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🌐 CODA (part 6): register the nuclear coordinates and construct 3D cell matrix | HuBMAP | JHU-TMC

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ABSTRACT

Using the registration transformations calculated in low-resolution (as described CODA (part 2): calculate registration on low-resolution tissue images), register the nuclear coordinates generated in CODA (part 5): nuclear coordinate generation

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We use this protocol and it's working

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register the nuclear coordinates

- 1 As input, this function needs the path to the low-resolution images that were registered in [CODA-part2](#) (pth1x), the path containing the coordinates calculated in [CODA-part4](#) (pthcoords), and the scale factor between these images. For cell detection performed in 10x and registration calculated in 1x,

scale=10;
register_cell_coordinates(pth1x,pthcoords,scale);

*if the yellow dots do not line up over the tissue, there is a problem in your code.
- 2 This function will create a subfolder inside pthcoords named 'cell_coordinates_registered.' Inside that subfolder will be a mat file corresponding to each tif image. Inside each mat file will be a variable named 'xy' (unregistered cell coordinates), 'xyg' (globally registered cell coordinates), and 'xye' (elastically registered

cell coordinates), all saved at the same low-resolution as the original registration images used in [CODA-part2](#).

```
pthcoordsE=[pthcoords,'cell_coordinates_registered'];
```

construct 3D cell matrix

- Next, create **volcell**, a matrix containing the nuclear labels in a volumetric matrix the same size and resolution of the tissue matrix constructed in [CODA-part4](#). To call this function, you need to define the folder containing the registered, classified images created in [CODA-part4](#), the folder containing the registered cell coordinates, the folder containing the tissue matrix created in [CODA-part4](#), and the scale between the high-resolution, classified images and the high-resolution images used for cell detection (this is probably 1).

- Define the folders containing the registered, classified images, the registered nuclear coordinates, and the tissue volume matrix:

```
pthclassifiedE=[pthclassified,'registeredE'];
pthcoordsE=[pthcoords,'cell_coordinates_registered'];
pthvolume=[pth,'lung_data'];
```

- Next, define the scale between the high-resolution images used for tissue classification in [CODA-part3](#) and the high-resolution images used for cell detection in [CODA-part5](#). If the classification and cell detection were applied to the same resolution images (this is likely), the scale is 1.

```
scale=1;
```

- Finally, **nwhite** is the whitespace class from the deep learning model. Here, let's again use **nwhite=6** (the background class number from the sample lung model).

```
nwhite=6;
```

- Now, call the **make_volcell** function:

```
build_cell_volume(pthclassifiedE,pthcoordsE,pthvolume,scale,nwhite)
```

This function will create a 3D matrix containing nuclear coordinates. The matrix will be at the same resolution as the tissue volume matrix, and automatically crop out the same background space that was manually selected for removal in [CODA-part4](#).

- To validate that the function creates a variable **volcell** that is the same resolution and crop as the tissue matrix **vol**, the function will display two images.

1. the function will display an overlay of the center image from **vol** and **volcell**.

2. the function will display an overlay of the z-projections of **vol** and **volcell**.

If either of these images do not appear overlaid (you see two separate, unaligned tissues), there was a problem with the construction of either **vol** or **volcell**.

9 Beyond the scope of this guide, consider **volshow** or **patch**

Notes on quantification

10 Primarily beyond the scope of this guide, as most spatial calculations must be custom to the biological question. Some basic considerations are in the following sections

Smooth data

11 To smooth your data. Your segmentation algorithm and 3D volumetric matrix will likely contain small false positive and false negative data. Try smoothing your volumetric data with the following.

vol = volumetric matrix of size [m n z]
containing 6 labels for 5 tissue structures plus background

ws = 6; % background class for volumetric matrix

to lightly smooth the tissue labels inside vol:

volS=vol; %

create a smoothing variable

for b=1:max(vol(:))

tmp=vol==b;

tmp=imclose(tmp,strel('sphere',2));

% morphological closing to fill small holes

tmp=imopen(tmp,strel('sphere',1));

% morphological opening to remove small noise

```
tmp=bwareaopen(tmp,500);
% delete 3D objects that are fewer than 500 voxels;
volS(volS==b)=ws;
volS(tmp==1)=b;
end
```

Note: be VERY CAUTIOUS using **imopen** to smooth thin structures (glands and vasculature), as **imopen** may eliminate “true positive” thin walls. Consider skipping **imopen** when you smooth

Quantify volume

12 To quantify volumes. volume of tissue class 2 (use **vol** or your smoothed **volS**):

```
sxy=4; %
resolution of 'vol' matrix in xy in units of micron / pixel
sz=4; %
distance between serial histological sections
volume_type_2=sum(vol(:)==2)*sx*sx*sz;
% volume of type 2 in units of micron3
volume_type_2=volume_type_2/(10^9);
% volume of type 2 in units of mm3
```

Quantify cellularity

13 To quantify cellularity. number of cells in class 2 (use **vol** or your smoothed **volS**):

```
tmp=double(volcell).*double(vol==2);
cell_type_2=sum(tmp(:));
```

And more

14 For more complex quantifications:

Consider the functions **bwdist**, **regionprops3**, and custom functions you create.