



# Robust Angle Invariant 1D Barcode Detection

Alessandro Zamberletti

A. Zamberletti, I. Gallo and S. Albertini

Department of Theoretical and Applied Science (DiSTA)  
University of Insubria

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## Goals:

- detection of 1D barcodes from real world images
- fast execution on a low computational power device
- robust to lighting variations, background clutter, rotation and distortions

## Existing 1D barcode detection methods:

- gradients analysis
  - O. Gallo and R. Manduchi, "Reading 1-D barcodes with mobile phones using deformable templates", IEEE PAMI, 2011.
- scan lines
  - S. Wachenfeld, S. Terlunen and X. Jiang, "Robust 1-D barcode recognition on camera phones and mobile product information display", Mobile Multimedia Processing, 2010.
  - R. Adelman, M. Langheinric, and C. Floerkemeier, "A toolkit for barcode recognition and resolving on camera phones", MEIS, 2006.
- canny edge detector
  - E. Basaran, O. Ulucay and S. Erturk, "Reading barcodes using digital cameras", IMS, 2006.

## Issues:

- slow
- rotation





## Novel detection approaches:

- optical character recognition methods
  - numbers are not part of the barcode standard
- Hough transform for lines detection
  - successfully used for reading 1D barcodes
  - R. Muniz, L. Junco and A. Otero, "A robust software barcode reader using the Hough Transform", IISA, 1999.

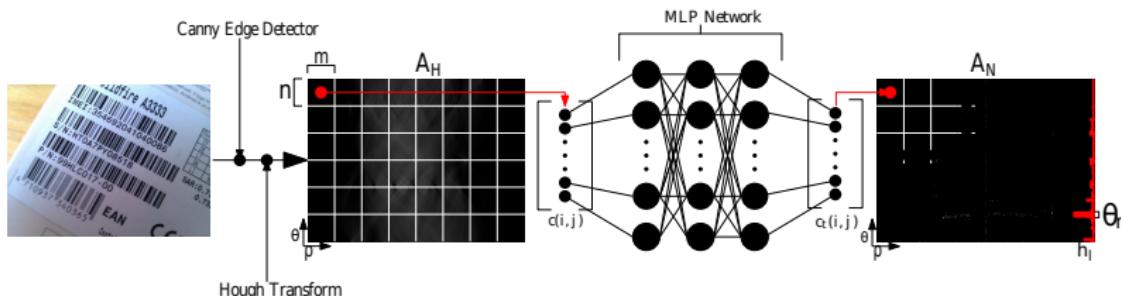


Detection of 1D barcode patterns in the Hough Transform space using a simple MLP network:

- **pros**
  - rotation invariant
  - fast
- **cons**
  - sensitive to noise
  - requires training



# Supervised Angle Detection

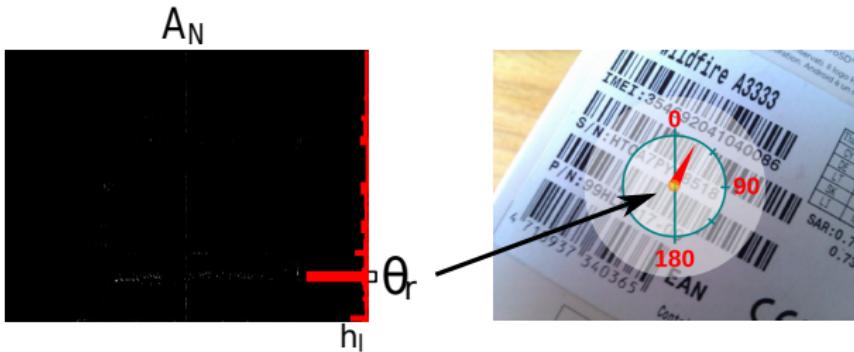


- 1 a regular grid  $n \times m$  of cells  $c_1, \dots, c_{n \times m}$  is superimposed over the Hough accumulator matrix  $A_H$
- 2  $c$  is given as input to the MLP
- 3 the MLP produces a new cell  $c_t$

$$c_t(i,j) = \begin{cases} 1 & \text{if } c(i,j) \text{ denotes a barcode bar in } I. \\ 0 & \text{otherwise.} \end{cases}$$

- 4 the processed cells are combined together to generate  $A_N$

# Supervised Angle Detection



- 5 given a row  $r \in A_N$ , the sum of its elements denotes the likelihood  $l_r$  of a barcode appearing in  $I$  rotated by  $\theta_r$ .
- 6 we build the histogram of likelihoods  $h_l$
- 7 rows associated with max elements of  $h_l$  denote barcodes



# Supervised Angle Detection - Details

## MLP:

- 3 layers
- size of each layer:  $n \times m$
- trained using *rprop* (parameters from Igel *et al.* 2005)
- 200 training patterns (150 bgs, 50 fgs)

**Training patterns** -  $\forall c_i \in A_H, (in_{c_i}, out_{c_i})$ :

- $in_{c_i}$  - vector representation of  $c_i$
- $out_{c_i}$  - elements of  $in_{c_i}$  denoting barcode bars are assigned 1, the others are assigned 0



# Supervised Angle Detection - Datasets

Dataset	No. of Imgs	Device	Barcode Rotation
WWU Muenster Barcode DB*	1055	Nokia N95	$\pm 30^\circ$
ArTe-Lab 1D Medium Barcode**	215	Nokia 5800	//
Rotated Barcode DB**	368	Various	$\pm 180^\circ$

All the datasets are available online for download:

- \* <http://cvpr.uni-muenster.de/research/barcode/Database/>
- \*\* <http://artelab.dista.uninsubria.it/download/>



# Supervised Angle Detection - Evaluation

## Results:

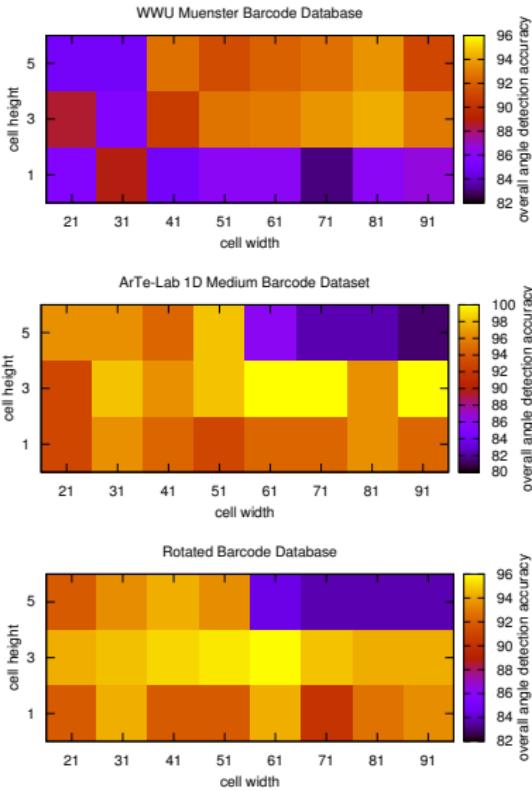
- ArTe-Lab 1D Medium Barcode  $\simeq 100\% OA^\theta$
- WWU Muenster Barcode DB & Rotated Barcode DB  $\simeq 95.5\% OA^\theta$

## Time:

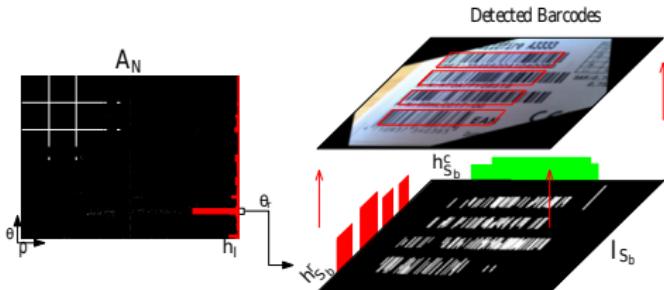
- $\sim 200$  ms per image

## Metric:

- $OA^\theta = \frac{tp}{tp + fn + fp}$

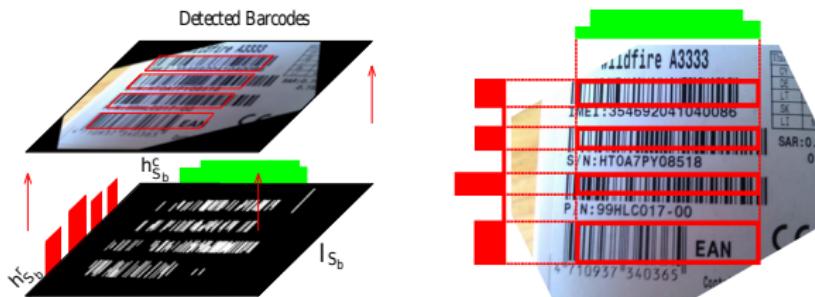


# Bounding Box Detection



- 1 rotate  $I$  by  $\theta_r$  (detected during the previous phase)
- 2 get the segments parallel to the vertical, as in Mataszyz *et al.*, 1999
- 3 obtain a binary image  $I_{S_b}$
- 4 define two histograms  $h_{S_b}^r$  and  $h_{S_b}^c$  describing the intensity profiles of the rows and columns of  $I_{S_b}$  respectively

# Bounding Box Detection



- 5 smooth the histograms to remove low value bins corresponding to isolated non-barcode segments
- 6 the detected bounding boxes correspond to the intersection area between the rows and the columns associated with the non-zero bins of  $h_{S_b}^r$  and  $h_{S_b}^c$

# Bounding Box Detection - Evaluation

Dataset	$OA^{bb}$
WWU Muenster Barcode DB	0.83
ArTe-Lab 1D Medium Barcode	0.86
Rotated Barcode DB	0.84

## Time:

- ~ 70 ms per image

## Metric:

- $OA^{bb} = \frac{tp}{tp + fn + fp}$
- $tp$  = # of barcode bounding boxes correctly detected
- a bounding box  $d_b$  is correctly detected iff  $\frac{bb_b \cap d_b}{bb_b \cup d_b} \geq 0.5$



Dataset	Reading Accuracy		
	ZXing	Our $\cup$ ZXing	$\Delta_{acc}$
WWU Muenster Barcode DB	0.73	0.81	+0.08
ArTe-Lab 1D Medium Barcode	0.82	0.85	+0.03
Rotated Barcode DB	0.61	0.82	+0.19



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## Thank You!

Department of Theoretical and Applied Science (DiSTA)  
University of Insubria  
Varese, Italy

email: [ignazio.gallo@uninsubria.it](mailto:ignazio.gallo@uninsubria.it)  
web: <http://artelab.dista.uninsubria.it/>

