

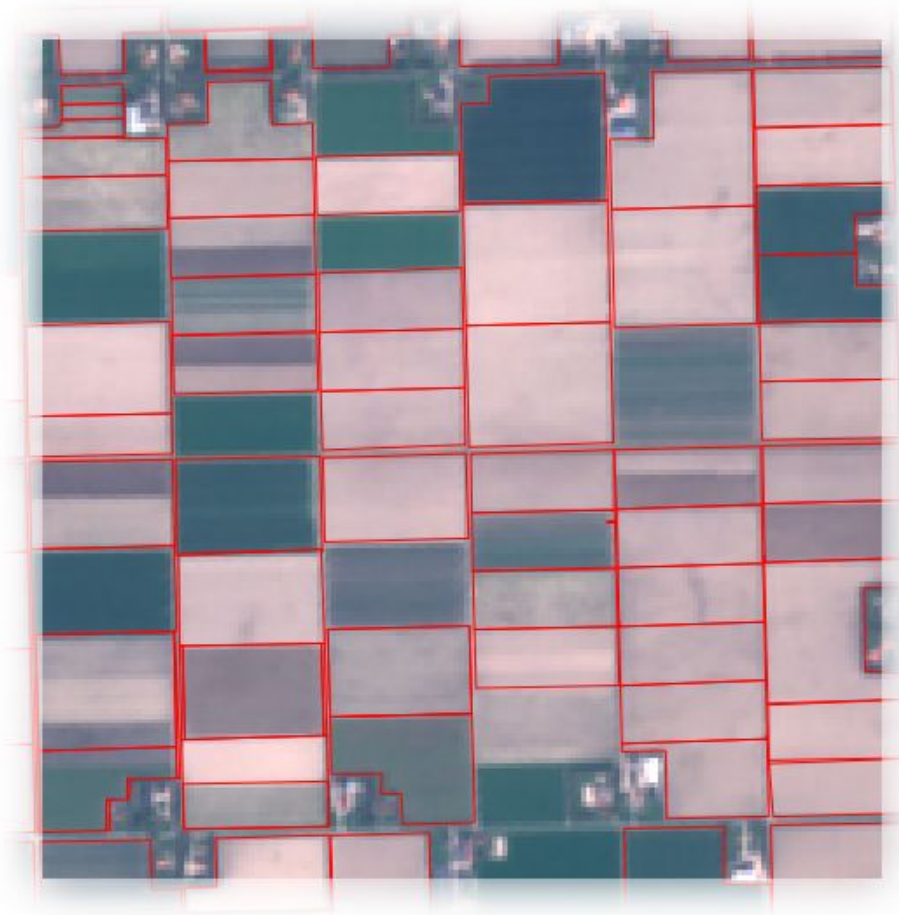
Deep Learning for Instance Segmentation of Agricultural Fields

supervised by

Prof. Dr. Christiane Schmullius (FSU Jena)

Adam Erickson (MPI BGC Jena)





Applications:

- ▶ Crop type / yield monitoring
- ▶ Subsidy management
- ▶ Precision farming

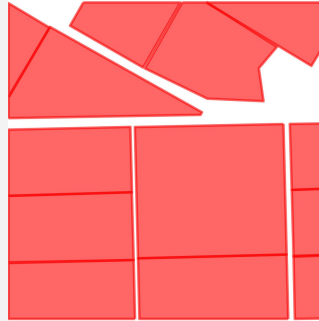
Existing datasets:

- ▶ LPIS (EU)
- ▶ CLU (US)

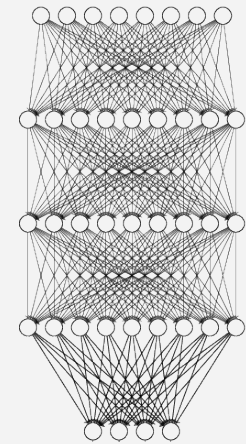
Training



Train Image

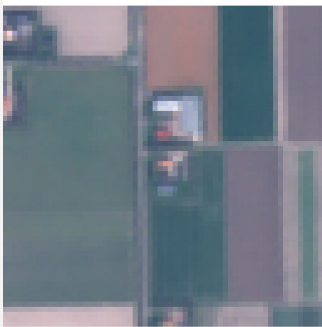


Train Fields

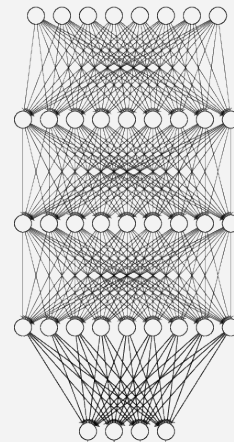


Trained Model

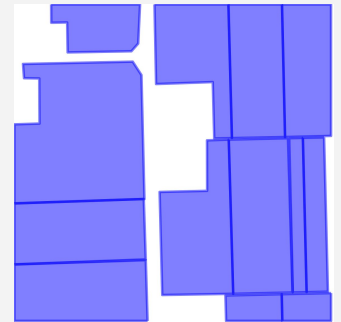
Prediction



Test Image



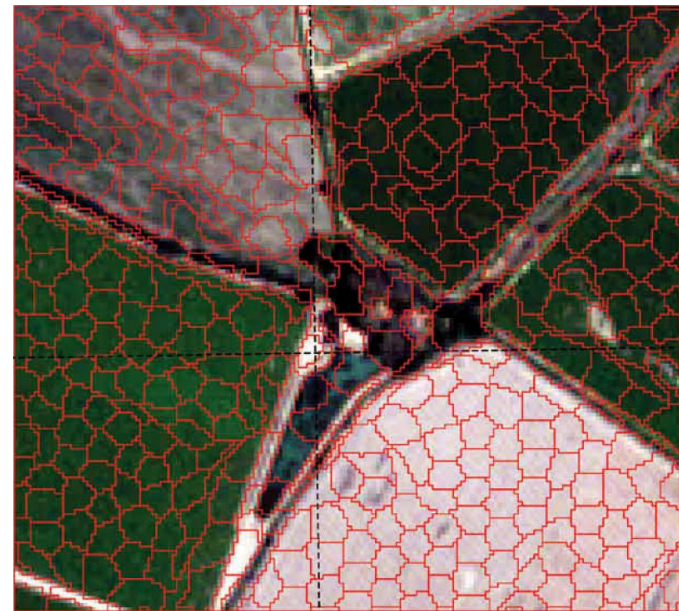
Trained Model



Predicted Fields

Traditional

- ▶ Homogenous regions
- ▶ Human engineered features & tresholds
- ▶ Iterative refinement
- ▶ Trained SVM (classifier)



© García-Pedrero et al. (2017)

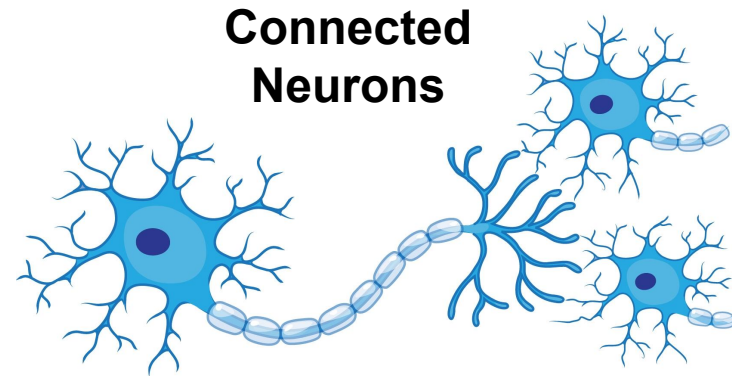
Challenges For Automatic Methods

5

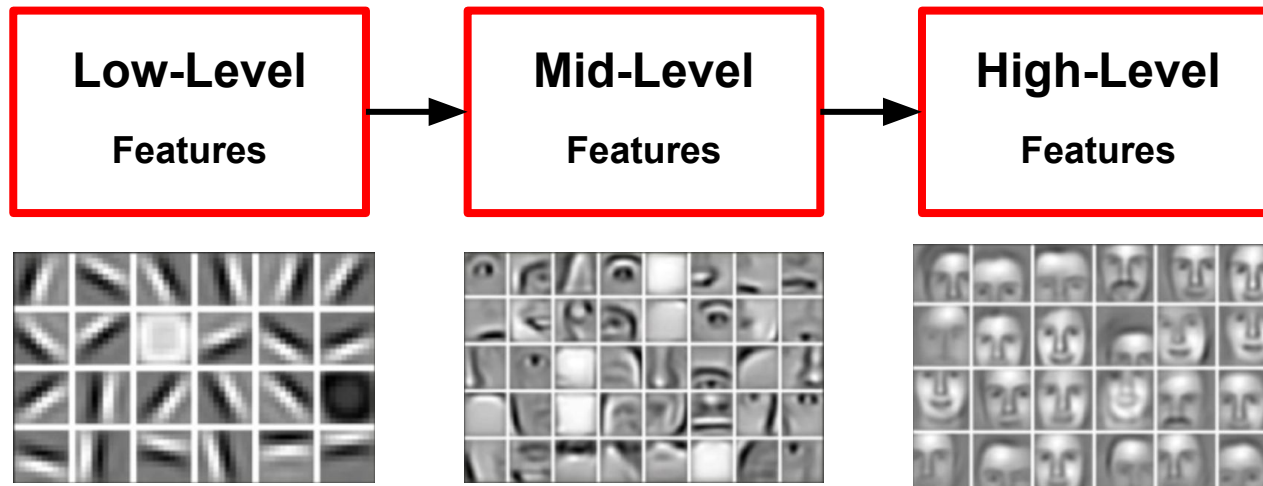




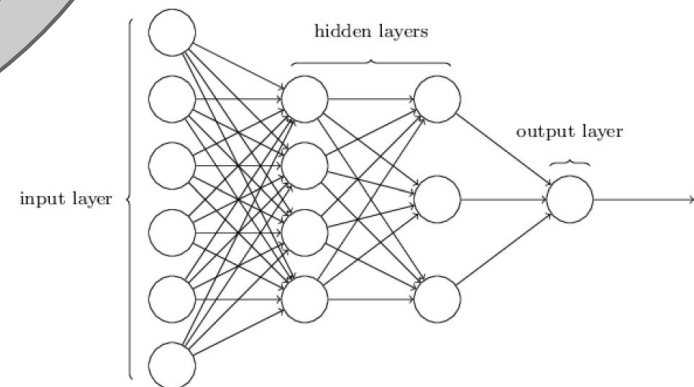
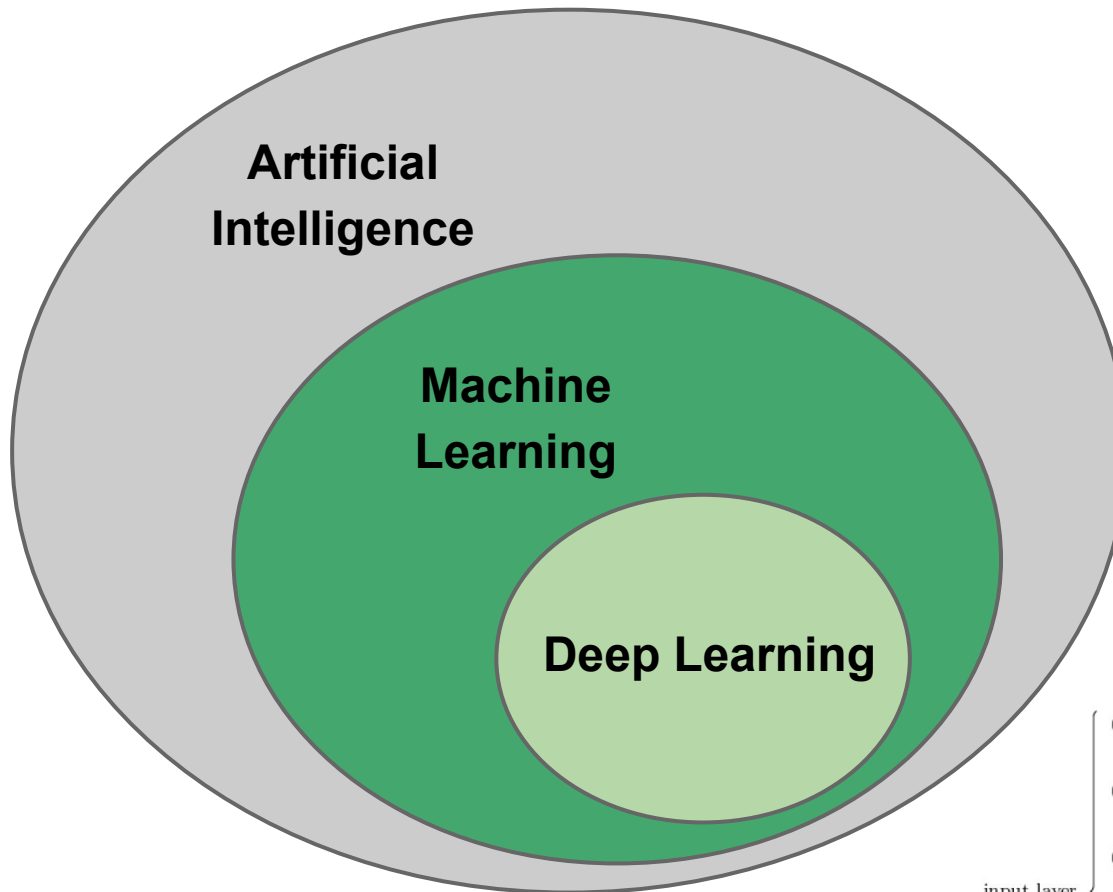
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© asu.edu.com



Adapted © Jones (2014)



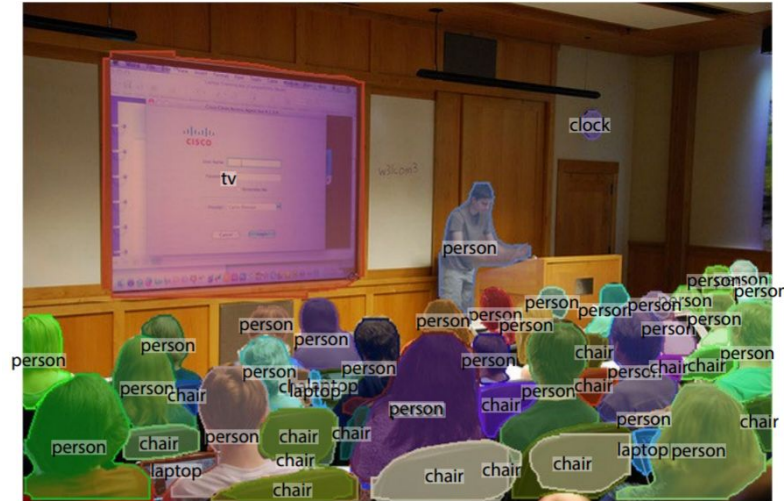
Neural Network with two hidden layers (Nielsen 2015)

Convolutional Neural Network

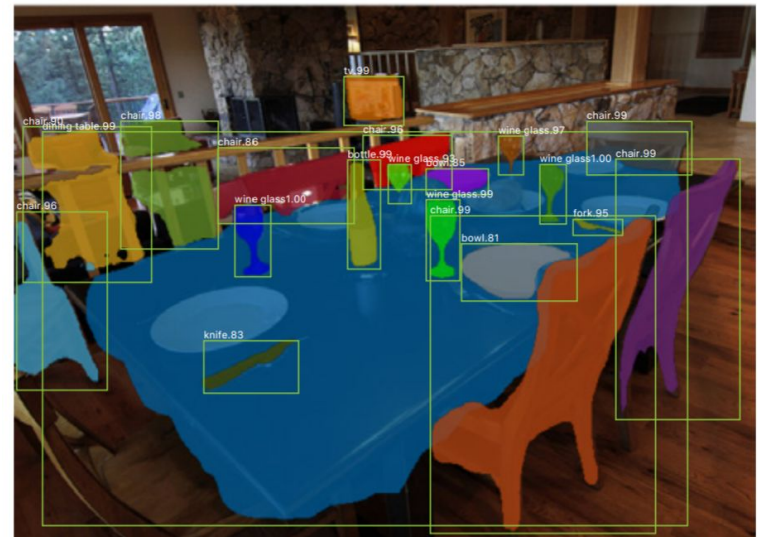
- ▶ Feature hierarchy
- ▶ Computer learned features
- ▶ Abstract features



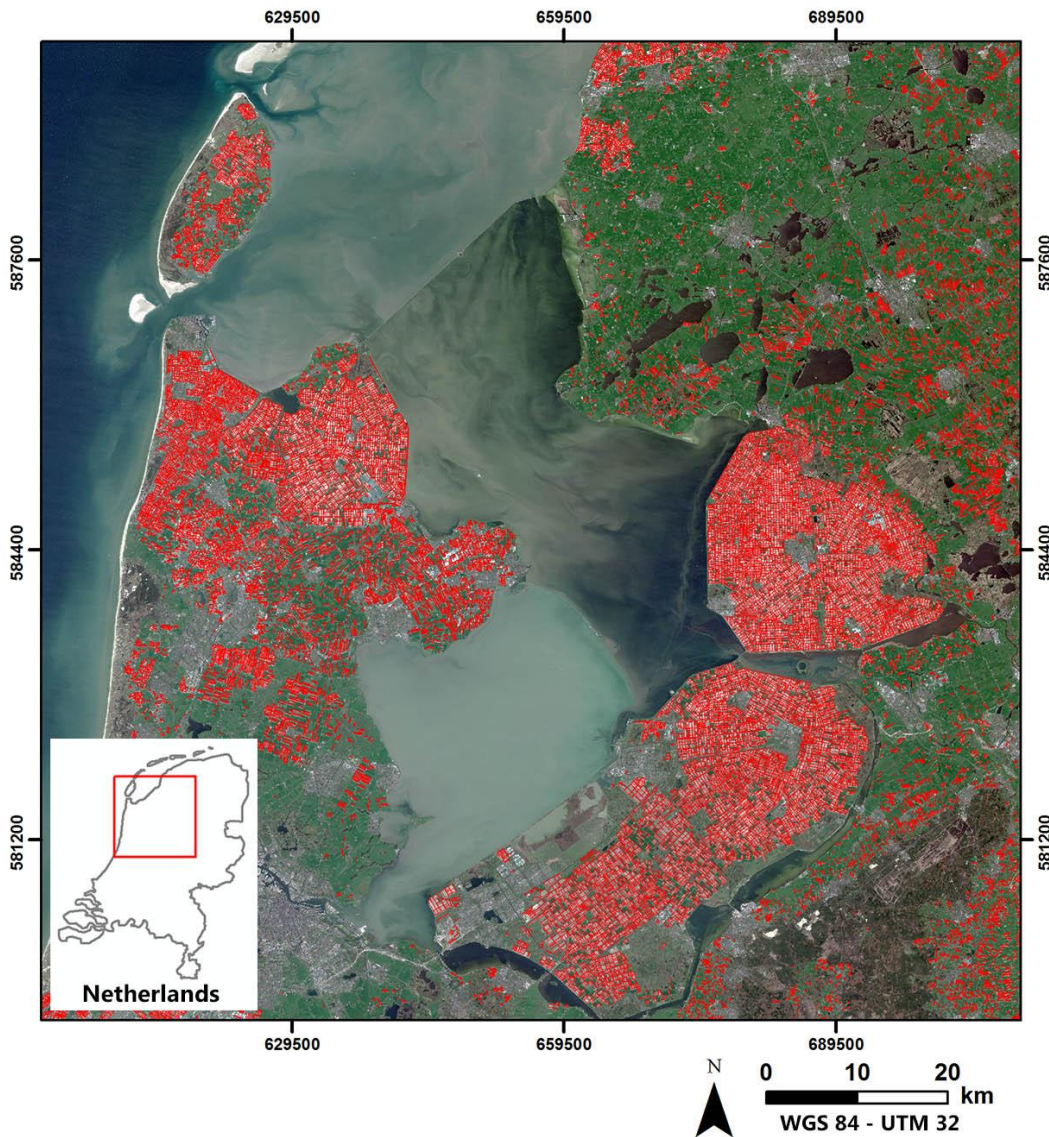
**Basis for
Instance Segmentation**



© Li et al. (2016)



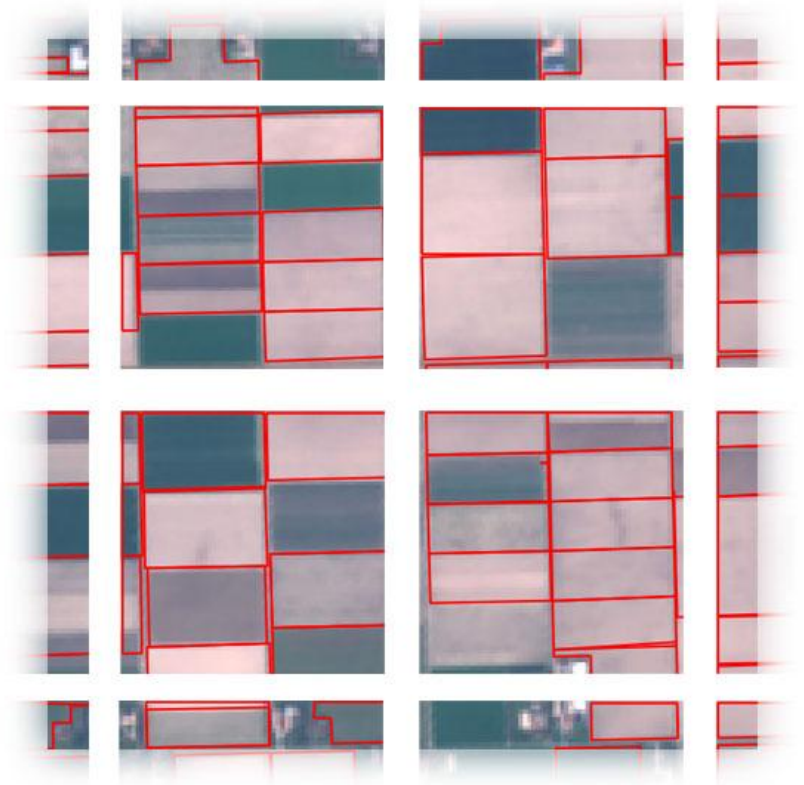
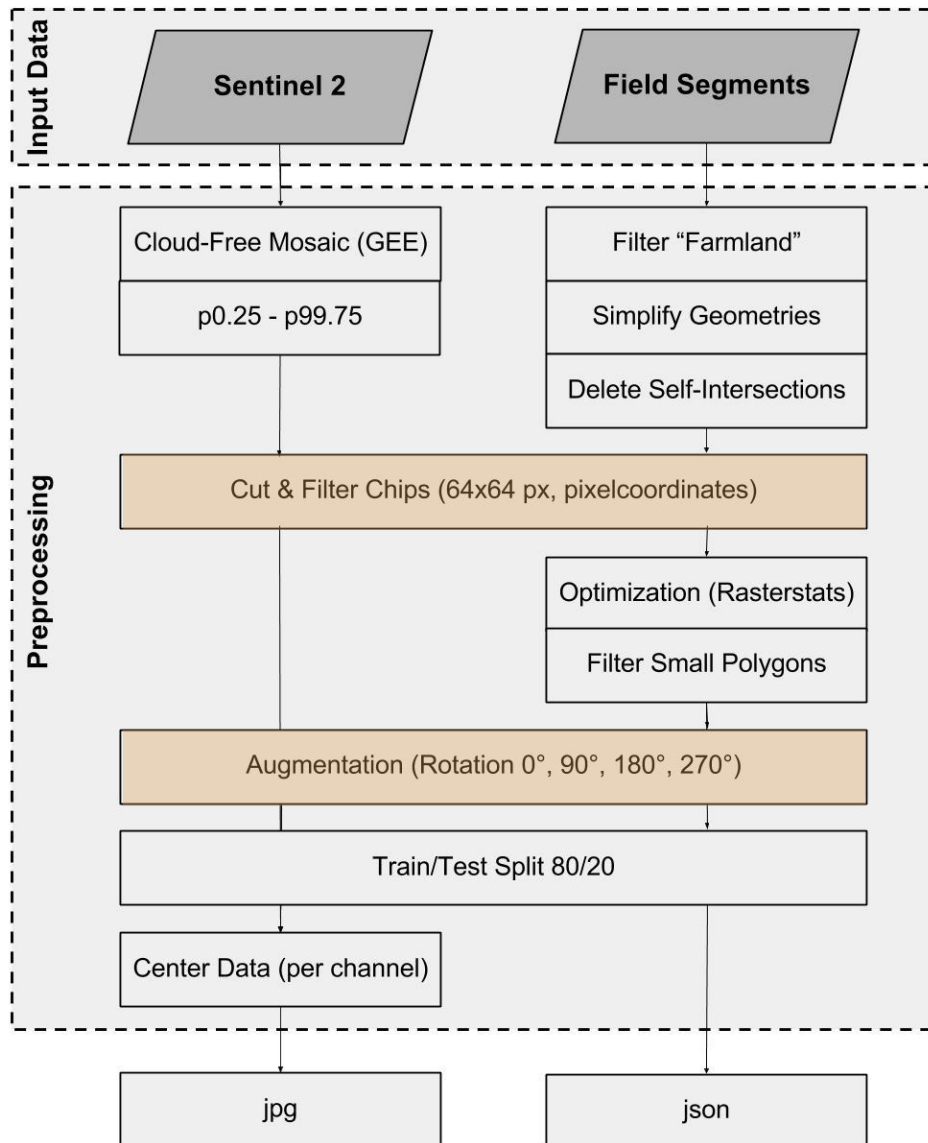
© He et al. (2017)



**2016 Basis Registration
Parcels (BRP)**

AOI:

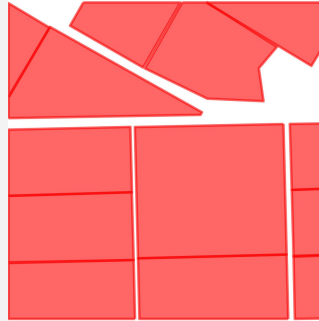
~ 40k farmland polygons



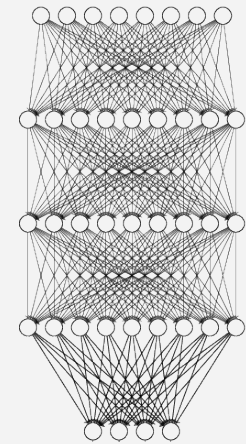
Training



Train Image

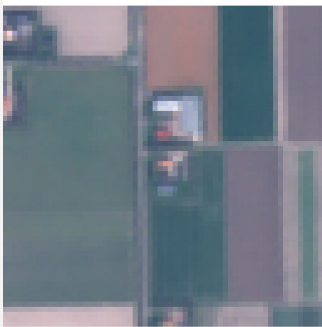


Train Fields

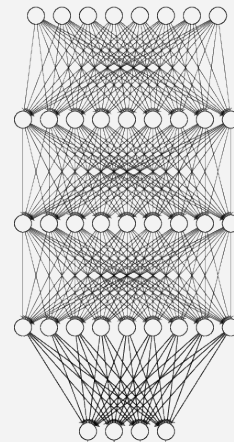


Trained Model

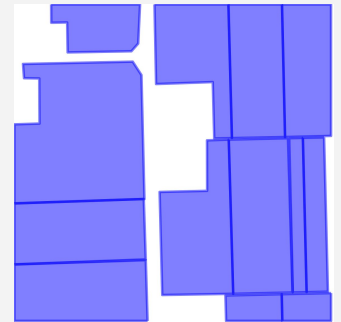
Prediction



Test Image



Trained Model



Predicted Fields

Deepmask

(Pinheiro et al. 2015)

**Predicts
Segment Proposals**

Sharpmask

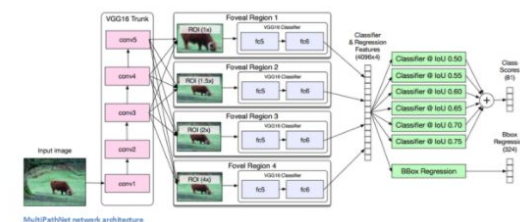
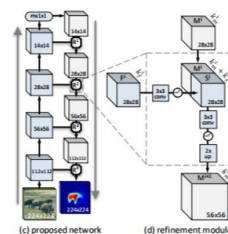
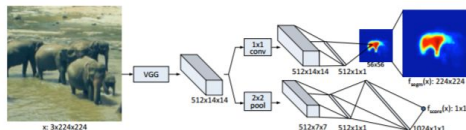
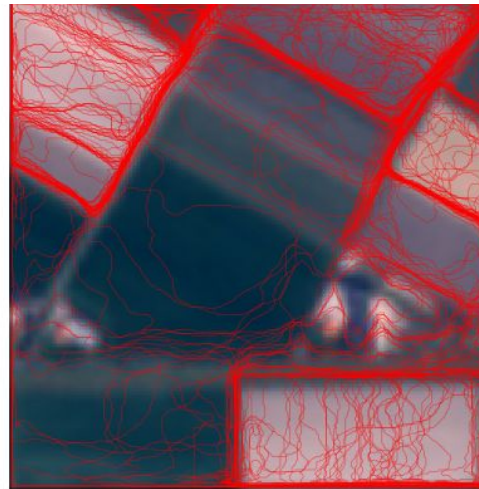
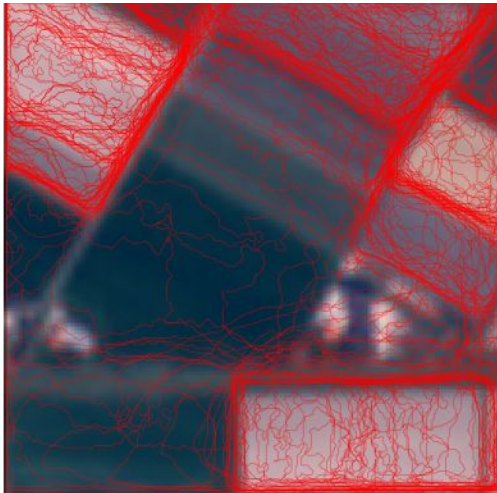
(Pinheiro et al. 2016)

Improves

MultiPath-Net

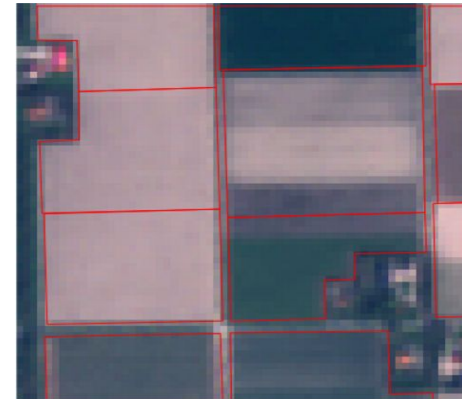
(Zagoruyko et al. 2016)

**Suppresses &
Classifies**



Challenges:

- ▶ Complex Training
- ▶ Partly property boundaries
- ▶ GPU cost > AWS



Outlook:

- ▶ Finish Multipathnet / Switch model (MNC, FCIS, Mask-RCNN)
- ▶ Crop classes
- ▶ Transferability (Thuringia usage mapping ?)
- ▶ (RGB vs. 8 bands)

Sources:

Girshick, R. & Donahue, J. & T. Darrell & Malik, J. (2014): Rich feature hierarchies for accurate object detection and semantic segmentation. IEEE Conference on Computer Vision and Pattern Recognition (CVPR). arXiv:1311.2524.

Pinheiro, P. & Collobert, R. & Dollar, P. (2015): Learning to Segment Object Candidates. arXiv:1506.06204.

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Zagoruyko, S. & Lerer, A. & Lin, T. & Pinheiro, P. & Gross, S. & Chintala, S. & Dollar, P. (2016): A MultiPath Network for Object Detection. arXiv:1604.02135.

Figures:

A. García-Pedrero, C. Gonzalo-Martín & M. Lillo-Saavedra (2017): A machine learning approach for agricultural parcel delineation through agglomerative segmentation. International Journal of Remote Sensing. 38. 1809 – 1819.

Jones, N. (2015): The learning machines. Nature 505, S. 146-148.

Nielsen, M. (2015). Neural Networks and Deep Learning. Determination Press.

He, K. & Gkioxari, G. & Dollár, P. & Girshick, R. (2017): Mask R-CNN. arXiv:1703.06870


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The banner features a dark blue header with the Topcoder logo on the left and navigation links 'THE CHALLENGE', 'PRIZES', 'REGISTER', and 'RESOURCES' on the right. The main background is a satellite image of a city with purple building footprints. The SpaceNet logo is centered above the main title. A large blue button with white text is positioned in the lower center. At the bottom, logos for CosmiQ Works, DigitalGlobe, and NVIDIA are displayed. A paragraph of text explains the challenge's purpose.

topcoder

THE CHALLENGE PRIZES REGISTER RESOURCES

SpaceNet

The SpaceNet™ Challenge Round 2

Compete to Create Next-Gen Geospatial Computer Vision Algorithms

Challenge is LIVE !!!

cosmiQ works

DigitalGlobe

NVIDIA

CosmiQ Works, DigitalGlobe and NVIDIA are challenging the Topcoder Community to develop automated methods for extracting building footprints from high-resolution satellite imagery. Such automated methods will help create more accurate maps, more rapidly.

<https://crowdsourcing.topcoder.com/spacenet>



CS231n: Convolutional Neural Networks for Visual Recognition

Spring 2017

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Neural Networks and Deep Learning is a free online book. The book will teach you about:

- Neural networks, a beautiful biologically-inspired programming paradigm which enables a computer to learn from observational data
- Deep learning, a powerful set of techniques for learning in neural networks

neuralnetworksanddeeplearning.com

Fully Convolutional Instance-aware Semantic Segmentation

Yi Li, Haozhi Qi, Jifeng Dai, Xiangyang Ji, Yichen Wei

(Submitted on 23 Nov 2016 (v1), last revised 10 Apr 2017 (this version, v2))

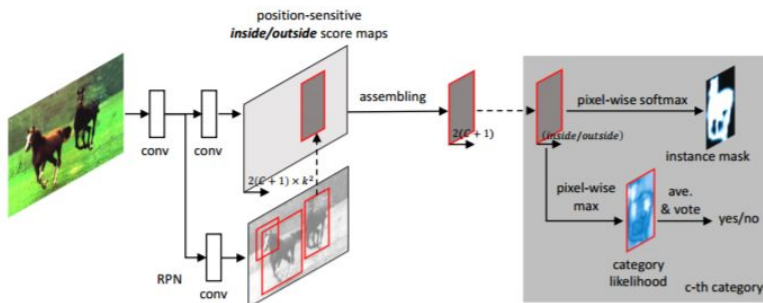


Figure 3. Overall architecture of FCIS. A region proposal network (RPN) [34] shares the convolutional feature maps with FCIS. The proposed region-of-interests (ROIs) are applied on the score maps for joint object segmentation and detection. The learnable weight layers are fully convolutional and computed on the whole image. The per-ROI computation cost is negligible.

Mask R-CNN

Kaiming He, Georgia Gkioxari, Piotr Dollár, Ross Girshick

(Submitted on 20 Mar 2017 (v1), last revised 5 Apr 2017 (this version, v2))

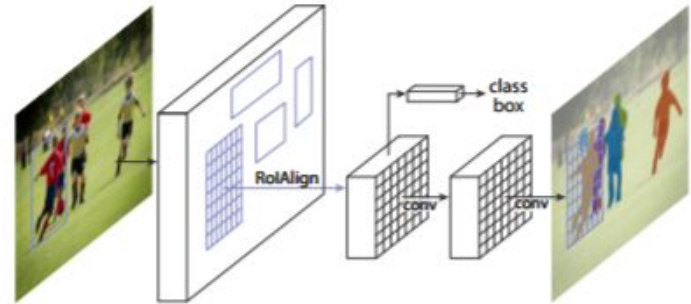
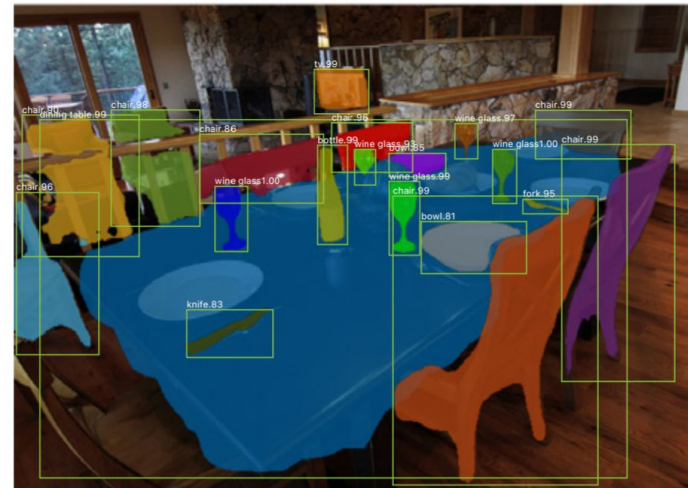
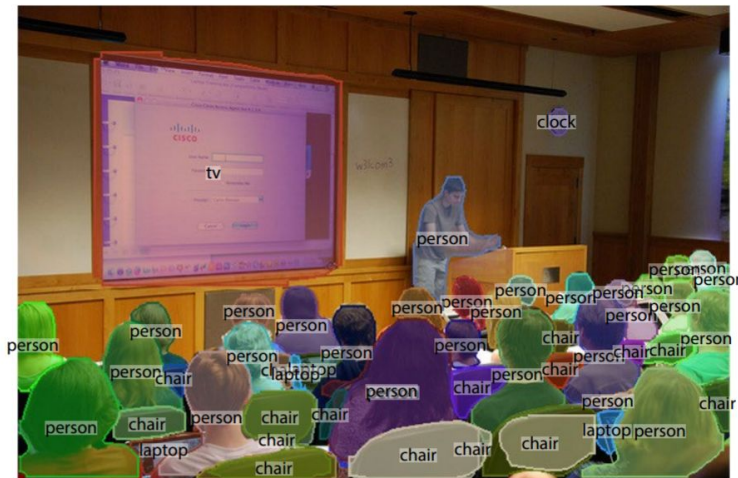
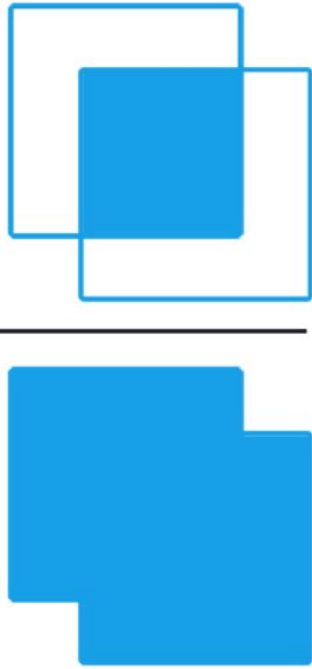
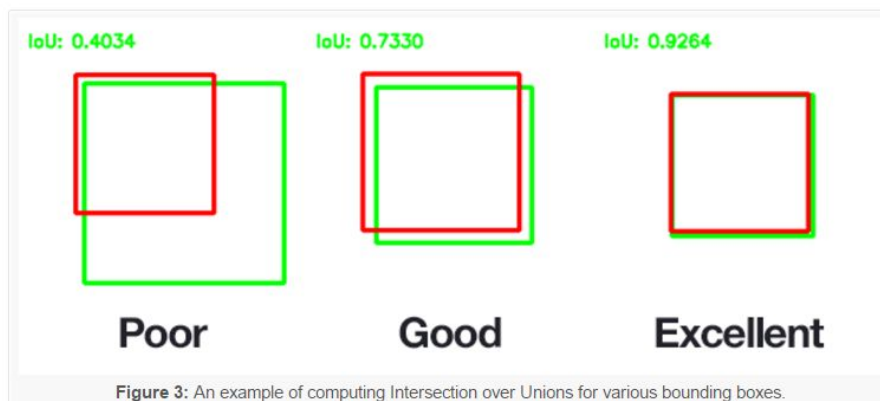


Figure 1. The Mask R-CNN framework for instance segmentation.



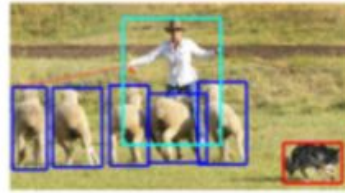
$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}}$$




<http://www.pyimagesearch.com/2016/11/07/intersection-over-union-iou-for-object-detection/>



(a) classification



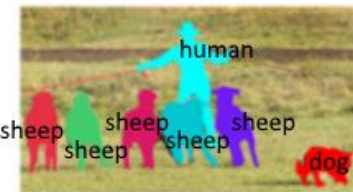
(b) detection



Semantic segmentation



Object Segment Proposal



Instance aware Semantic Segmentation

