

Evaluating Efficiency vs. Effectiveness for Vague Queries and Similarity Search in Digital Libraries

Norbert Fuhr

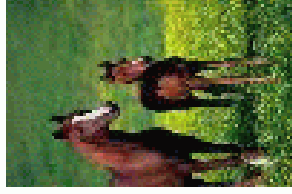
University of Dortmund, Germany

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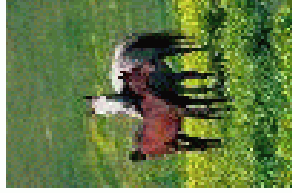
Motivating example



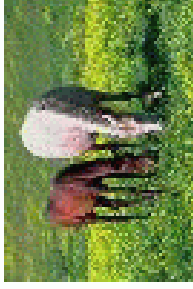
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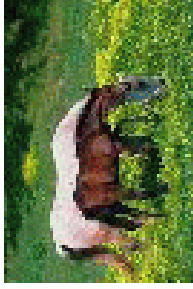
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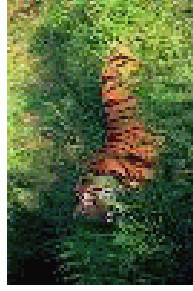
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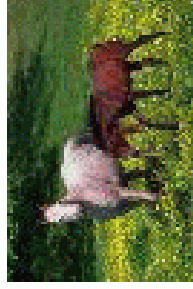
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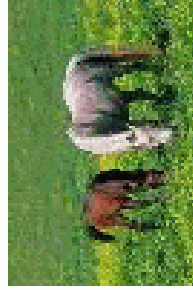
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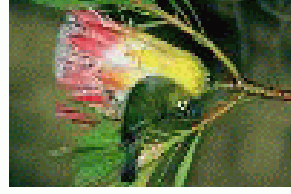
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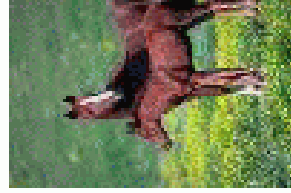
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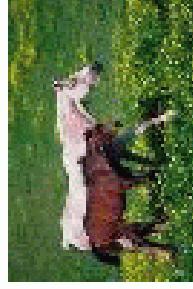
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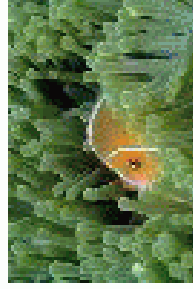
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[6239](#) 18.03 4



[35465](#) 18.05 3

Effectiveness

$$\text{effectiveness} = \frac{\text{result quality}}{\text{user's effort}}$$

user's effort:

≈ time for task completion

= (response time + thinking time) * #iterations

Increasing effectiveness

$$\text{effectiveness} = \frac{\text{result quality}}{\text{user's effort}}$$

- increase result quality

(may require more computations → longer response times → increased user effort)

- reduce user effort

(reduce response time → may decrease result quality)

→ tradeoff between result quality and response time

A new quality paradigm

example task: find 1 relevant document

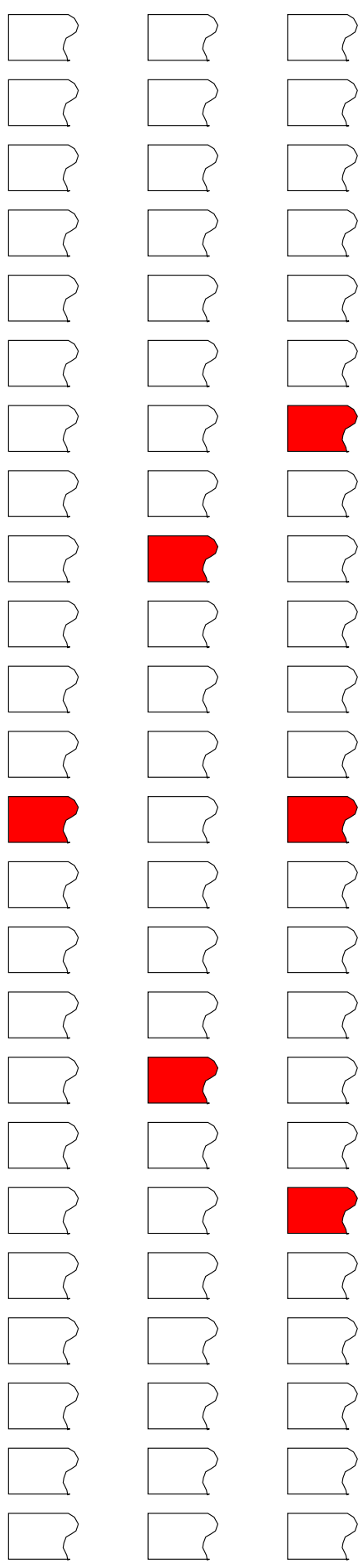
- I. 10 relevant documents in database,
system A succeeds 50% of the time
- II. 100 relevant documents in database,
system B succeeds 50% of the time

Which system is better?

Ideal system

- perfect matching function
- linear scan through the database

task complexity = expected # processing steps for task completion



Task complexity

expected search length $esl(j, r, s) = \frac{j}{r+1}s$

r # relevant documents in collection

s # nonrelevant documents in collection

j # relevant documents requested

N collection size, $N \approx s$

task complexity $tc(j, r, N) = \frac{j}{r+1}N$

Example (cont'd)

- I. 10 relevant documents in database,
system A succeeds 50% of the time

$$t_{CI} = \frac{1}{10 + 1}N$$

- II. 100 relevant documents in database,
system B succeeds 50% of the time

$$t_{CII} = \frac{1}{100 + 1}N$$

→ system A does the harder task

Retrieval quality

define retrieval quality as relative quality:

$$\text{relative quality} = \frac{\text{complexity of task performed by real system}}{\text{complexity of task performed by ideal system}}$$

(standard recall and precision measures also compare performance with ideal system)

Find n documents

Task: User wants n documents
(e.g. Web retrieval: $n = 10$)
assume $r \geq n$ relevant documents in the database

- ideal system: retrieves only relevant docs

$$tC_i = \frac{n}{r+1}N$$

- real system: finds only k relevant ones

$$tC_r = \frac{k}{r+1}N$$

$$\frac{tC_r}{tC_i} = \frac{k}{n}$$

relative quality:

→ **Precision measure**

Find k relevant documents

Task: User wants k *relevant* documents

- ideal system: retrieves k documents, all relevant
- real system: retrieves n documents s.th. k relevant ones included
(assumes all n documents are relevant)

$$\frac{tC_r}{tC_i} = \frac{\frac{k}{r \cdot n/k + 1} N}{\frac{k}{r+1} N} = \frac{r+1}{r \cdot n/k + 1} \approx \frac{k}{n}$$

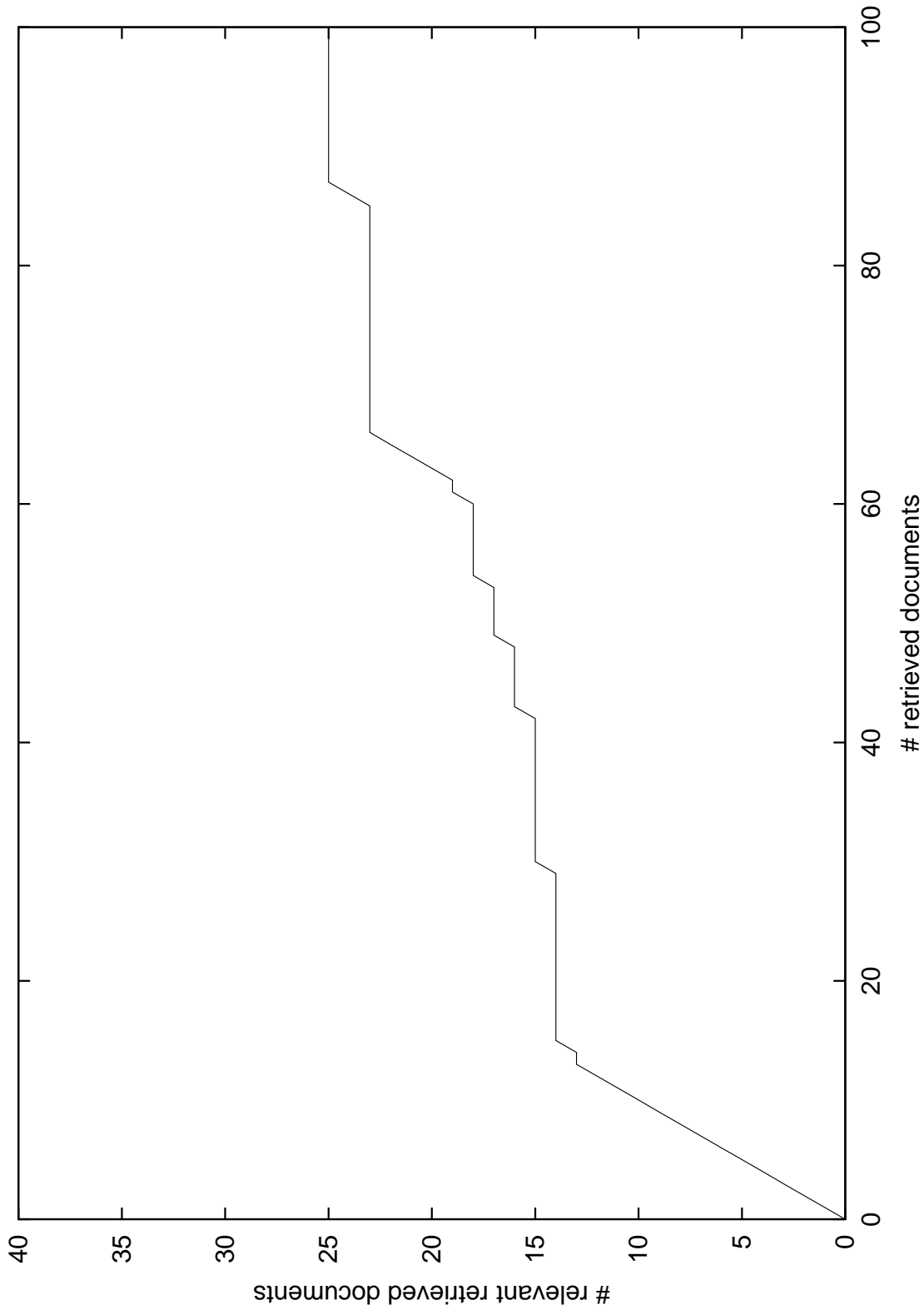
kNN search

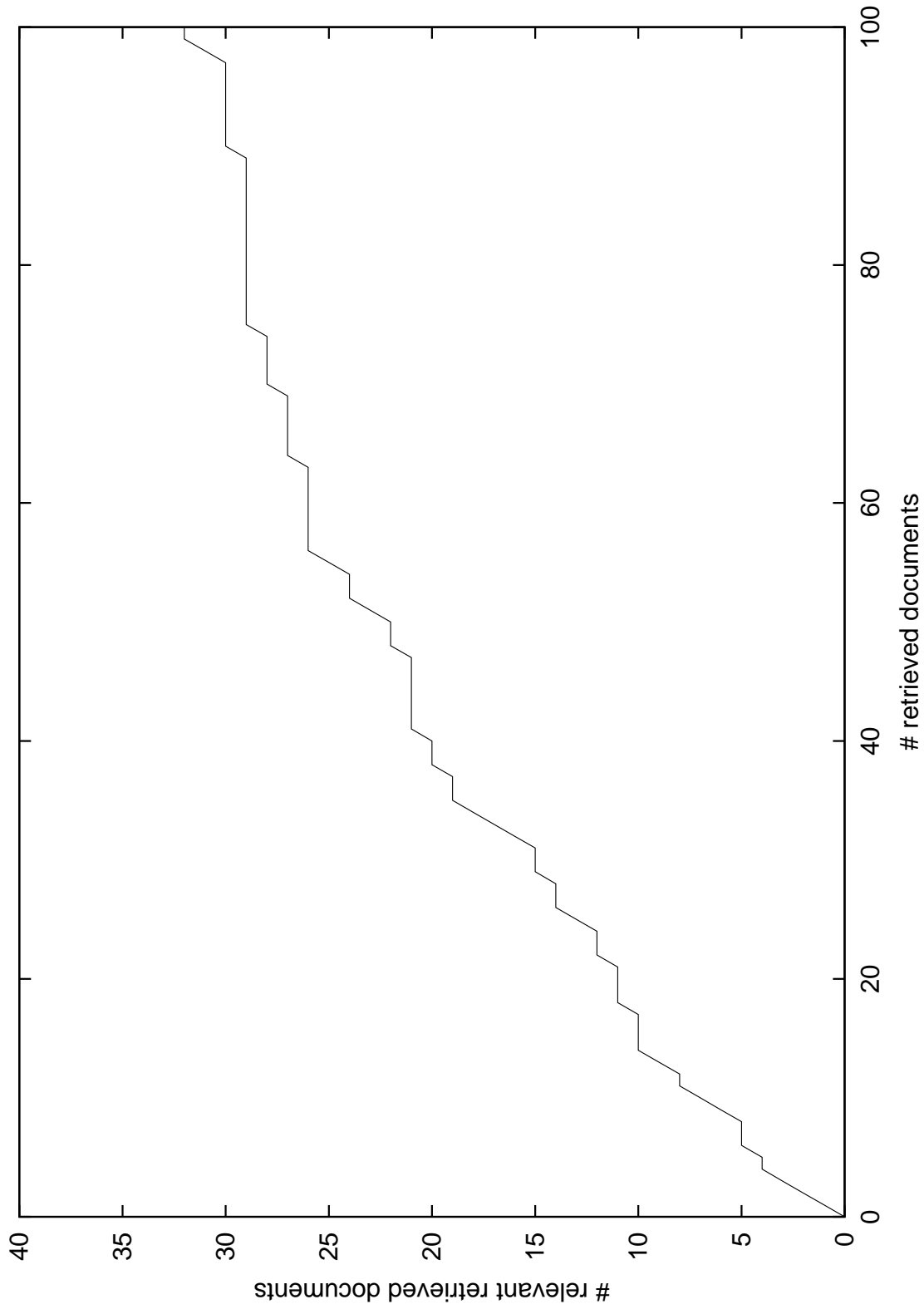
Task: User wants k closest matches

- ideal system: retrieves exactly k closest matches
- real system: retrieves k documents, k th document has real rank n
→ real system selects k documents from the top n ones

$$\frac{tc_r}{tc_i} = \frac{\frac{k}{n+1}N}{\frac{k}{k+1}N} = \frac{k+1}{n+1}$$

Similarity-based vs. relevance-based evaluation





Summary and conclusions

- new paradigm for measuring retrieval quality
- → new justification for recall and precision measure
- → new measure for evaluating approximate kNN searches

Open issues

- User's perception of absolute quality?
- Application to complex structures in digital libraries?