

Grundlagen des Information Retrieval

Vorbesprechung - 188.977

GIR – 2019/20

Lecturers

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Contact / Questions

Use the TUWEL Forum

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About 46.400.000 results (0,43 seconds)

Dictionary



information retrieval

noun **COMPUTING**

the tracing and recovery of specific information from stored data.
"an information retrieval system"



Translations, word origin, and more definitions

[Feedback](#)

Information retrieval - Wikipedia

https://en.wikipedia.org/wiki/Information_retrieval ▼

Information retrieval (IR) is the activity of obtaining **information** system resources relevant to an **information** need from a collection of **information** resources. Searches can be based on full-text or other content-based indexing.

[Overview](#) · [History](#) · [Model types](#) · [Timeline](#)

Information Retrieval – Wikipedia

https://de.wikipedia.org/wiki/Information_Retrieval ▼ [Translate this page](#)

Information Retrieval [ˌɪnfəˈmeɪʃən ɹɪˈtʃiːvəl] (**IR**) oder Informationsrückgewinnung, gelegentlich ungenau Informationsbeschaffung, ist ein Fachgebiet, ...

[Geschichte](#) · [Grundbegriffe](#) · [Relevanz und Pertinenz](#) · [Typologie von ...](#)

[PDF] Introduction to Information Retrieval - Stanford NLP Group

<https://nlp.stanford.edu/IR-book/pdf/01bool.pdf> ▼

Information retrieval (IR) is finding material (usually documents) of an unstructured nature (usually text) that satisfies an **information** need from within large collections (usually stored on computers).



Information retrieval



Information retrieval is the activity of obtaining information system resources relevant to an information need from a collection of information resources. Searches can be based on full-text or other content-based indexing. [Wikipedia](#)

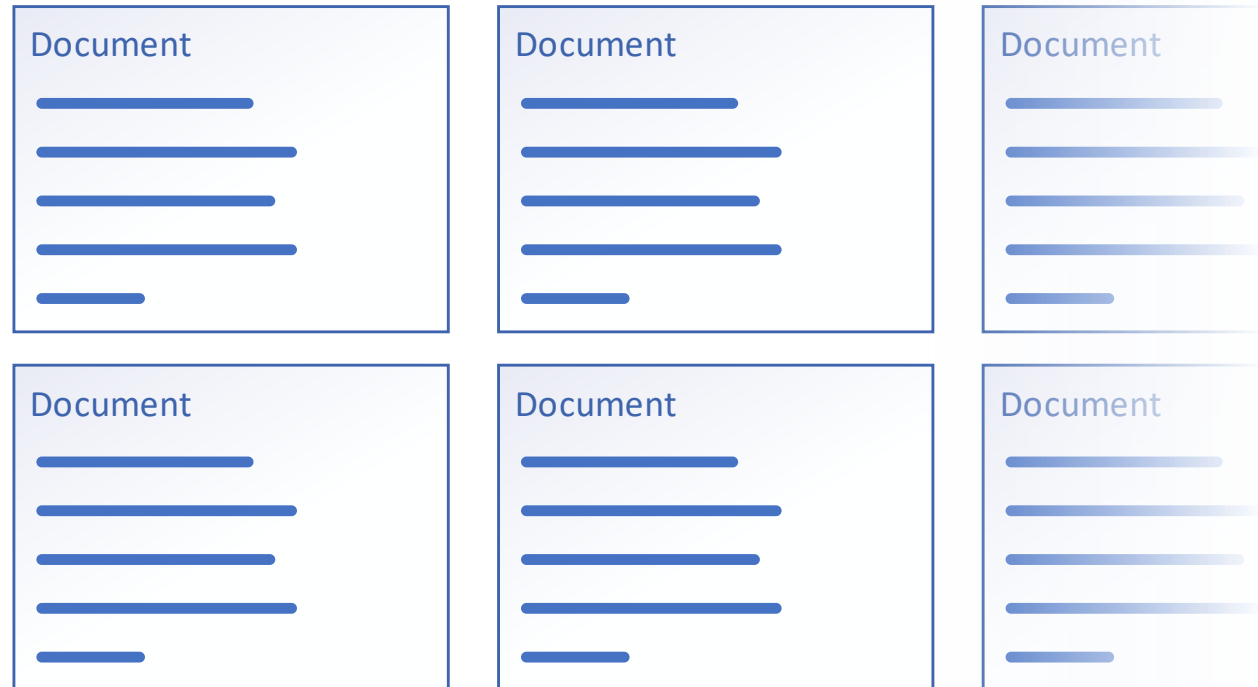
[Feedback](#)

Information Retrieval (Finding the needle in the haystack)

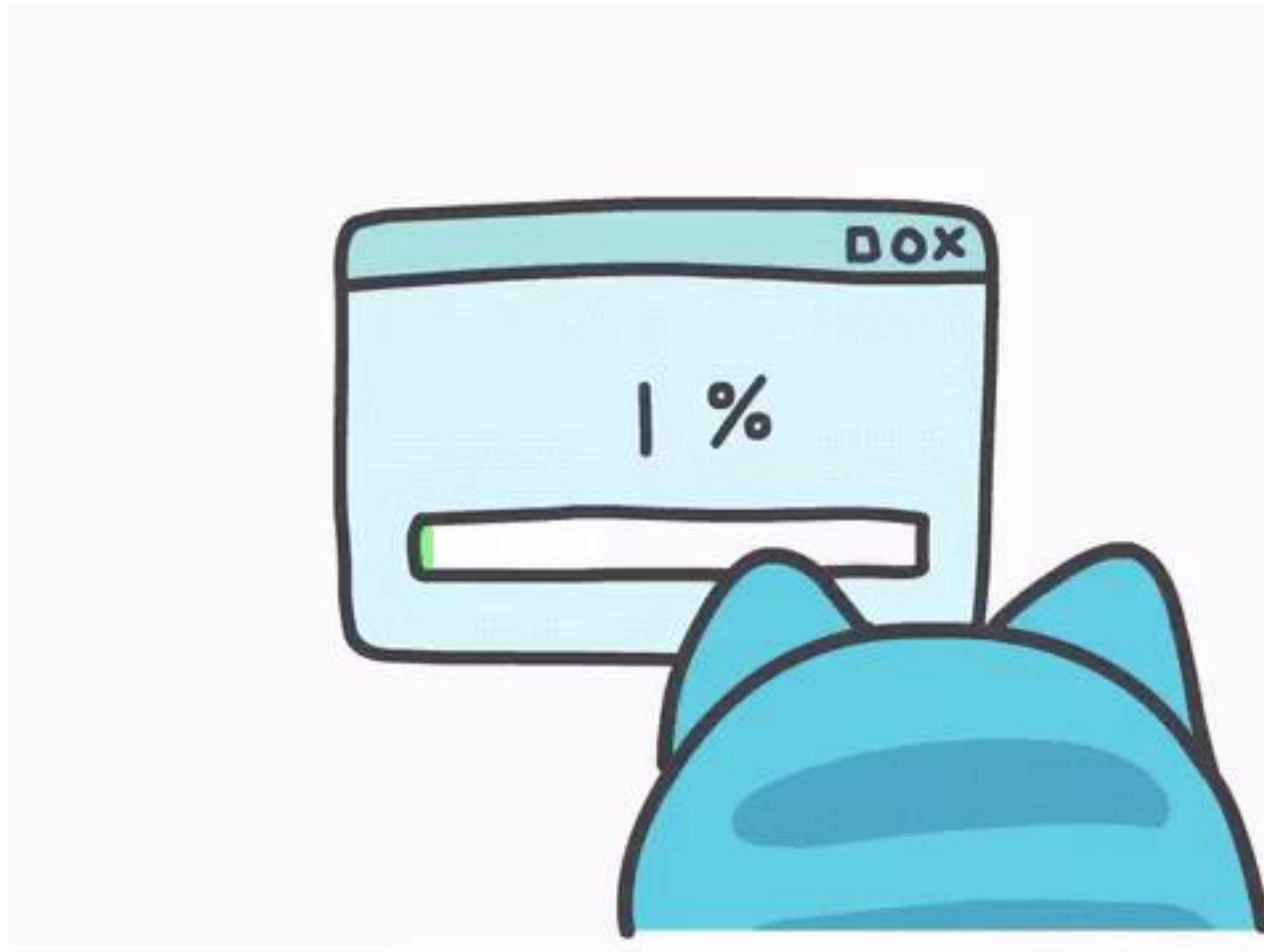
“ Baby kittens ”



**How
Relevant?**



Reality ...



Organization

Lectures, Exercises, Grading

Lectures / Content

Thursday, 16:00 – 18:00, EI 3A

- 3.10. Vorbesprechung
- 10.10. Foundations of Information Retrieval
- 17.10. Efficient text processing + Exercise 1 Infos
- 24.10. Scoring and search
- 31.10. Evaluation
- 7.11. Web search
- 14.11. Search interfaces
- TBD Music and image retrieval

Exercises

- Two exercise in total (#1: search engine, #2: music retrieval)
- Shared group for all exercises
 - 2 persons per group (managed via TUWEL)
 - First exercise will be evaluated in an interview (Abgabegespräch)
- All exercises share 1 private GitHub repository (via GitHub classroom)
- Lot's of bonus point opportunities in #1 🎉 🙌 100 ✨

Exercise #1

- Implement your own search engine!
 - With lots of freedom and lots of bonus points 🙌 100
- Index a subset of Wikipedia and make it searchable
 - And evaluate with given queries
- Focus on efficiency and correctness
- **More on this in the lecture @ 17.10.** 👍
 - Practical how to: write efficient text processing code
 - Fixing teams
 - Presentation of the exercise and QA

Exam

- Style: open answer questions
- We provide 2 dates:
 - **12.12.19** 13:00 FAV 1 (Hörsaal Favoritenstraße)
 - **9.1.20** 10:00 FAV 1 (Hörsaal Favoritenstraße)

Grading

Exercise 1 (Search engine):	50%
Exercise 3 (Music IR):	10%
Exam:	40% (min 30% to pass)
Total	100% (min 50% to pass)

Sounds cool?

Advanced Information Retrieval

- Available next semester!
- Content (Very fancy machine learning 😊)
 - Word Embeddings
 - Neural Networks for NLP
 - Neural IR
 - State of the art developments
- Exercise:
 - Implement different Neural IR models in PyTorch & create creative visualizations

Responsibility

+ Classroom Discussion

Responsibility – Social Impact of Ranking

- Recommendations are optimized for time spent on a platform (= revenue)
- Easy to fall down the rabbit hole – because scandalous & “click here to find the truth” videos keeps you on the platform
- Multilingual problems: Manually blocking English content does not automatically translate to other languages

How YouTube Radicalized Brazil <https://www.nytimes.com/2019/08/11/world/americas/youtube-brazil.html>

Zhe Zhao, Lichan Hong, Li Wei, Jilin Chen, Aniruddh Nath, Shawn Andrews, Aditee Kumthekar, Maheswaran Sathiamoorthy, Xinyang Yi, Ed Chi. 2019. Recommending What Video to Watch Next: A Multitask Ranking System. RecSys '19

Responsibility – Facial Recognition Error

Table 1: Overall Error on Pilot Parliaments Benchmark, August 2018 (%)

Company	All	Females	Males	Darker	Lighter	DF	DM	LF	LM
Target Corporations									
Face ++	1.6	2.5	0.9	2.6	0.7	4.1	1.3	1.0	0.5
MSFT	0.48	0.90	0.15	0.89	0.15	1.52	0.33	0.34	0.00
IBM	4.41	9.36	0.43	8.16	1.17	16.97	0.63	2.37	0.26
Non-Target Corporations									
Amazon	8.66	18.73	0.57	15.11	3.08	31.37	1.26	7.12	0.00
Kairos	6.60	14.10	0.60	11.10	2.80	22.50	1.30	6.40	0.00

Table 2: Overall Error Difference Between August 2018 and May 2017 PPB Audit (%)

Company	All	Females	Males	Darker	Lighter	DF	DM	LF	LM
Face ++	-8.3	-18.7	0.2	-13.9	-3.9	-30.4	0.6	-8.5	-0.3
MSFT	-5.72	-9.70	-2.45	-12.01	-0.45	-19.28	-5.67	-1.06	0.00
IBM	-7.69	-10.74	-5.17	-14.24	-1.93	-17.73	-11.37	-4.43	-0.04

Buolamwini, J., Gebru, T. "Gender Shades: Intersectional Accuracy Disparities in Commercial Gender Classification." Proceedings of Machine Learning Research 81:1–15, 2018 Conference on Fairness, Accountability, and Transparency

Raji, I & Buolamwini, J. (2019). Actionable Auditing: Investigating the Impact of Publicly Naming Biased Performance Results of Commercial AI Products. Conference on Artificial Intelligence, Ethics, and Society.

Responsibility – Word Embedding Bias

- Word embeddings are trained on large scale unlabeled text
 - For example: Wikipedia
- If training data is biased, vectors are biased as well

- Word2Vec trained on Wikipedia contains significant gender bias

Gender stereotype *she-he* analogies

sewing-carpentry	registered nurse-physician	housewife-shopkeeper
nurse-surgeon	interior designer-architect	softball-baseball
blond-burly	feminism-conservatism	cosmetics-pharmaceuticals
giggle-chuckle	vocalist-guitarist	petite-lanky
sassy-snappy	diva-superstar	charming-affable
volleyball-football	cupcakes-pizzas	lovely-brilliant

Gender appropriate *she-he* analogies

queen-king	sister-brother	mother-father
waitress-waiter	ovarian cancer-prostate cancer	convent-monastery

Bolukbasi, Tolga, et al. "Man is to computer programmer as woman is to homemaker? Debiasing word embeddings." *Advances in neural information processing systems*. 2016.

Classroom Discussion

- What should we do as computer scientists?
 - Just measure existing data bias?
 - Pro-actively improving information systems?
- Do you have an example you encountered in your work?
- Who is responsible for fair machine learning / search engines?

See you next week :)