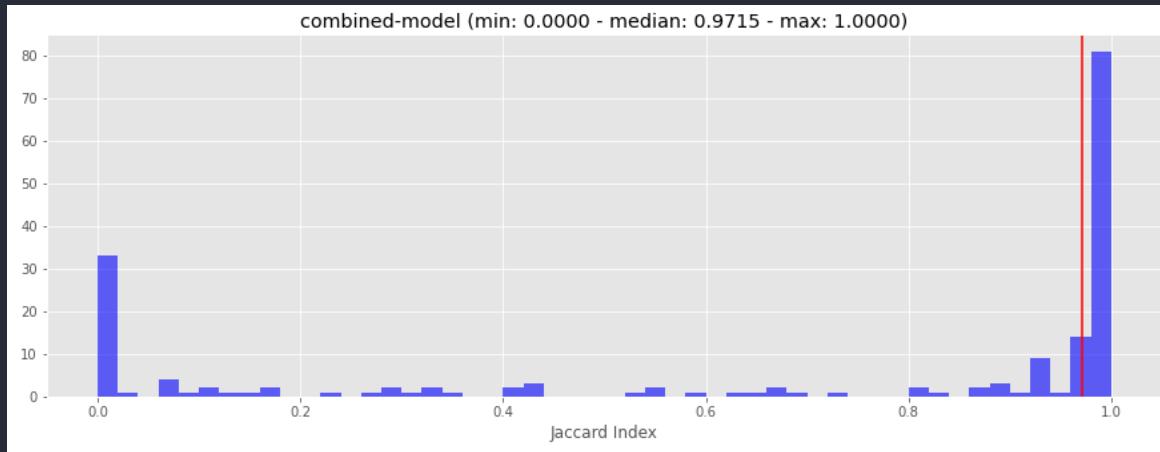
Test set contains 182 images from Sentinel-2 cloudless map service.

Test results II - U-net residual large with CRF model



Test set contains 182 images from Sentinel-2 cloudless map service.

Test results III – Ensemble model



Test set contains 182 images from Sentinel-2 cloudless map service.

DEEP WATER DASHBOARD

Next Steps

- Post-processing techniques like defrosting;
- Work with images with clouds;
- Use a pre-trained network to extract features;
- Collect more data from Sentinel-2 API;
- Calculate the volume of a given lake.

Repository

<https://github.com/maxbeber/deep-water>

Special thank you for our mentor [Marcus Jones](#) for the guidance, support and kindness.

Dependencies:

- Cython
- Dash
- Matplotlib
- NumPy
- Pillow
- Pydensecrf
- Rasterio
- Requests
- Tensorflow 2.4