Historical document image analysis: a structural approach based on texture



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Outline

- Introduction
- Proposed approach : a structural signature based on texture
 - Digitized historical book page characterization
 - Digitized historical book page categorization
- Experiments
- Evaluation and Results
- Discussion
- Future Work

Introduction – Context

- Rapid growth of digital libraries worldwide
 - access to large sets of cultural heritage documents
 - indexing tool of digitized resources
 - information retrieval in digital libraries

• Historical document image analysis

- reliable historical document image (HDI) interpretation system
- efficient and fast computer-aided indexing tool of HDIs

• Objectives

- segment and characterize a HDI as easily, quickly and automatically as possible
- find homogeneous regions (graphic and different font text regions)
- to characterize the HDI content and layout
- propose a signature
- to categorize digitized historical book (DHB) pages according to several criteria (layout structure, graphical or typographical characteristics)

Introduction – DIGIDOC (document image digitization with interactive description capability)

- design "smart" digitizers
- limit manual intervention
- perform easy and high quality digitization of HDIs
- assist the digitization operator to adjust the best set of parameters (e.g. resolution, lightening, color calibration),
- detect errors in the digitization process (e.g. blur, skewed or folded pages),
- provide appropriate assistance for document indexing (e.g. recognize page types or breaks in a sequence of pages), etc [LeBourgeois04]

[[]LeBourgeois04] LeBourgeois, F., Trinh, E., Allier, B., and Emptoz, H., "Document images analysis solutions for digital libraries", DIAL, 2004.

Introduction – Goal

 identify similar groups of pixels sharing similar visual properties to characterize and compare the layout and content of HDIs by means of a graph-based signature



Overview of the context and an example of signature-based applications

Introduction – Historical document images



Some particularities of HDIs: superimposition of information layers (e.g. stamps, handwritten notes, noise, back-to-front interference) and page skew [Gallica]

[Gallica] French digital library Gallica: http://gallica.bnf.fr

Introduction – Historical document images

- Properties
 - large variability of page layout, complex layouts, random alignment
 - use of multiple fonts and illustration styles
 - large variability of editorial style and logical structure
- Life cycle
 - noise and degradation caused by copying, scanning, or aging
 - superimposition of information layers (e.g. stamps, handwritten notes at the margins, noise, back-to-front interference)
- Digitization
 - page skew
 - capture defects (e.g. curvature, light), black borders, etc. [Coustaty11]

[Coustaty11] Coustaty, M., Raveaux, R., and Ogier, J. M., "Historical document analysis: A review of French projects and open issues", EURASIP, 2011.

- a signature for DHB page characterization and categorization
- based on the use of texture and graph algorithms
- varying low-level features characterizing the HDI content (i.e. different text fonts, or graphic regions)
- structural information describing the HDI layout
- provide a rich and holistic description of the layout and content of the DHB pages
- applicable to a large variety of ancient books
- without hypothesis on the document layout and content



Overview of the different steps of this work

- **1. DHB page characterization**
- 2. DHB page categorization
 - a. Unsupervised DHB page classification
 - b. DHB page stream segmentation

DHB page characterization



Detailed schematic block diagram of the proposed approach used to generate the graph-based signature for DHB page characterization

DHB page characterization

- Pre-processing and Gabor feature extraction (Step 1),
- Estimation of the number of DHB content types (Step 2),
- Pixel-clustering and labeling (Step 3),
- Pixel-labeling refinement (Step 4), [Mehri13, Mehri14, Mehri15]
- Post-processing (Step 5),
- Extraction of representative homogeneous regions (Step 6),
- Generation of a structural signature per page (Step 7).

[Mehri13] Mehri, M., Héroux, P., Gomez-Krämer, P., Boucher, A., and Mullot, R., "A pixel labeling approach for historical digitized books", ICDAR, 2013.

[Mehri14] Mehri, M., Mhiri, M., Héroux, P., Gomez-Krämer, P., Majoub, M. A., and Mullot, R., "Performance evaluation and benchmarking of six texture-based feature sets for segmenting historical documents", ICPR, 2014.

[Mehri15] Mehri, M., Gomez-Krämer, P., Héroux, P., Boucher, A., and Mullot, R., "A texture-based pixel labeling approach for historical books", PAA, 2015.

Step 1

Pre-processing and Gabor feature extraction

Multi-channel Gabor Filtering

- magnitude response of the output of Gabor filters
- space of Gabor filter is set as: $\sigma = \sigma_x = \sigma_y$
- 6 spatial frequencies
 - {2√2, 4√2, 8√2, 16√2, 32√2 and 64√2}
- 4 orientations
 - $\{0, \pi/4, \pi/2 \text{ and } 3\pi/4\}$



24 Gabor filters

Gabor feature extraction & Multi-scale Analysis



Figure (a) show the original image with a selected pixel position Figure (b) zoom in selected region (red dot) Figures (c), (d), (e) and (f) illustrate (16 x 16), (32 x 32), (64 x 64) and (128 x 128) windows

192-D feature vector:

(6 spatial frequencies) X (4 different orientations) X (2 Gabor indices) X (4 sliding windows)

Step 2

Estimation of the number of DHB content types

Consensus Clustering



Step 3

Pixel-clustering and labeling

Hierarchical Agglomerative Clustering (HAC)& Nearest Neighbor Search (NNS) algorithms



Step 4

Pixel-labeling refinement

Multi-Scale Majority Voting Technique



Step 5

Post-processing

Connected component (CC) & Majority Voting (MV) analysis



CC & Color Layer Separation (CLS) analysis



Labeled CCs

Blue CCs (graphics)

Green CCs (text)

Adaptive run-length smearing algorithm (ARLSA) Blue CCs (graphics)



Adaptive run-length smearing algorithm (ARLSA) Green CCs (text)



Adaptive run-length smearing algorithm (ARLSA)

Merge with OR



Step 6

Extraction of representative homogeneous regions

Extraction of Homogeneous regions



Resulting image of ARLSA => Extracted CCs

+MV=> Resulting image of MV =>Labeled CCs Output image =>Extracted and labeled Homogeneous regions

Extraction of Representative Homogeneous regions



Resulting image of ARLSA => Extracted CCs

+MV=> Resulting image of MV =>Labeled CCs

Extracted and labeled representative homogeneous regions

Step 7

Generation of a structural signature per page

Graph-Based Signature – Approach

- vertex attributes
 - 192 Gabor attributes
 - 46 shape, geometric, and topological attributes
 - ✓ centroid position,
 - ✓ number of pixels,
 - ✓ gray-level average,
 - ✓ contour area and perimeter,
 - ✓ Hu, spatial, central, and central normalized moments, etc.
- edge attributes
 - absolute differences between the two extracted region centroids in the x- and y-axis
 - edge force

Graph-Based Signature – Edge Force

- emphasize on the most representative, largest and spatially closest regions
- deduced from Newton's law of universal gravitation
- proportional to the number of pixels of the destination vertex of the graph
- inversely proportional to the square of the Euclidean distance between the two graph vertices

$$F_e^{s,d} = \frac{N_{G_v^d}}{(ED_{G_v^{s,d}})^2} \qquad F_e^{s,d} \ge Th_e$$

 $N_{G_v^d}$: number of pixels of the destination vertex (G_v^d) of the built directed graph, $ED_{G_v^{s,d}}$: Euclidean distance between the two graph vertices: source (G_v^s) and destination (G_v^d) , Th_e : edge force and threshold.

Graph-Based Signature – Results (1/2)



Graph-Based Signature – Results (2/2)



Input image

Labeled CCs

Graph-based signature

- 1. DHB page characterization
- 2. DHB page categorization
 - a. Unsupervised DHB page classification
 - b. DHB page stream segmentation

DHB page categorization – GED

- measure the (dis)similarity between the obtained graphbased DHB page signatures
- computation of the minimum-cost sequence (e.g. insertion, deletion and substitution of vertices or edges) [Bunke07]
- This work
 - an optimized binary linear programming [GEM++]
 - a statistical analysis of the feature variations to determine weight of each feature
 - computational complexity of the GED is reduced (i.e. up to 11 vertices in the obtained graphs)
 - compute a distance matrix, whose elements represent the dissimilarity between the compared graphs

[Bunke07] Bunke, H. and Riesen, K. "Towards the unification of structural and statistical pattern recognition", PRL, 2012. [GEM++] http://litis-ilpiso.univ-rouen.fr/ILPIso/gem++.html

Unsupervised DHB page classification

 Unsupervised DHB page classification to group or gather similar layout and/or content DHB pages



DHB page stream segmentation

 DHB page stream segmentation to generate automatically a table of content/summary of the analyzed DHB



Experiments – Corpus

 a printed monograph which is dated 1596, titled "II mondo nuovo, del sig. Giov. Giorgini da Jesi" and written in Italian, which is composed of 322 ground-truthed onepage color HDIs [Gallica]



[Gallica] http://gallica.bnf.fr/ark:/12148/bpt6k132294p/f5.planchecontact.r=.langFR

Experiments – Ground truth

- Ground-truthing Environment for Document Images [GEDI]
- By assigning predefined content types to rectangular regions in a document image, GEDI generates an XML schema representing:
 - location on the page
 - height
 - width
 - label



[GEDI]: http://gedigroundtruth.sourceforge.net/

Evaluation – Accuracy metrics

- Evaluation of the different steps of the proposed algorithm
 - > Per-pixel classification accuracy (CA)[Baird07]
- Evaluation of the homogeneous region extraction step
 - Jaccard (J_{AR}): overlap ratio between the number of foreground pixels defined in the two areas: B_{gt} and B_r [Brunessaux14]

where

 $\mathbf{B}_{\mathbf{gt}}$: bounding box of the ground-truth ,

B_r: bounding box of the result block.

[Baird07] Baird, H. S., Moll, M. A., An, C., and Casey, M. R., "Document image content inventories", DRR, 2007. [Brunessaux14] Brunessaux, S., Giroux, P., Grilheres, B., Manta, M., Bodin, M., Choukri, K., Galibert, O., and Kahn, J. "The Maurdor project: Improving automatic processing of digital documents", DAS, 2014.

Evaluation – Results (1/5)

• DHB page characterization



Evaluation – Results (2/5)

- DHB page characterization
- Step 3: Pixel-clustering and labeling
- Step 4: Pixel-labeling refinement
- Step 5: Post-processing
- Step 6: Extraction of representative homogeneous regions

		Step 3	Step 4	$Step \ 5$
CA	μ	0.977	0.983	0.987
	σ	0.066	0.085	0.076

		Step 6
Inp	μ	0.952
JAR	σ	0.174

Unsupervised page classification

Evaluation – Results (3/5)

• DHB page categorization

WP3-DIGIDOC	o x
Load Clear Summary	
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Unsupervised page classification

Evaluation – Results (4/5)

• DHB page categorization

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	0000241	0000242	0002243	00002441	0000245	0000246		0000248	0000249	0002250	accessi	0000252	0000253	0000254	
	00002266	10000257	000238	0000259			0000262	000263	0000254	0001285	0000265	0000267	630,000	0000269	E
20270 A			8000273	0000274	0000275	0090275	0000377	0000278	0090279	0000280	9000281	0000 282		5000.284	

Page stream segmentation

Evaluation – Results (5/5)

• DHB page categorization



Page stream segmentation

Discussion

- good discrimination of foreground layers
 - particularly between text and graphics
- simple GUI tool for characterization and categorization of DHB pages
- numerous signature-based applications for managing effectively a corpus or collections of books
- > 90% classification accuracy is noted
- table of contents of the DHB by detecting different or dissimilar pages

Future Work (1/2)

- use the proposed signature on a larger databases with more variable content
- Find pages in a DHB or HDI corpus (contain a particular content component or a group of patterns)
- retrieve similar pages in a HDI corpus query tool
- > detect the scanning failure
- investigate a finer book page classification

Future Work (2/2)

- > analyze the impact of different feature weighting schemes in the cost of the GED operations
- > assess other state of the art graph dissimilarity techniques
- improve the designed GUI tool for characterization and categorization of DHB pages



Thanks

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Introduction – State of the art

- Contemporary document image analysis
 - textual content (e.g. OCR) [Google, "smartFIX", "A2IA Document Reader"]
 - interest point detection (e.g. BoW, BoVW) [Bouguelia13, Augereau14]
- State-of-the-art
 - pixel-based algorithms to extract regions [Baird07,An07,An10]
 - texture features to analyze pages with overlapping layers [Okun99,Kise14]
 - poor performance of textual content based approaches due to many particularities of HDIs [Grana14]

[Bouguelia13] Bouguelia, M. R., Belaid, Y., and Belaid A., "A stream-based semi-supervised active learning approach for document classification", ICDAR, 2013.

[Augereau14] Augereau, O., Journet, N., Vialard A., and Domenger J. P., "Improving classification of an industrial document image database by combining visual and textual features", DAS, 2014.

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[An07] An, C., Baird, H. S., and Xiu, P., "Iterated document content classification", ICDAR, 2007.

[An10] An, C., Yin, D., and Baird, H. S., "Document segmentation using pixel-accurate ground truth", ICPR, 2010.

[Okun99] Okun, O. and Pietikäinen, M., "A survey of texture-based methods for document layout analysis", TAMV, 1999.

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[Grana14] Grana, C., Serra, G., Manfredi, M., Coppi, D., and Cucchiara, R., "Layout analysis and content enrichment of digitized books", MTA, 2014.