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- 1
- 2
- 3
- 4

**Active Organs Segmentation in Metastatic Breast Cancer Images combining Superpixels and Deep Learning Methods**

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**Hypothesis**

In the clinical follow-up of metastatic breast cancer patients, semi-automatic measurements are performed on <sup>18</sup>F<sup>18</sup>FDG PET/CT images to monitor the evolution of the main metastatic sites. Apart from being time-consuming and prone to subjective approximations, semi-automatic tools cannot make the difference between cancerous regions and active organs, presenting a high <sup>18</sup>F<sup>18</sup>FDG uptake.

In this work, we develop and compare fully automatic deep learning-based methods segmenting the main active organs (brain, heart, bladder), from full-body PET images.

**Methods**

We combine deep learning-based approaches with superpixels segmentation methods. In particular, we integrate a superpixel SLIC segmentation at different depths of a convolutional neural network, i.e. as input and within the optimization process. Superpixels reduce the resolution of the images, keeping sharp the boundaries of the larger target organs while the lesions, mostly smaller, are blurred. Results are compared with a deep learning segmentation network alone.

The methods are cross-validated on full-body PET images of 36 patients from the ongoing EPICURE<sub>seinmeta</sub> study. The similarity between the manually defined ground truth masks of the organs and the results is evaluated with the Dice score. Moreover, these methods being preliminary to tumor segmentation, the precision of the networks is defined by monitoring the number of segmented voxels labelled as "active organ", but belonging to a lesion.

## **Results**

Although the methods present similar high Dice scores ( $0.96 \pm 0.006$ ), the ones using superpixels present a higher precision (on average 6, 16 and 27 selected voxels belonging to a tumor, for the CNN integrating superpixels in input, in optimization and not using them, respectively).

## **Conclusion**

Combining deep learning with superpixels allows to segment organs presenting a high  $^{18}\text{F}$ FDG uptake on PET images without selecting cancerous lesions. This improves the precision of the semi-automatic tools monitoring the evolution of breast cancer metastasis.