



Text Technologies for Data Science INFR11145

Introduction

Instructor: Youssef Al Hariri

slides credit to Dr Walid Magdy

20/09/2023

Lecture Objectives

- Know about the course:
 - Topic
 - Objectives
 - Requirements
 - Format
 - Logistics

- Note:
 - No much technical content today
 - Don't assume next lectures would be the same!



Text Technologies for Data Science = documents, words, terms, ... ≠ images, videos, music (with no text) **Search Engines Information Retrieval Technologies** Text Classification





Youssef Al Hariri, TTDS 2023/2024

Text Analytics

What is Information Retrieval (IR)?

IR is NOT just



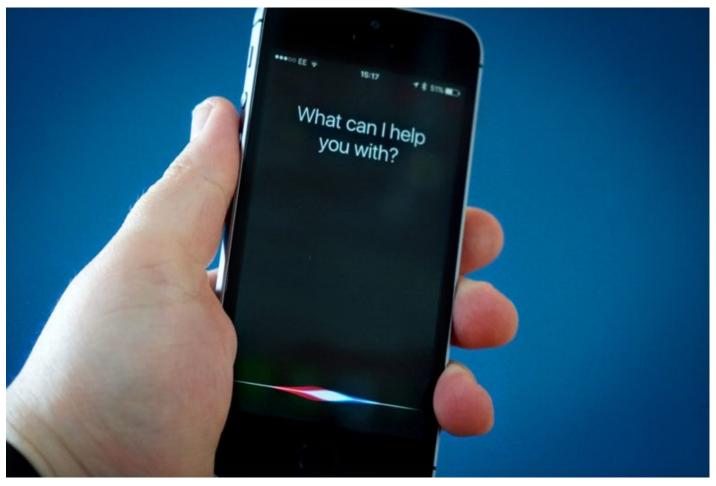
Google Search

I'm Feeling Lucky

Web search



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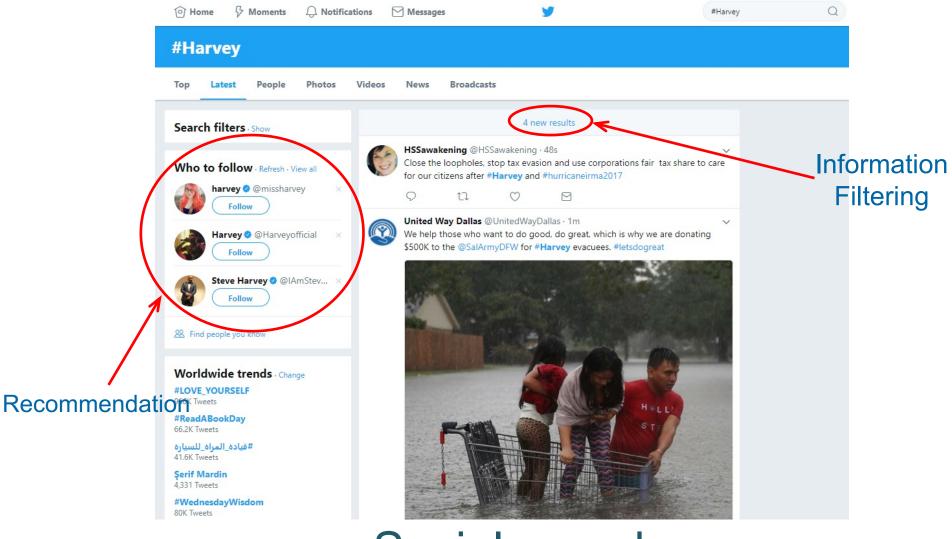


Speech - QA



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Social search



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DON ANTONIO LOPEZ

de Santa-Anna, Perez de Lebron, Caballero de número de la órden imperial de Guadalupe, Brigadier de los Egércitos nacionales, Comandante general, Gobernador y Gefe político de esta Plaza y su Provincia, condecorado con la cruz de Córdova, &c.

El Escmo. Sr. Secretario del Ministerio de Hacienda, con fecha 3 de setiembre me dice lo siguiente.

Los Esemos. Sres. secretarios del soberano Congreso constituyente, me dicen en oficio de ayer lo que sigue.

"Esemo. Sr.—Al resolver el soberano Congreso las dudas que han ocurrido á la Direccion general de alcabalas en el cobro del dos por ciento impuesto al oro, plata y cobre acuñado que salga de todas las aduanas terrestres para pago del préstamo de seiscientos mil pesos, tuvo á bien dictar: Que cesase dicho gravamen desde luego que llenase su objeto. En la órden en que aquella resolucion se comunicó á V. E., se omitió por olvido este requisito, y ahora lo avisamos á V. E. para que en su caso tenga el debido cumplimiento."

Y habiendo dado cuenta con ℓl al Emperador, se ha servido S. M. I. mandar se cumpla, publique y circule; y \acute{a} V. S. lo comunico de su órden para los fines que le tocan.

Dios guarde á V. S. muchos años. Mégico 3 de setiembre de 1822.-Medina.

Y para que llegue á noticia de todos, mando se publique por bando para su inteligencia y cumplimiento. Dado en Veracruz á 15 de octubre de 1822.

> Antonio Lopez de Santa-Anna. Por mandado de S. S. José Gimenez.

Library (book) search 1950's



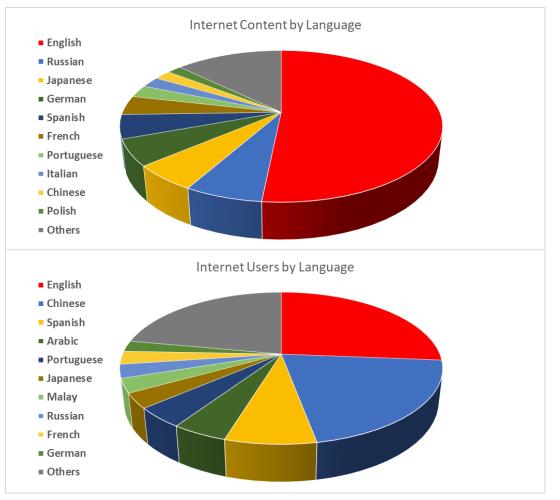






Legal search





Cross-Language search

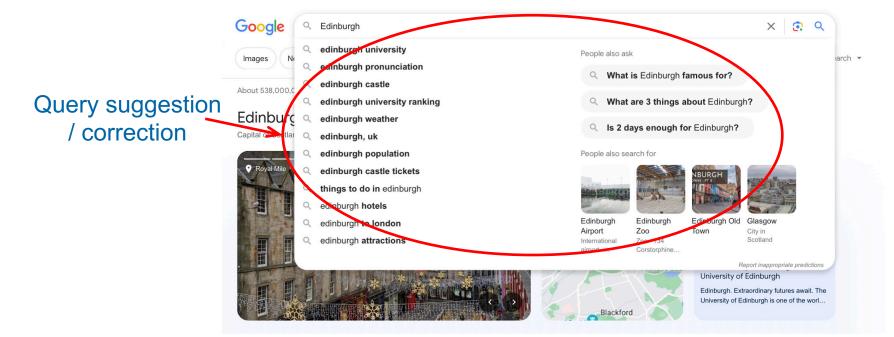






Content-based music search





Things to do



Edinburgh Castle 4.6 ★ (87K) Castle £22.00



National Museum of Scotland 4.8 ★ (42K) National museum

Free



Palace of Holyroodhouse 4.6 ★ (17K) Castle £18.00

More things to do $\,\,
ightarrow\,$

About

Edinburgh is Scotland's compact, hilly capital. It has a medieval Old Town and elegant Georgian New Town with gardens and neoclassical buildings. Looming over the city is Edinburgh Castle, home to Scotland's crown jewels and the Stone of Destiny, used in the coronation of Scotlish rulers. Arthur's Seat is an imposing peak in Holyrood Park with sweeping views, and Calton Hill is topped with monuments and memorials. — Google

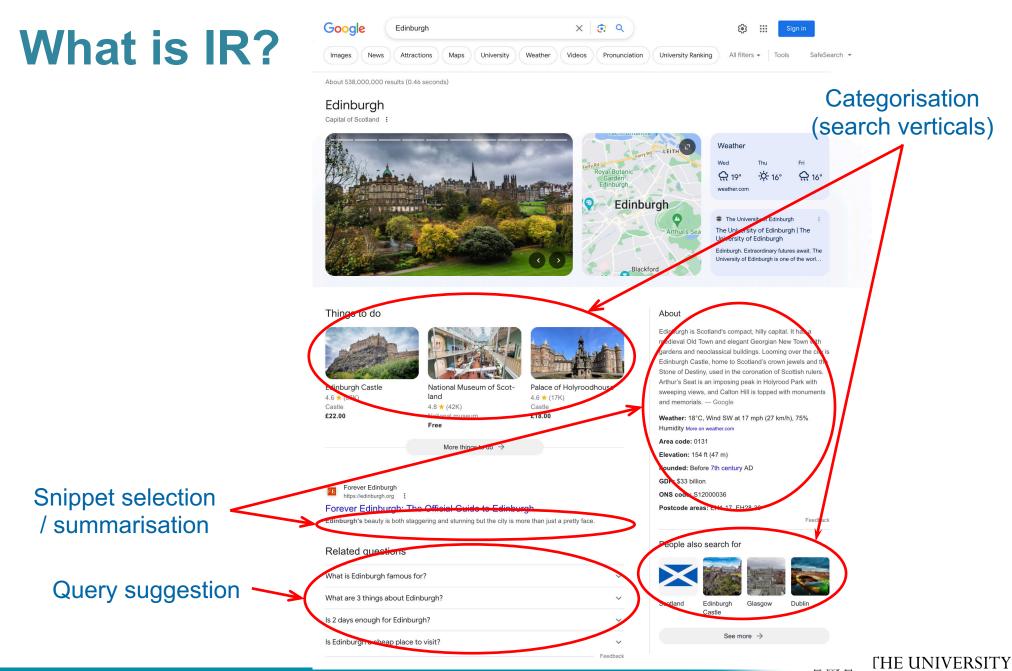
Weather: 18°C, Wind SW at 17 mph (27 km/h), 75% Humidity More on weather.com Area code: 0131

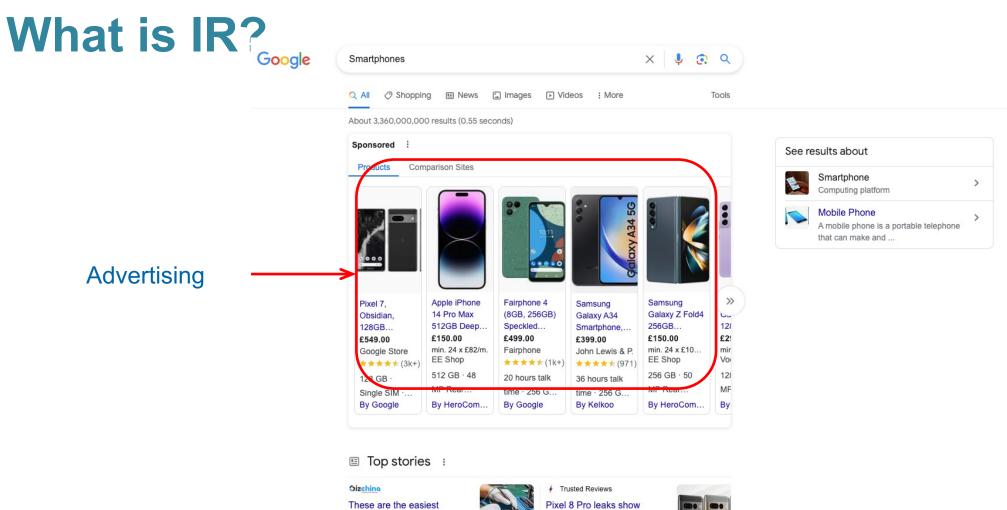
Elevation: 154 ft (47 m)



*Source: Matt Lease (IR Course at U Texas)

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smartphones to repair right now





20 hours ado

we've hit peak smartphone



3 days ago



How to set volume limits for audio on your smartphone



20 hours ano



What is IR? Find?

File Edit View Window Help Home Tools

QU-RT.pdf

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terent enough from any of the pushed tweets), otherwise, the system does not consider pushing it to the user.

×

2.2 Push Notifications Scenario

The push notifications scenario simulates a recommender system that sends pop-up messages to users on their mobile phones after capturing tweets that match their interests. The task design restricts the number of pushed tweets per profile to a maximum of 10 tweets per day to avoid overwhelming the users. Having such constraint on the number of tweets to push, the system should wisely select the best candidate tweets to elect to the user in a timely fashion. We explain next how we used tweet freshness to nominate tweets to be pushed for an interest profile.

2.2.1 Tweets Nomination

While tracking all interest profiles simultaneously and monitoring the tweets stream, the system maintains, for each of the interest profiles, a list of candidate tweets that contains the tweets that were found relevant and novel so far. The RTS system periodically nominates a tweet to push to the

$IR \neq Find$

- Sequential
- Exact match

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After scoring all terms, the top Find pansion terms, are added to the tor drift, the topic vector is reset to the vector) before each expansion, as s

 $\vec{q}' = \vec{q} + \beta *$ Previous Next

retrieval

where \vec{e} is the normalized vector of the k expansion terms, and β is a parameter used to restrict the influence of expansion terms on the new topic vector.

2.3 Periodic E-mail Digest Scenario

In this scenario, the RTS system is required to compile a daily list of a maximum of m tweets per interest profile and send it as an email digest to the user. For that, we adopted a similar but even simpler approach than the approach for push notification scenario. At the end of each day of the evaluation period, the system issues the title of the interest profile against the local tweet index that is incrementally updated over time. We experimented with three retrieval

¹https://dev.twitter.com/rest/public/search

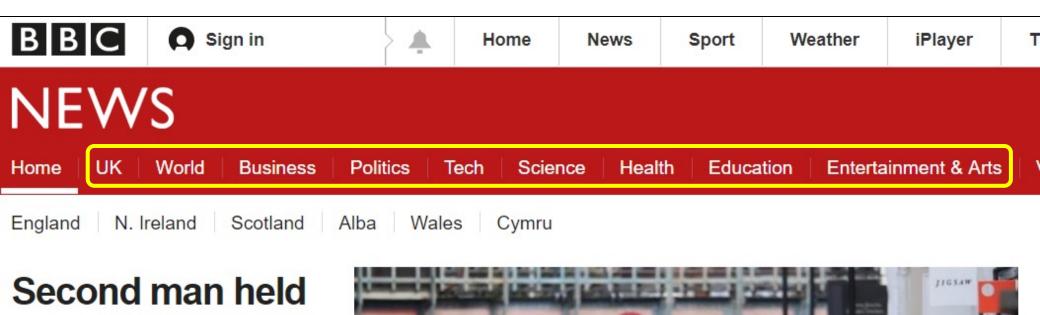


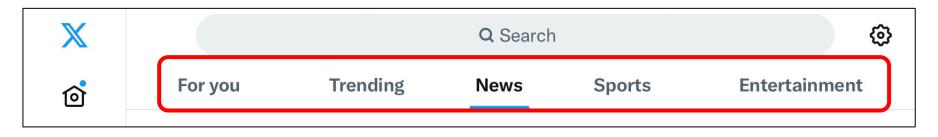


- IR is <u>finding</u> material of an <u>unstructured</u> nature that <u>satisfies</u> an <u>information need</u> from within large collections
- Find \rightarrow Task
- Unstructured \rightarrow Nature
- Information need \rightarrow Target
- Satisfies \rightarrow Evaluation



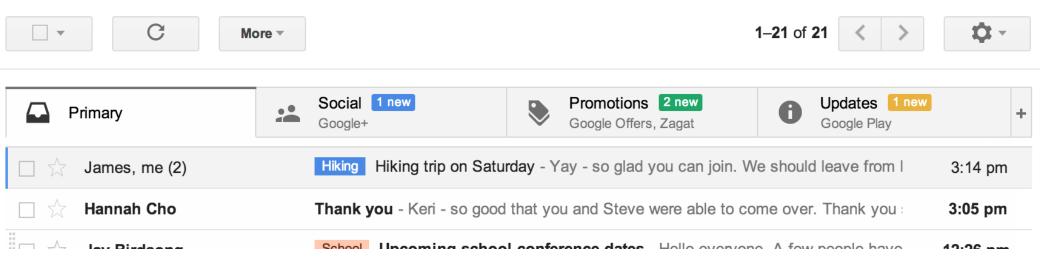
Text classification







Text classification



Sent Mail Sent Max Spam (372) Trash



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Text classification

- (12) United States Patent Magdy et al.
- (54) PERSONALIZED EVENT NOTIFICATION USING REAL-TIME VIDEO ANALYSIS
- (75) Inventors: Walid Magdy, Giza (EG); Motaz El-Saban, Giza (EG)
- (73) Assignee: Microsoft Corporation, Redmond, WA (US)
- (*) Notice: Subject to any disclaimer, the term of this

- (10) Patent No.:(45) Date of Patent:
- US 8,881,191 B2 Nov. 4, 2014
- (51) Int. Cl. *H04H 60/65* (2008.01) *H04H 60/48* (2008.01) *G06F 17/30* (2006.01)

US008881191B2





What is text classification?

 Text classification is the process of <u>classifying</u> documents into <u>predefined categories</u> based on their <u>content</u>.

- Input: Text (document, article, sentence)
- Task: Classify into one/multiple categories
- Categories:
 - Binary: relevant/irrelevant, spam .. etc.
 - Few: sports/politics/comedy/technology
 - Hierarchical: patents



In this course, we will learn to

- How to build a search engine
 - which search results to rank at the top
 - how to do it fast and on a massive scale
- How to evaluate a search algorithm
 - is system A really better than system B
- How to work with text
 - two tweets talk about the same topic?
 - handle misspellings, morphology, synonyms
- How to classify text
 - into categories (sports, news, comedy, ...)
 - features to use
 - evaluate classification quality
- Apply text analytics
 - Find what makes a set of document different from others



How this course is different from others?

- ANLP, FNLP
 - Some text processing
 - Text laws
 - No NLP (word/phrase level vs document level)
- ML practical
 - Text classification
 - No ML (using off-the-shelf ML tool)
- It does <u>not</u> overlap with others on:
 - Search engines
 - IR methods/models
 - IR evaluation
 - Text analysis
 - Processing large amount of textual data



Some terms you will learn about

- Inverted index
- Vector space model
- Retrieval models: TFIDF, BM25, LM
- Page rank
- Learning to rank (L2R)
- MAP, MRR, nDCG
- Mutual information, information gain, Chi-square
- binary/multiclass classification, ranking, regression



This Course is Highly Practical

- 70% of the mark is on practical work
- You will implement 50+% of what you learn
- By W5, you should have developed a basic working <u>Search Engine</u> from scratch
- Practical Lab <u>every week</u>
- Two coursework, mostly coding
- A course group project to develop a full system



Pre-requests (1/3)

- Maths requirements:
 - Linear algebra: vectors/matrices (addition, multiplication, inverse, projections ... etc).
 - Probability theory: Discrete and continuous univariate random variables. Bayes rule. Expectation, variance. Univariate Gaussian distribution.
 - Calculus: Functions of several variables. Partial differentiation. Multivariate maxima and minima.
 - Special functions: Log, Exp, Ln.

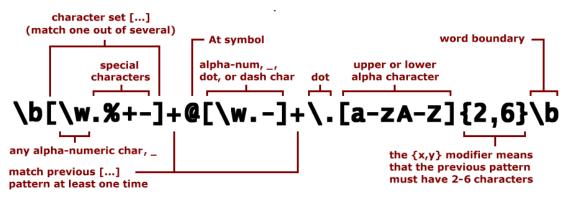
$$\texttt{BM25}(D,Q) = \sum_{i=1}^n \ \log \frac{N - n(q_i) + 0.5}{n(q_i) + 0.5} \cdot \left[\frac{f(q_i,D) \cdot (k_1 + 1)}{f(q_i,D) + k_1 \cdot \left(1 - b + b \cdot \frac{|D|}{\text{avgdl}}\right)} + \delta \right]$$



Pre-requests (2/3)

python

- Programming requirements:
 - <u>Python</u>
 - Knowledge in <u>regular expressions</u>
 - Shell commands (cat, sort, grep, uniq, sed, ...)
 - Data structures and software engineering for course project.
- We <u>DO NOT</u> teach coding skills in this course! We assume you can code!



Parse: username@domain.TLD (top level domain)



Pre-requests (3/3)

- Team-work requirement:
 - Final course project would be in groups of 5-6 students.
 - Working in a team for the project is a <u>requirement</u>.
 - No exceptions will be allowed!







Skills to be gained !!!

- Working with large text collections
- Few shell commands
- Some Python programming
- Software engineering skills
- Build text classifier in few mins
- TEAM WORK
 - Project management
 - Time management
 - Task assignment + system integration



Course Structure

- 20 Lectures:
 - 2 lectures \rightarrow Introduction (today)
 - 14 lectures \rightarrow IR (50% practical lectures)
 - 4 lectures → Text Analytics/Classification
- 8-10 Labs:
 - Practice what you learn
- No Tutorials
- Some self-reading
- Lots of system implementation
- Few online videos



Course Instructors



- Youssef Al Hariri Lecturer (15 lectures)?
- + 1 guest lecture

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Bjorn Ross Lecturer (4 lectures)



Lecture Format

- 2 Lectures at a time
- Questions are allowed any time. Feel free to interrupt
- 5-10 mins break after L1
 - Feel free to go out and come back
 - Discuss 1st lecture with friends
 - Questions on L1 are allowed before starting L2
- Some lectures are interactive. Please participate
- Some lectures will include demos (running code)



Labs

- How it works:
 - Relevant lab will be announced with each lecture on Wednesday
 - You should implement lab directly after lecture
 - Any issues \rightarrow ask on Piazza (tag question by lab number)
 - Produced output \rightarrow Share on Piazza (publicly)
 - Demonstrators \rightarrow answer questions + validate your output
 - TA \rightarrow answer questions about the course
 - DO NOT ask a question before checking if it was asked before
 - Tuesdays \rightarrow Optional in-person labs for those still require support
- Optional in-person labs:
 - Location: AT 6.06 (TBC)
 - Times: Tuesday, 10:00, 11:00, 13:00 (TBC)
- Demonstrators: TBA



Lab Zero (Lab 0)

- Please check Lab 0 before next week lectures
- Lab 0 is designed for one purpose: Help you decide to take TTDS or not
- Lab content:
 - Read a text file word by word, lower-case letters, print
 - Count the number of occurrence of few words
- If Lab 0 <u>challenging</u> \rightarrow
 - \rightarrow Probably, TTDS would be <u>very challenging</u> to you
 - \rightarrow You will need much <u>extra effort</u> to implement labs and CW
 - \rightarrow Think <u>wisely</u> before you decide to take the course



Assessments

- Coursework 1: 10%
 The same as labs 1-3 → Build your first search engine
- Coursework 2: 20%
 IR Evaluation, Text classification/analytics
- Group project: **40%** A full running search engine supported by text technologies
- Final Exam: 30%

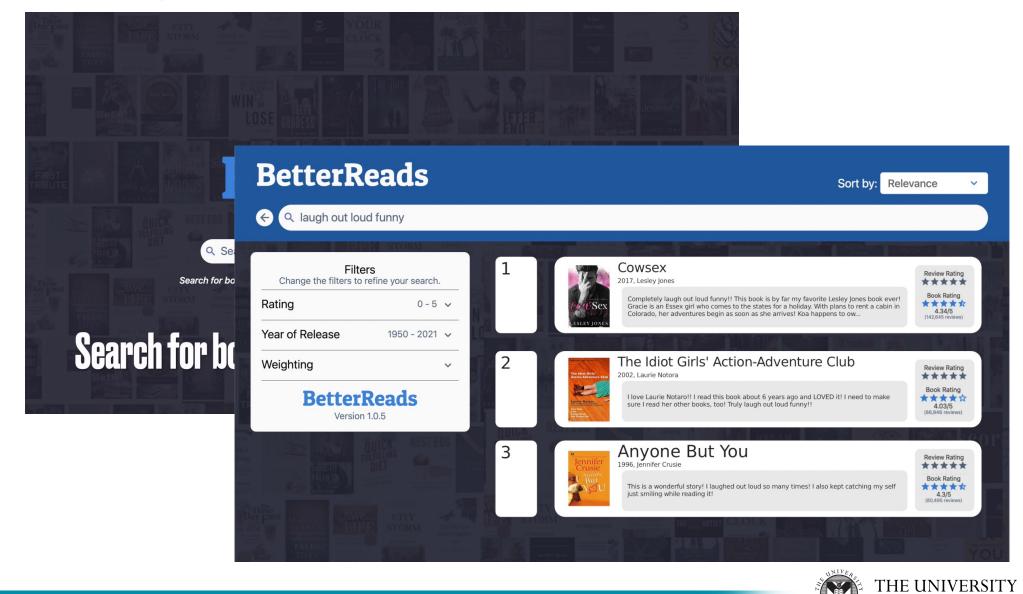


Group Project

- The largest weight: 40% of the total mark
- Teamwork \rightarrow Group 5-6 (<u>you select your own group</u>)
- Design a full end-to-end search engine that searches a large collection of documents with many functionalities.
- Mark = Mark_{project} x weight_{individual}
 - Mark_{project} \rightarrow the same for all team members
 - How complete/effective/fast/nice is your search engine?
 - weight_{individual} → weight for individual contribution.
 ranges from 0 to 1. It should be 1.0 by default but can be different for each member according to their contribution.
- Project prize \rightarrow a prize will be awarded to best project



Example: BetterReads



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Example: BetterReads

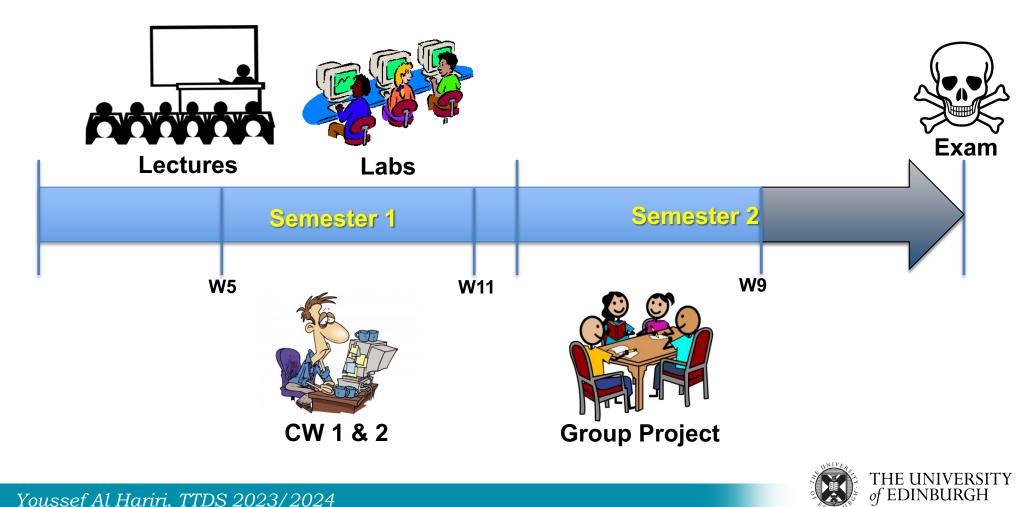
- 11.5M Book reviews from Good reads
- Average query time: 1 secs
- New reviews are crawled and indexed automatically every day
- Ranking: Relevance + Sentiment
- Engine hosted on Google cloud compute

• Note: we will provide credit to Google cloud to host your engine



Timeline

• 2 Semesters (or one?)





- Lectures:
 - Two lectures on Wednesdays, 15.00-17.00
 - Recording will be available
 - Handouts to be posted on the day of the lecture
- Course webpage:
 - Link: https://opencourse.inf.ed.ac.uk/ttds/
 - Handouts, Labs, CW details
- Learn:
 - Lecture recordings
 - Deadlines
- Note: all course materials are made public, including recordings. Feel free to share with anyone interested.





- All communication will be there
- Questions about lectures/labs/CW are there
- Feel free to answer each other questions
- Lab support will be mainly there
- Please share your lab answers there
- Tag each question/post by its relevant topic (lab, CW ... etc)
- Join <u>NOW</u>: <u>link</u>





- How the project would be managed? What if one member does not work?
- I am not that solid in programming, should I take this course?
- Can I audit the course?

• Anything else?



Next Lecture

• Definitions of IR main concepts (more introduction)





• These slides are originally created by previous lecturers Walid Magdy and Bjorn Ross.

