Accelerating Microstructure Recognition of Nickelbased Superalloy Data by UNet++

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Background

- Nickel-based superalloy materials have been widely used in many fields, and high-throughput preparation has been realized.
- However, the traditional methods to study its microstructure are no longer applicable.
- Here we use deep learning methods: U-Net, UNet++ to improve the accuracy and speed of materials segmentation.

Methods & Results



Figure 1 U-Net architecture, the black and blue arrows indicate the down-sampling and up-sampling, respectively. The red dotted curve denotes the skip connection. The X_j^i indicates the convolutional and pooling operations.

Figure 2 UNet++ :Different levels of pruning: (a) UNet++ L^4 , (b) UNet++ L^3 , (c) UNet++ L^2 , (d) UNet++ L^1 . The black and red arrows indicate the down-sampling and up-sampling, respectively. The blue dotted line denotes the skip connection. The $x^{i,j}$ indicates the convolutional and pooling operations. The green indicates deep supervision.



almost has the same performance between the train set and validation set while the U-Net is a little over fitting.





Figure 5 Example image and segmentation results using different methods. The segmentation results of the neural network models achieve the near-expert performance, and outperform the conventional method.

Conclusions

- The microstructure segmentation of superalloy materials prepared by high-throughput experiment is an essential work.
- Deep neural network models, like U-Net, UNet++, are more efficient and accurate than conventional approach, which reduce the cost of sample preparation significantly. Meanwhile, Unet++ performs better than U-Net.