

# Volume-Based Shape Analysis for Internal Microstructure of Steels

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# Importance of microstructure



Micro-structures significantly affect properties.

# New global trend

### **Computer-aided** material designing:

Simulate and design microstructures and properties

#### Materials genome initiative in USA [OSTP 2011]



Japan has also started project "Materials integration". [2014]

# What's our chance?

# Image-based materials genome



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We propose a new framework for 3D micro-structure evaluation based on image processing.

# Focus on bearing steel

**Bearing** : Mechanical element to support rotating shaft.



High strength requirement than normal steel.

Fatigue failures of bearing High stress condition Contamination

→ Crack initiation & propagation



Huge real-world applications.

# What shape features of inclusion will affect?





Complex shapes: Anisotropy and surface roughness of inclusions will cause crack initiation. [Cyril2008]

Our goal : 3D imaging and shape analysis technique.

# Related imaging technique of steels

Require 100 nano resolution & millimeter region !



#### Difficulty

3D imagings have been developed.

But, not enough in resolution, view range, or labor cost.

For example, X-ray CT is unavailable to steels unfortunately.

# Related 3D shape analyses of steels

Topology [Adachi 2010]



**3D morphology** [Morito2013]



genus, Euler number

volume ratio of micro-structures

#### Difficulty

3D shape characterization of steels are at early stage. Only a few frameworks exist for dealing with steel volumes.

# Our approach

### Framework



Imaging device Segmentation 3D shape analysis



3D, Precise Automatic Automatic labeling

Multi-resolution analysis

Shape:

Anisotropy, Roughness, + Other features

# Our imaging technique

# 3D internal structure microscope



Automatic serial-sectioning device

### Automatic serial-sectioning



Feature: Automatic: Labor-saving Precise: High resolution Sufficient image alignment Efficient : Months → Hours or days





# Segmentation



## Limited in simple method.

Threshold of intensity & Manual segmentation

## Future work to incorporate, ...

#### Supervised Segmentation

- Pattern recognition
- Machine learning



[Straehle 2013]



#### **Unsupervised Segmentation**

- Otsu's method [Otsu 1979]
- Snake (Active Contour)
- Graph Cuts
- Mean Shift
- Water Shed (Region Growing)

# **Multi-resolution analysis**

Labeling and analysis in each resolution.



#### Multi -resolution & -material methods

Laplacian pyramid [Burt 1983] + Sethian's method [Sethian 1996] Detail to Global feature

# Anisotropy analysis

Principal component analysis (PCA)



PCA results corresponds to axes of ellipsoid fitting.

Eigen values:  $\lambda_1$ ,  $\lambda_2$ ,  $\lambda_3$ Eigen vectors:  $\mathbf{e}_1$ ,  $\mathbf{e}_2$ ,  $\mathbf{e}_3$ 

### Application

Ratios of Eigen values  $(\lambda_1 / \lambda_2, \lambda_1 / \lambda_3, \lambda_1 / \lambda_3) \longleftrightarrow$  Crack give anisotropy of inclusions.

# **Roughness analysis**

### Surface curvatures

- > Principal curvatures  $\kappa_1$  and  $\kappa_2$
- > Gaussian curvature  $\kappa_1 \kappa_2$
- Mean curvature
- Total curvature

$$(\kappa_1 + \kappa_2)/2$$
  
 $\kappa_1^2 + \kappa_2^2$ 

+ Their statistical values (Max, Min. Mean, Histograms, etc)

e.g. Gaussian curvature and 3D shapes



$$\kappa_1 \kappa_2 < 0 \quad \kappa_1 \kappa_2 = 0 \quad \kappa_1 \kappa_2 > 0$$

### **Discrete algorithm**

Least-squares polynomial fitting Method [Yoshizawa 2008]

# Application

Inclusions Surface roughness



Initiating points Existence of cracks

Cracks

# Experiment

- 3D Imaging & segmentation
- Labeling
- Anisotropy analysis (PCA)
- Roughness analysis (Curvature)

# of

real-world bearing steels provided by a steel company.



Detailed images of Inclusions and micro-cracks. Automatically labeled images.

# Experiment -Anisotropy (PCA results) -

### Axis length ratio of inclusions



Anisotropy was characterized from three axes.

### Discussion

Previous literatures use anisotropy,

but difficult to characterize crack-initiating inclusions.

# Experiment -Roughness-

### Principal curvature (maximum)



#### Multi-resolution

Roughness was calculated from 3D images.

#### **Statistics**

Maximum, minimum, mean, median, histograms are also calculated.

## Experiment -Histogram of max principal curvature-





Specimen with cracks

slope in negative region  $\rightarrow$  concave



Specimen without cracks

Almost no slope in negative region

New hypothesis based on quantitative analysis Additional histogram analysis of curvature implied concave regions were important for initiating cracks.

# Limitations & Future work

# Limitations

- Simple segmentation
- Limited shape parameters

# Future work

- Incorporating sophisticated segmentation
- Using other shape parameters & features



Distance field Streamline Min

Min. distance Multi-resolution

More specimens, other materials

etc.

# Conclusion

- > We proposed image-based materials genome.
- We constructed a novel framework with efficient
  3D imaging and shape characterization of steels.









Labeling Shape analysis

Imaging Segmentation







Curvature analysis implied concave regions was important for initiating cracks.