

Master Thesis

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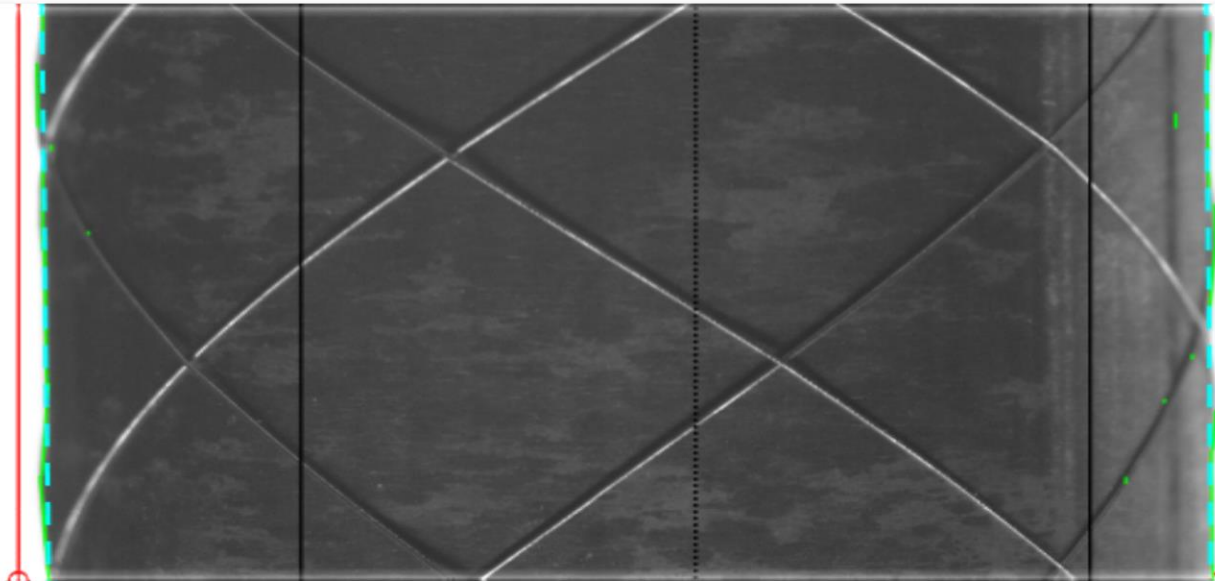
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Evaluation of Image-Based Approaches Extracting a Mandrel's Borders Using Machine Vision

cand. el. Dinggen Dai



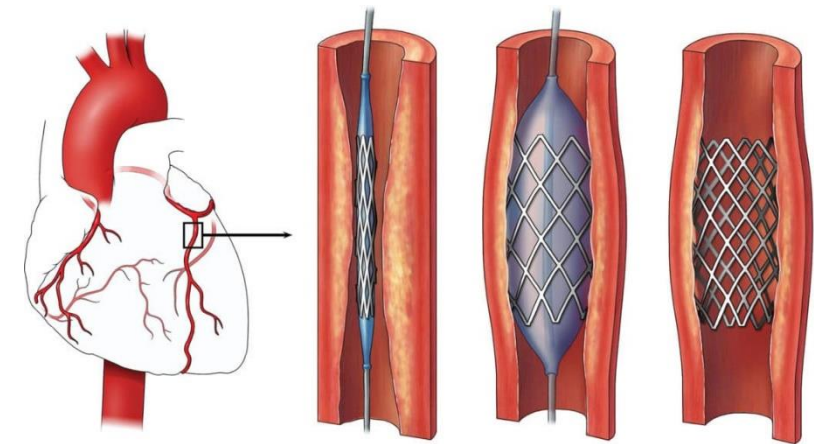
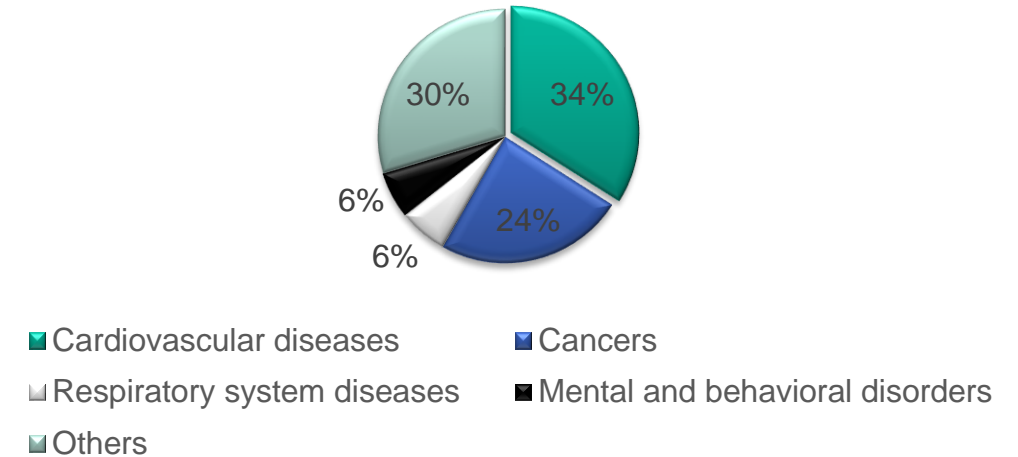
Structure

- 1 Motivation and Objective
- 2 Concept
- 3 Methods
- 4 Result and Evaluation
- 5 Conclusion and Outlook

1 Motivation and Objective

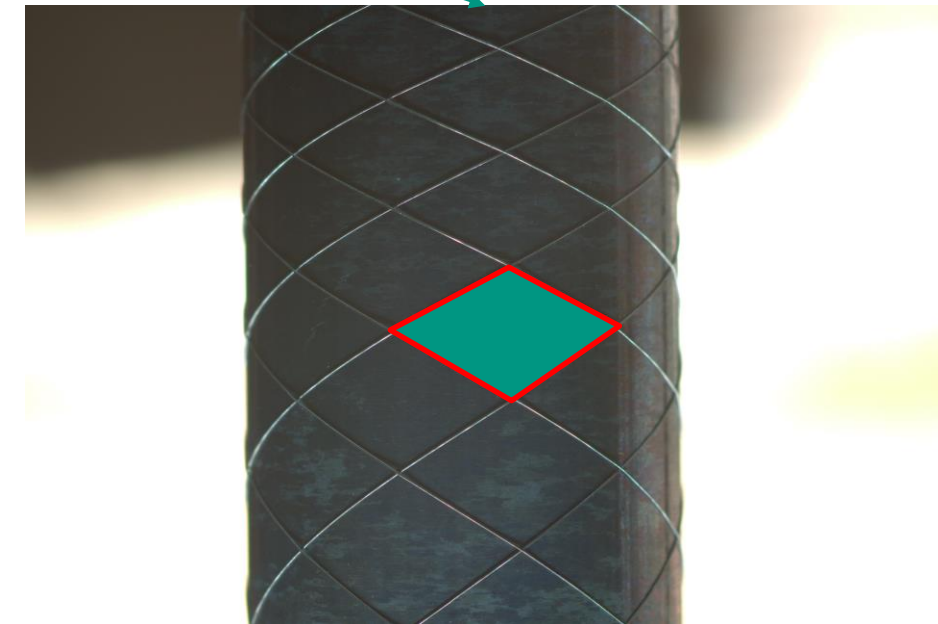
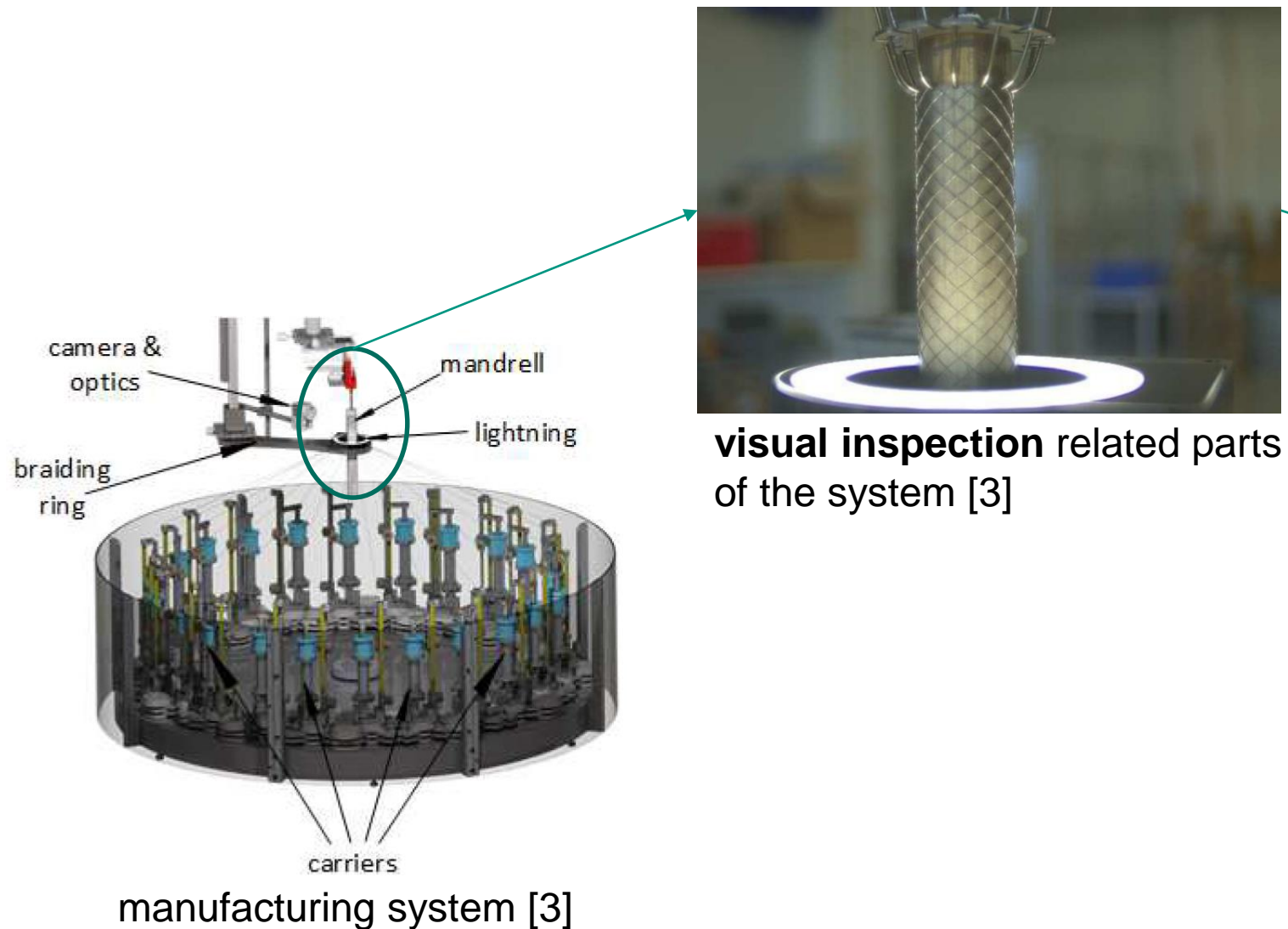
- Cardiovascular diseases caused 338,000 deaths [1]
 - 22.3% due to narrowing of heart disease vessels
- Treatable by stent implementation [2]
- Stent quality control in manufacturing is essential
 - manual inspection VS. **automatic inspection**

Causes of death in Germany 2020



Stent implementation process [2]

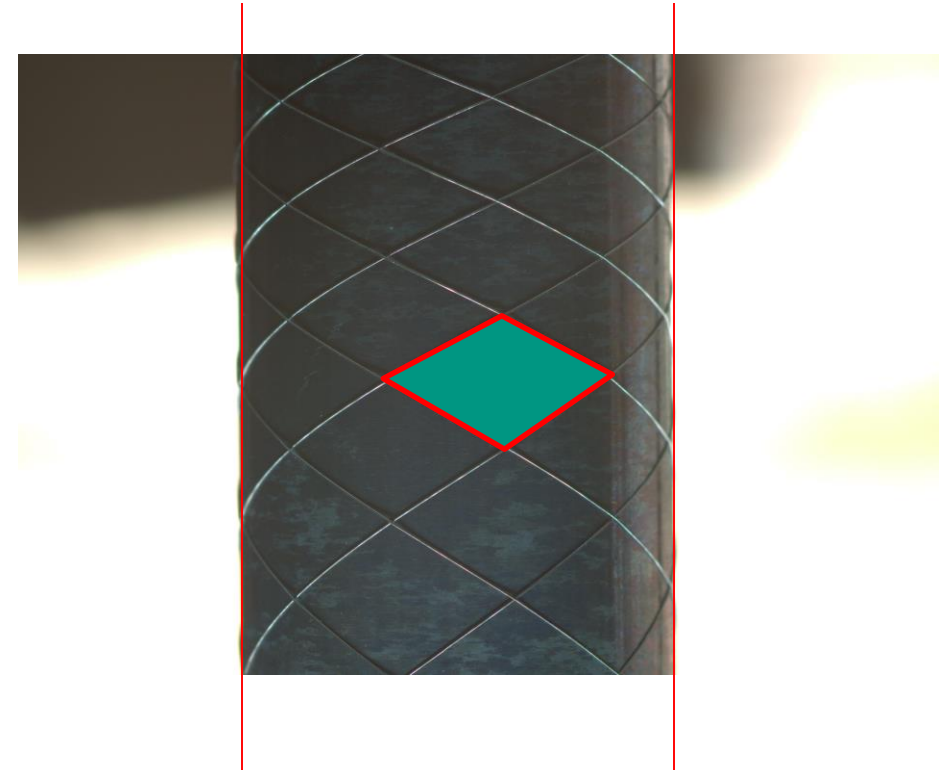
Motivation: Stent4Tomorrow



focus on the stent

Objective: Borders extraction

- Issue 1: the pitch size
 - how to “enlarge”?
- Issue 2: the background useless
 - how to remove?
- **Solution:** borders extraction!
- **Extra:** the diameter of the mandrel already known (in **mm**)
 - calculate the distance of the borders
 - build mapping relations between image (in **pixel**) and human perception (in **mm**)



2 Concept

■ Assumption:

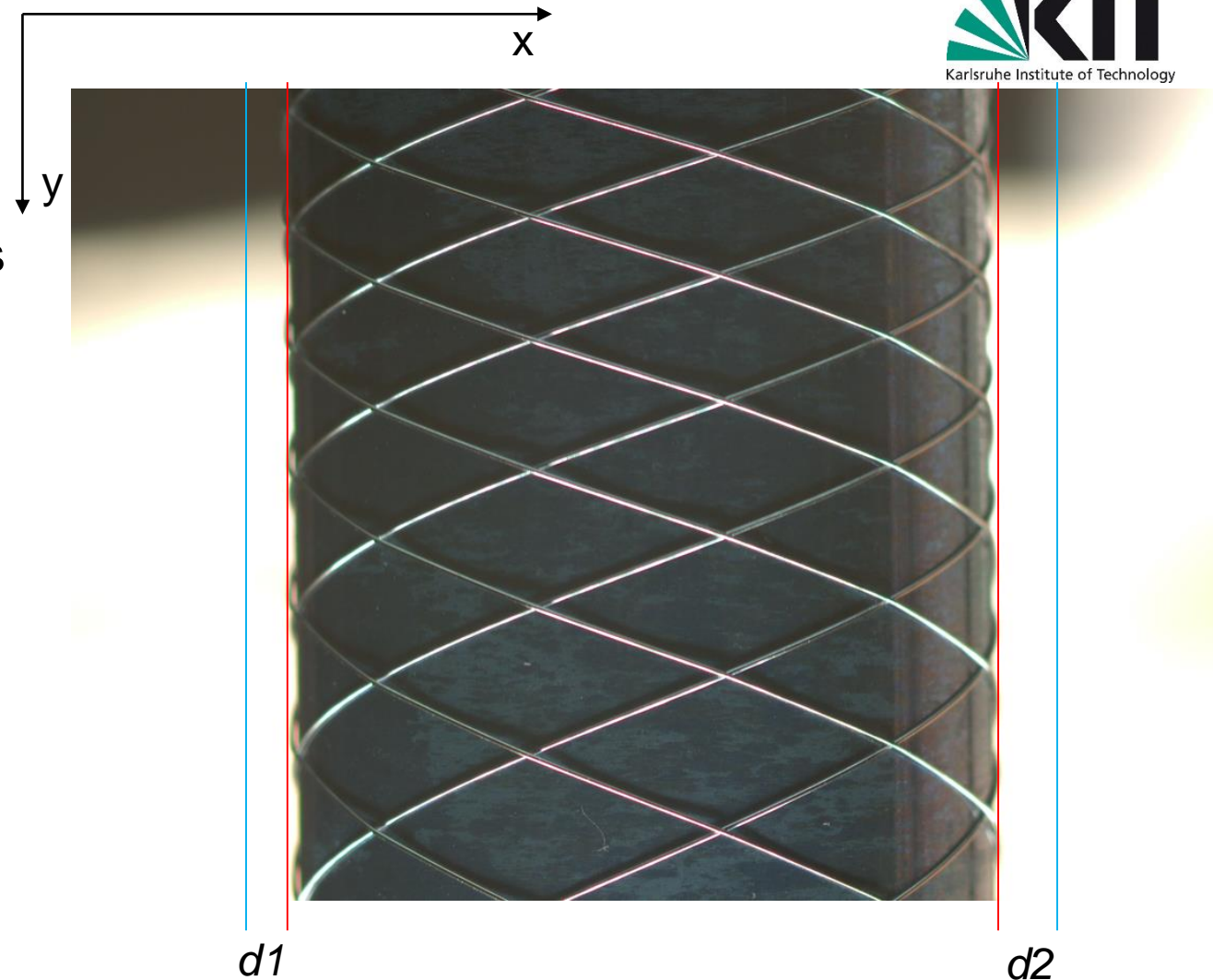
- the border lines are vertical to the x-axis of image

■ Input: RGB image

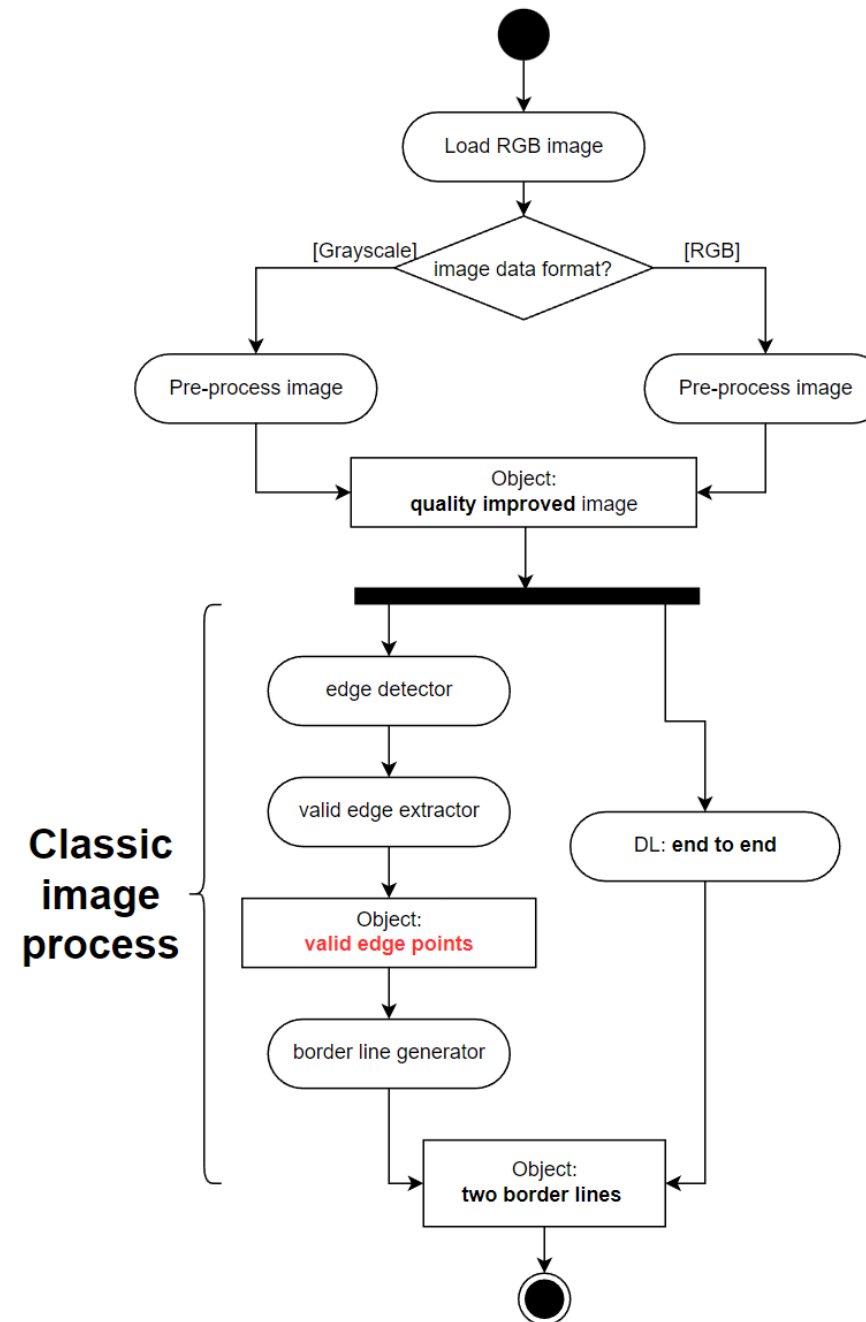
■ Output: two border lines

■ Metrics:

- 1. RSE (root square error): $\sqrt{d1^2 + d2^2}$
- 2. Execution time



Concept: diagram

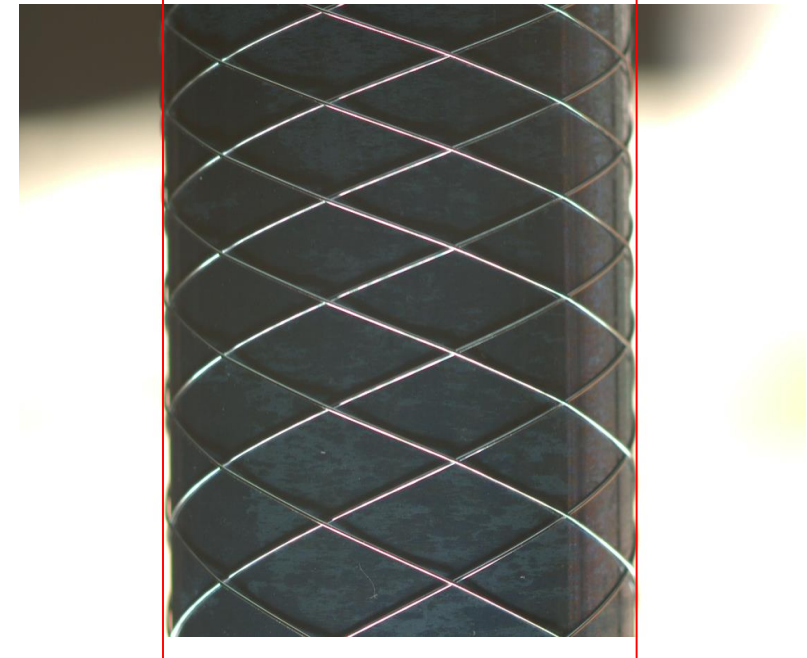


3 Methods: Dataset

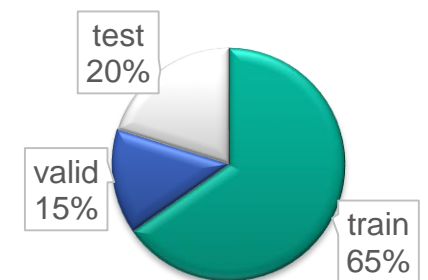
- Original Image (RGB):
 - size: 2064*3088*3 (> 5 MB)

- The labels:
 - **two border lines** vertical to the x-axis
 - format: [x_left_border, x_right_border]

- Dataset splitting:



	Training	Validation	Test
Total	5529	1224	1694
Usable ¹	4881	1085	1508



■ train ■ valid ■ test

Method: classic image process

■ Step 1. edge detector

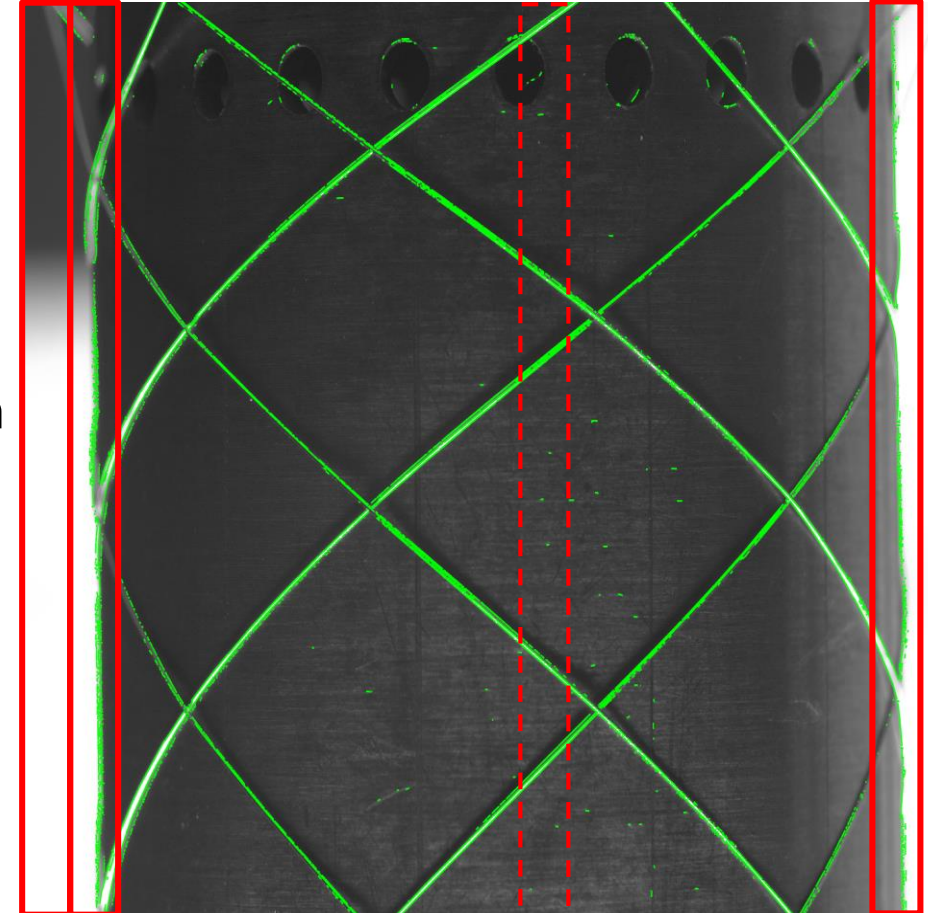
- Bottom-up approaches:
 - LSD (2010): still typical today [4]
 - EDLines (2011): 10 times faster than LSD [5]
 - Canny^{PF} (2015): parameter free [6]
- Top-down approaches: based on hough transformation
 - MCMLSD (CVPR 2017) [7]

■ Step 2. valid edge extractor

- filter: moving window

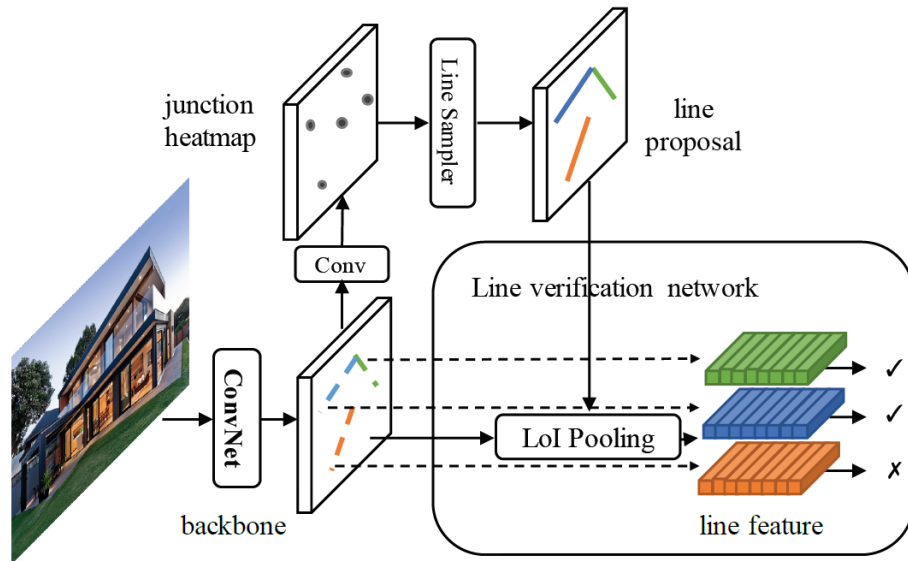
■ Step 3. border line generator

- least square method



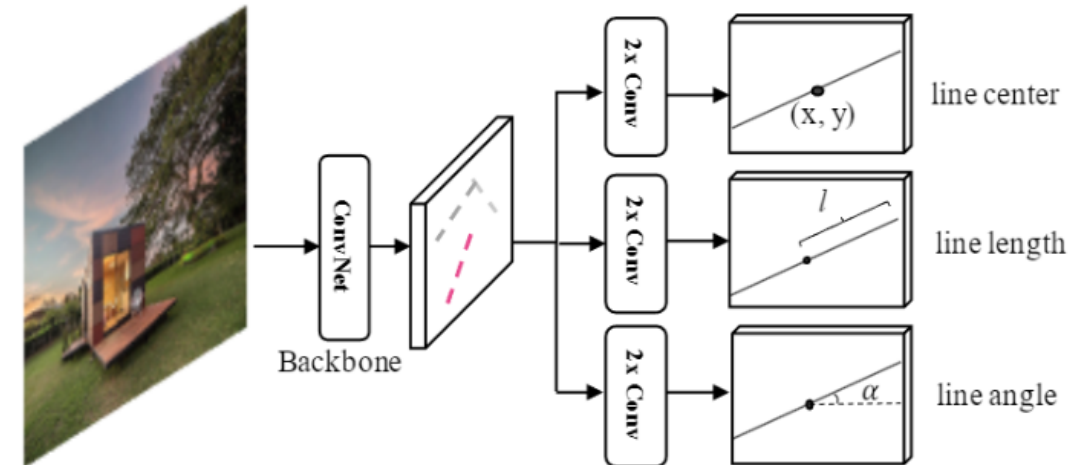
Method: deep learning

- HT-LCNN (ECCV 2020) [8]
 - based on LCNN (ICCV 2019)
 - add a trainable Hough transform block into a deep network



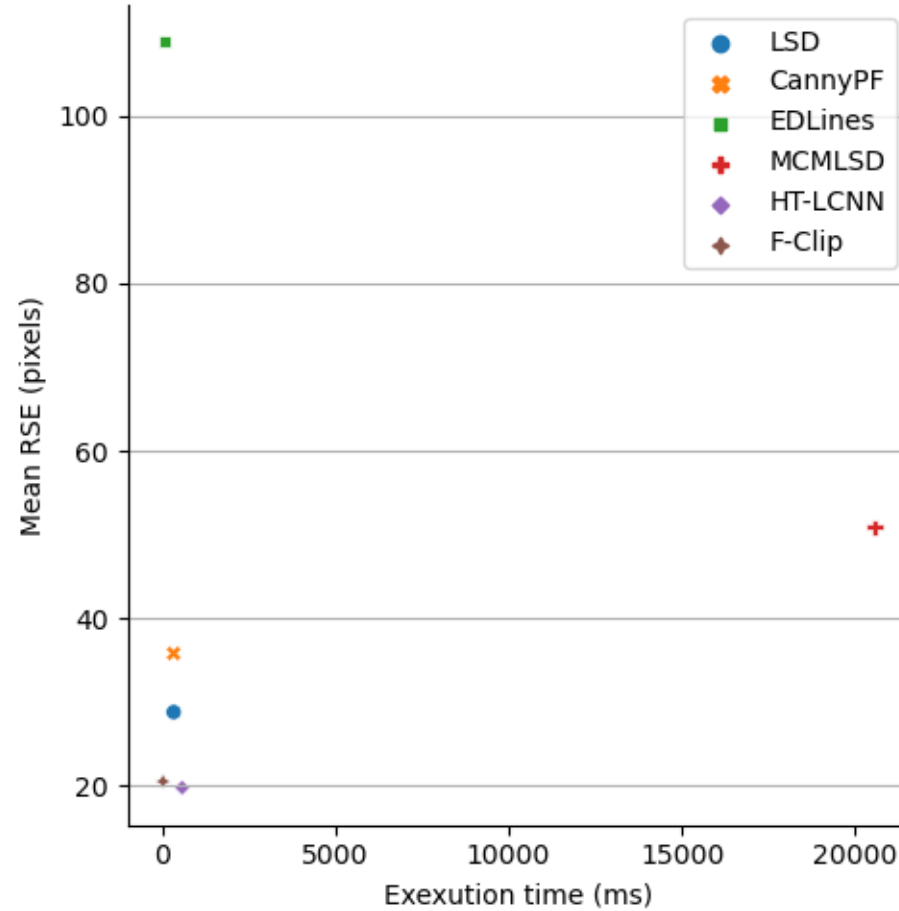
An overview of LCNN architecture

- F-Clip [9]
 - SOTA of line detectors (2021)
 - one stage: much more faster
 - the representation of a line

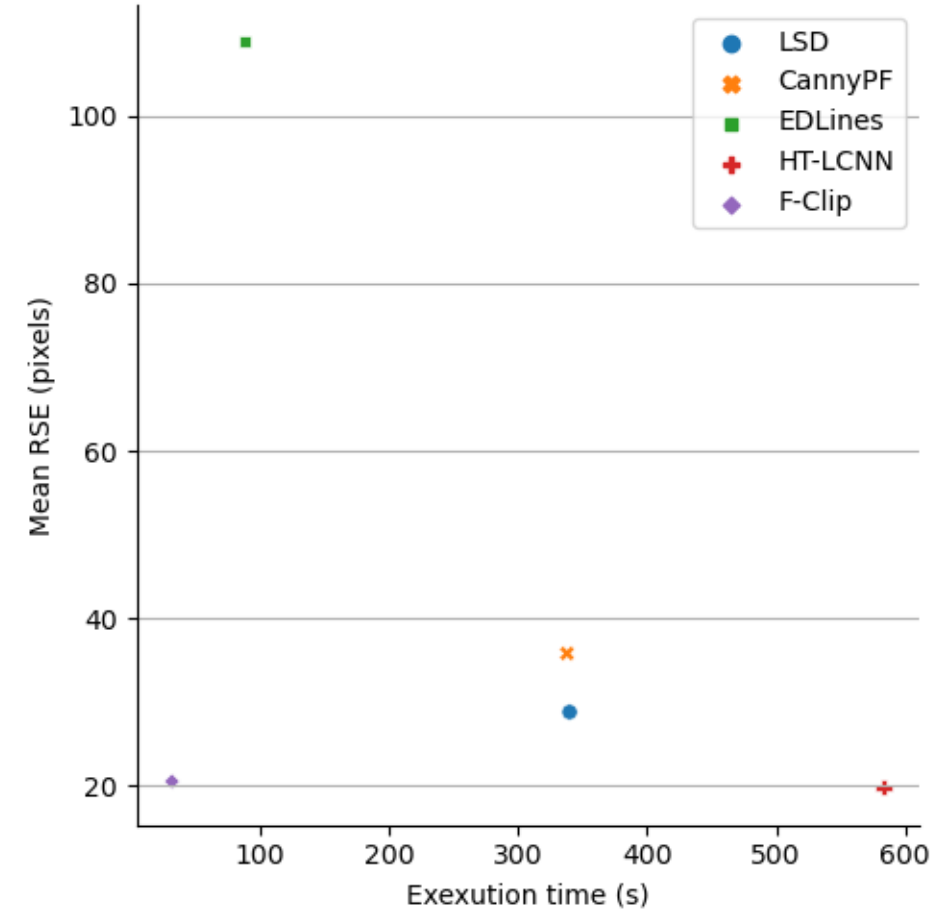


An overview of F-Clip architecture

4 Result & Evaluation

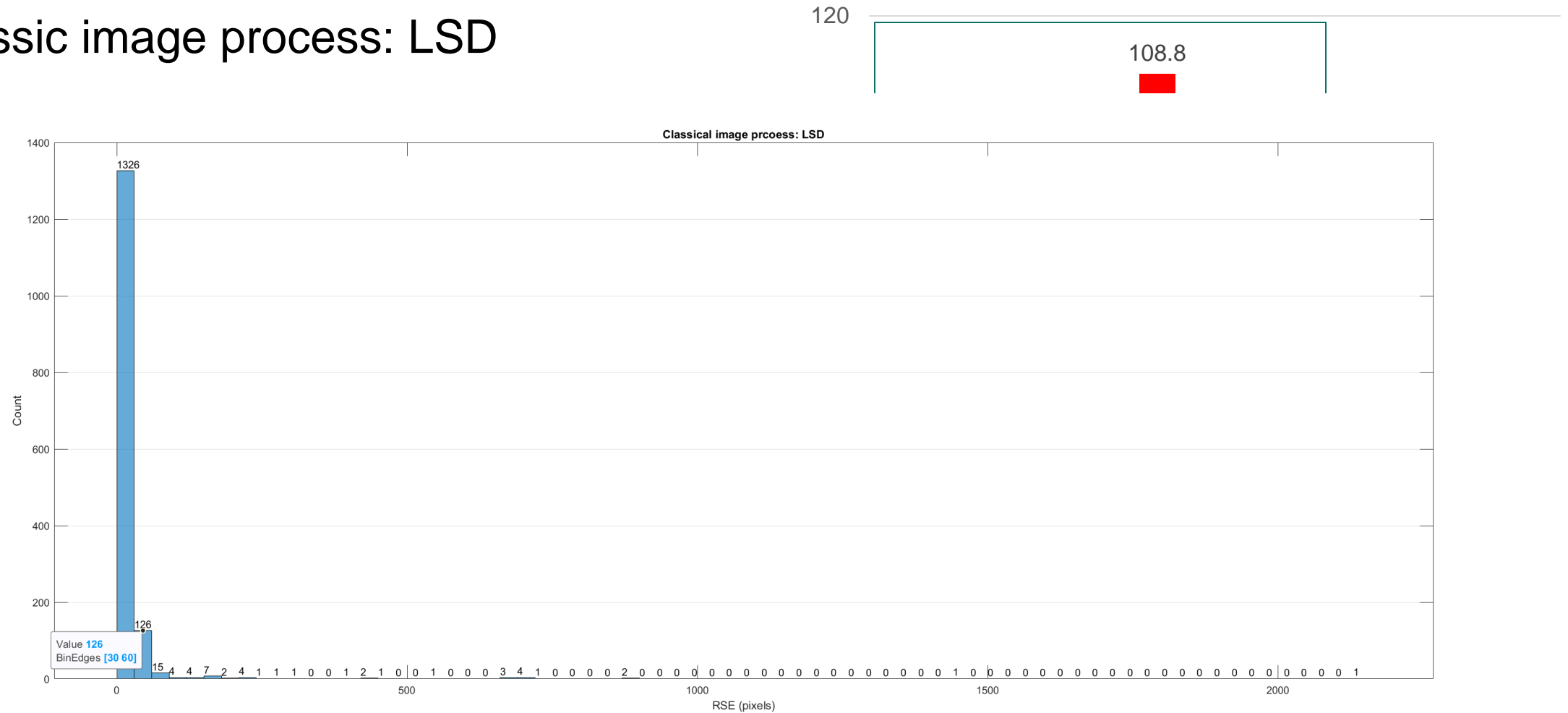


Without MCMLSD



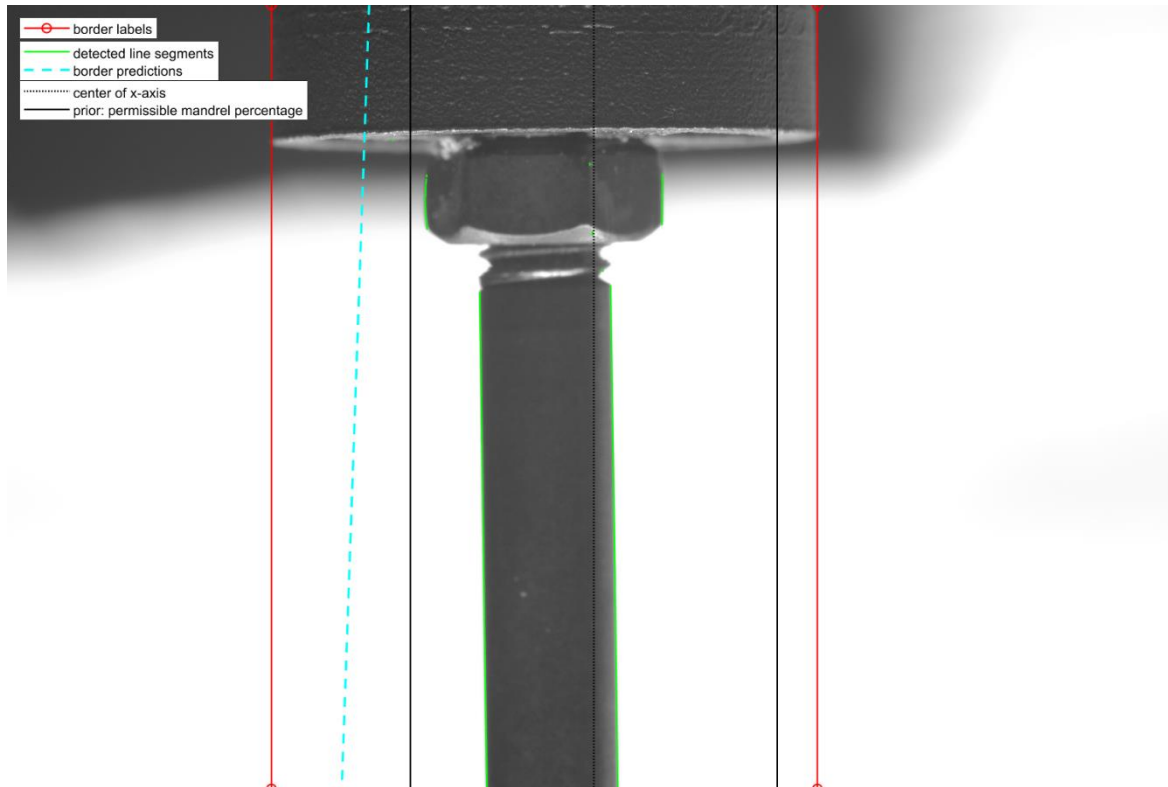
Evaluation: RSE

■ Classic image process: LSD

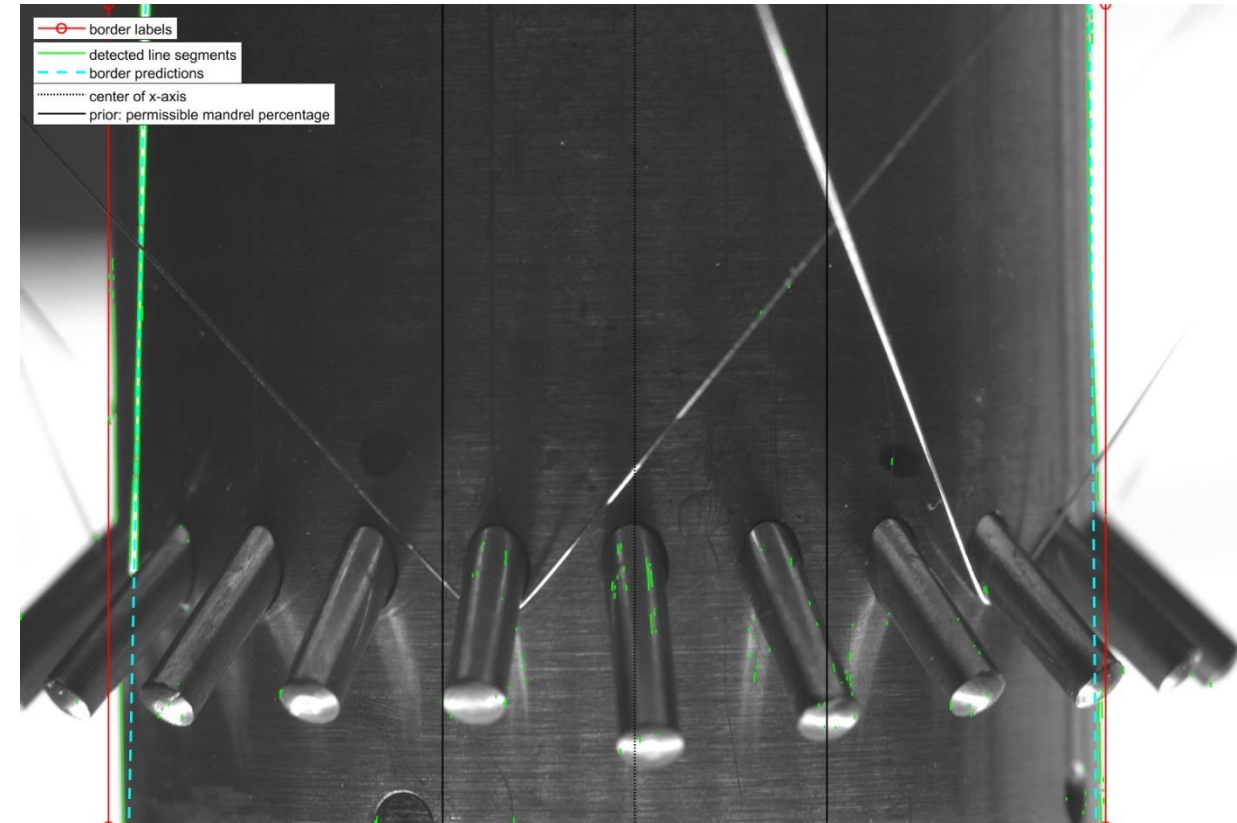


Evaluation: RSE

■ LSD: 2 kinds of bad cases



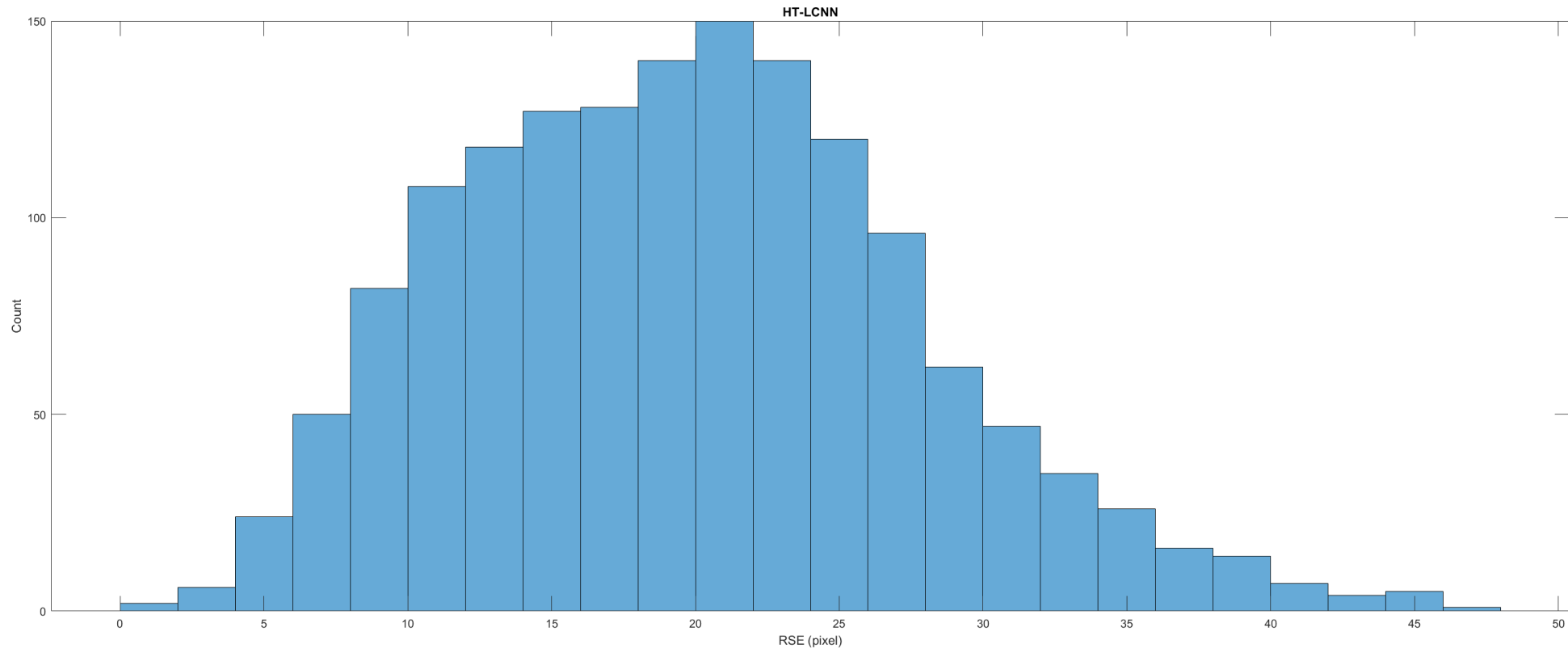
Case 1: no edge detected around the borders



Case 2: noise (stent wire) around the left border

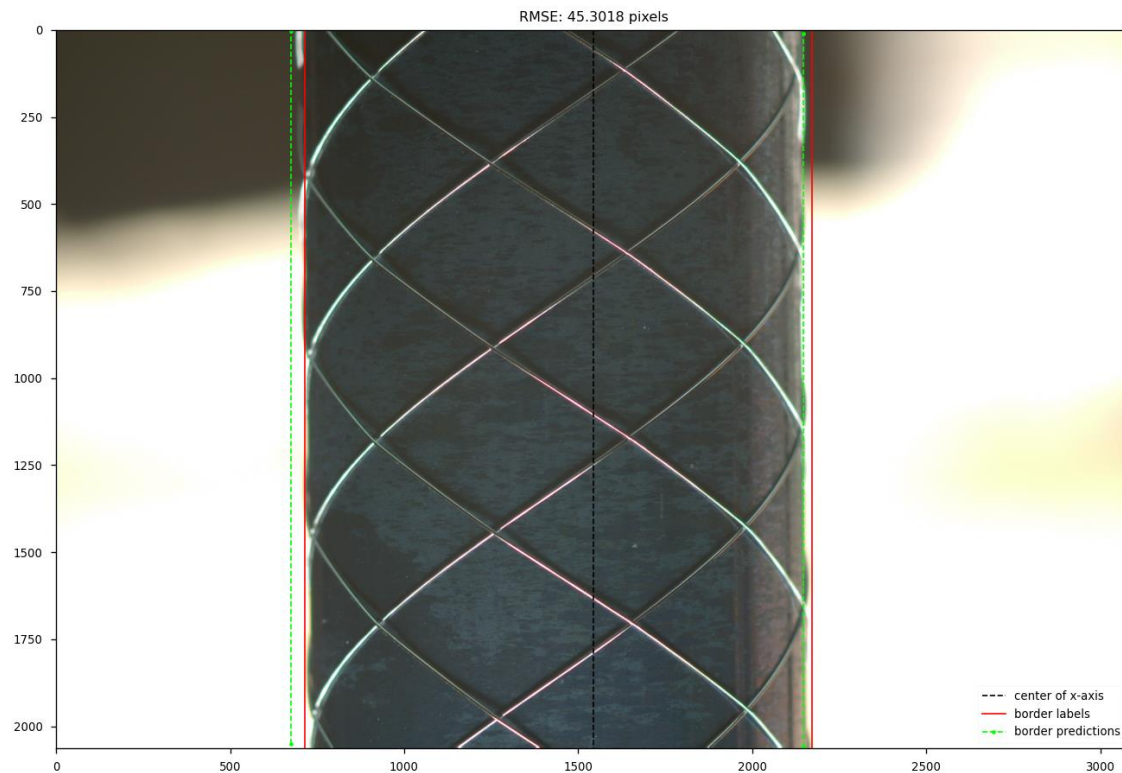
Evaluation: RSE

■ Deep learning: HT-LCNN

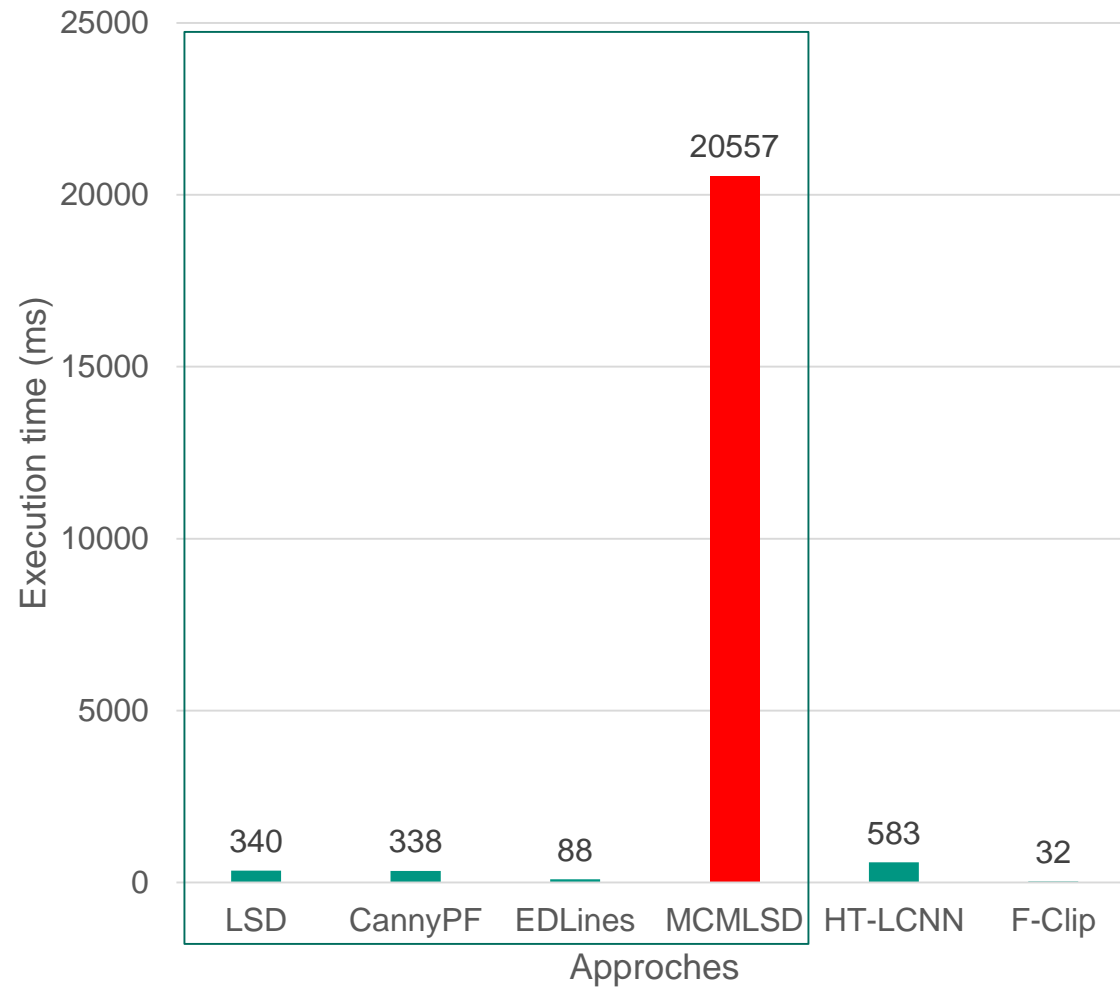


Evaluation: RSE

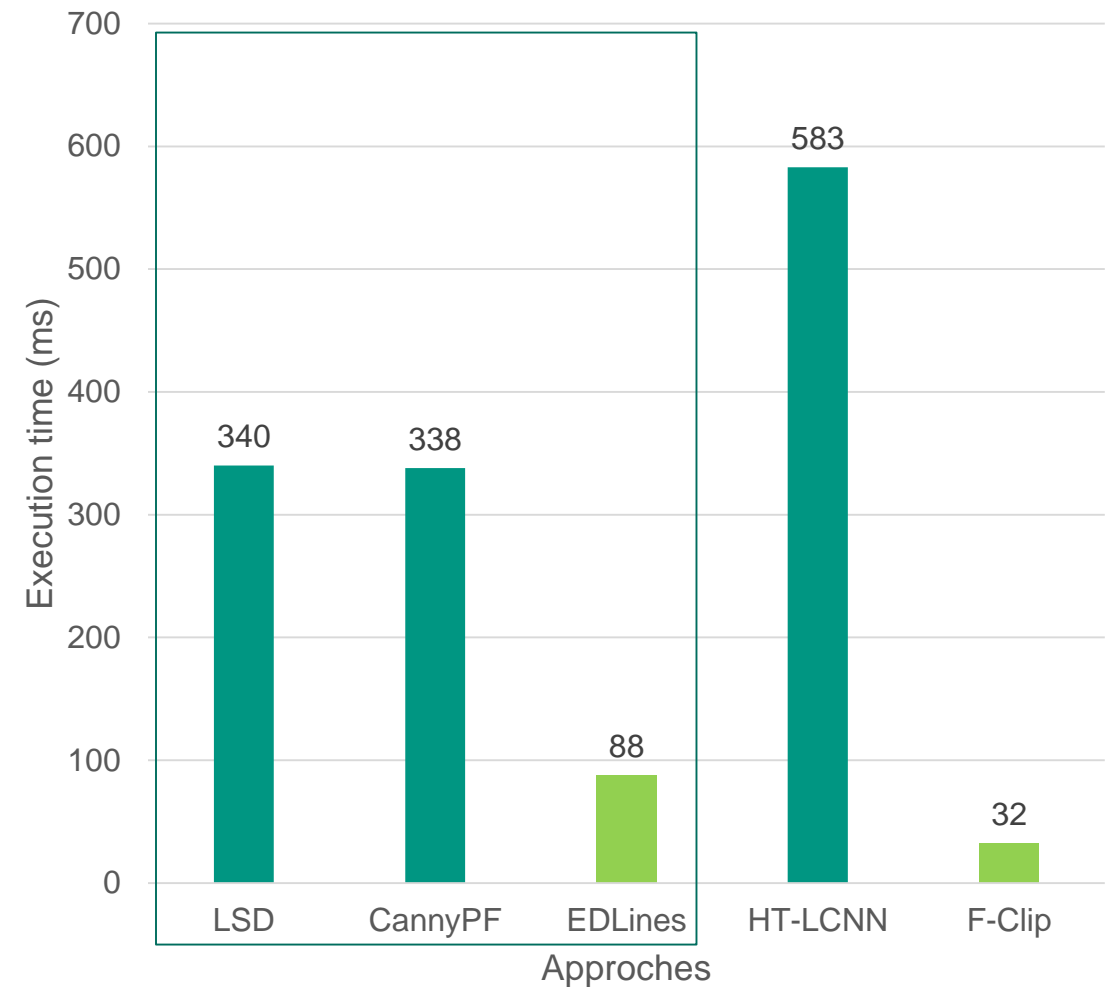
- HT-LCNN: the **predicted** left border line in the left side of the **labeled**



Evaluation: execution time



Without MCMLSD



5 Conclusion and outlook

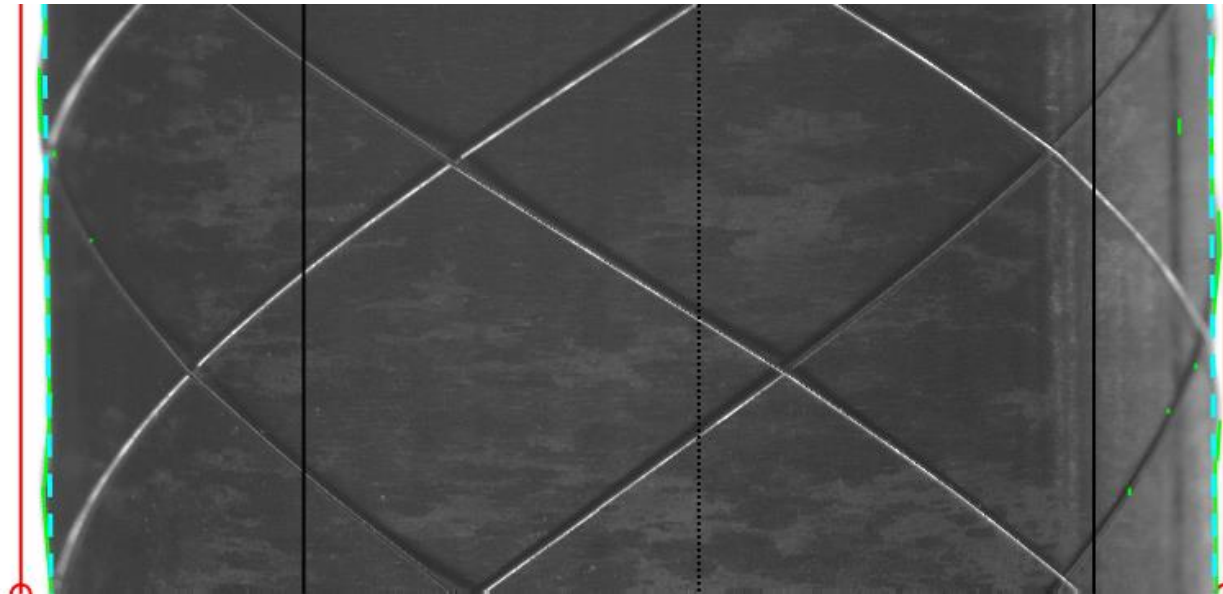
■ Conclusion:

- Implemented / reproduced 6 approaches to detect the border lines
 - 4 in classic image process: LSD, EDLines, CannyPF, MCMLSD
 - 2 in deep learning: HT-LCNN, F-Clip
- Metric RSE:
 - **DL models** show better performance: more **robust** to the noise
 - Lowest of all: **HT-LCNN** 19.77 pixels
 - Lowest of classic methods: **LSD** 28.8 pixels
- Metric execution time: F-Clip is the fastest with 32 (ms / image)

■ Outlook:

- Classic image process: based on LSD, deal with the two bad cases (robust to noise)
- Deep learning: try to use two junctions to represent a border line instead of the vertical line
 - compared to the top point of the left border (obstruction from the background), easy to percept for human being

Thank you for your attention!



Appendix - Reference

[1]