

Handwritten \rightarrow \LaTeX

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Objectives

Recognize handwritten math symbols and combine them to a \LaTeX -formula

- Generate handwritten formula by single symbols
- Detection and classification of symbols
- Generating a \LaTeX -sequence

Data

- > 350,000 images of 82 different math symbols
- Combining 23 different symbols to easy formula
- Normalization of symbols
 - Removing the border
 - Scale to at most 40×40
 - Center the mass in a 48×48 image
 - Subtract mean and divide by standard deviation

Figure 1: Generated formula

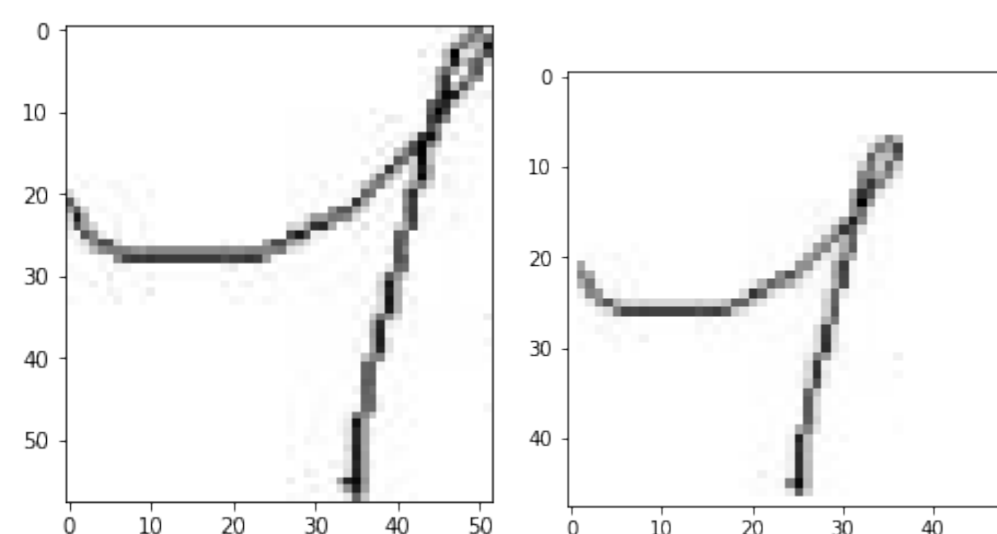


Figure 2: Crop and centered single math symbol

Approaches

Image classification and sliding window

- Image classification using CNN
- Sliding window for object detection
- Very slow
- Low accuracy due to prediction if overlapping (no false examples)

Object recognition in one step

- Detection and classification using Tensorflow API
- Training phase too long
- Too complicated for this problem

Final approach

Image classification using CNN

OpenCV for object detection

- 3 Layer CNN trained on > 100,000 images
- Classification with > 99% accuracy
- OpenCV contour finding and bounding boxes
- Combining overlapping bounding boxes
- fast and reasonable for this simple data

Sequence 2 Sequence Model

- Use predictions and positions
- Simple greedy sequence to sequence model
- Accuracy around 85%

Pipeline

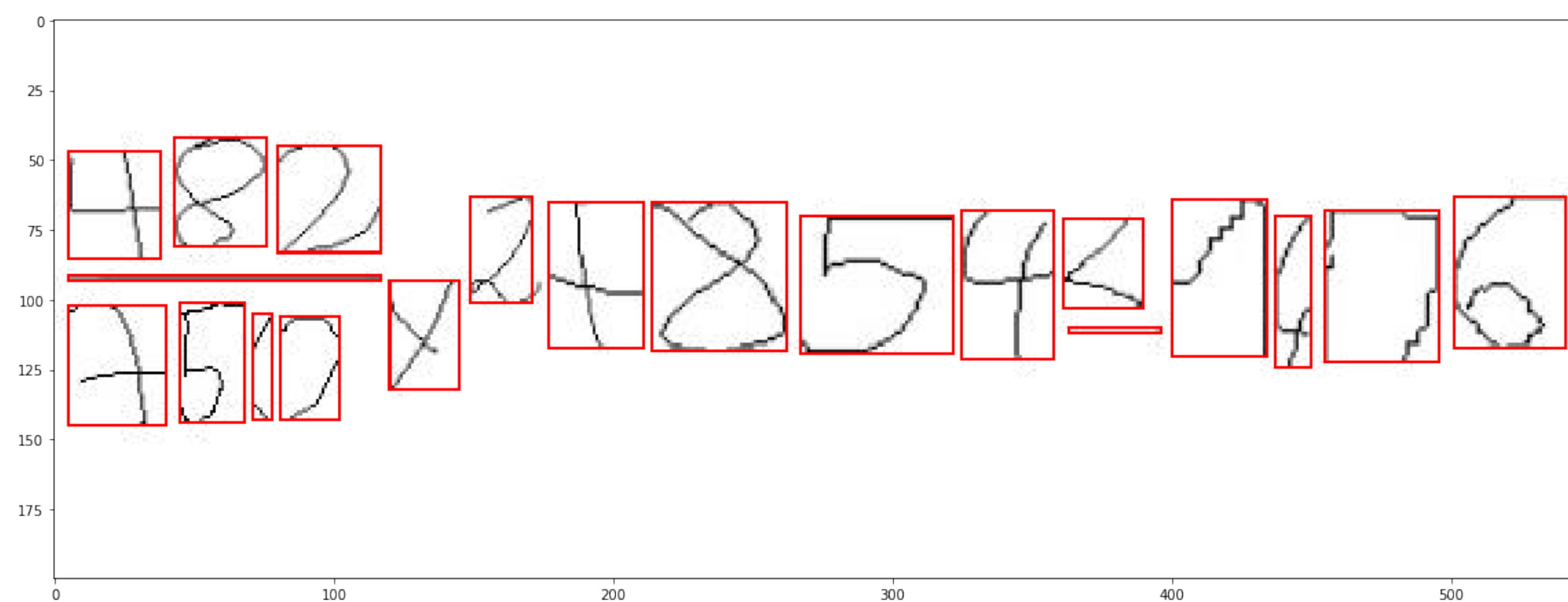


Figure 3: Find single bounding boxes using OpenCV

Use predictions and positions of bounding boxes (relative and absolute) to generate a sequence.

Training data:

- 30.000 images of simple one level formula as shown above
- 30.000 images of fractions in a simple formula

$$\frac{482}{750}y^2 + 854 \leq 1406$$

Results

- $\approx 85\%$ of the symbols in the sequences are correct
- only $\approx 50\%$ of test sequences are completely correct

Conclusion

Combination of three steps

- Detection:
Relatively good for black symbols on white background but symbols like \geq and $=$ get two bounding boxes.
- Classification:
Very good classification accuracy using an easy to train 3 Layer CNN.
- Generating a \LaTeX -sequence:
Slower to train and classification accuracy isn't as good.
Possible improvements:
 - Sequence model with attention
 - Beamsearch
 - More training data

References

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Accessed: 07/07/2018.
- [2] Jeffrey Ling Alexander M. Rush Yuntian Deng, Anssi Kanervisto.
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