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Objectives

- Digital transformation engineering ΟΤ documents
- Extraction of textual entities on large and unstructured documents
- Short text contextualisation and correct OCR predictions of Tags.

Related works

Post-OCR Correction

[ICADL2020] When to Use OCR Post-correction for Named Entity Recognition?, [ICDAR2019] A Cost Efficient Approach to Correct OCR Errors in Large Document Collections, [2019] Leveraging text repetitions and denoising autoencoders in OCR postcorrection, [ACL2017] OpenNMT: Open-source toolkit for neural machine translation

Text Detection

[CVPR2017] EAST: An Efficient and Accurate Scene Texte Detector, [ICPR2020] DUET: Detection Utilizing Enhancement for Text in Scanned or Captured Documents, [CVPR2019] Character Region Awareness for Text Detection, [ICDAR2021] Context Free TextSpotter for Real-Time and Mobile End-to-End Text **Detection and Recognition**

Text detection and post-OCR correction in Engineering Documents

Contribution

Detection :

- Based on the FCN model EAST with pre-trained Affinity Propagation Resnet V1.
- NMS part is removed by filtering own our method.
- Adaptation of the system to the use case

Post-OCR Correction:

- Identification
- Proposal characters

Experimentations

- Industrial data are sensitive and cannot be shared, so we have tested on our own dataset
- Calculation of the Precision, Recall & F1 Score for the detection module
- The Affinity Propagation clusters are the tags of the plan "shot by shot"
- Difference between the WER of the OCR output and the post-OCR Correction : 7% (75% OCR-82% post-

Conclusion & future works

- Focus on graphic symbols to help the post-OCR correction
- Association of symbolic and textual entities
- Improve initial step of text recognition by OCR competition

Clustering of Tags using of potential errors made by the OCR by analyzing the tag structure to correct misrecognized



P0	P1	OCR	Post-OCR
(15,200)	(36,212)	AA-2504-X <mark>x</mark>	AA-2504-XX
(135,200)	(148,212)	AA-25 <mark>Q</mark> 7-XX	AA-25 <mark>0</mark> 7-XX
(264,108)	(278,120)	AA-2513-XX	AA-2513-XX







recision	Recall	F1 Score
0.818	0.619	0.705
0.824	0.863	0.843

Evaluation of the detection system with our own dataset