Document semantic hashing for hybrid security

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Context: document security

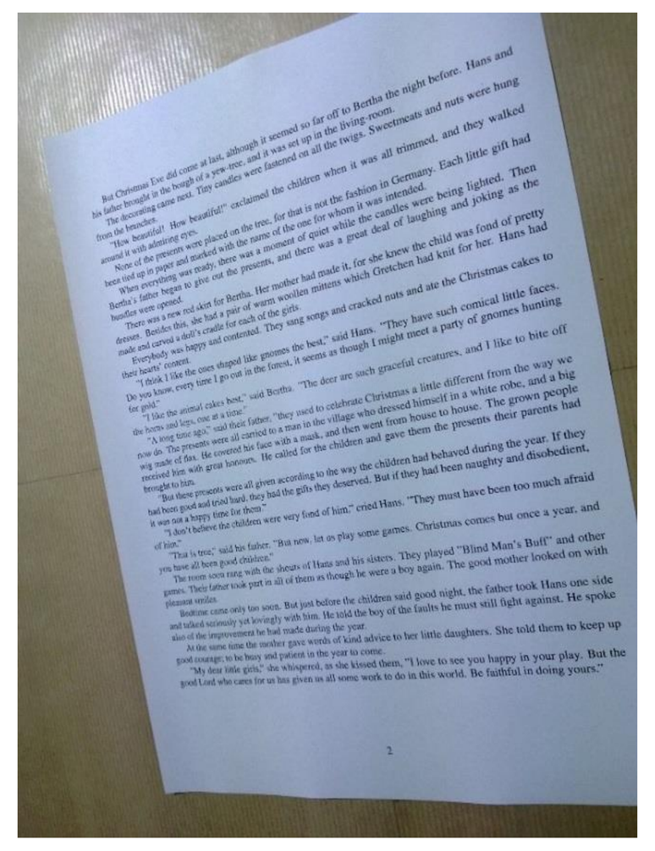
Paper document → Watermark, fingerprint → Secure paper document

Digital document → Cryptographic hash → Secure digital document

Hybrid document → Semantic hash → Secure hybrid document

Hash computation process

Document image



Geometric correction

Segmentation

Table analysis | Image analysis | Text OCR processing

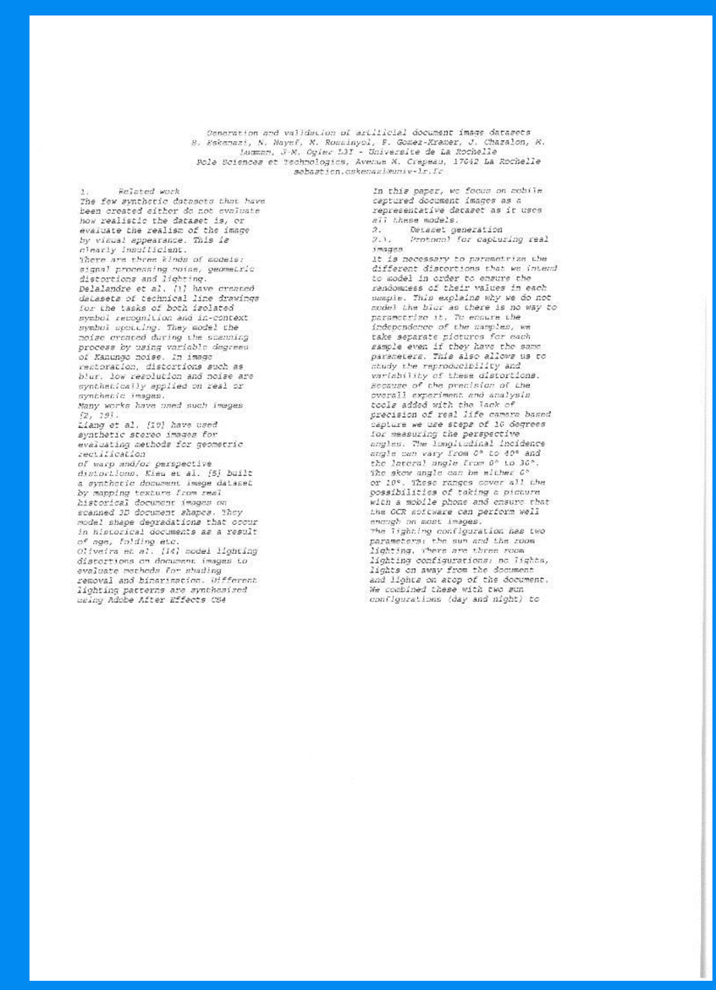
Document robust reconstruction

Document hash computing

Document signature

954f7d96502b5c5fe2e98a5045bca7f5

What is a semantic hash ?



Cryptographic hash (such as SHA-256)

00110101001 → 954f7d965

Fuzzy hash (such as ssdeep)

0011 0101 001 → 954f7d 96502b 5c5fe2

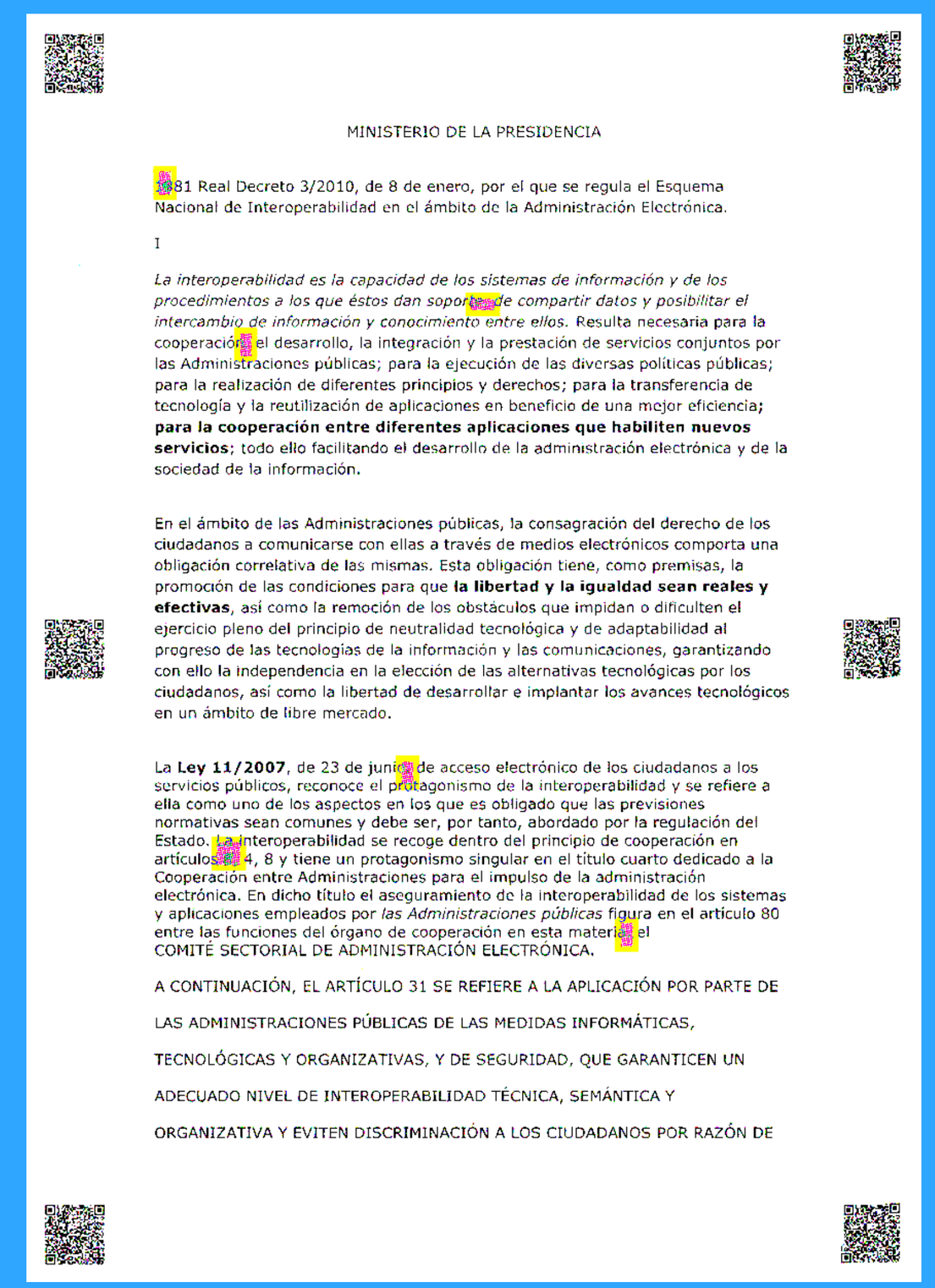
Semantic hash

Lorem ipsum... → 98a5045bca7f5

State of the art and remaining challenges

SIGNED project [1]

- Cut the document image in squares of 64 by 64 pixels
- Apply a Haar Discrete Wavelet Transform on each square
- Create a fuzzy hash
- Other undisclosed pre- and post-processing steps.



Document secured by the SIGNED project, it takes 6 2D-barcodes to embed the signature

Strengths	Weaknesses
Probability of false alarm <0,001	Cannot detect the replacement of dots by commas with an error <0,001
Probability of missed detection <0.001 for the replacement of digits	Cannot detect a manipulation smaller than 64x64 pixels at 600dpi (42x42 required)
Collision probability <0,001	Throughput >5s per page
Compatible with current scanners and printers	Hash size >4kB (between 4,8 and 170 kB)

References:

[1]A. Malvido Garcia: Secure Imprint Generated for Paper Documents (SIGNED). Technical Report December 2010, Bit Oceans (2013)

[2] S. Eskenazi, P. Gomez-Krämer, J.-M. Ogier: When document security brings new challenges to document analysis. International Workshop on Computational Forensics (IWCF), (2014)

For further information

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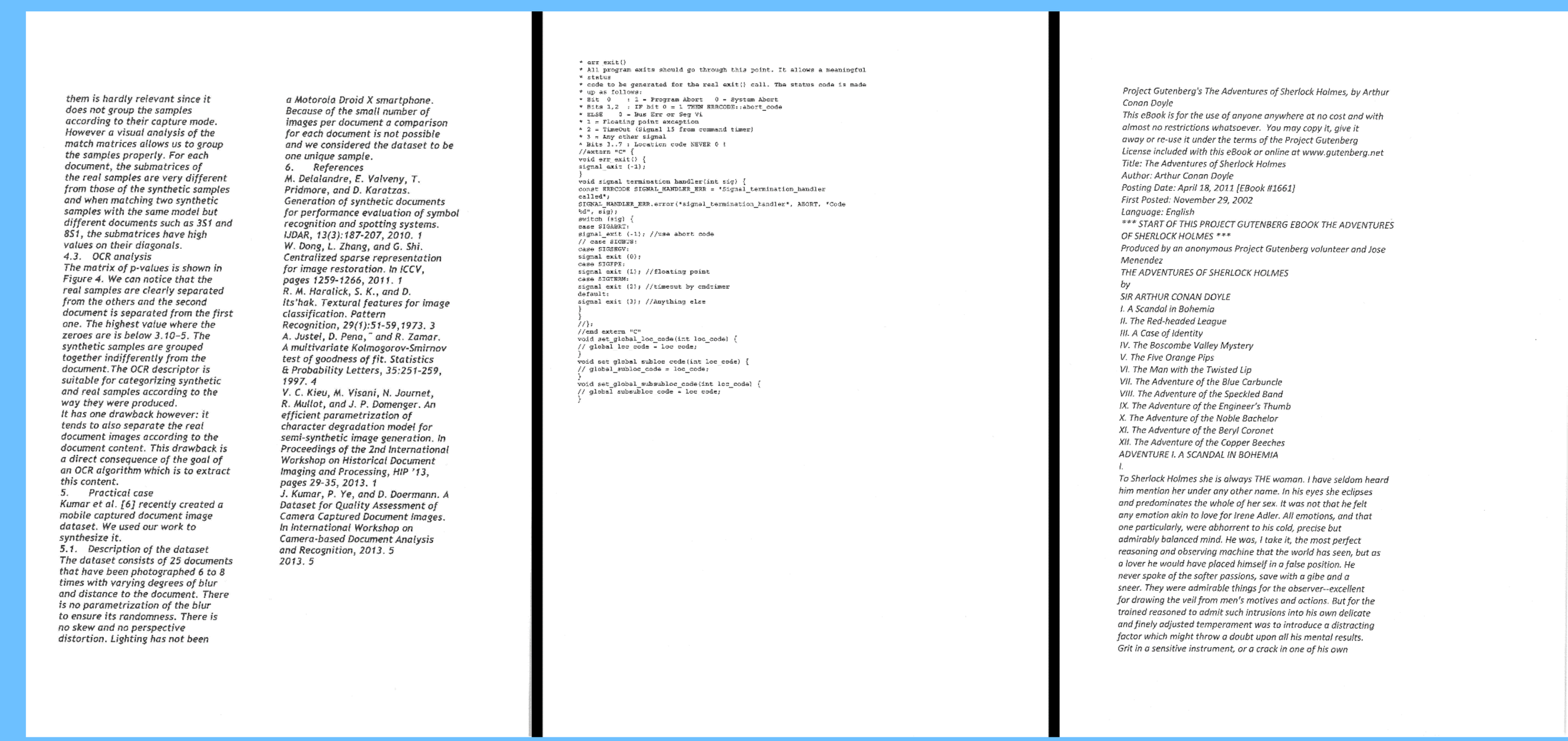
Conclusion: We need to study and improve the robustness of document analysis algorithms.

Text OCR processing [2]

Test of Tesseract on 28512 document images



Dataset:

- 22 texts
- 6 fonts
- 3 font sizes
- 4 font emphasis
- 3 printers
- 3 scanners
- 3 scanning resolutions

Test protocol:

- Run Tesseract on the document image
- Post processing : alphabet reduction
- Compute the SHA-256 hash of the document

Alphabet reduction	
Character	**Replacement**
Empty line	Removed
Tabulation and space	Removed
— (long hyphen)	- (short hyphen)
’, ‘ (left and right apostrophes)	' (centered apostrophe)
“,”,'' (left and right quotes, double apostrophe)	" (centered quote)
I, l, 1 (capital i, 12th letter of the alphabet, number 1)	| (vertical bar)
O (capital o)	0 (zero)
ﬁ (ligature)	fi (two letters f and i)
ﬂ (ligature)	fl (two letters f and l)

Best case scenario results:

- Accuracy : 99,83%
- Probability of false positive: 53% !!!
- Collision probability : 0,2%
- Other criteria are OK (throughput, digest size…)

False positives:

- Occurs when two identical documents have different digests
- Related to the robustness of the OCR algorithm
- Computed as a random draw of two copies of the same document:

$$P_i = \sum_{j=1}^{n_i} \left(\frac{n_{ij}}{\sum_{k=1}^{n_i} n_{ij}} \times \frac{n_{ij}-1}{\sum_{k=1}^{n_i} n_{ij}-1} \right)$$

- n documents
- n_i digests per document
- Each digest is present n_{ij} times