



Grain Size Analysis Using Machine Learning

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Contribution

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- **Nayudu Ravindranath Chowdary** – Multi-stage processing and outline filtering
- **Poola Harish.** – Utilizing Gray Level Co-occurrence Matrix
- **Kakarla Sashank Chowdary** – Integration of image preprocessing

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This project aims to develop an advanced image processing system for precise microscopic particle characterization in grain images, utilizing innovative techniques such as precision cropping, sharpening kernels, 3D scatter plots, and GLCM-based texture analysis. The goal is to offer a robust methodology for detailed particle analysis and contribute valuable insights to microscopic image processing.



Abstract

This work introduces a sophisticated method for categorizing particles in microscope images, employing editing, multi-stage processing, and particle segmentation. It conducts detailed analysis of particle sizes, locations, and aspect ratios, utilizing 3D scatter maps and GLCM texture analysis. The technique offers a valuable framework for particle identification in various scientific fields.



Introduction

This project explores microscopic image analysis, employing innovative techniques like format conversion, precision cropping, and multi-stage processing to enhance particle visibility and segmentation. It utilizes multidisciplinary approaches, 3D scatter plots, and GLCM texture analysis to gain insights into particle composition, structure, and distribution, culminating in comprehensive findings and future research recommendations.

System Requirements

Hardware requirements:

8GB RAM
256 or above ssd
Multicore Processor
Preinstalled OS(Windows, macOS, or Linux).

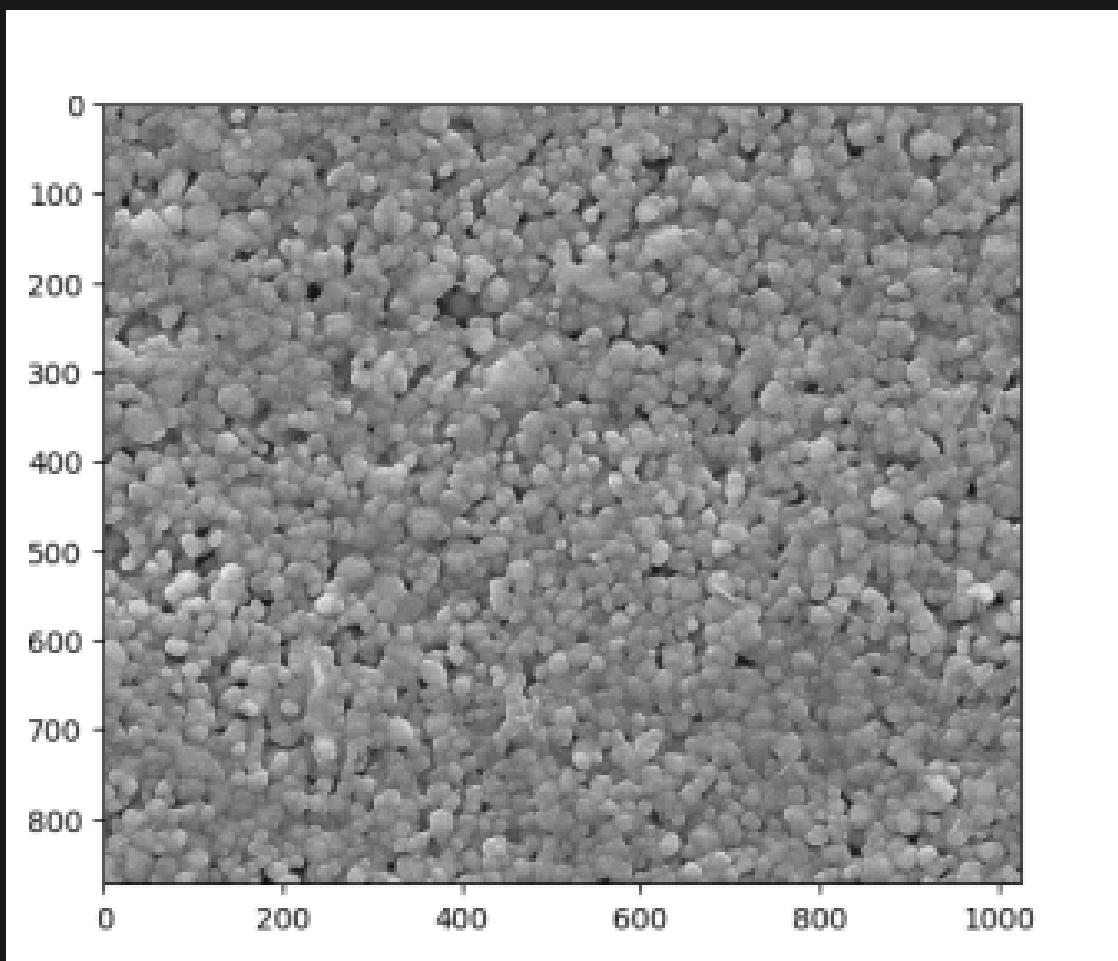
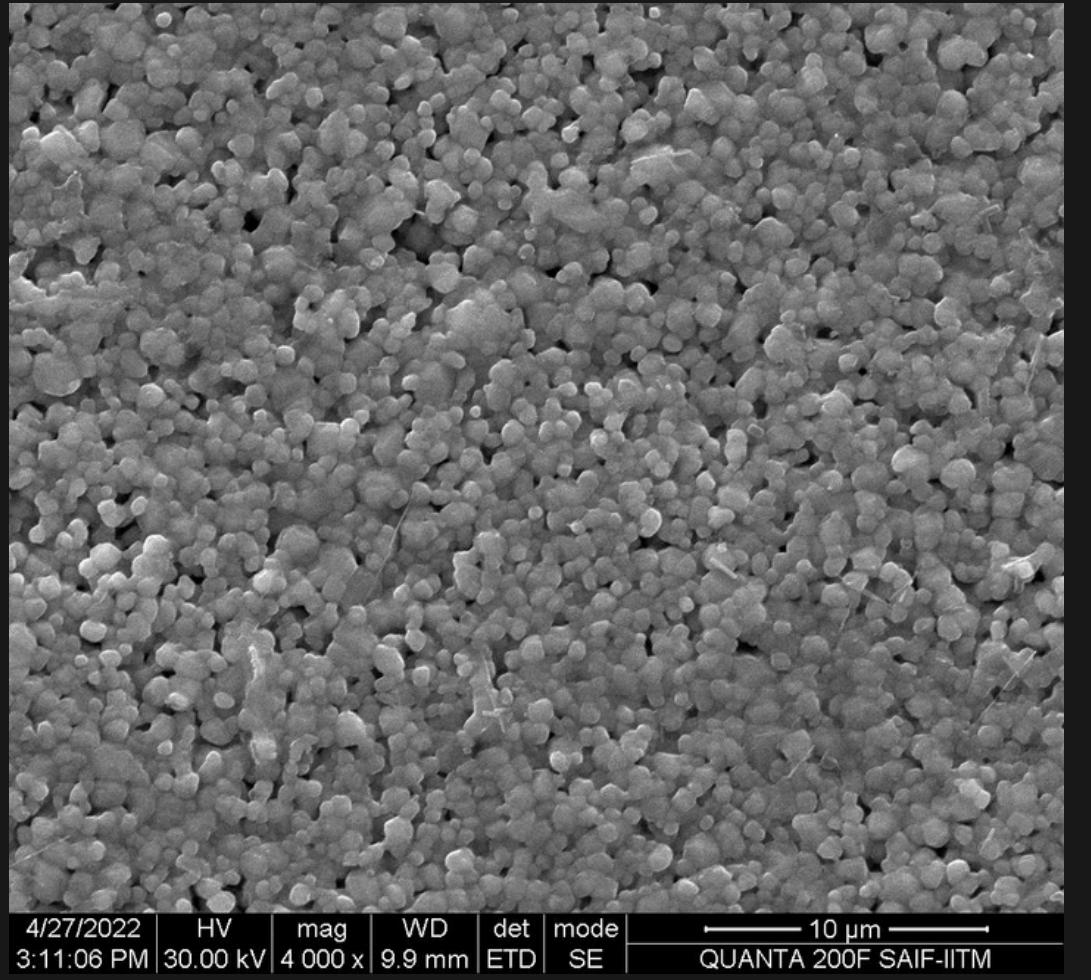
Software requirements:

An IDE such as Jupyter Notebook or Visual Studio Code is recommended
Python 3.x, along with data analysis libraries like Pandas, NumPy, Matplotlib, and Seaborn are necessary
windows XL or any other preferred apps for data sets
ImageJ.JS Tool for size detection analysis



```
react.Fragment>
  <div className="py-5">
    <div className="container">
      <Title name="our" title="product">
        <div className="row">
          <ProductConsumer>
            {(value) => {
              console.log(value)
            }}
          </ProductConsumer>
        </div>
      </Title>
    </div>
  </div>
```

Image Processing



- 1. Load TIFF and Convert to JPEG:**
 - Open a TIFF image, convert it to JPEG, and save the result.

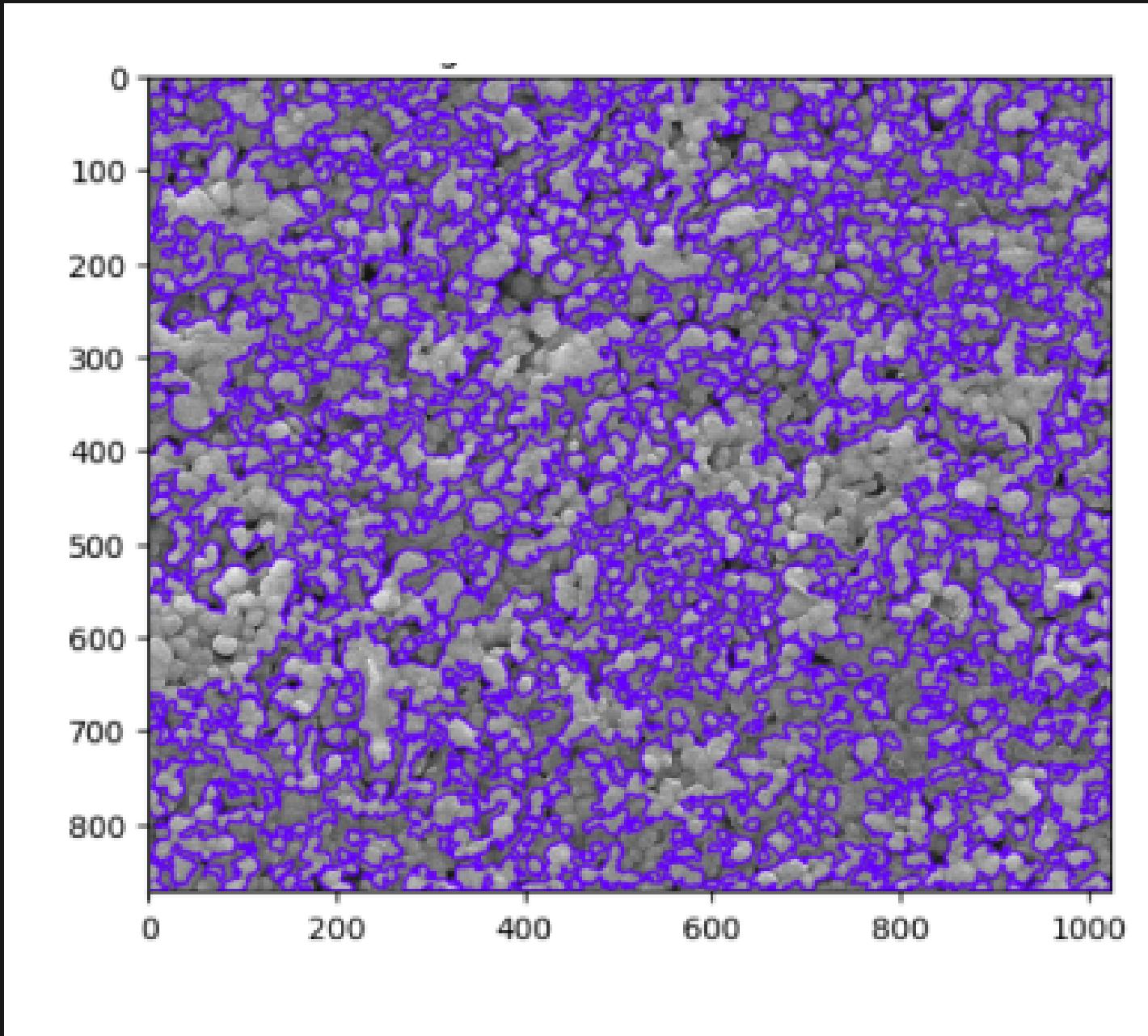
- 2. Open, Crop, and Save JPEG Image:**
 - Open a JPEG image, crop it to specified dimensions, and overwrite the original file.

- 3. Adjust Brightness and Contrast:**
 - Apply brightness and contrast adjustments to an image using OpenCV, and visualize the result.

- 4. Create and Apply Sharpening Kernel:**
 - Define a sharpening kernel, apply it to an image using OpenCV, and display the sharpened result.

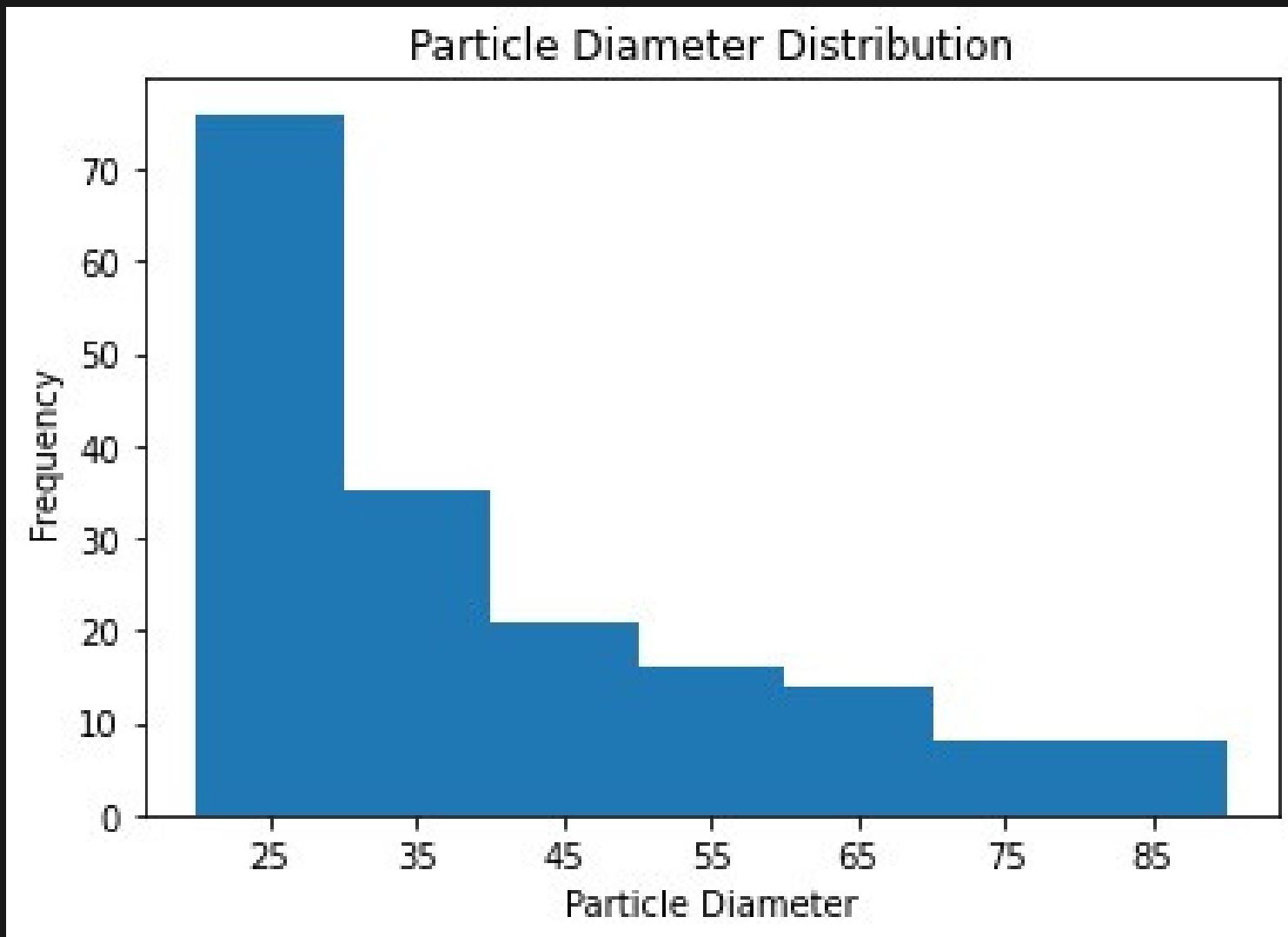
- 5. Show Images and Visualization:**
 - Use Matplotlib to display the original adjusted image and the final sharpened image. Note the potential overwrite issue in saving the cropped image.

Boundaries Detection

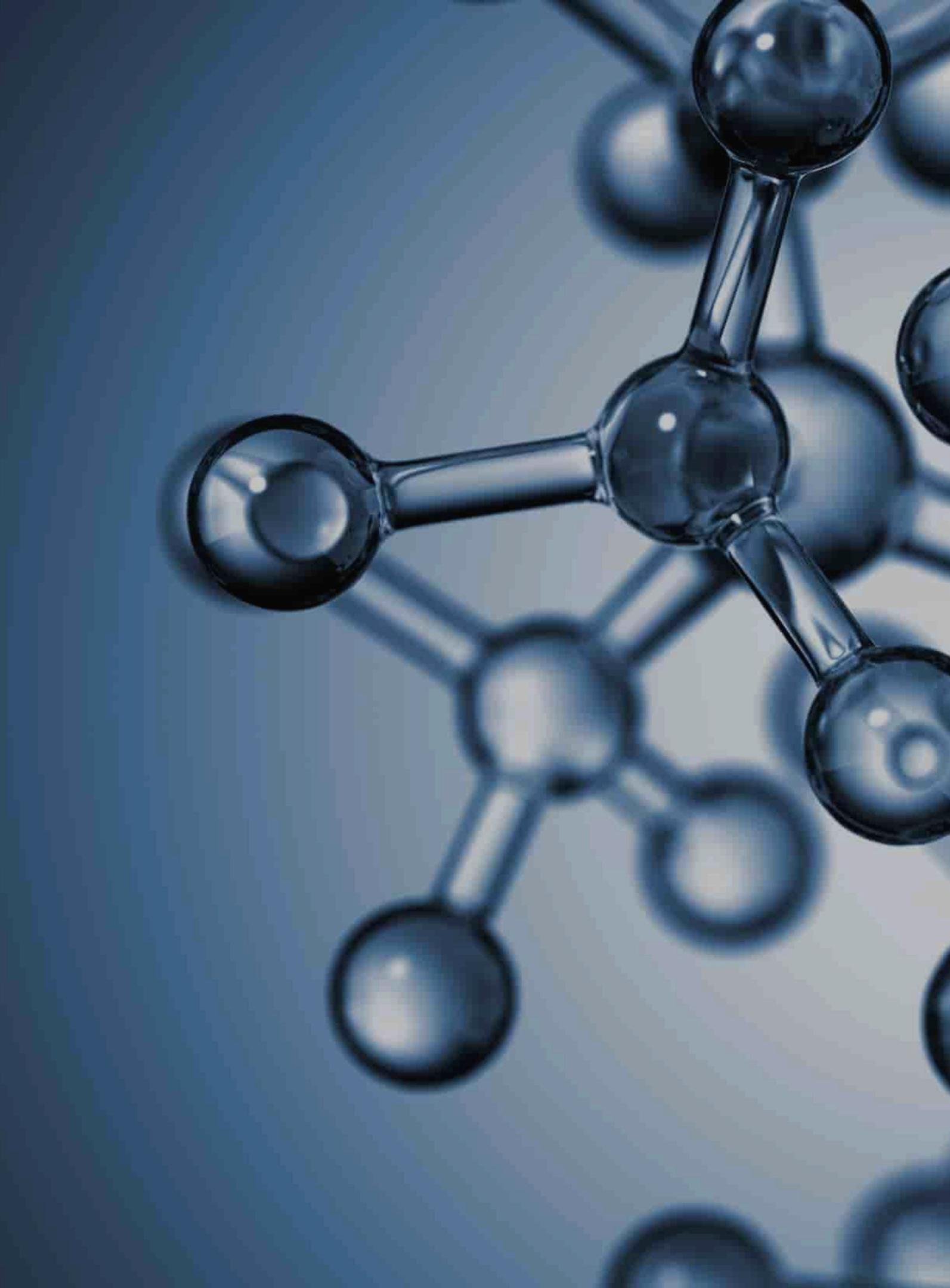


- 1. Grayscale Conversion and Gaussian Blur:**
 - Convert the cropped image to grayscale and apply Gaussian blur for noise reduction.
- 2. Particle Segmentation with Thresholding:**
 - Use thresholding to segment particles in the blurred grayscale image.
- 3. Contour Detection and Particle Diameter Calculation:**
 - Find contours in the thresholded image, calculate particle diameters, and store them.
- 4. Draw Particle Boundaries and Display Result:**
 - Draw particle boundaries on the original cropped image, calculate and print the average diameter, and display the modified image with Matplotlib.

Mean Grain Size



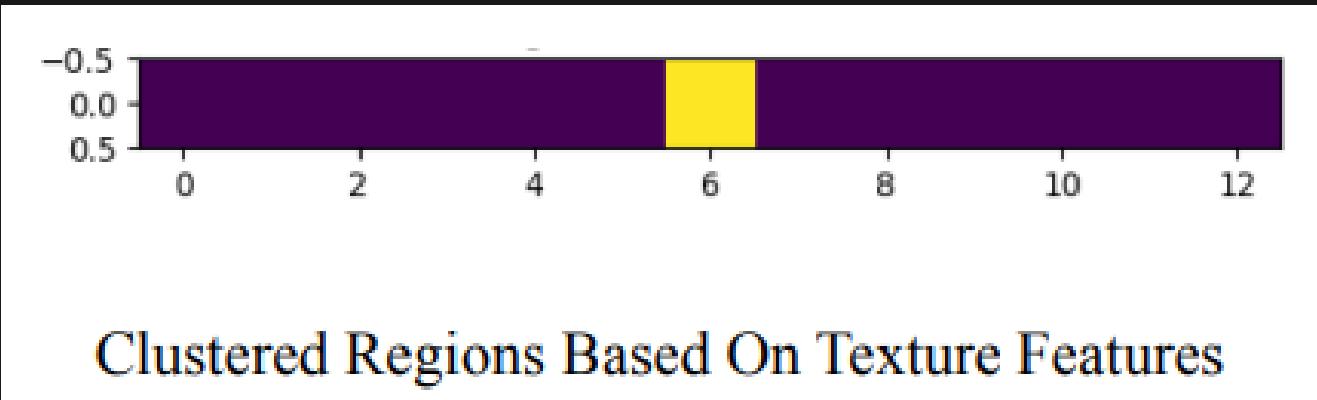
1. Round and Sort Particle Diameters:
 - Round each particle diameter to two decimal places and sort the list.
2. Remove Outliers and Inappropriate Diameters:
 - Remove diameters equal to 0.0 or outside the range [20, 90) from the sorted list.
3. Calculate Mean and Display Filtered Data:
 - Calculate the mean of the filtered particle diameters and print the result.
4. Visualize Particle Diameter Distribution:
 - Create a histogram using Matplotlib with defined bins and plot the frequency distribution of particle diameters.



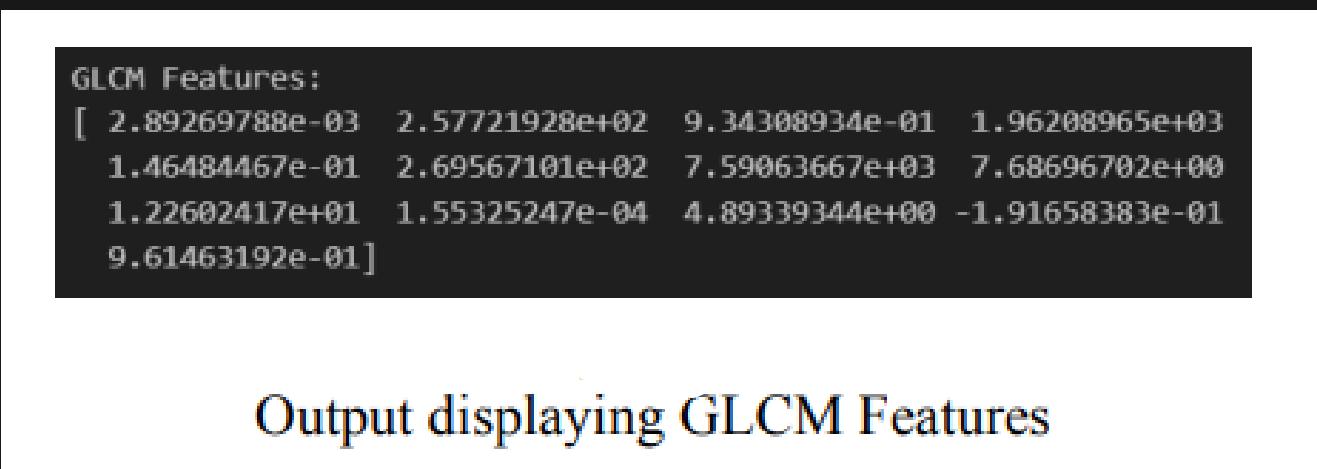
Applications

1. Materials science and engineering
2. Nanotechnology and Nano materials
3. Energy Storage Devices
4. Environmental Monitoring
5. Biomedical Research
6. Advanced Manufacturing
7. Geological and Earth Sciences
8. Pharmaceutical and Chemical Industries
9. Educational and Research Institutions

Results and Discussion



The recommended particle characterization approach improves visibility and segmentation, and enhances the visualization of size distribution patterns in a three-dimensional scatter plot.



Texture analysis using GLCM features enhances the ability to differentiate between particles based on their textural properties and employs GLCM-based KMeans clustering to detect plausible heterogeneities in the image.

Results and Discussion

```
Region Shape: (1, 1), Non-Zero Pixels: 255
Error computing haralick features: mahotas.haralick_features: the input is empty. Cannot compute features
This can happen if you are using 'ignore_zeros'
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Region Shape: (1, 1), Non-Zero Pixels: 255
Error computing haralick features: mahotas.haralick_features: the input is empty. Cannot compute features
This can happen if you are using 'ignore_zeros'
Region Shape: (1, 2), Non-Zero Pixels: 510
Error computing haralick features: mahotas.haralick_features: the input is empty. Cannot compute features
This can happen if you are using 'ignore_zeros'
Region Shape: (1, 1), Non-Zero Pixels: 255
Error computing haralick features: mahotas.haralick_features: the input is empty. Cannot compute features
This can happen if you are using 'ignore_zeros'
Region Shape: (1, 1), Non-Zero Pixels: 255
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Region Shape: (1, 1), Non-Zero Pixels: 255
Error computing haralick features: mahotas.haralick_features: the input is empty. Cannot compute features
This can happen if you are using 'ignore_zeros'
Region Shape: (1, 2), Non-Zero Pixels: 510
...
Region Shape: (27, 68), Non-Zero Pixels: 121380
Region Shape: (25, 22), Non-Zero Pixels: 36210
Region Shape: (38, 22), Non-Zero Pixels: 45135
Region Shape: (7, 6), Non-Zero Pixels: 4335
```

Output displaying texture features for each contour region

Quantitative analysis utilizes a violin plot to visually represent particle diameters and mean diameter, complemented by modern technologies like edge detection and GLCM-based clustering to further enhance the inquiry.

Morphological image processing theories by Sciaiani and Fritsch are utilized to identify contours on edge-detected images, with subsequent texture analysis enabling a deeper understanding of the spatial structure of the data within each contour area.

Mean Diameter of Particles: 59.917178957294134 (pixels)

Sorted Particle Diameters: [13.83998574 14.45306015 14.77899456 15.22924042 15.23174667

| | | | | |
|---------------|--------------|--------------|--------------|--------------|
| 15.44304466 | 15.52437496 | 15.81158829 | 15.85589886 | 15.87128448 |
| 16.1590023 | 16.17395782 | 16.41689873 | 16.6496848 | 17.28485115 |
| 17.26287651 | 17.26287651 | 17.27584839 | 17.49305534 | 17.49305534 |
| 17.51092148 | 17.69200786 | 17.82909966 | 17.90940475 | 18.05908585 |
| 18.58143242 | 18.78849411 | 19.03969193 | 19.10085678 | 19.18517311 |
| 19.10517311 | 19.24655151 | 19.31678391 | 19.31863594 | 19.41668892 |
| 19.41668892 | 19.52396811 | 19.64708328 | 19.65820694 | 19.83530045 |
| 20.09995079 | 20.09995079 | 20.12481117 | 20.22394943 | 20.46699715 |
| 20.59146118 | 20.59353828 | 20.81035423 | 21.0438633 | 21.09522247 |
| 21.10778809 | 21.26849232 | 21.480113449 | 21.75422096 | 22.02291489 |
| 22.04895782 | 22.10722351 | 22.12602843 | 22.13049889 | 22.20388482 |
| 22.31461716 | 22.3608799 | 22.3608799 | 22.47240448 | 22.71713638 |
| 22.80370903 | 23.08699226 | 23.4095993 | 23.47221184 | 23.68104752 |
| 23.60104752 | 23.85347557 | 23.94628716 | 24.04183006 | 24.04183006 |
| 24.20763779 | 24.35179138 | 24.51550182 | 24.69837761 | 24.9458865 |
| 25.00020027 | 25.06576157 | 25.08000724 | 25.08000724 | 25.11984634 |
| 26.00020027 | 26.03109932 | 26.0770092 | 26.20997047 | 26.40095711 |
| 26.41988945 | 26.6835289 | 26.90744781 | 27.018713 | 27.20314217 |
| 27.48758125 | 27.65883446 | 27.72887611 | 28.08558846 | 28.67417145 |
| 28.84461021 | 28.84461021 | 28.85869217 | 29.06908417 | 29.1648922 |
| 29.41108322 | 29.49799919 | 29.54677391 | 29.64186287 | 30.1721611 |
| 30.46653748 | 30.86502075 | 31.14502335 | 31.24119949 | 31.30515289 |
| 31.93328857 | 32.08691788 | 32.25271606 | 32.55783844 | 33.24173737 |
| ... | | | | |
| 168.52896118 | 186.74671936 | 191.9098271 | 196.65725708 | 201.10215759 |
| 203.00323486 | 262.12438965 | 269.02993774 | 269.09628296 | 285.61709595 |
| 326.55877686 | 334.13259888 | 335.94448853 | 343.65701294 | 376.38796997 |
| 621.51477051] | | | | |

Output displaying Mean Diameter Of Particles

1. Explore machine learning algorithms for automated particle categorization to enhance study efficiency.
2. Extend the system to handle three-dimensional data, enabling a more comprehensive exploration of particle interactions and structures.
3. Investigate state-of-the-art classification approaches for improved efficacy and scalability in particle identification and categorization.

Our integrated image processing and analysis approach successfully provides precise characterizations of microscopic particles, showcasing efficacy in format conversion, cropping, and pipeline processing. The results, including 3D scatter plots and texture analysis, affirm the precision and clarity achieved in understanding particle characteristics.



Reference :

Rosebrock

A. Mishra and T. Pathak, "Estimation of grain size distribution of friction stir welded joint by using machine learning approach," ADCAIJ: Advances in Distributed Computing and Artificial Intelligence Journal 10, 99–110(2021)

