# Image Registration for ACROBAT Challenge: Team MEDAL

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**Abstract.** We present a methodology of our submission for ACROBAT - a image registration challenge for histopathology whole slide images.

Keywords: Image registration  $\cdot$  Whole Slide Images  $\cdot$  Histopathology.

#### 1 Methods

We can divide our image registration pipeline in two stages, ie. local registration stage and local correction stage. We briefly describe working of both the stages in following subsections.

#### 1.1 Global Registration

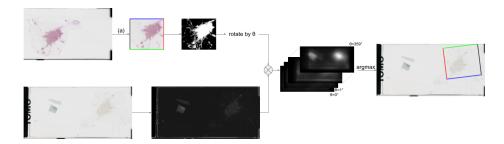


Fig. 1. Global registration estimates shift and rotation between HE and IHC using convolution between pre-processes IHC and rotated pre-processed HE crop. (a) Tissue detection for HE images is done using patch based classification model. Boundaries for HE tissue bounding box and IHC tissue bounding box are colored to highlight orientation.

We detect tissue in HE image using a patch classification model trained primarily on patches BRIGHT dataset. We re-train it with patches from AC-ROBAT dataset. We consider epithelial and stromal region as foreground. After tissue detection we detect tissue dense region in HE images as shown in Figure 2. Specifically, we create a density map of foreground in X and Y direction. We

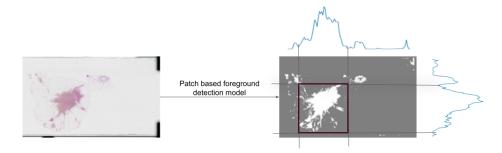


Fig. 2. Tissue detection and bounding box for tissue dense region

cut the HE whole slide image where density map crosses value lesser than 5% of maximum value for a density map.

Next, we estimate shift and rotation of tissue in IHC image with respect to tissue in HE image by template matching using convolution operation. We invert HE bounding box image to get  $H_i$ . Similarly we invert IHC image to get  $I_i$ . We replace black-like pixels in IHC with white pixels to avoid getting maxima at edges. We rotate  $H_i$  from 0° to 360° to get output tensor T of size  $360 \times I_{ih} \times I_{iw}$ , where  $I_{ih}$  and  $I_{iw}$  are height and width of  $I_i$  respectively. Finally we get global registration of HE and IHC images by taking argmax of T. All the operations at this stage are carried out using images down-sampled by 32 from level 0.

$$\theta^*, x^*, y^* = argmax(T) \tag{1}$$

#### 1.2 Local Corrections

After obtaining parameters for global registration, we try to estimate local correction to be applied across HE image. We estimate corrections to be applied in X-axis and Y-axis independently. First, we sample the points,  $H_p = h_{p1}, h_{p2}, \ldots$  on HE tissue edges using foreground mask obtained using foreground detection model. We then use global registration parameters  $\theta^*, x^*, y^*$  to get corresponding point  $i_p$  in IHC image.

$$i_p = f(h_p, \theta^*, x^*, y^*)$$
 (2)

We employ similar strategy as used in global registration to get local corrections. We take  $256 \times 256$  sized patch from HE image and  $640 \times 640$  sized patch from IHC image around  $h_p$  and  $i_p$  respectively. These patches are down-sampled to factor of 16 from level 0. We invert both of these patches and perform convolution operation to get output tensor of size  $384 \times 384$ . argmax operation on this tensor gives local correction to be applied for point  $h_p$ . For around 50% of points we do not get desired output from this operation. argmax of convolution outputs coordinates on the edges if tissue dense region present in IHC patch. We remove those points from further analysis. Specifically, if local correction in

any axis crosses 150 pixels of correction, we discard those points. Further, we remove points with sharp maximas.

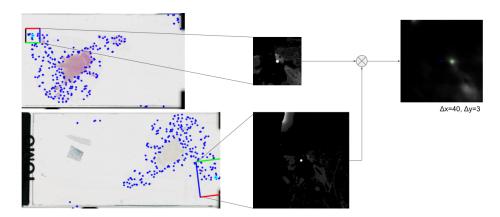


Fig. 3. Points on HE tissue edge are sampled using binary mask obtained from tissue detection model. Corresponding IHC point is obtained using global registration parameters. Square patch around sampled point in HE is inverted. Similarly, a bigger patch around a point from IHC is inverted. argmax of convolution gives local correction  $\Delta x$  and  $\Delta y$ 

**Interpolation:** We use radial basis function based interpolation for local corrections in X-axis and Y-axis independently to obtain local correction maps  $X_c$  and  $Y_c$  across HE image as shown in Figure 4. Finally, we use global registration parameters  $\theta^*, x^*, y^*$  along-with  $X_c$  and  $Y_c$  to output the registration map.

## References

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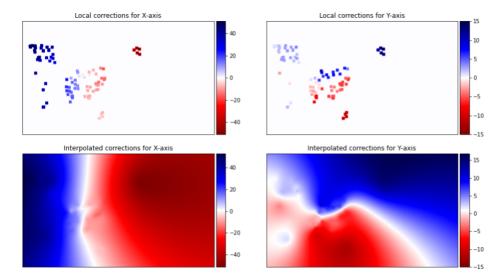


Fig. 4. Interpolation of local corrections in X-axis and Y-axis. Number in color-map correspond to down-sampled numbers by factor 16.