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CRF-Based Authors' Name Tagging for Scanned Documents

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Agenda

- Background
 - Motivation of our research
- Two-tier authors' names extraction
 - Authors' block extraction
 - **CRF-based author/delimiter tagging**
- Experiments
- Conclusions and future works

Motivation

- Digitizing process of printed docs for DL
 - Scan => Analyze => Recognize => Store
 - Construction of bibl. DB is labor-intensive
- Automatic extraction of bibl. data from scanned academic articles
 - Cost-effective
 - Need to be error-tolerant to OCR errors
- Why extract “authors’ names”?
 - Because among **the most critical** bibl. elements

Authors' names extraction

■ Two-tier authors' names extraction

□ Authors' block extraction

- Extract a block representing authors from a title page

□ CRF-based author/delimiter tagging

- Label every character as either author or delimiter

■ Authors' block example

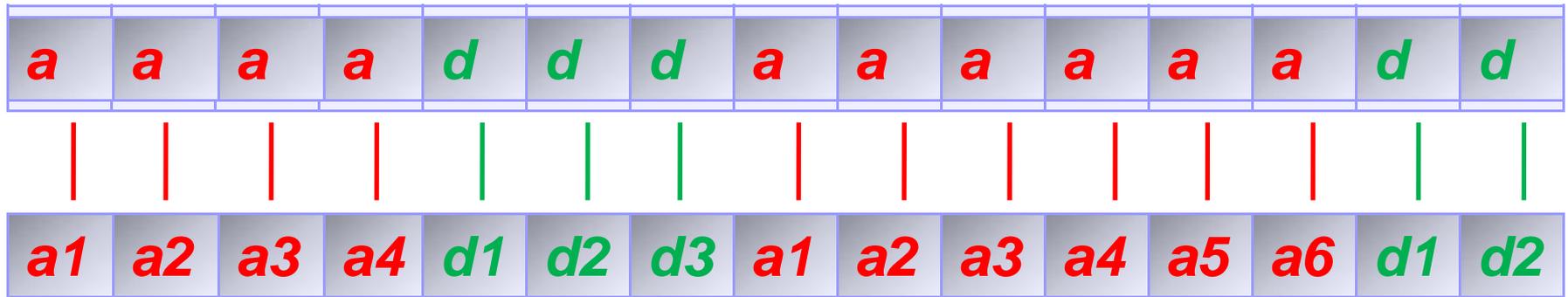
■ Our OCR system

□ English and Japanese OCR engines

□ Layout analysis + character recognition

□ Output bounding rectangles for chars, lines, blocks

Author/delimiter tagging



■ Tag sets

- 2-tag set: mere *author* or *delimiter*
- 2+pos-tag set: 2-tag set with **character positions**
 - The max positions of *a* and *d* determined by training

Conditional Random Fields (CRF)

$\mathbf{y} = t_1 t_2 \cdots t_n$: a tag sequence

$\mathbf{x} = c_1 c_2 \cdots c_n$: an input char sequence

□ Our formulation

Normalization factor

$$P(\mathbf{y} | \mathbf{x}) = \frac{1}{Z_{\mathbf{x}}} \exp\left(\sum_{i=1}^n \sum_k \lambda_k f_k(t_{i-1}, t_i, \mathbf{x})\right)$$

$$\hat{\mathbf{y}} = \arg \max_{\mathbf{y} \in Y(\mathbf{x})} P(\mathbf{y} | \mathbf{x})$$

Feature function

□ Feature functions

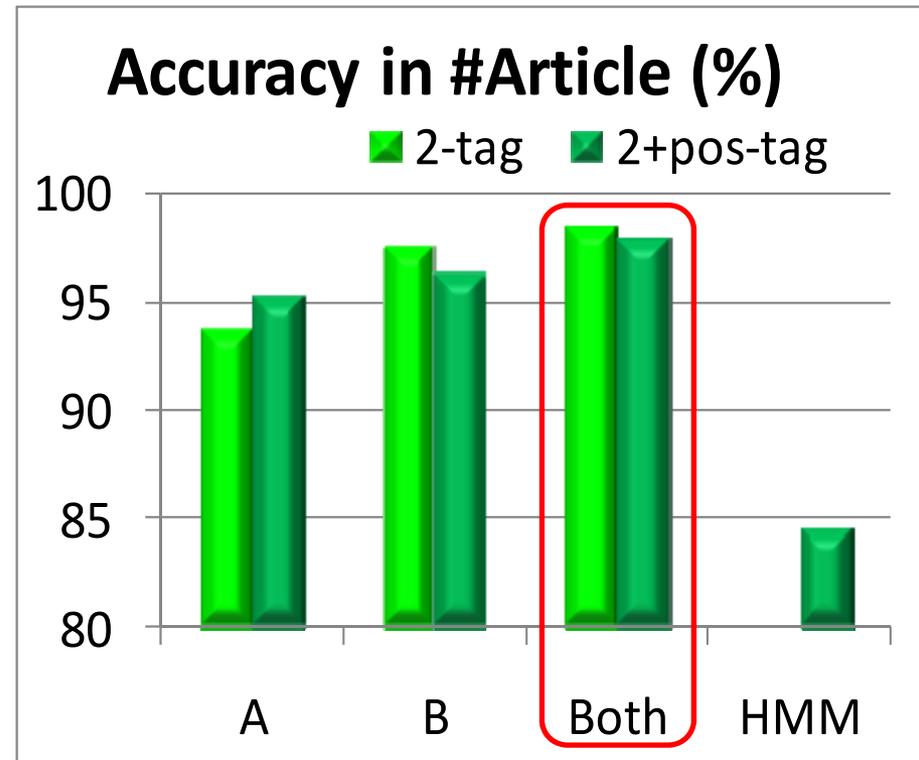
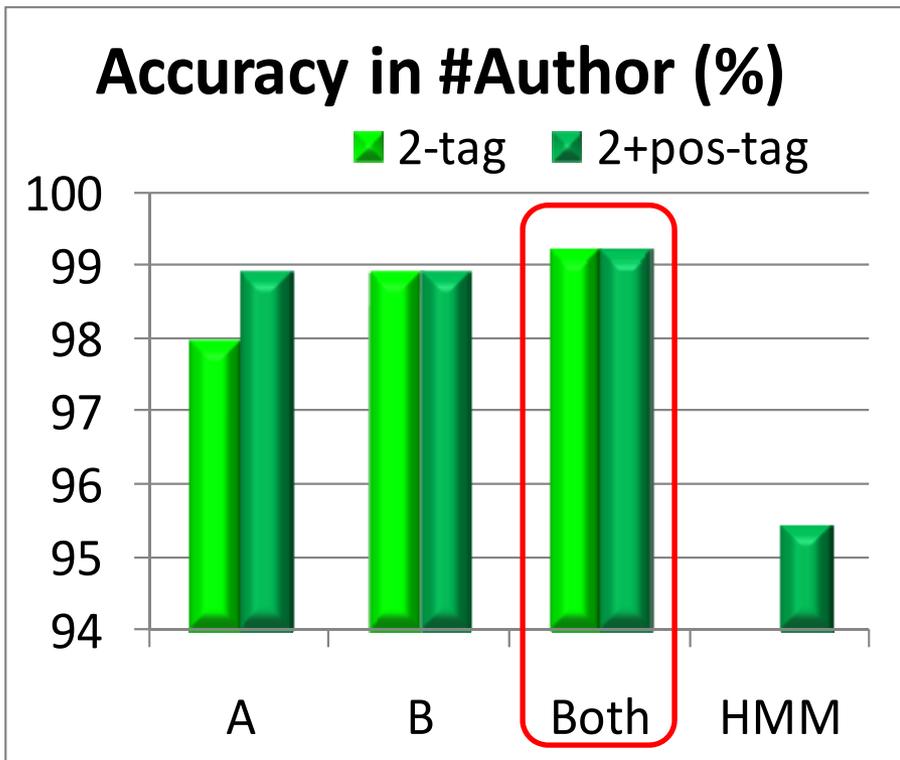
■ Features: chars (textual info.) & widths (layout info.)

■ E.g.
$$f_k(t_{i-1}, t_i, \mathbf{x}) = \begin{cases} 1 & \text{if } c_i = 'f', t_i = d \\ 0 & \text{otherwise} \end{cases}$$

Experiments

- Data : OCR-processed academic articles
 - 54 issues of TIPSJ in 2003(vol.44), 2004(vol.45)
 - Training: vol.44, Test: half of vol.45
- OCR accuracy
 - 99.00% for abstract, 97.01% for references
- Implementation: CRF++ 0.50
- Selected features for our CRF
 - $\langle c(0) \rangle$: character unigram
 - $\langle w(0) \rangle$: character's width unigram
 - $\langle t(-1), t(0) \rangle$: tag bigram

Tagging accuracy (test data)



A: $\langle c(0) \rangle + \langle t(-1), t(0) \rangle$

B: $\langle w(0) \rangle + \langle t(-1), t(0) \rangle$

Both: $\langle c(0) \rangle + \langle w(0) \rangle + \langle t(-1), t(0) \rangle$

HMM: our HMM-based tagger

Discussion

- The setting “Both”
 - Achieved **99.22% accuracy** (in #Author)
 - **Outperformed A, B, and our HMM-based tagger**
- 2-tag vs. 2+pos-tag sets
 - Almost no difference in this experiment
- Tagging errors
 - Caused by OCR errors & noises of documents
 - Often occur at the boundary between name and delimiter strings

Conclusions

- Proposed a CRF-based authors' name tagger
 - Applied it after extracting authors' (text) blocks
 - More than 99% tagging accuracy
 - It outperformed our HMM-based one
- Future works
 - Accuracy improvement with other features
 - Title page analysis system for automatic extraction
 - Extracting other bibl. data such as title, abstract, ...

Questions and Comments?

