

Sematic segmentation of bone implants – MDLMA project



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In cooperation with

 **Helmholtz-Zentrum
Geesthacht**
Zentrum für Material- und Küstenforschung



Multi-task Deep Learning for Large-scale Multimodal Biomedical Image Analysis

MDLMA Project

Motivation:

- characterization of degradable bone implants

Problem:

- many time-consuming steps to get final results:
image acquisition, -reconstruction, -enhancement
registration, segmentation and measurements

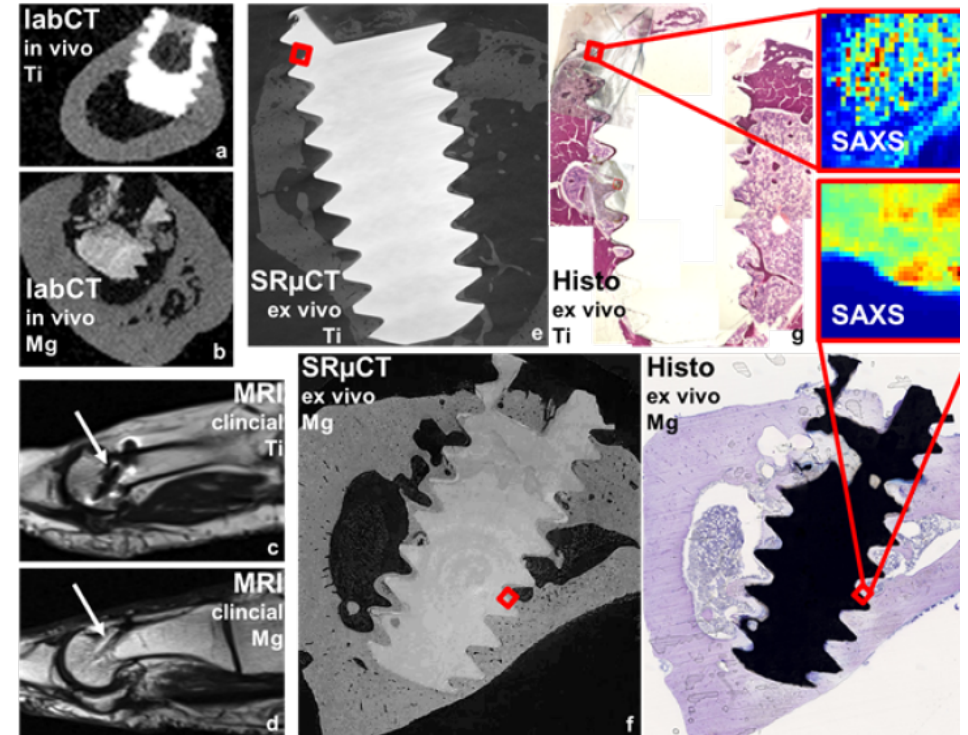
Goal:

- leverage multi-task to improve results of each individual task
- combine information of multimodal images

Project and industrial partners:



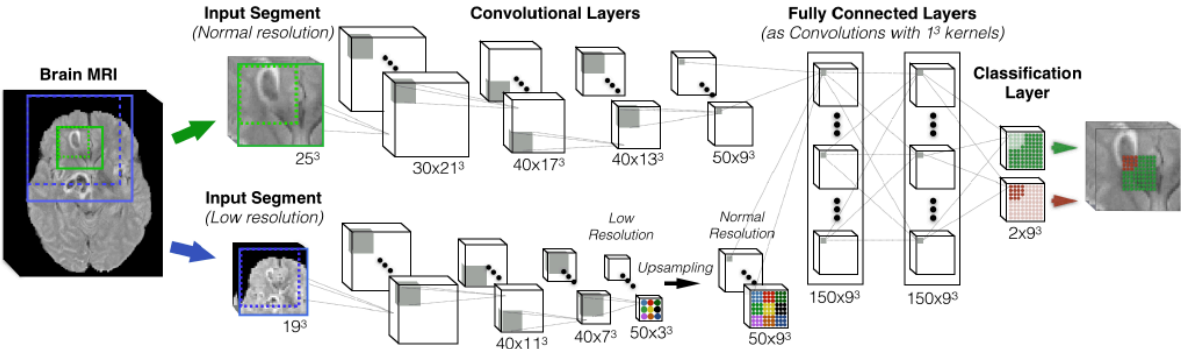
UNIVERSITÄT ZU LÜBECK



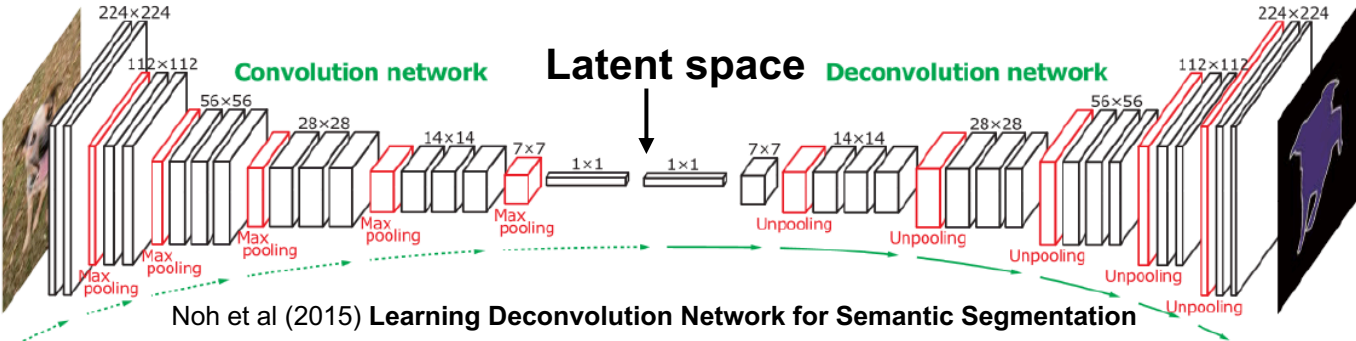
Semantic segmentation architectures

Multiple paths architecture:
local and global

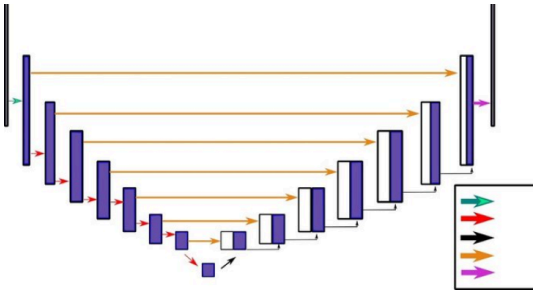
Encoder / Decoder architecture:
Latent space



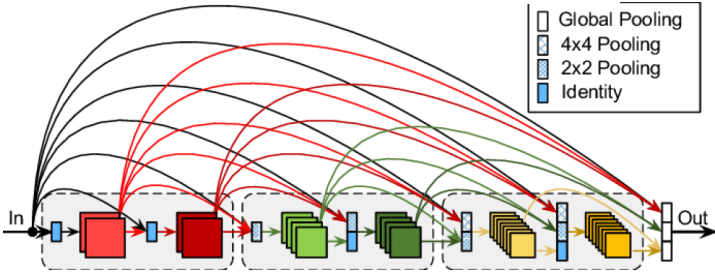
Kamnitsas et al (2016) Efficient Multi-Scale 3D CNN with Fully Connected CRF for Accurate Brain Lesion Segmentation



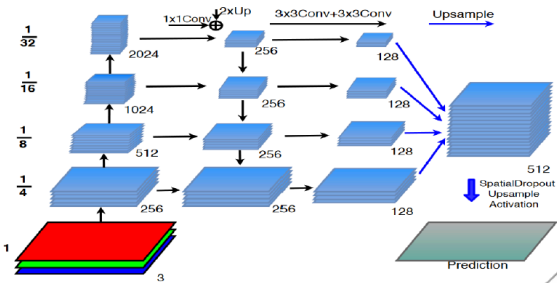
Noh et al (2015) Learning Deconvolution Network for Semantic Segmentation



U-Nets



Dense Nets as encoder



Feature Pyramid Nets

Rethinking receptive fields for high-dimensional SRμCT data

(baseline and work in progress)

- Classic U-Net architecture has a receptive field of **140 x 140 pixels** at the bottleneck layer
- Receptive field is the size of the region in the input that produces the feature
- In contrast the SRμCT data has $\approx 1500^3$ pixels -> one tenth of the input image contributes to each feature

Hypothesis:

- Simply increasing the input size is similar to increasing the batch size for small receptive fields
- To truly leverage the large input size (i.e. high resolution), the receptive field needs to be increased.
 - Stacking conv-layers, increase kernel size or employing dilated convolutions
 - Different combinations can increase the receptive field to 1080 x 1080 pixels



Very preliminary results

Dice:

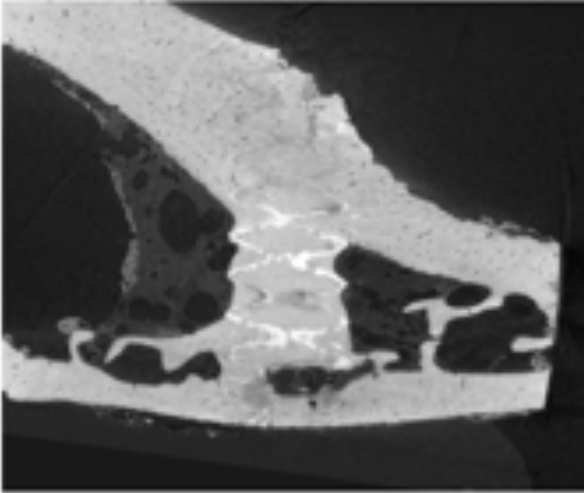
Avg = 0.933

Bone = 0.976

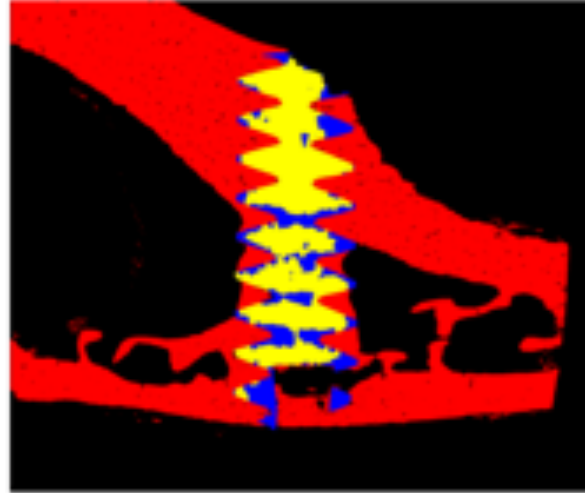
C. screw = 0.861

Screw = 0.961

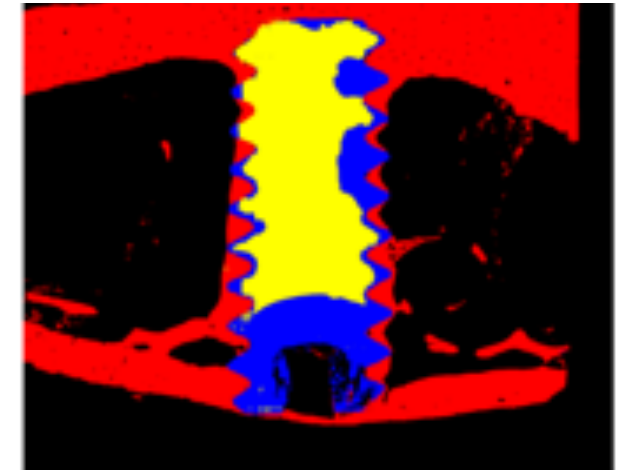
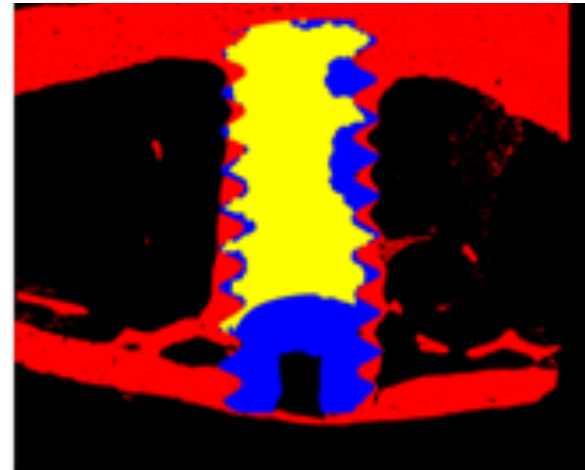
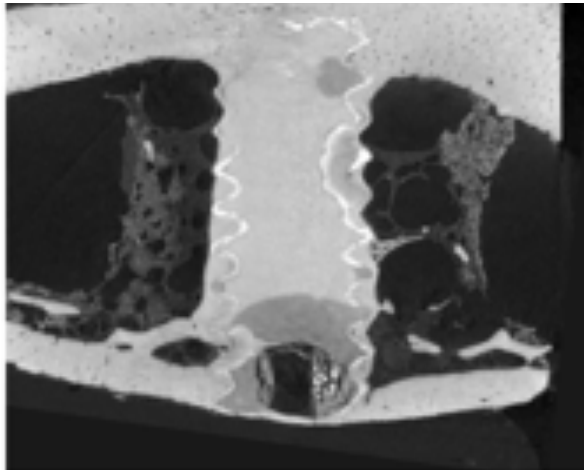
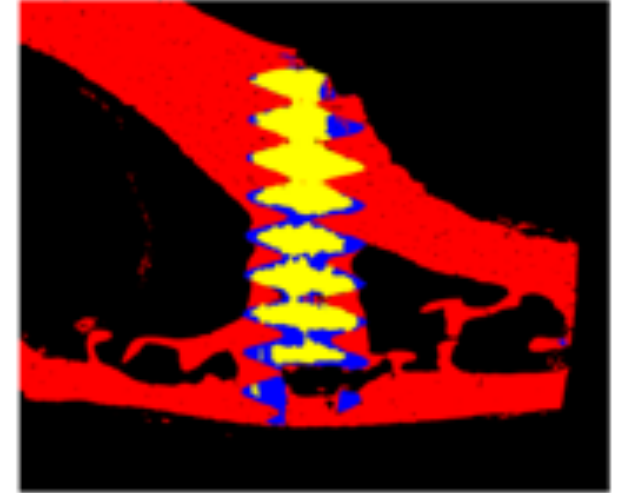
Input image slice



Workflow GT



Model prediction



Contact

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