

Search is not only the Web IR Applications

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Objectives

- Main objective of IR
- · Different tasks in IR
 - Printed documents search
 - Patent search
 - Social search
- This Lecture:
 High level Simplified Compressed



Information Retrieval Objective

- IR is finding material of an unstructured nature that <u>satisfies</u> an <u>information need</u> from within large collections.
- Information need
 - Expected search scenario?
 - Modeling the task?
- Data nature
 - Approach?
 - Scalable? Fast?
- User Satisfaction
 - More relevant documents?
 - Effective evaluation?



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Printed Documents Retrieval



Printed Documents Retrieval

- Documents: text on printed papers (books)
- Information need:
 Information within these books
- Challenge: It is an image of text
- Common Approach:
 OCR → Recognized text ← Search
- Challenges in Common Approach:
 OCR → Text with mistakes (WER_{Ar} ≈ 40%)
 OCR → Not available for all languages

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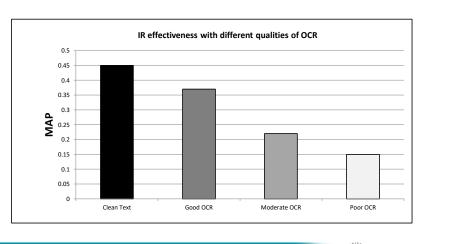
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Problem

• Text with errors (sometime many errors)





n-gram Char Representation of OCR

• Original:

example sentence

OCR output:

example senicnce

• 3-gram char representation:

\$ex exa xar arn rnp npl ple le\$ \$se sen enl nlc lcn cnc nce ce\$

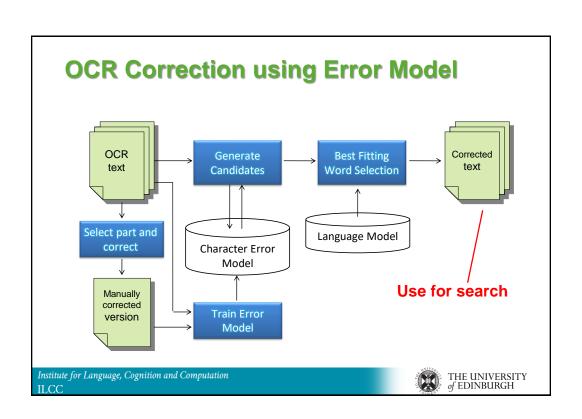
• Query:

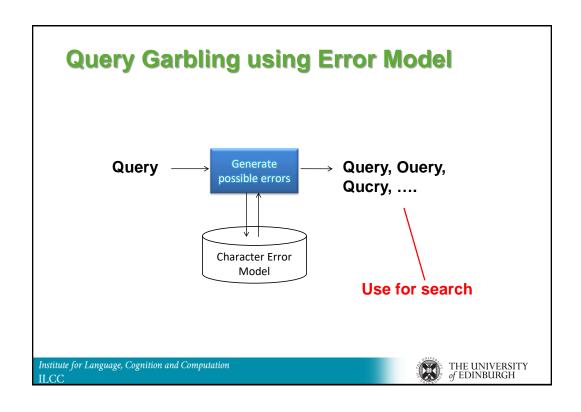
example sentence \rightarrow \$ex exa xam amp mpl ple le\$ \$se sen ent nte ten enc nce ce\$

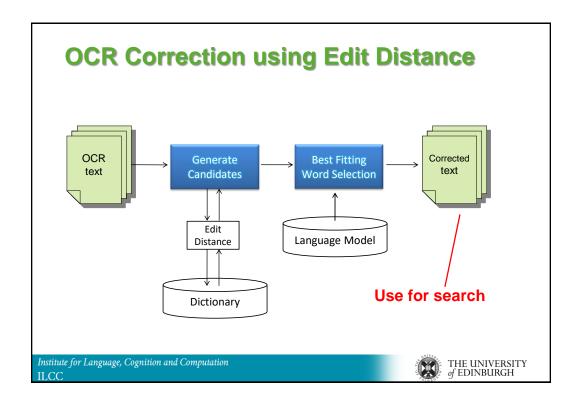
• Matching:

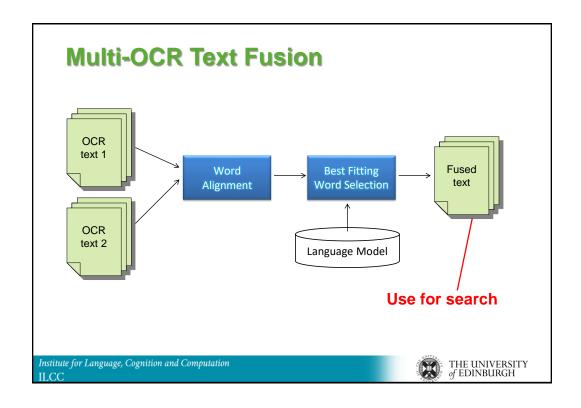
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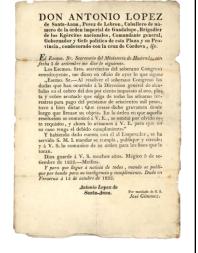
OCR Search

- Recognition errors in OCR text degrades retrieval
- Different methods of text processing can overcome the negative effect on retrieval and improves search
- n-gram character representation improves retrieval, but not that much
- Some training and resources are needed which can be manual correction, trained language model, or both
- Previous methods fail when errors are large (WER>50%)

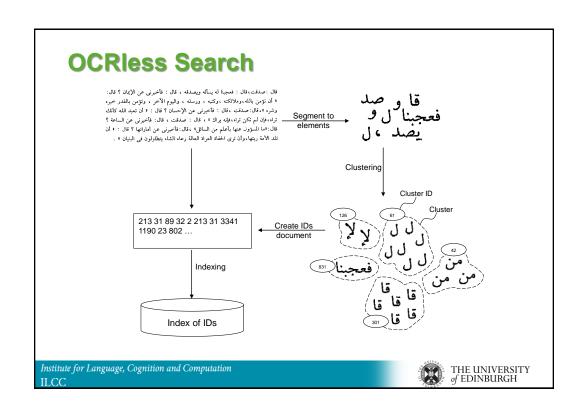


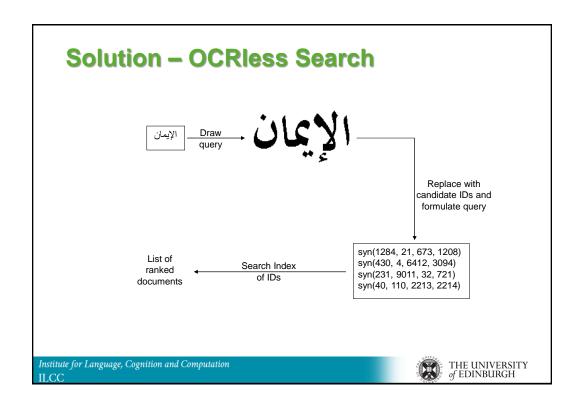
Solution - back to Information Need

- Information need: the printed papers
- Question: Why convert image to text?
- Related work: Word Spotting









Solution - OCRIess Search

- Effective and fast
- Robust to OCR errors (v1de0)
- No training resources required
- Language independent



English 鮭貓



Microsoft TechFest Demo

The same engine for searching printed documents in: Arabic, English, Chinese, Hebrew, and Hieroglyphic



Printed Documents Retrieval

- Text-based solutions: correction
- Image-based: clustering

 Current State-of-the-art: CAPTCHA



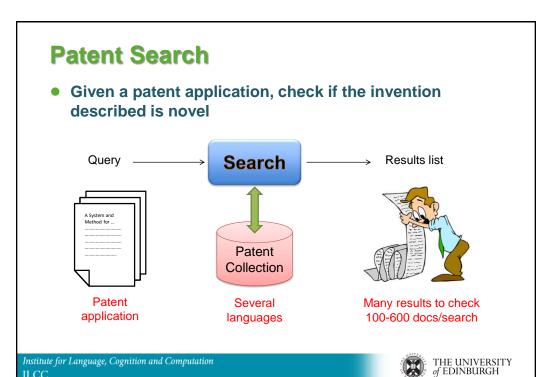
Information need → Approach

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Patent Search





Patent Search - User Satisfaction

NTCIR, CLEF, TREC

ILCC

- Recall-oriented → Try not to miss a relevant document
 - Recall is the objective
- Precision is also important
- Huge # documents checked (100-600 documents)
- Evaluation: average precision (AP)!!
 - Focuses on finding relevant docs early in ranked list
 - Less focus on recall



Example

For a topic with 4 relevant docs and 1st 100 docs to be examined:

System1: relevant ranks = {1}

System2: relevant ranks = {50, 51, 53, 54}

System3: relevant ranks = $\{1, 2, 3, 4\}$

 $AP_{system1} = 0.25$

 $R_{\text{system1}} = 0.25$

 $AP_{system2} = 0.0481$

 $R_{\text{system2}} = 1$

 $AP_{system3} = 1$

 $R_{\text{system3}} = 1$

 We need a metric that reflects <u>recall</u> and <u>ranking quality</u> in one measure

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PRES: Patent Retrieval Evaluation Score

$$PRES = 1 - \frac{\sum_{i} r_{i}}{N_{\text{max}}} - \frac{n+1}{2}$$

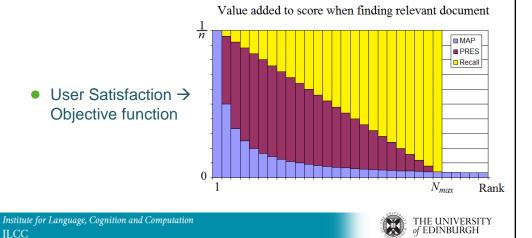
n: number of relevant docs r_i : rank of the i^{th} relevant document N_{max} : max number of checked docs

- Derived from R_{norm} (Rocchio, 1964)
- Gives higher score for systems achieving higher recall and better average relative ranking
- Dependent on user's potential/effort (N_{max})
- Robust to incomplete relevance judgements



PRES: as a cumulative gain

- Official score in CLEF-IP since 2010
- Adapted in many Recall-oriented IR tasks



Patent Search - CLIR

- Query: Full patent application
- Common approach: MT (the best)
- Challenge: training recourses + speed!
- Ideal: Query + Document translation



Patent Search - CLIR - Objective?

Manual translation

It is a great idea to apply stemming in information retrieval

MT output

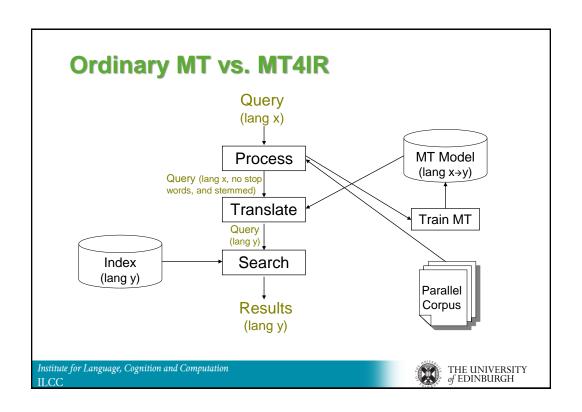
he are an great ideas to applied stem by information retrieving

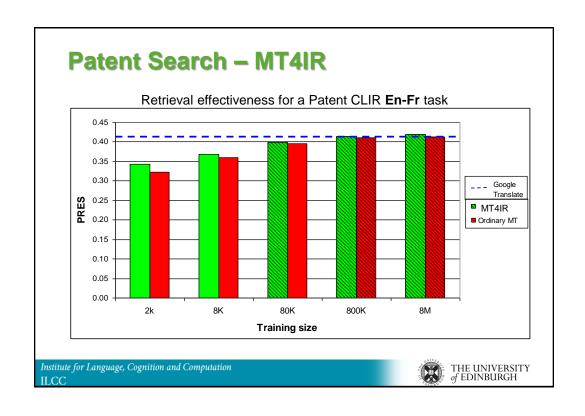
MT evaluation: MT sucks

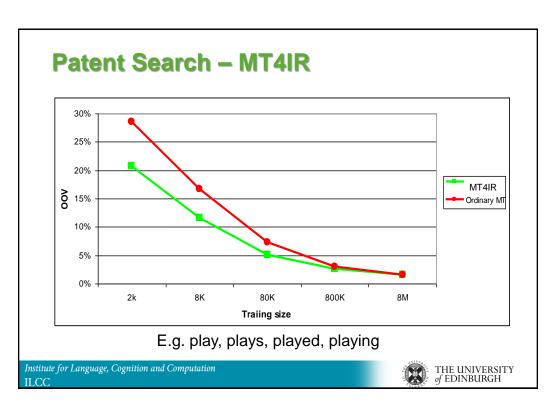
IR evaluation: MT rocks ©

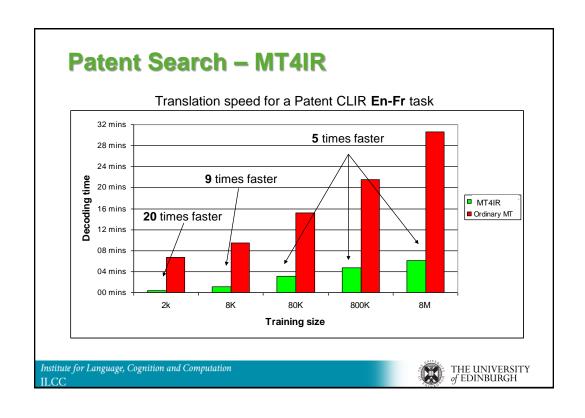
 MT4IR: An efficient MT that neglects morphological and syntactic features of output













Social Search

TREC Microblog track → Ad-hoc, filtering

- User's information need?
- Search scenario? Task?
- Boolean? News updates?



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Social Media & News

News websites are biased





Telegraph

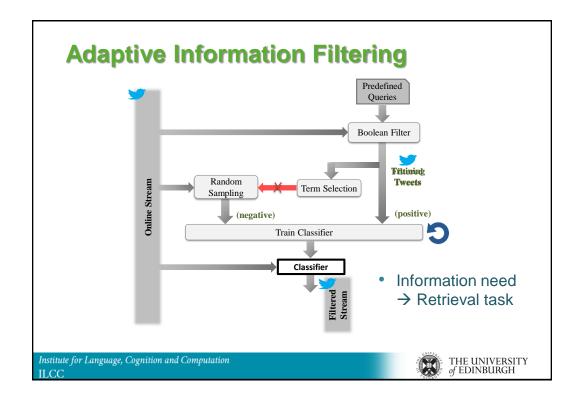
People use social media to

- Report news
- Comment on news
- Discuss different views on events



- Discussions on social media, reflects public interest
- Can social media answer the question: "What is happening in <region>?"





Summary

- The objective is IR is "User Satisfaction"
- Understand the user needs well
- Design the IR task carefully
- You do not have to stick to the path in the literature
- Are you sure performance is measured correctly?
- Beating the baseline is always desirable, just be sure you are moving in the right direction



Readings

- Magdy W. and T. Elsayed. Unsupervised Adaptive Microblog Filtering for Broad Dynamic Topics. IP&M 2016
- Magdy W. and G. J. F. Jones. Studying Machine Translation Technologies for Large-Data CLIR Tasks: A Patent Prior-Art Search Case Study. Springer, Information Retrieval, 2013
- Magdy W. and G. J. F. Jones. PRES: A Score Metric for Evaluating Recall-Oriented Information Retrieval Applications. SIGIR 2010
- Magdy W., K. Darwish, and M. El-Saban. Efficient Language-Independent Retrieval of Printed Documents without OCR. SPIRE 2009
- Magdy W. and K. Darwish. Effect of OCR Error Correction on Arabic Retrieval. Springer, Information Retrieval, 2008

