

A Fair Evaluation of Various Deep Learning-based Document Image Binarization Approaches

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Basic Principle

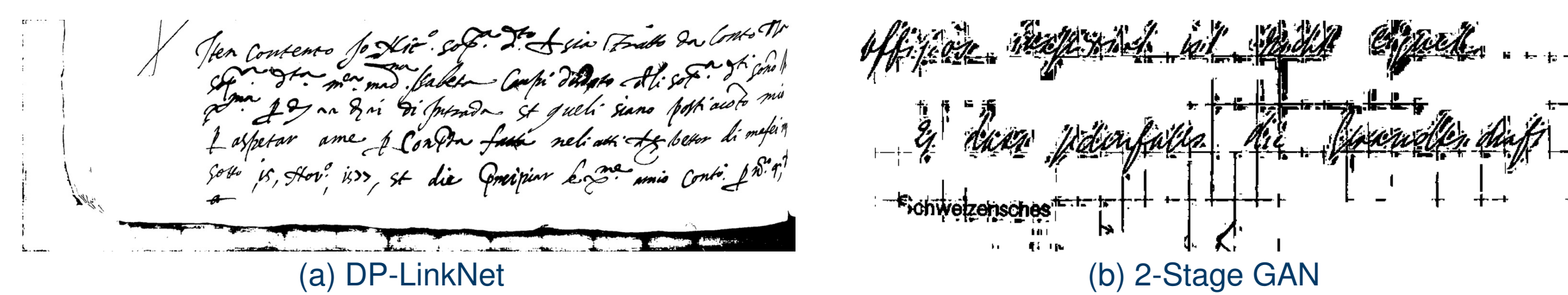
Motivation

- Binarization of document images is an important pre-processing step in the field of document analysis
- Learning-based approaches often incorporate extra data for training
- No fair comparison

Idea: Evaluation of different deep learning-based binarization methods using

- same evaluation protocol
- same training data

Failure Examples



Experimental Setup

Datasets

- Training: DIBCO2009, DIBCO2010, DIBCO2011, DIBCO2012, DIBCO2014 and DIBCO2016
- Evaluation: DIBCO2013, DIBCO2017, DIBCO2018 and DIBCO2019 [1]

Metrics

- F-measure: $FM = \frac{2 \times \text{Recall} \times \text{Precision}}{\text{Recall} + \text{Precision}}$
- Pseudo F-measure: $pFM = \frac{2 \times p\text{Recall} \times p\text{Precision}}{p\text{Recall} + p\text{Precision}}$
- Peak signal to noise ratio: $PSNR = \log_{10} \left(\frac{C^2}{MSE} \right)$
- Distance reciprocal distortion: $DRD = \frac{\sum_k DRD_k}{NUBN}$

Training Protocol

- All training datasets are used
- Augmentation: horizontal & vertical flipping and rotations
- Hyper-parameters optimized using "optuna" [2]
- Early stopping based on validation performance

Code and models available below:

https://github.com/RichSu95/Document_Binarization_Collection

Evaluation

Table 1: Average over (a) all metrics and (b) all ranks. Runtime: NVIDIA RTX 2060 GPU (12 GB RAM), note: DeepOtsu and 2-Stage GAN were limited by available memory.

| Model | PSNR \uparrow | FM \uparrow | pFM \uparrow | DRD \downarrow | img/sec \uparrow | Avg. rank \downarrow |
|---------------|-----------------|---------------|----------------|------------------|--------------------|------------------------|
| DE-GAN | 18.37 | 85.25 | 87.54 | 4.64 | 0.67 | 2.44 |
| Robin (U-Net) | 18.24 | 82.87 | 84.08 | 5.95 | 1.99 | 4.19 |
| DeepOtsu | 16.69 | 79.97 | 81.89 | 13.96 | 0.01 | 5.50 |
| 2-Stage GAN | 19.07 | 87.20 | 88.39 | 4.54 | 0.01 | 3.25 |
| DP-LinkNet | 19.10 | 83.70 | 84.36 | 5.79 | 0.49 | 3.38 |
| SAE | 16.15 | 79.14 | 80.77 | 9.41 | 0.68 | 6.63 |
| SauvolaNet | 18.67 | 84.91 | 86.83 | 4.86 | 0.37 | 2.63 |

(a) Average metrics

(b) Average ranks

- Performance dependent on different datasets
- Best average rank: DE-GAN [3] followed by SauvolaNet [4]
- Best metrics: 2-Stage GAN [5] followed by DE-GAN

Conclusion

- No clear winner
- Optimal binarization approach is dependent on the underlying data, amount, and quality of training samples

Qualitative Results

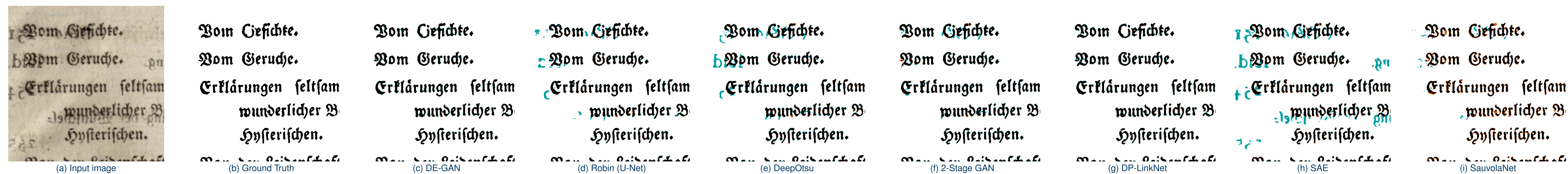


Figure 2: Illustration of some results for an image from DIBCO-2017. Pixels in cyan are false positives. The few pixels in orange are false negatives. Pixels in white or black match the ground truth.

References

- [1] I. Pratikakis, K. Zagoris, X. Karagiannis, L. Tsochatzidis, T. Mondal, and I. Marthot-Santaniello, "Icdar 2019 competition on document image binarization (dibco 2019)," in *2019 International Conference on Document Analysis and Recognition (ICDAR)*, pp. 1547–1556, 2019.
- [2] T. Akiba, S. Sano, T. Yanase, T. Ohta, and M. Koyama, "Optuna: A next-generation hyperparameter optimization framework," in *25th ACM SIGKDD International Conference on Knowledge Discovery & Data Mining, KDD '19*, p. 2623–2631, 2019.
- [3] M. A. Souibgui and Y. Kessentini, "De-gan: A conditional generative adversarial network for document enhancement," *IEEE Transactions on Pattern Analysis and Machine Intelligence*, vol. 44, pp. 1180–1191, March 2022.
- [4] D. Li, Y. Wu, and Y. Zhou, "Sauvolanet: Learning adaptive sauvola network for degraded document binarization," in *The 16th International Conference on Document Analysis and Recognition (ICDAR)*, p. 538–553, 2021.
- [5] S. Suh, J. Kim, P. Lukowicz, and Y. O. Lee, "Two-stage generative adversarial networks for document image binarization with color noise and background removal," *CoRR*, vol. abs/2010.10103, 2020.

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