

Automatic Segmentation of Head and Neck Tumors in PET-CT scans

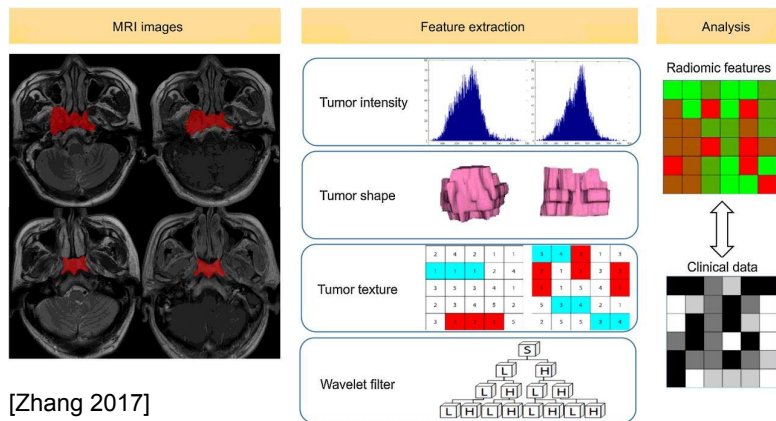
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MIDL 2020



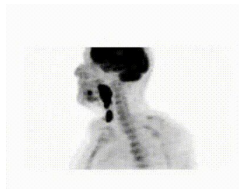
Introduction

- **Radiomics**: prediction of disease characteristics using quantitative image biomarkers from medical images



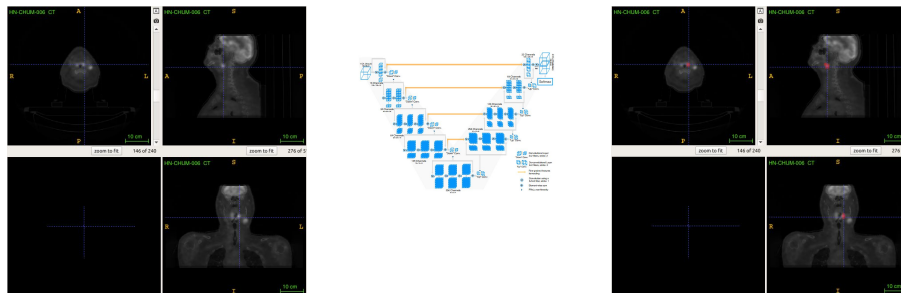
Example: Pre-treatment signature of a tumour region to predict response to treatment and survival time

Introduction



- Head and Neck (H&N) cancers: 5th leading cancer by incidence (Parkin et al. 2005)
- Radiomics studies based on PET/CT -> predict patients prognosis in a **non-invasive** fashion (Vallières et al. 2017),(Bogowicz et al. 2017),(Castelli et al. 2017)
- **Limitations:** Validated on 100-400 patients -> larger cohorts required for estimating generalization
- Manual annotations in 3D are tedious and error-prone

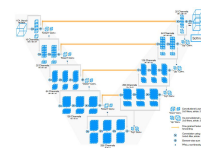
We need **automatic segmentation** of H&N tumor



Experiments design

Automatic H&N tumor segmentation in PET-CT images

- 203 PET-CT volumes with ground truth annotation (Vallières et al. 2017)
- Multi-centric (4 centers)
- Leave-one-center-out cross-validation
- U-Net (2D) vs V-Net (3D): CNNs for image segmentation
- PET vs CT vs PET/CT



Results

Table 1: Comparison of single modality and bimodal network performance of 2D and 3D V-Nets. The average DSCs, sensitivity and specificity (%) are reported with 95% CIs.

model	modality.	DSC	precision	recall
2D/3D	CT	48.7% ± 2.2 / 49.2% ± 2.2	52.7% ± 2.9 / 48.6% ± 2.6	54.1% ± 2.8 / 65.0% ± 2.6
2D/3D	PET	58.2% ± 2.3 / 58.6% ± 2.5	59.7% ± 3.0 / 59.1% ± 2.9	66.7% ± 2.9 / 70.2% ± 3.1
2D/3D	early fusion	58.5% ± 2.2 / 58.9% ± 2.3	58.1% ± 2.9 / 59.0% ± 2.9	70.2% ± 2.7 / 70.8% ± 2.8
2D/3D	late fusion	60.6% ± 2.1 / 59.7% ± 2.2	69.4% ± 2.6 / 62.8% ± 2.8	62.1% ± 2.9 / 69.1% ± 2.8

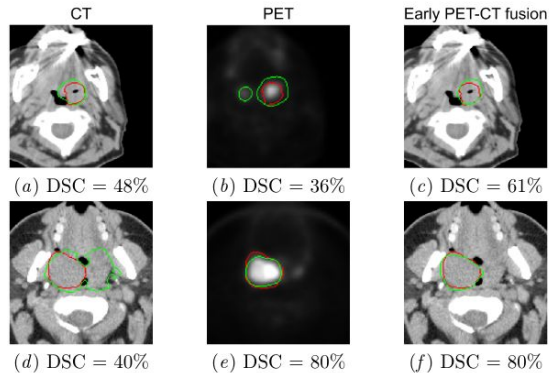


Figure 3: 2D axial examples of qualitative segmentation results (green) and ground truth (red) with their corresponding DSC (evaluated on the entire 3D volumes). Top row illustrates an example in which better prediction is obtained on the CT. Bottom row illustrates an example in which better prediction is obtained on the PET. (a,d) CT segmentation, (b,e) PET segmentation, (c,f) early fusion PET/CT segmentation.

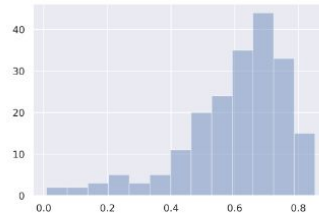


Figure 1: Count histogram of DSCs obtained by late 2D PET-CT fusion for the 4-center cross-validation.

Conclusion

- Automatic segmentation necessary for large scale radiomics studies
- Promising results obtained, more details in the paper: <https://openreview.net/forum?id=1QI71nEERx>
- HECKTOR challenge at MICCAI 2020: Cleaned and added data

<https://www.aicrowd.com/challenges/miccai-2020-hecktor>

Annotations to clean

