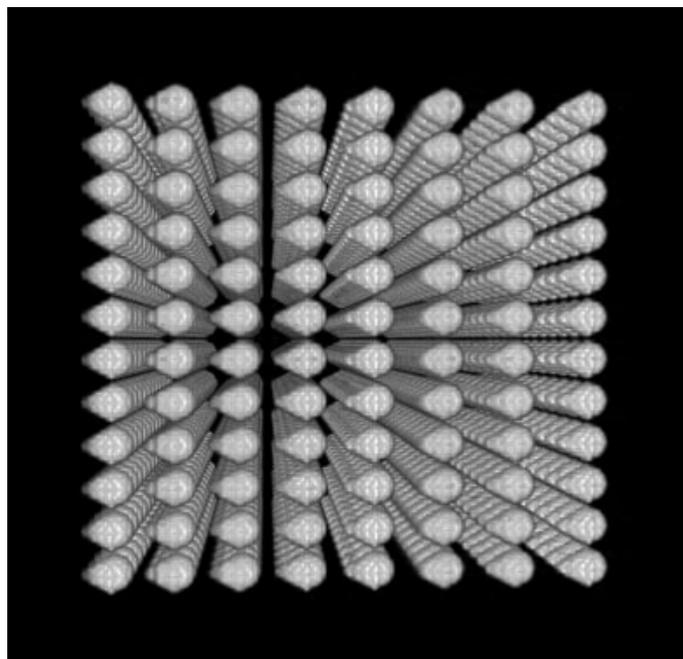


## Supplementary Material

### 1 CONCEPT OF ANISOTROPY

Validation of machine learning model in our case is a difficult process as we do not have ground truth data available. As nanoporous gold is self-similar, we should see isotropy in all directions. We used two functions, namely the two-point correlation function and the lineal path function, to measure anisotropy in the microstructures. The two-point correlation function is the autocorrelation function showing the probability of finding another point of the same phase in the neighbourhood. Similarly, the lineal path function shows the probability of having two points in the same cluster. For the isotropic structures, both functions in all directions should remain very similar.

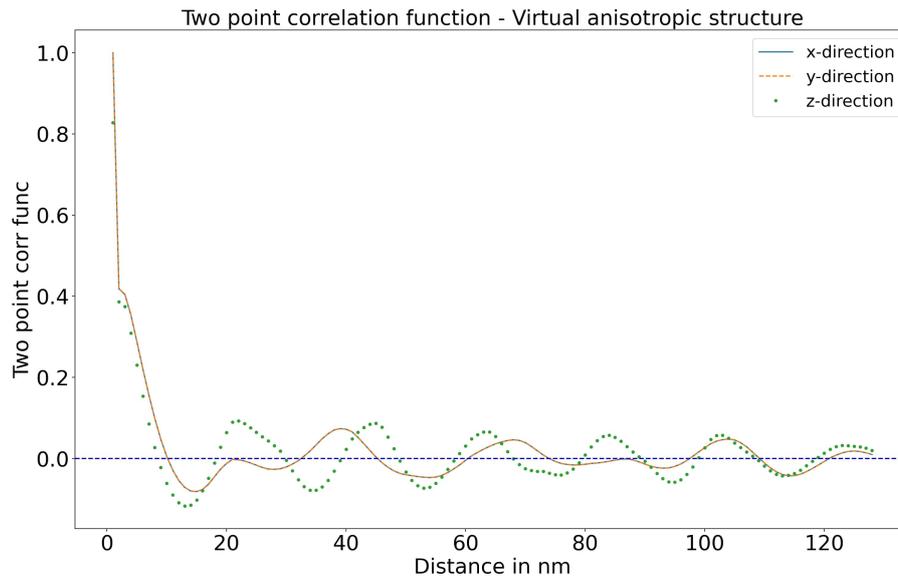
To test how well both functions work to detect anisotropy, we prepared one virtual structure with controlled anisotropy in the z-direction. This structure contained spheres stretched in the z-direction, making them elliptical in the +z direction. The intensity of the stretched part increases by going deeper in the z-direction to mimic the shine-through effect. Distances between each sphere layer in z-direction also change (Figure S1). These changes make the virtual structure isotropic in x and y-direction while anisotropic in the z-direction.



**Figure S1.** 3D view of virtual anisotropic structure

#### 1.1 Test of isotropy – Two point correlation function

We calculated two-point correlation functions (Berryman, 1985) for prepared virtual structure in x, y and z-directions. As expected, we get the same two-point correlation functions in x and y-directions (yz and xz planes), but we get significantly different functions in the z-direction (xy plane) (Figure S2).



**Figure S2.** Two-point correlation function plots of virtual anisotropic structure in z-dir (xy plane), y-dir (xz plane) and x-dir (yz plane)

## 1.2 Test of isotropy – Lineal path function

(Torquato and Haslach Jr, 2002) proposed inserting lines of arbitrary length and orientation in the image and then counting the fraction of these lines that lie wholly within a single phase to calculate the Lineal path function. This function can also be useful to calculate average ligament size.

We have calculated a total of two lineal path functions whose directions are perpendicular to each other. Different lengths and errors between two functions suggest anisotropy in the structure. We repeated this procedure for all three planes.

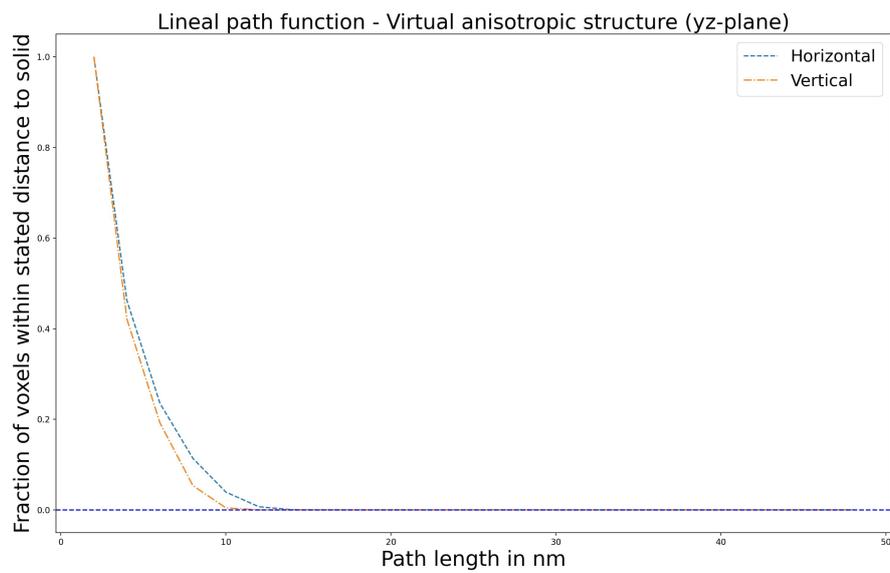
As we know, there is some anisotropy in the z-direction; we can see length differences in the function in yz-plane (Figure S3).

## 2 SUPPLEMENTARY TABLES AND FIGURES

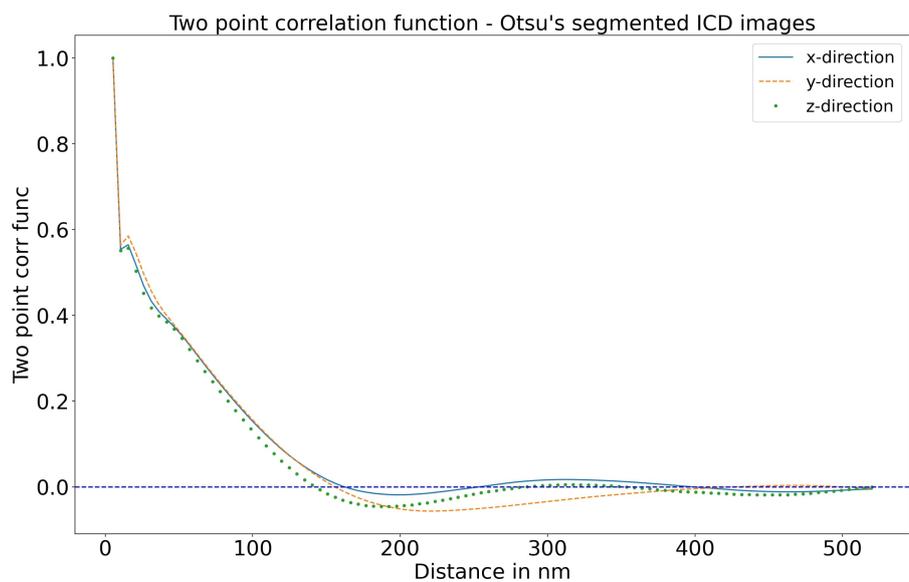
### REFERENCES

- Berryman, J. G. (1985). Measurement of spatial correlation functions using image processing techniques. *Journal of Applied Physics* 57, 2374–2384
- Torquato, S. and Haslach Jr, H. (2002). Random heterogeneous materials: microstructure and macroscopic properties. *Appl. Mech. Rev.* 55, B62–B63

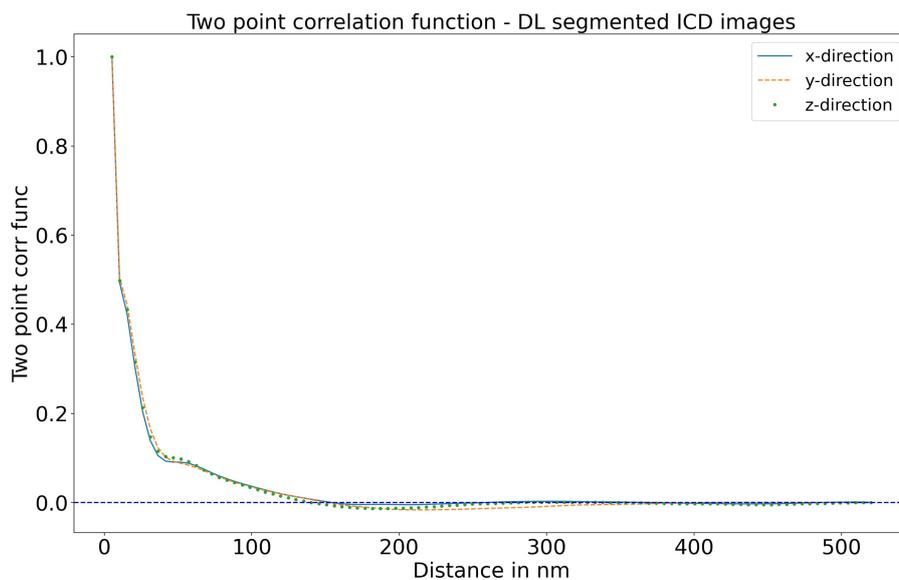
### 2.1 Figures



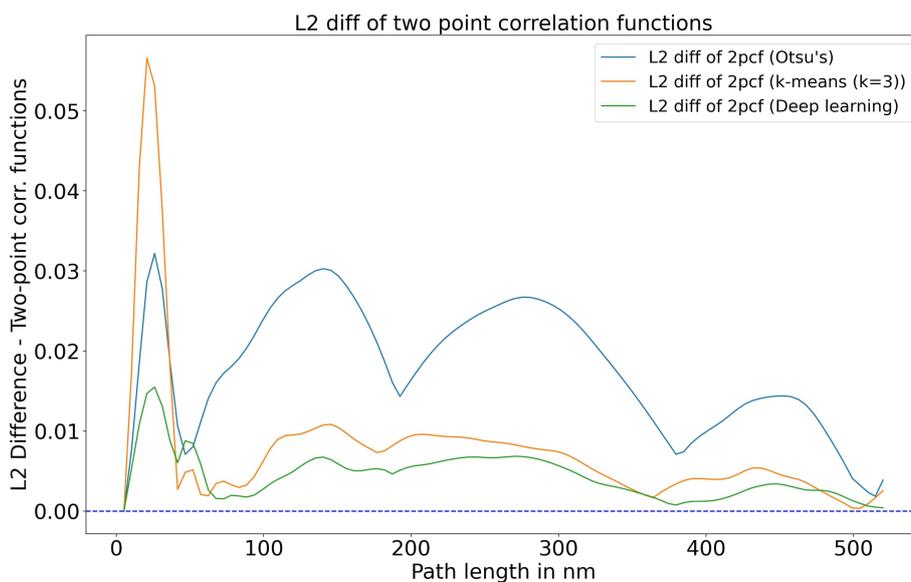
**Figure S3.** Lineal path function plots of virtual anisotropic structure in yz-plane



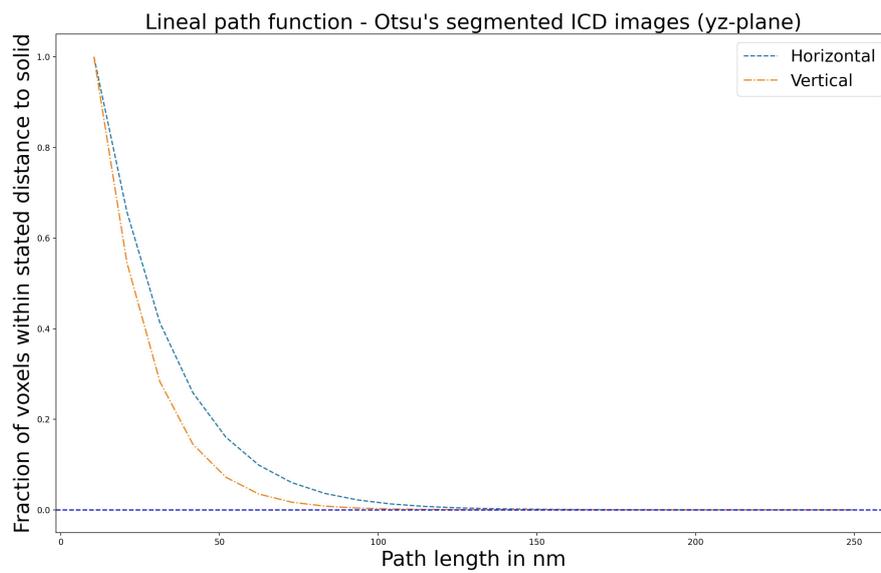
**Figure S4.** Two-point correlation function plots of segmented real hnpgepoxy ICD dataset (using Otsu's method) in z-dir (xy plane), y-dir (xz plane) and x-dir (yz plane)



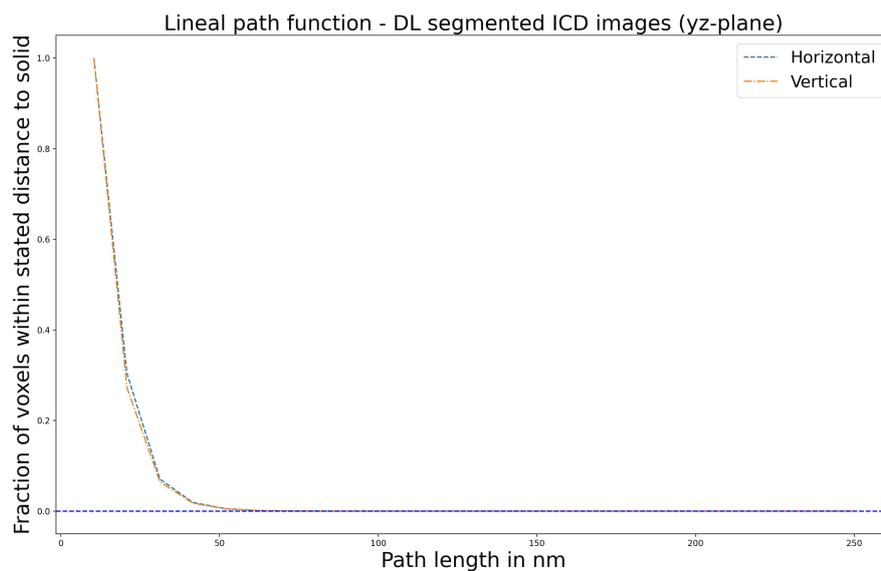
**Figure S5.** Two-point correlation function plots of segmented real hnp<sub>g</sub>-epoxy ICD dataset (using our deep learning method) in z-dir (xy plane), y-dir (xz plane) and x-dir (yz plane)



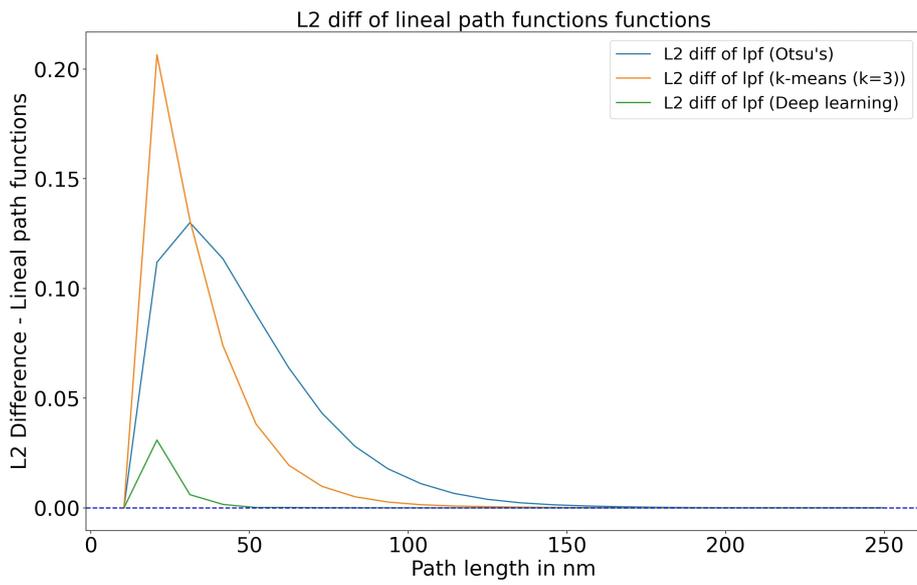
**Figure S6.** L2 differences of two-point correlation functions of segmented structures using Otsu's, k-means and our deep learning method



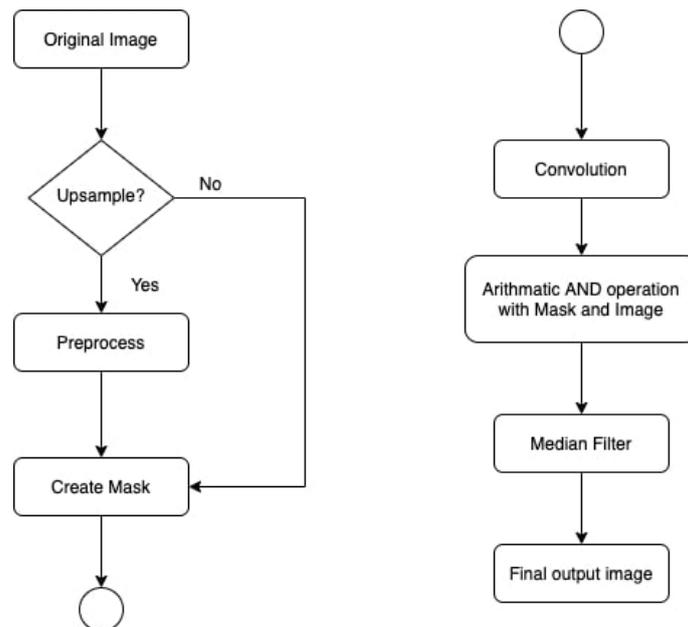
**Figure S7.** Lineal path function plots of segmented real hnp<sub>g</sub>-epoxy TLD dataset (using Otsu's method) in yz-plane



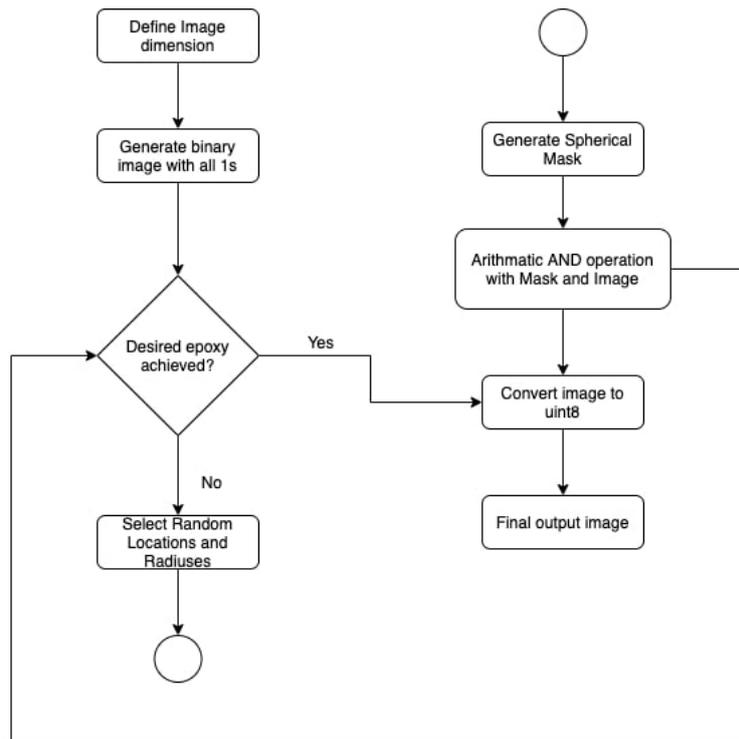
**Figure S8.** Lineal path function plots of segmented real hnp<sub>g</sub>-epoxy ICD dataset (using our deep learning method) in yz-plane



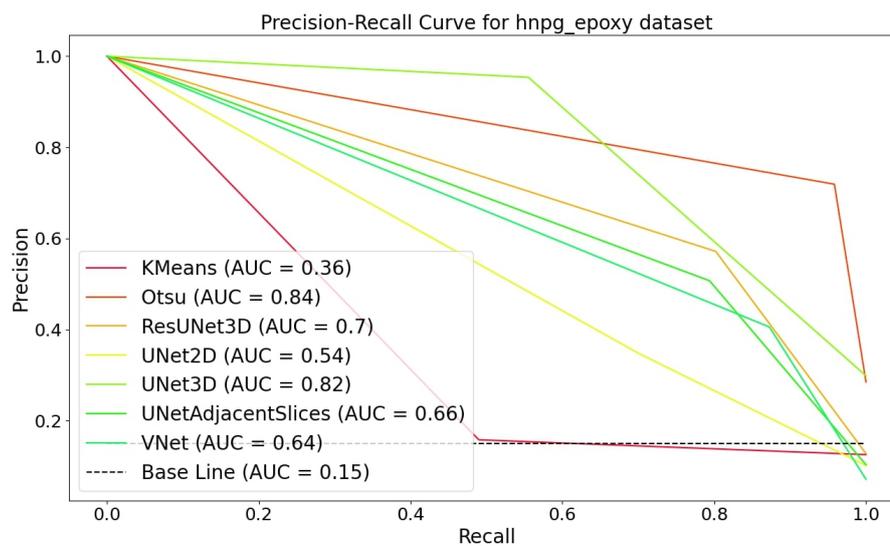
**Figure S9.** L2 differences of lineal path functions of segmented structures using Otsu's, k-means and our deep learning method



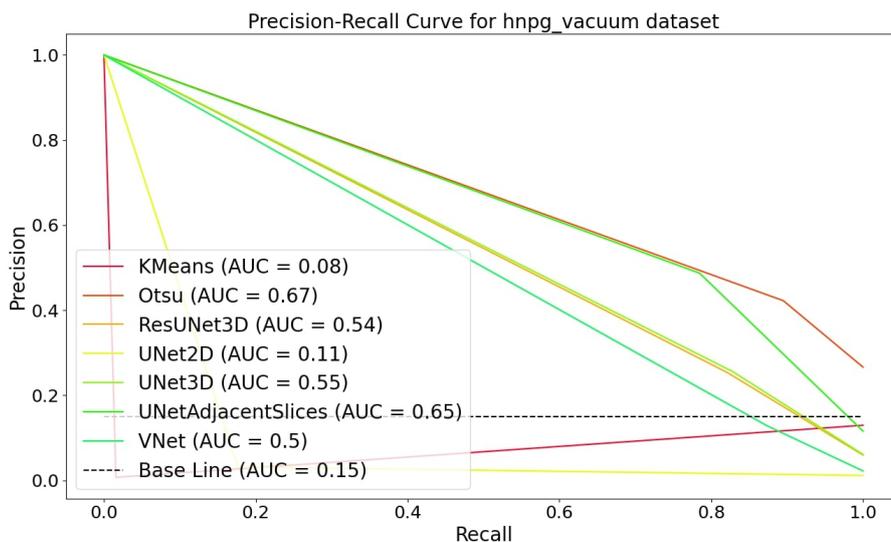
**Figure S10.** Methodology of SSM



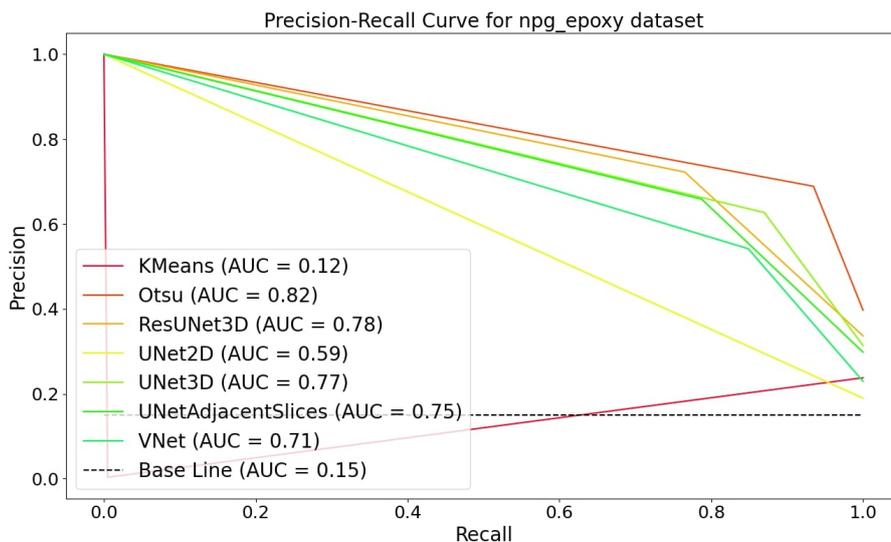
**Figure S11.** Methodology of RPGM



**Figure S12.** Precision-Recall curves for dataset pairs hnp\_g\_epoxy\_TLD and hnp\_g\_epoxy\_ICD. Labels of colored lines in the figure denotes method used to segment dataset.



**Figure S13.** Precision-Recall curves for dataset pairs hnp\_g\_vacuum\_TLD and hnp\_g\_vacuum\_ICD. Labels of colored lines in the figure denotes method used to segment dataset.



**Figure S14.** Precision-Recall curves for dataset pairs npg\_TLD and npg\_ICD. Labels of colored lines in the figure denotes method used to segment dataset.