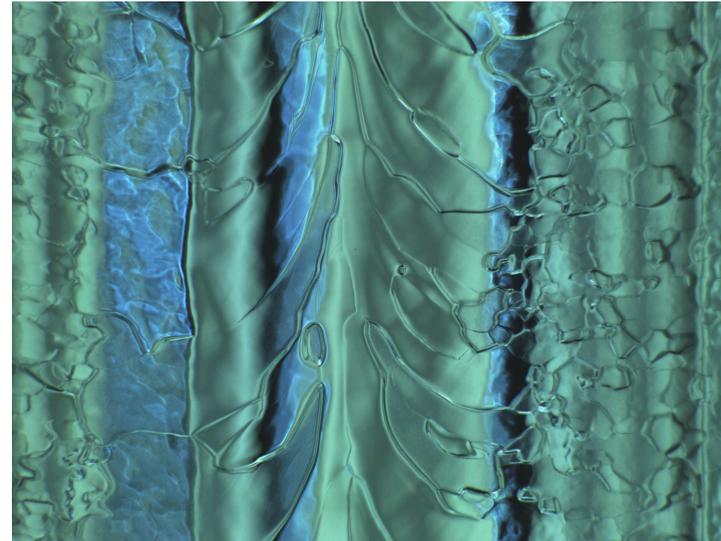


Optical analysis of SCRF niobium cavities

Status Report

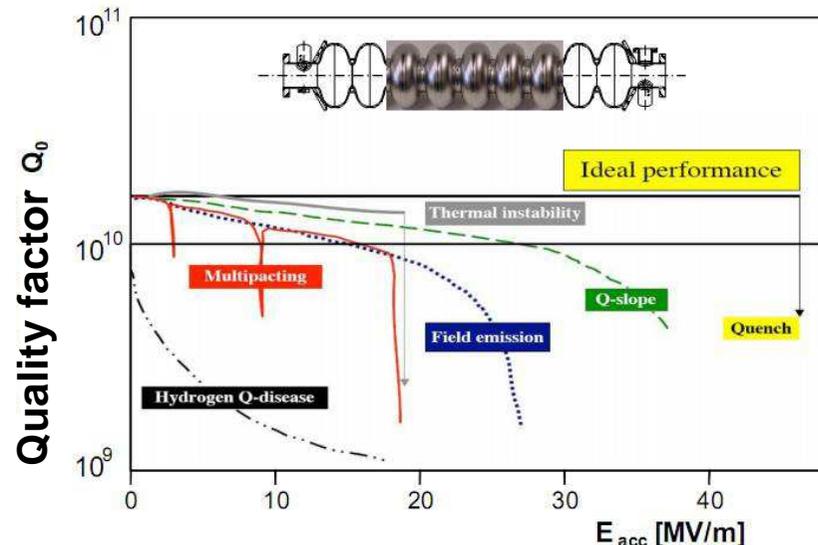


Marc Wenskat

6th Annual Terascale Alliance Workshop
Hamburg, 3. 12. - 5. 12. 2012

SCRF Fundamentals

- Electromagnetic field oscillates inside volume bounded by superconducting surface
- Surface currents shield the external fields from bulk niobium (~30 nm skin depth).
- The highest magnetic field is located at equator region → highest current densities occur in this region.
- Surface roughness, a geometrical defect or an impurity lead to a local breakdown of superconductivity



Goals

> Surface classification

- Deduce a minimal set of image variables to
 - Understand the influence of a chemical polishing of the cavity surface
 - Connect surface properties with operational behavior of the cavities
- Define a specification of surface and welding seam properties for quality assurance during industrial mass production of cavities

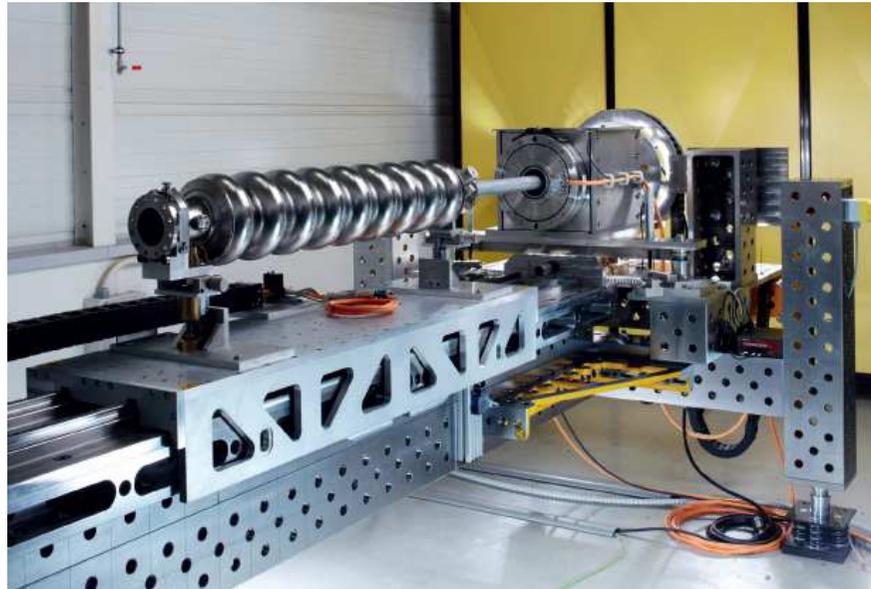
> Defect detection

- Development of automated detection and classification of irregularities as possible operational limiting heat spots (quench, Q-drop)



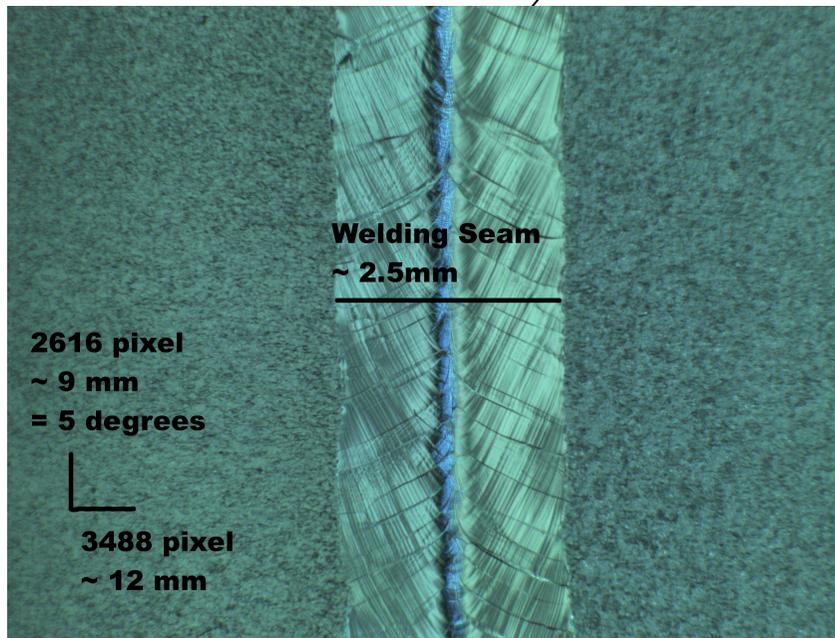
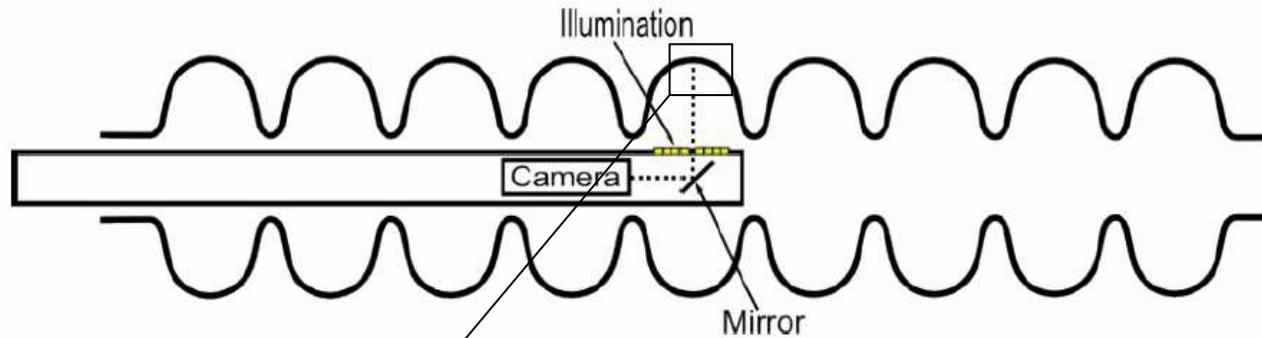
OBACHT

- > **Optical bench** for **automated cavity inspection** with **high resolution** and **short timescales**



- > Fully automated optical inspection: camera position, illumination, image taking and image storing, (auto focus)
- > Timescale for single inspection decreased from the order of days to 8-10h
- > Image processing for variable deduction runs in parallel using stored images

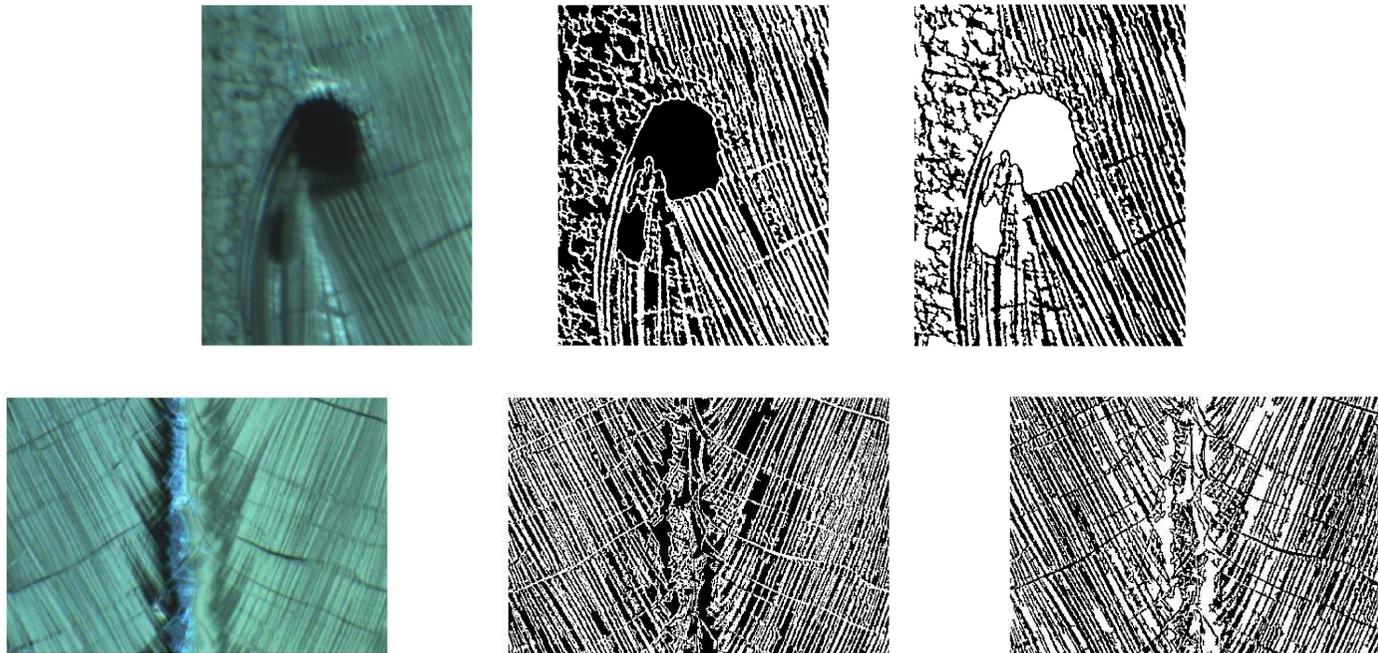
Optical inspection



- > Pixelsize of $3.5 \mu\text{m} \times 3.5 \mu\text{m}$
- > ~ 20 MB per image
- > 2790 images per cavity
9 equator regions + 10 irides

Image processing

- Image processing steps are applied to original color image
- Two binary images are derived (edge and object representation)

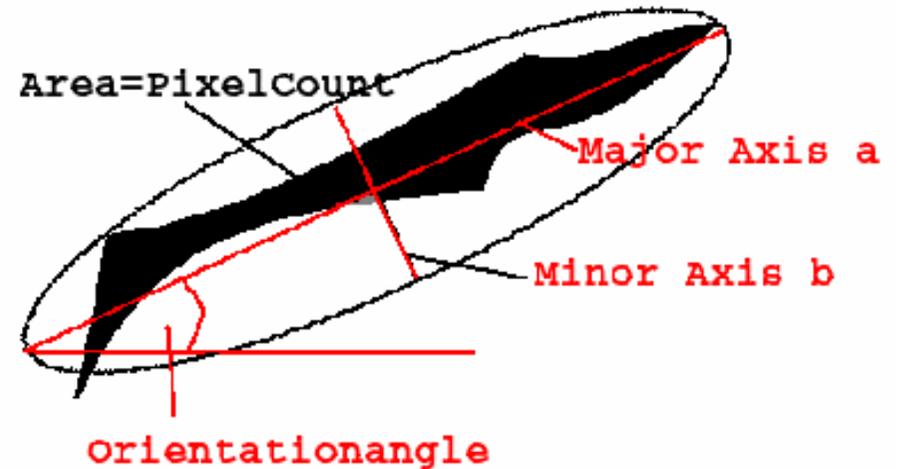


- Grayscale image is used and surface profile is optional
- Sets of variables are deduced using the image representations

Key variables

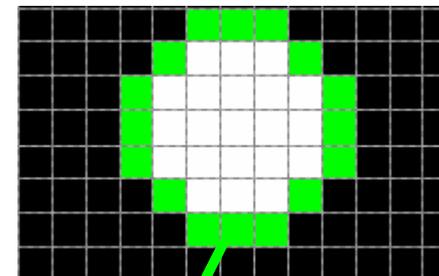
> Derived from the binary image

- Area
- Major & minor axis length
- Perimeter
- Orientation



> Derived with binary and grayscale image

- Surface Roughness R_{dq}

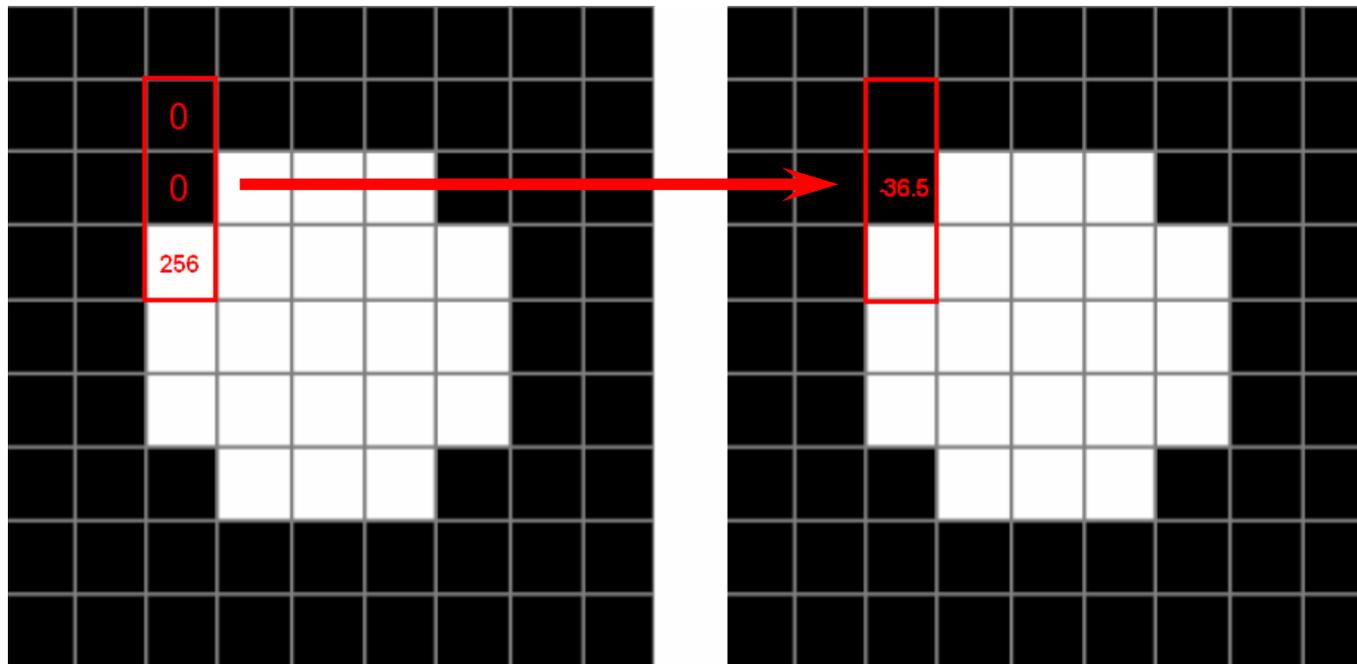


Perimeter

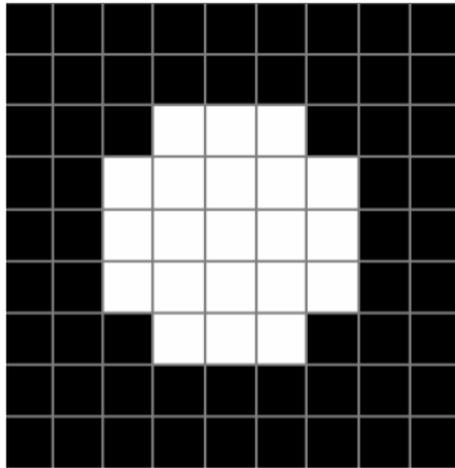
Surface Roughness

- > Pixel to pixel intensity difference is calculated (intensity gradient) for the x and for the y direction using the finite central difference

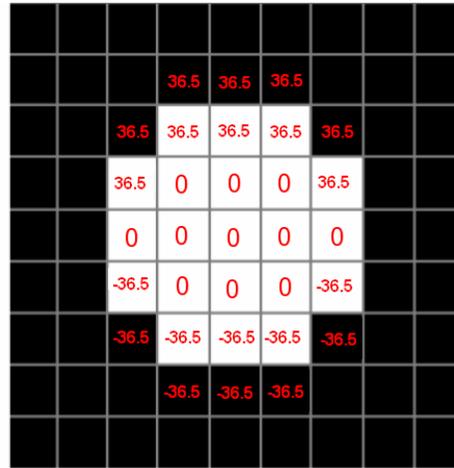
$$\Delta_{y_i} = \frac{I(x, y+1) - I(x, y-1)}{(y+1) - (y-1)} = \frac{256 - 0}{7} \frac{\text{Bit}}{\mu\text{m}} = 36.57 \frac{\text{Bit}}{\mu\text{m}}$$



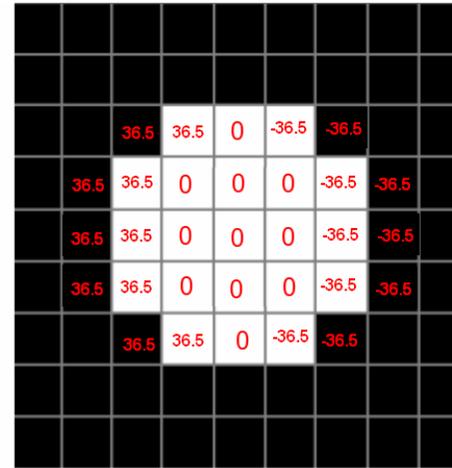
Surface Roughness



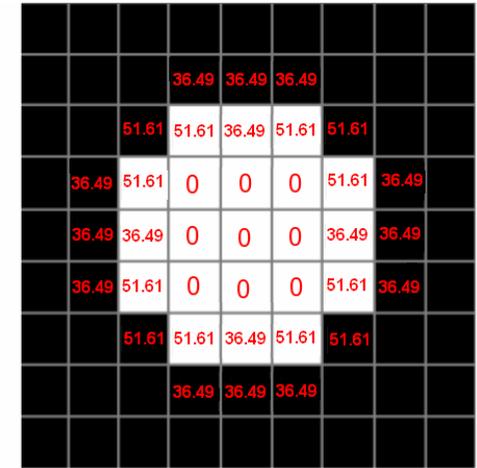
(1) Object



(2) Gradient in x direction



(3) Gradient in y direction



(4) Δ_i per pixel

- > x and y values for each pixels are used to derive a single scalar value Δ_{P_i} (4)

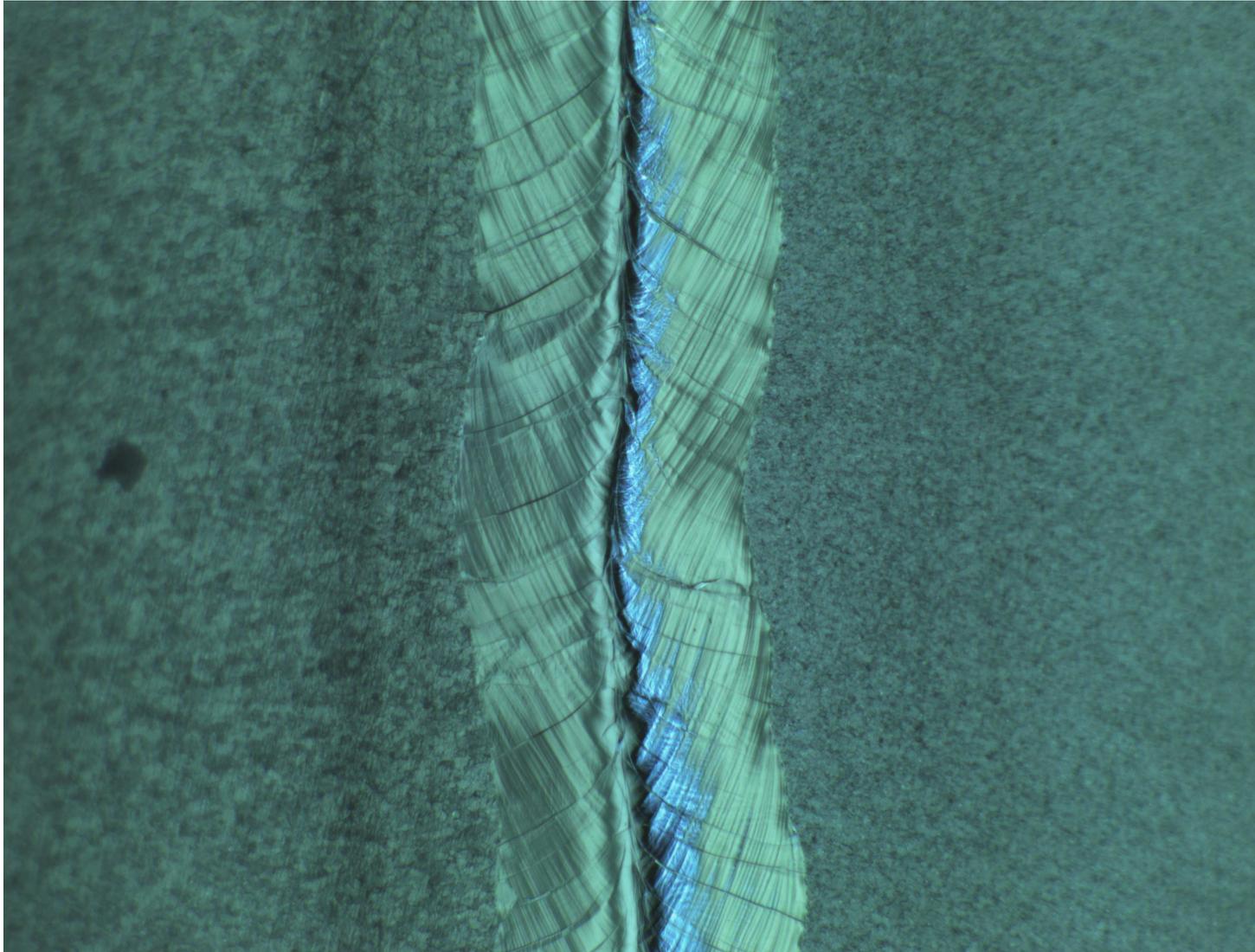
$$\Delta_{P_i} = \sqrt{\Delta_{x_i}^2 + \Delta_{y_i}^2}$$

- > Define object based R_{dq} :
$$R_{dq} = \frac{1}{N} \sum_p \Delta_{P_i}$$

Example: $R_{dq} = 26.61 \text{ Bit}/\mu\text{m}$



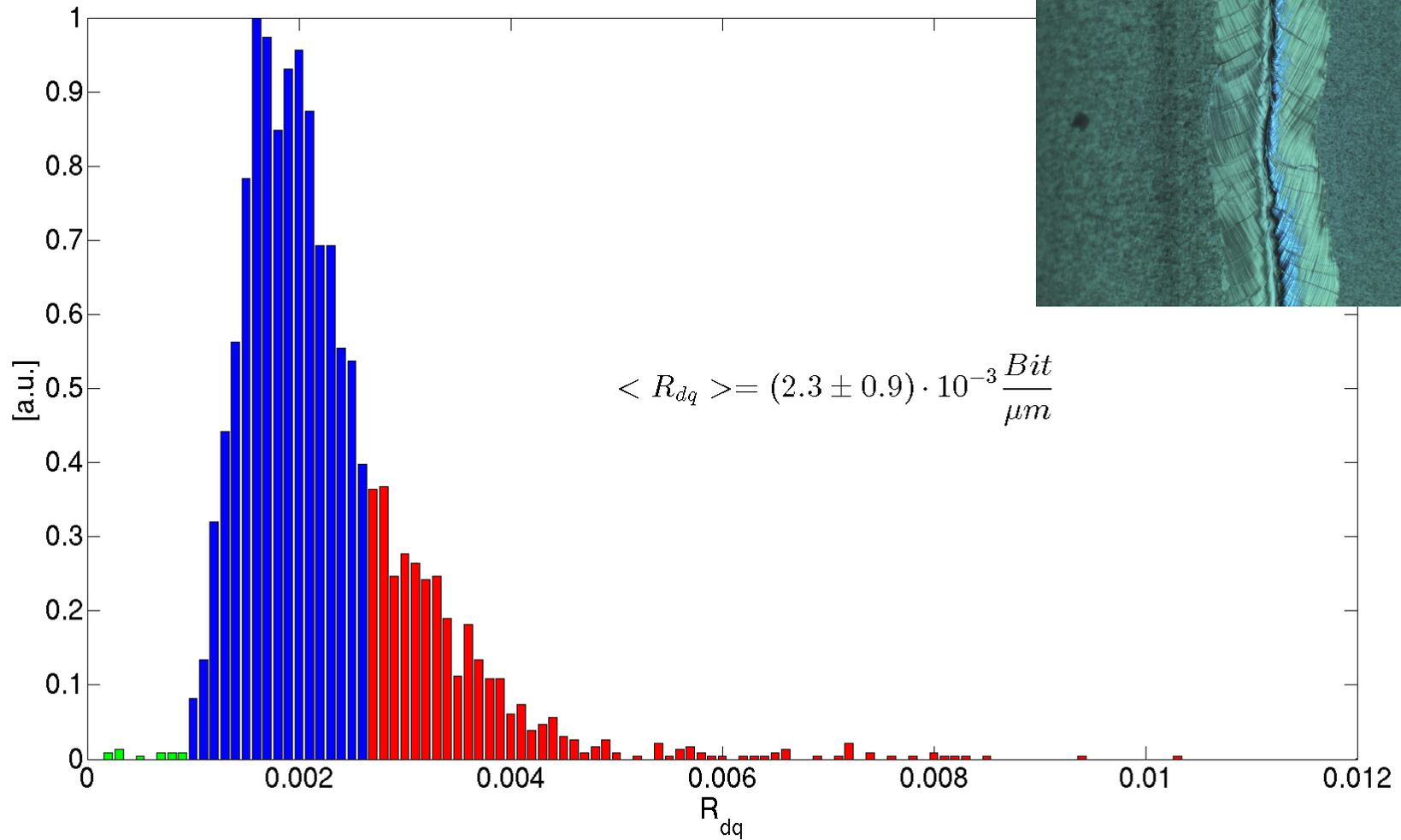
Surface Roughness - before electro polishing (ep)



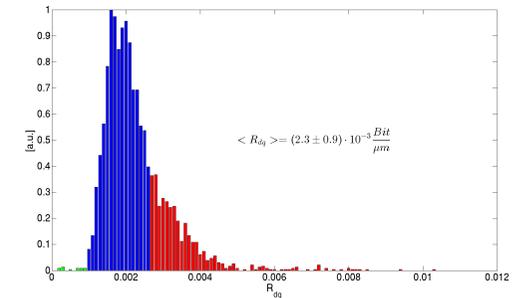
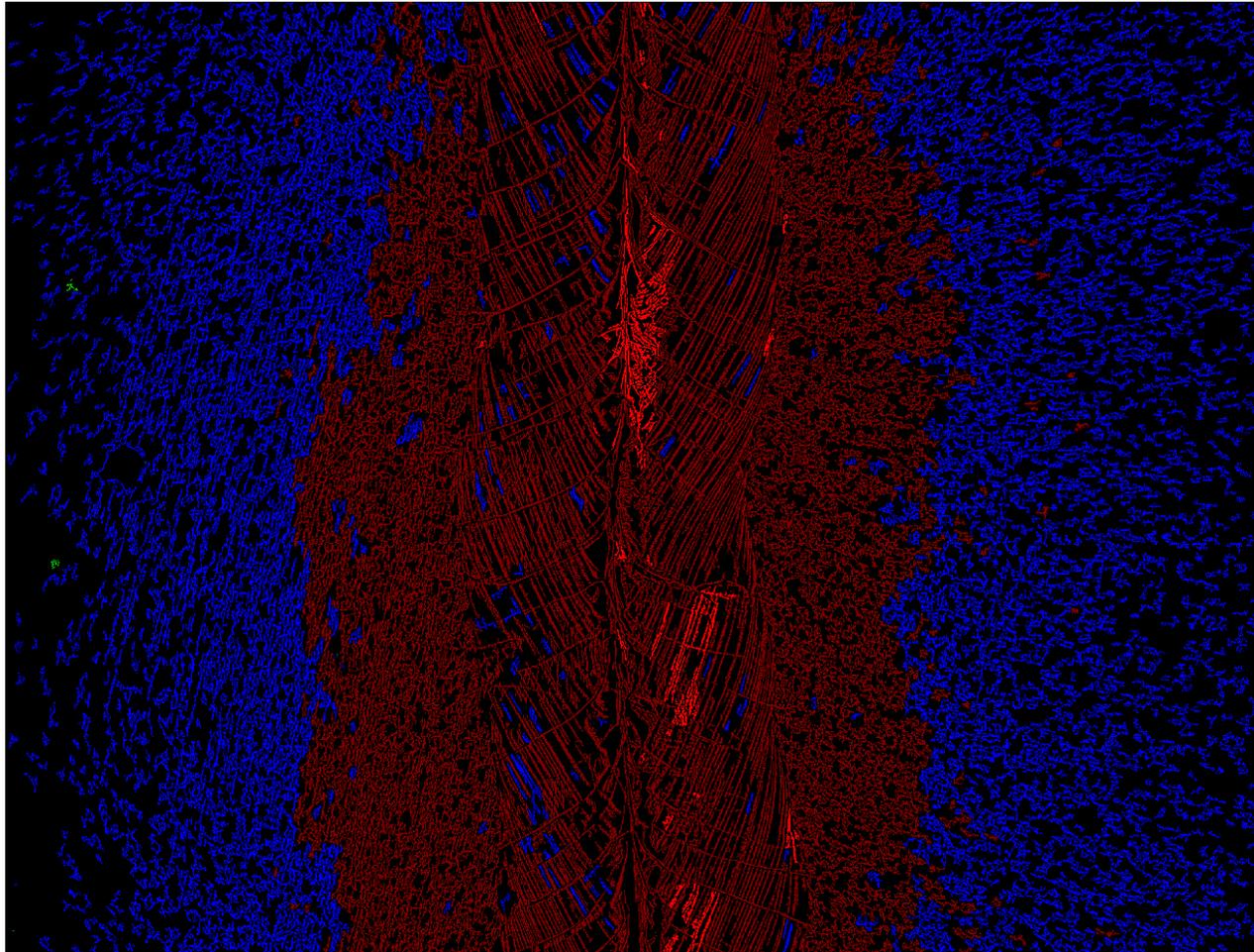
Z161 – E3 – 117°



Surface Roughness - before ep



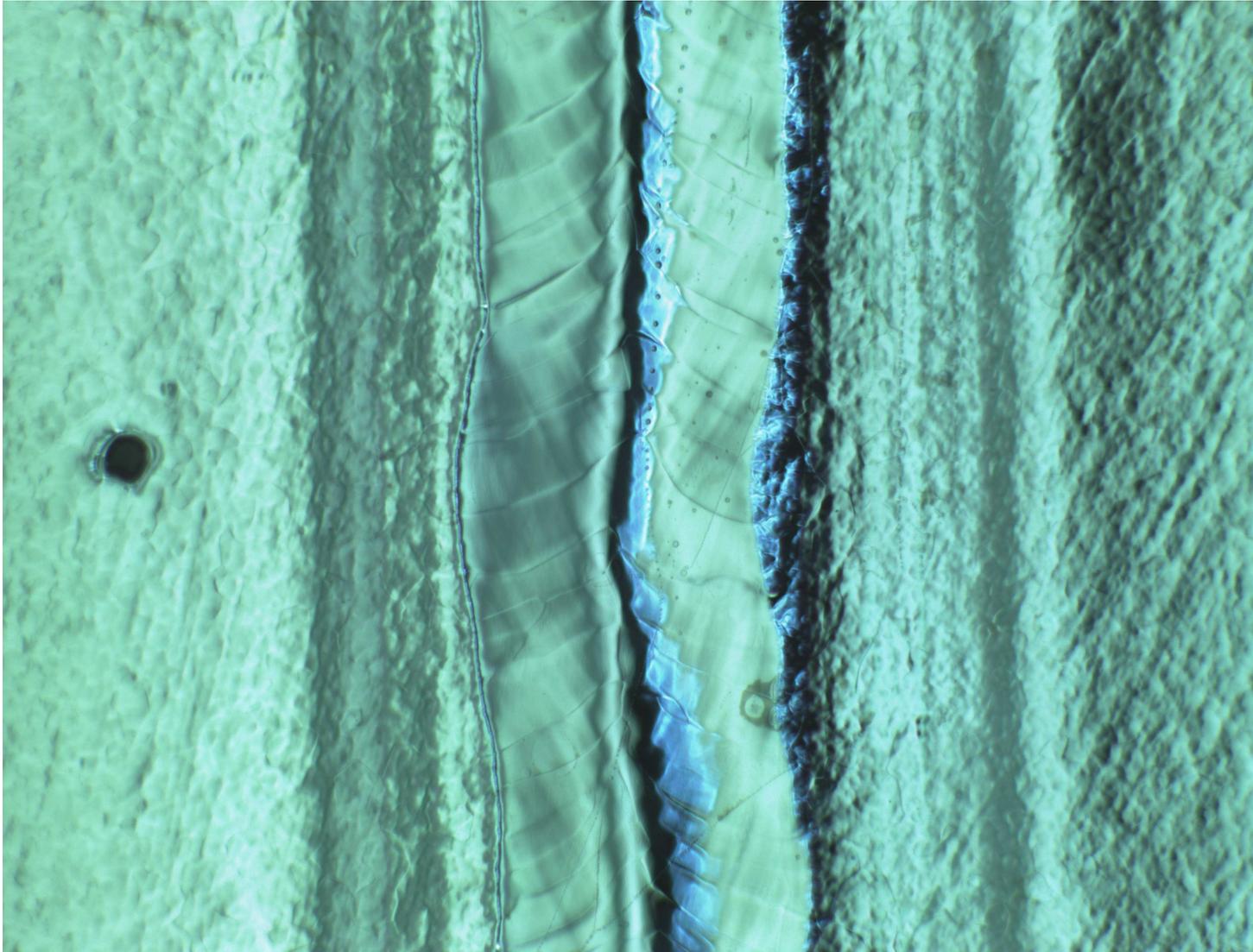
Surface Roughness + edge length - before ep



**Total length of edges
found: 437 cm**

**Length of edges in
welding seam region:
112 cm**

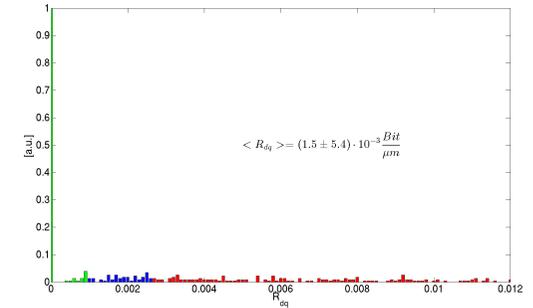
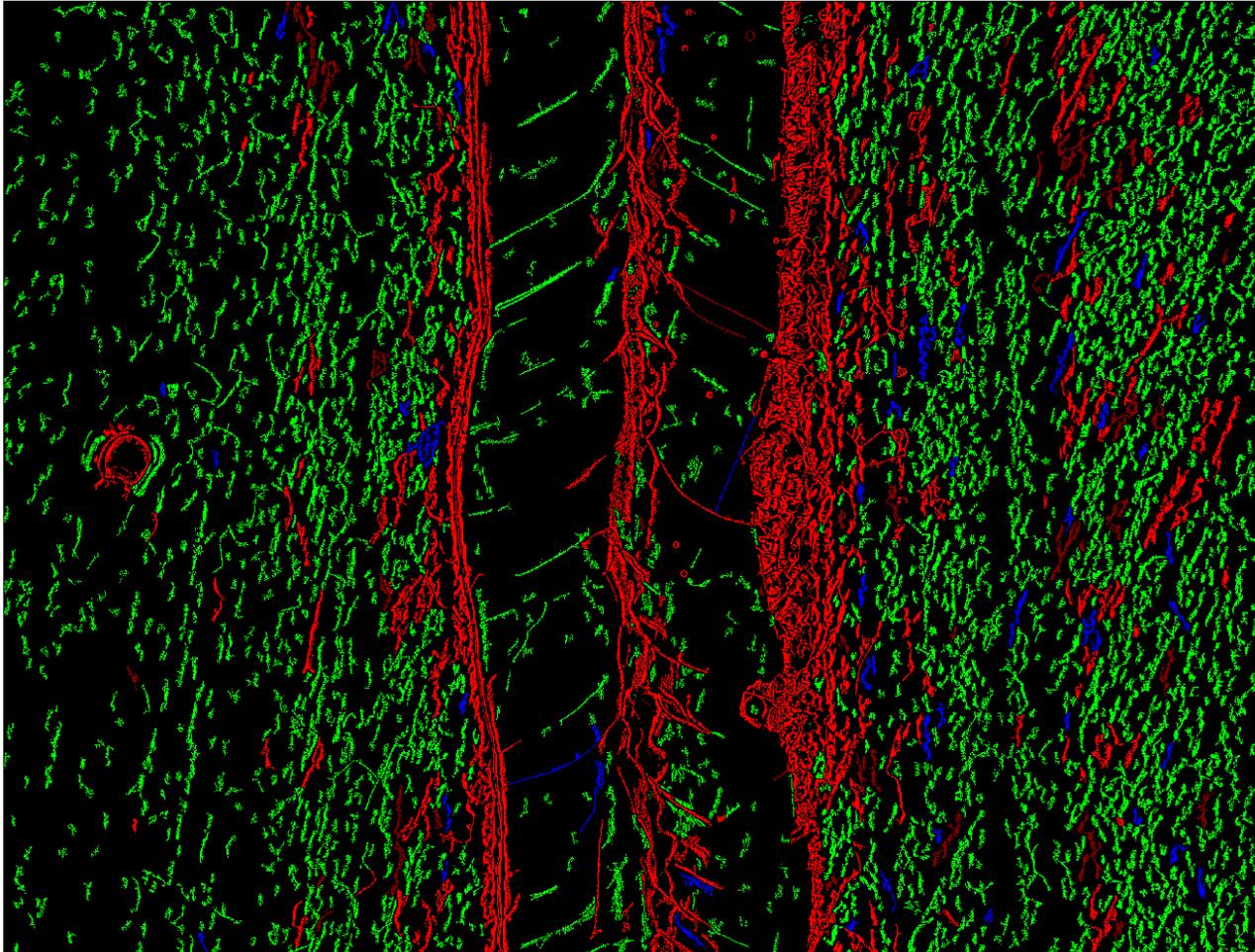
Surface Roughness - after ep



Z161 – E3 – 117°



Surface Roughness + edge length - after ep



After ep:

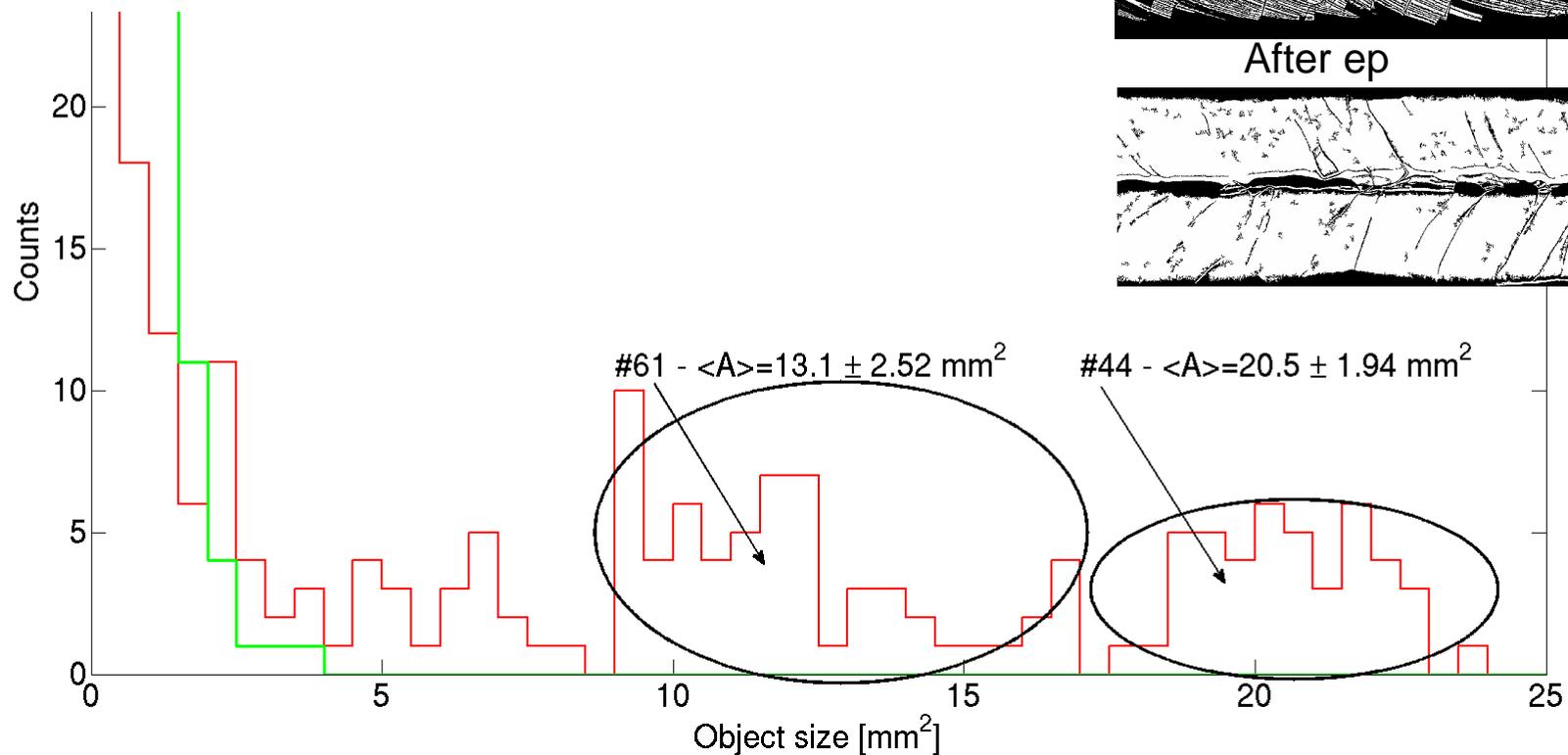
**Total: 257 cm
or 41.1%**

**Welding seam region:
42 cm
or 62.5%**

$\langle R_{dq} \rangle$ reduced by 34 %

Object size during treatment

- Histogram shows object size on welding seam region for complete equator (90 pics) before and after ep



Total welding seam area found: 1607 mm² - total welding seam area expected: 1622 mm²

Defect detection

- > n=4 variables used to compare an object with the neighborhood

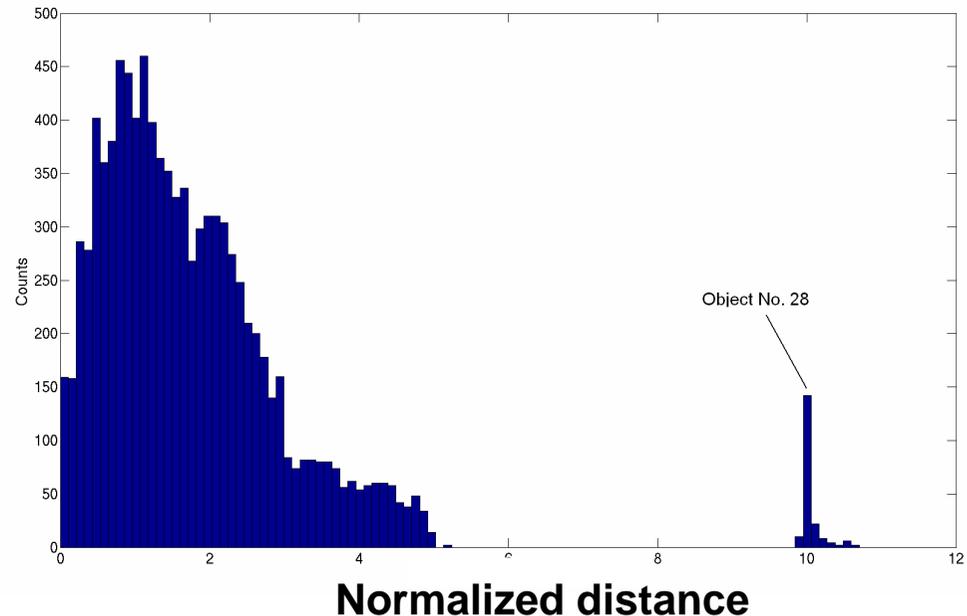
- $\vec{x} = [\text{Major axis length, eccentricity, area, color}]$

- > Introduce normalized Euclidean metric with σ_i variance of the variable

$$d(\vec{x}, \vec{y}) = \sqrt{\sum_{i=1}^4 \frac{(x_i - y_i)^2}{\sigma_i^2}}$$

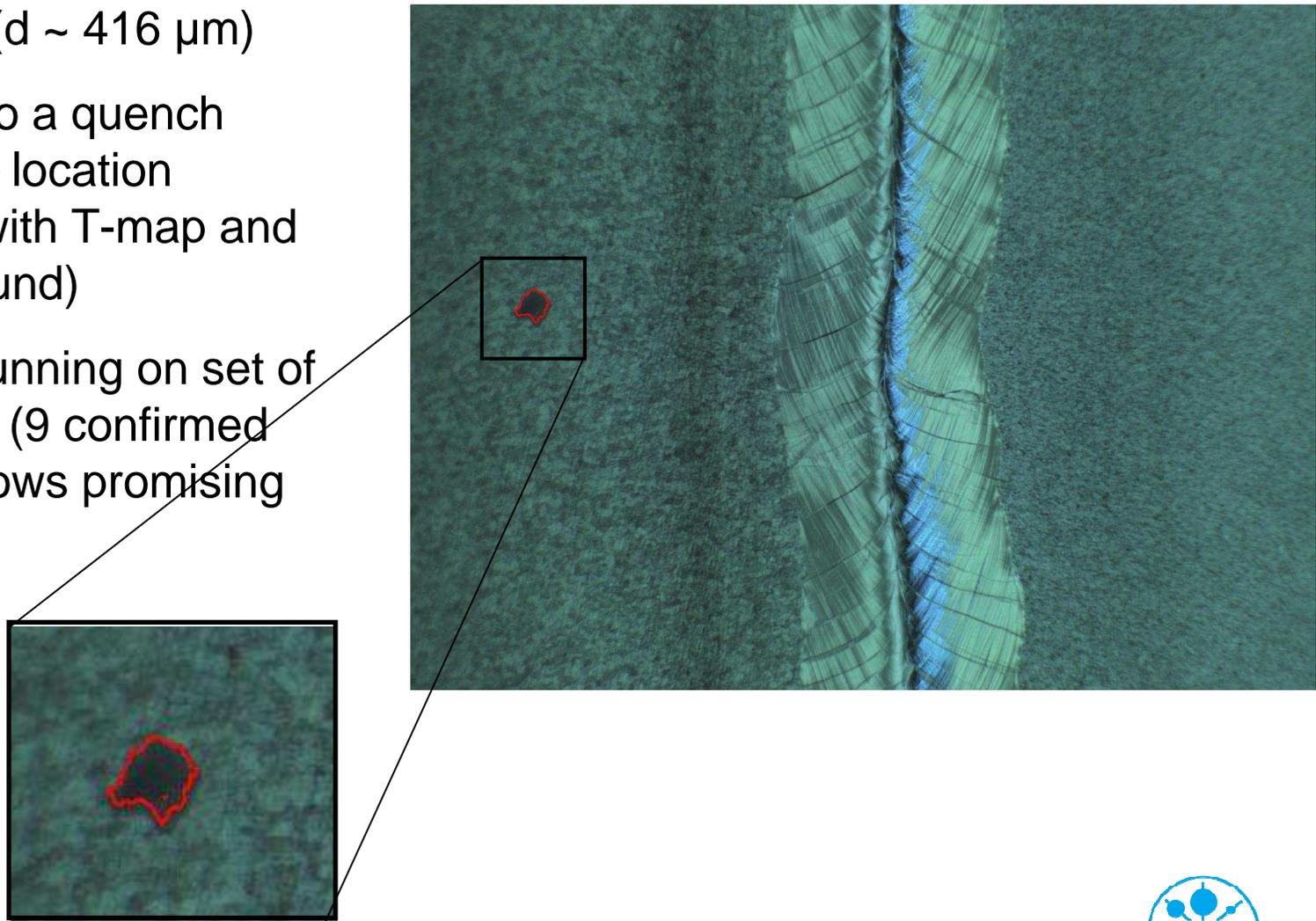
- > Object is compared to its neighborhood: objects inside a circle of radius $3.5 * \text{major axis length}$ (red circle)

- > Calculate distance d of object to neighborhood objects. If its above threshold, mark as irregularity



Defect detection

- One object was identified as irregularity ($d \sim 416 \mu\text{m}$)
- Defect led to a quench (13 MV/m – location confirmed with T-map and Second Sound)
- Algorithm running on set of test images (9 confirmed defects) shows promising results



Summary

- > Optical inspection has become an established tool for quality control of SCRF cavities and is key in recognizing gradient limiting defects
- > Automatic processing of cavity images
 - Provides an objective measurement of cavity surface properties
 - Allows for search of explicit surface defects
- > Examples were shown for surface roughness and measurements of feature sizes and length
- > Chemical surface treatment of cavities reduces number of features and diminishes feature size and area, which helps to support higher accelerating fields.



Outlook

- > Statistics will be improved with the 24 HiGrade cavities with same production scheme like the 800 XFEL cavities
 - Influence of treatment onto image variables should become more obvious
 - Correlations between image variables and performance of cavities should become clearer and a prediction of performance after inspection could become possible

- > A minimal set of variables describing/classifying the cavity surface will be defined
 - Vendor classification
 - Treatment step identification
 - Prediction for operational key values (max. Eacc / Quality-factor performance)

- > Algorithm for defect detection and classification will be improved
 - Find set of variables to minimize false classification
 - Predict performance of irregularity

