

# DETECTION OF BURIAL MOUNDS IN HIGH-RESOLUTION SATELLITE IMAGES OF AGRICULTURAL LAND

Øivind Due Trier (NR), Anke Loska (NDCH),  
Siri Øyen Larsen (NR), and Rune Solberg (NR)

## Collaborators:

The Norwegian Directorate for Cultural Heritage, NDCH (also funding)

The Norwegian Space Centre (also funding)

Vestfold County Administration

Museum of Cultural History - University of Oslo



# Outline

- ▶ Background
- ▶ Methods
  - contrast enhancement
  - template matching
  - classification experiments
- ▶ Results
- ▶ Conclusion



# CultSearcher

- ▶ Software for computer assisted detection of potential cultural heritage sites
- ▶ Agricultural fields
- ▶ Soil marks
- ▶ Crop marks
- ▶ Circular patterns – could be remains of burial mounds



# Quickbird images



# Cross section of ditch that has surrounded bronze age burial mound



# Detail of Lågendalen image



**RGB**

**2.4 m**



**Near infrared**

**2.4 m**



**Panchromatic**

**0.6 m**

# Where to look

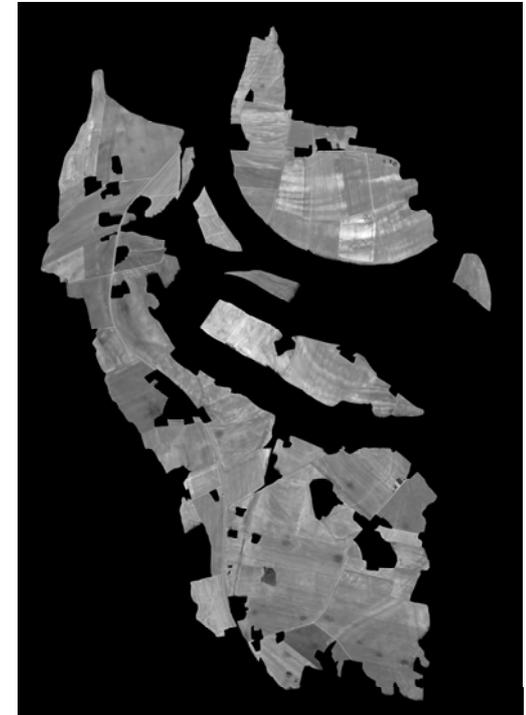
- ▶ We only consider agricultural fields



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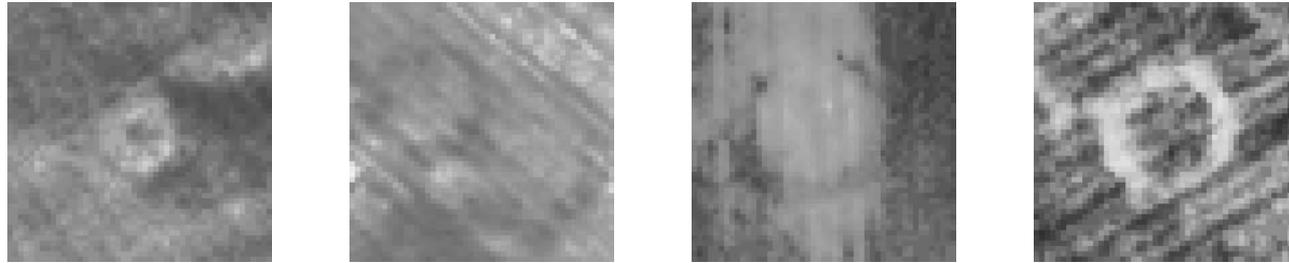
# Test data

- ▶ Altogether 35 rings in the two images:
  - 15 strong rings (clear visibility)
  - 10 fairly strong rings (moderate visibility)
  - 10 weak rings (poor visibility)
- ▶ The appearance of different rings varies greatly:
  - radius,
  - thickness,
  - gray tone intensity,
  - degree of completeness,
  - contrast to the local background

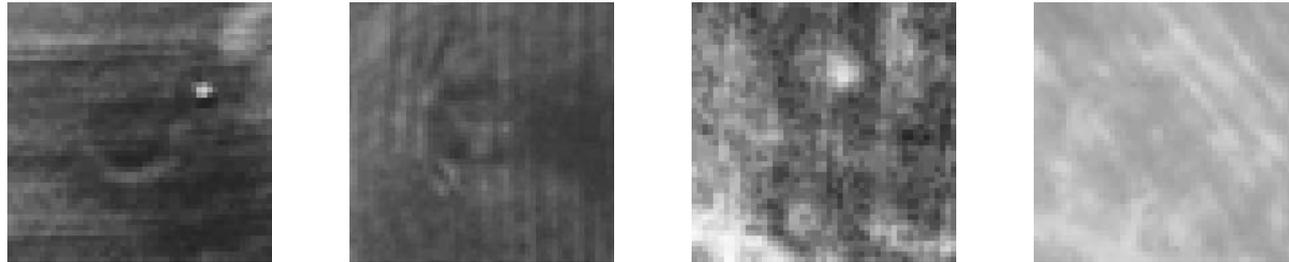


# Example ring marks

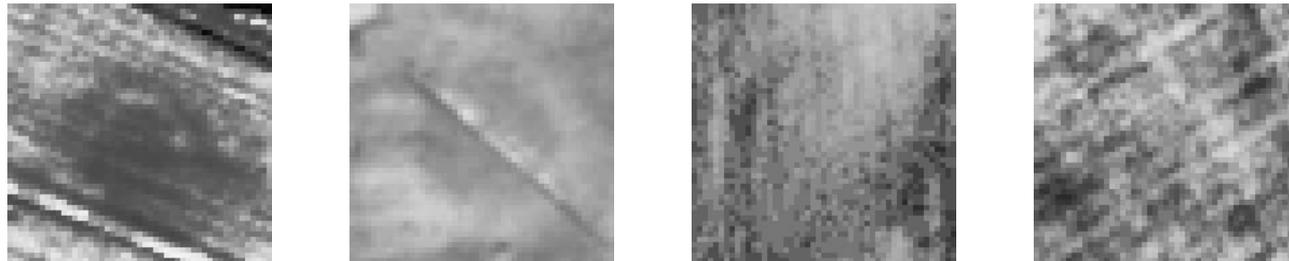
**Strong  
rings:**



**Fair  
rings:**



**Weak  
rings:**



**The contrast has been adjusted to highlight the rings**

# Local contrast enhancement

- ▶ The pixel value  $p_{CE}(x, y)$  in the contrast enhanced image is computed as

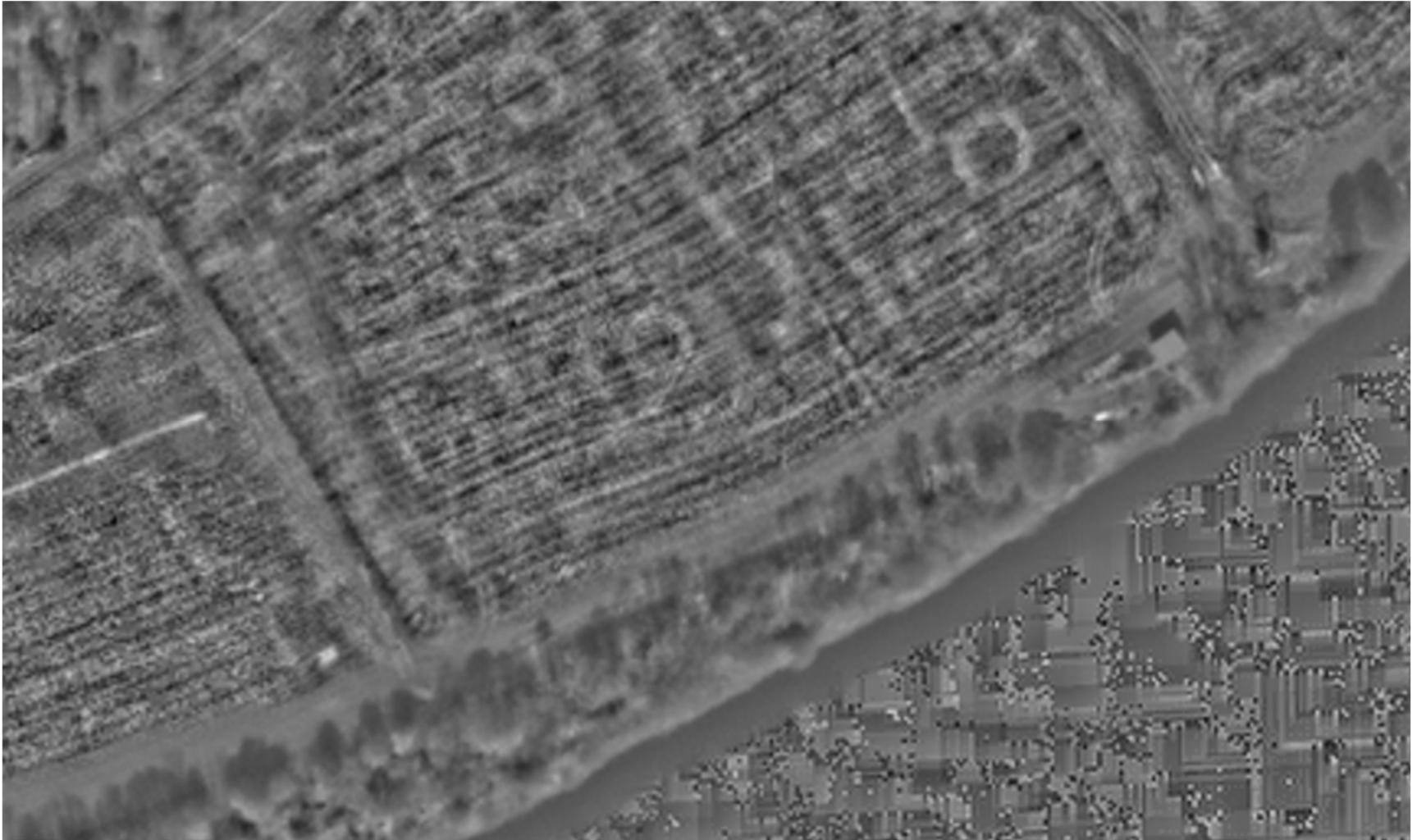
$$p_{CE}(x, y) = \frac{p(x, y) - \mu(x, y, N)}{\sigma(x, y, N)}$$

in an  $N \times N$  neighbourhood centered on  $(x, y)$  .

- ▶ Achieves more or less constant local contrast over the entire image.

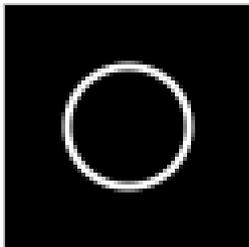


# Local contrast enhancement

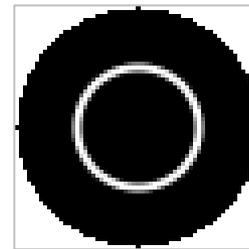


# Template matching

- ▶ A ring filter is convolved with the image
- ▶ Correlation image; pixel value indicate how well the ring filter agrees with the image when centered on the respective location



**Square boundary**



**Circular boundary**

# Locating potential ring sites

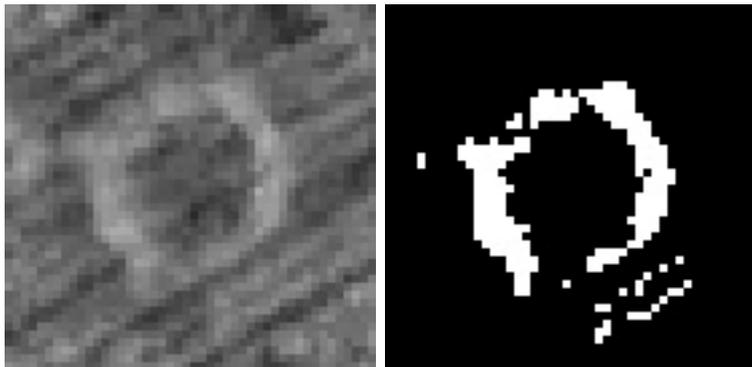
- ▶ Locations with high correlation
- ▶ Threshold  $\tau$ :
  - correlation image  $> \tau \Leftrightarrow$  bright ring
  - correlation image  $< -\tau \Leftrightarrow$  dark ring
- ▶ The threshold  $\tau$  may be adjusted by the user
  - influence true ring recognition rate vs. number of false detections



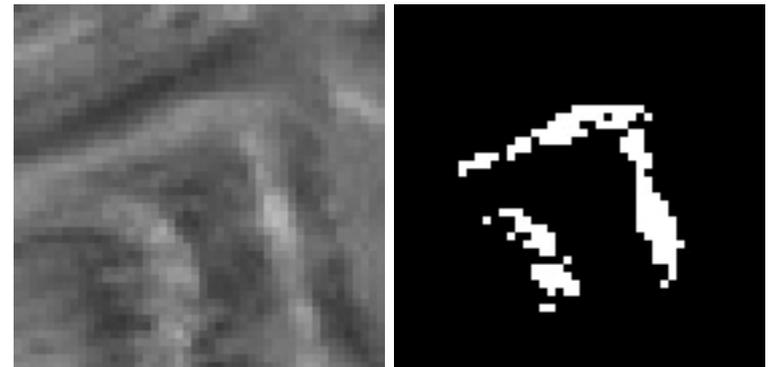
# Classification experiments

- ▶ Are there features that can discriminate false positives from true rings?
- ▶ Features extracted from  $4r \times 4r$  sub images (panchromatic + binary) containing ring candidates

**True ring**



**False ring**

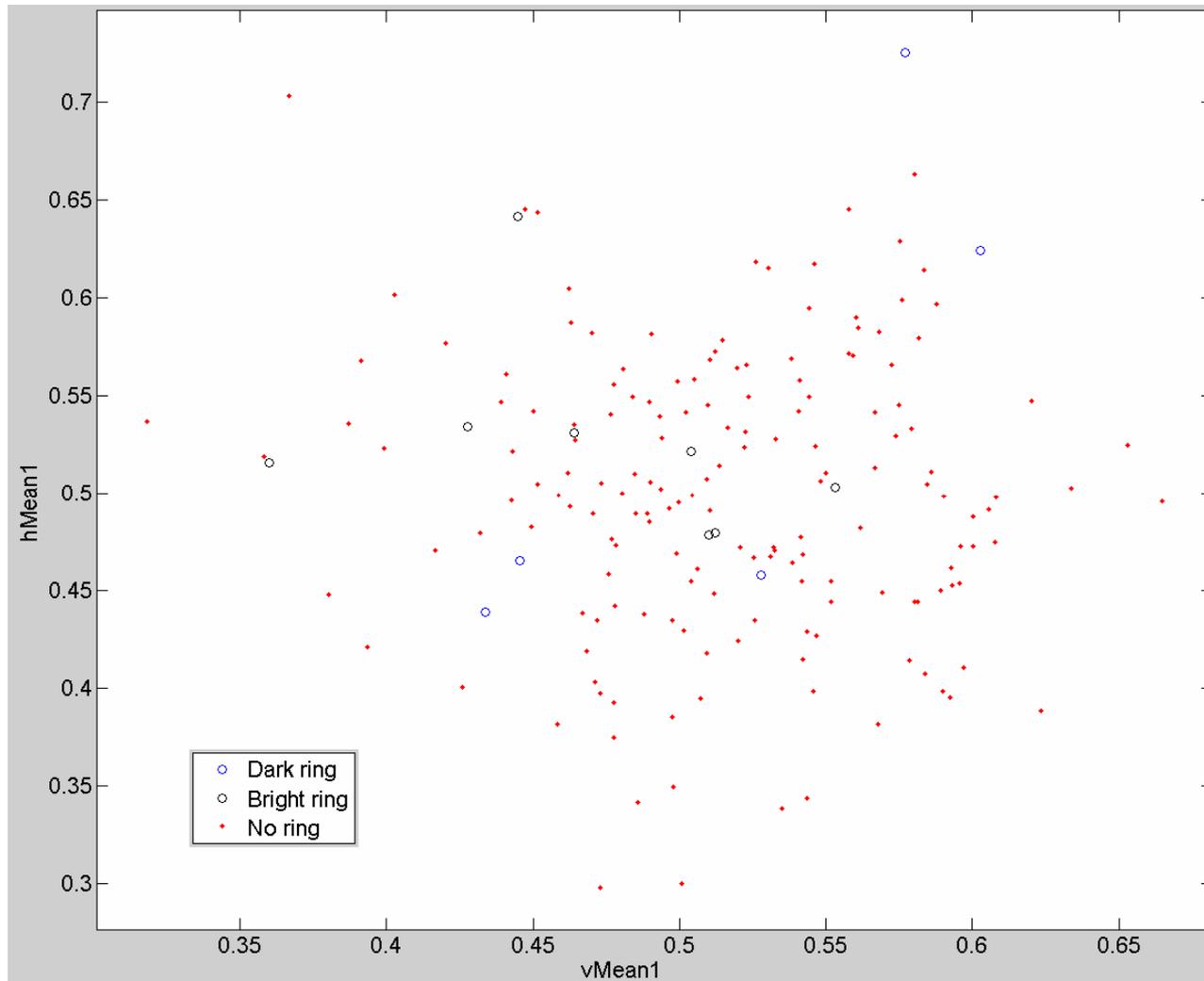


# Classification experiments

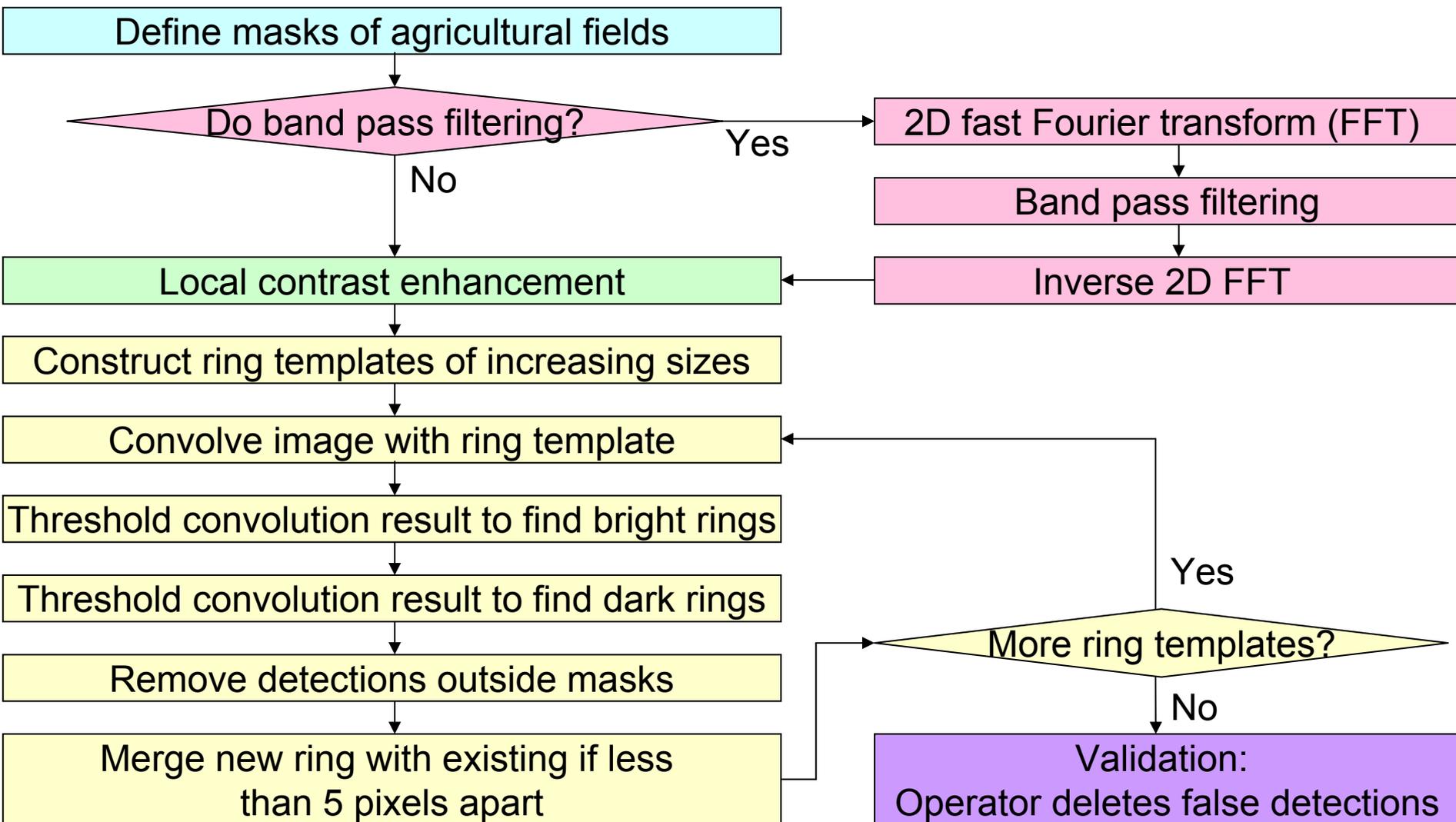
- ▶ Features include:
  - ring cover; overlap between binarized sub image and binary version of the ring filter
  - mean x- and y-coordinates of binarized image
  - Hu moment invariants
  - Real weighted Fourier moments
- ▶ Decision tree classifier
- ▶ Results discourage the use of classification



# Scatter plots



# Flowchart of algorithm



# Results

band pass	corr. thresh.	strong rings	fair rings	weak rings	true rings	false rings
no	0.30	11	5	0	16	450
no	<b>0.33</b>	<b>11</b>	<b>5</b>	<b>0</b>	<b>16</b>	<b>109</b>
no	0.35	10	2	0	12	39
no	0.40	8	0	0	8	3
yes	0.35	12	3	0	15	174
yes	0.38	11	2	0	13	48
yes	0.39	10	2	0	12	31
yes	0.40	9	1	0	10	12
ground truth		15	10	10	35	



# Verification

## Detected rings



## Verified rings



**Bright  
rings**

**Dark  
rings**

# Conclusion

- ▶ Detection of rings is a challenging task.
- ▶ Local contrast enhancement
- ▶ Template matching
- ▶ Archaeologists state that the software tool is helpful.
  - Avoid manual inspection of entire images.
  - Easy to remove false detections.



# Thank you for your attention

