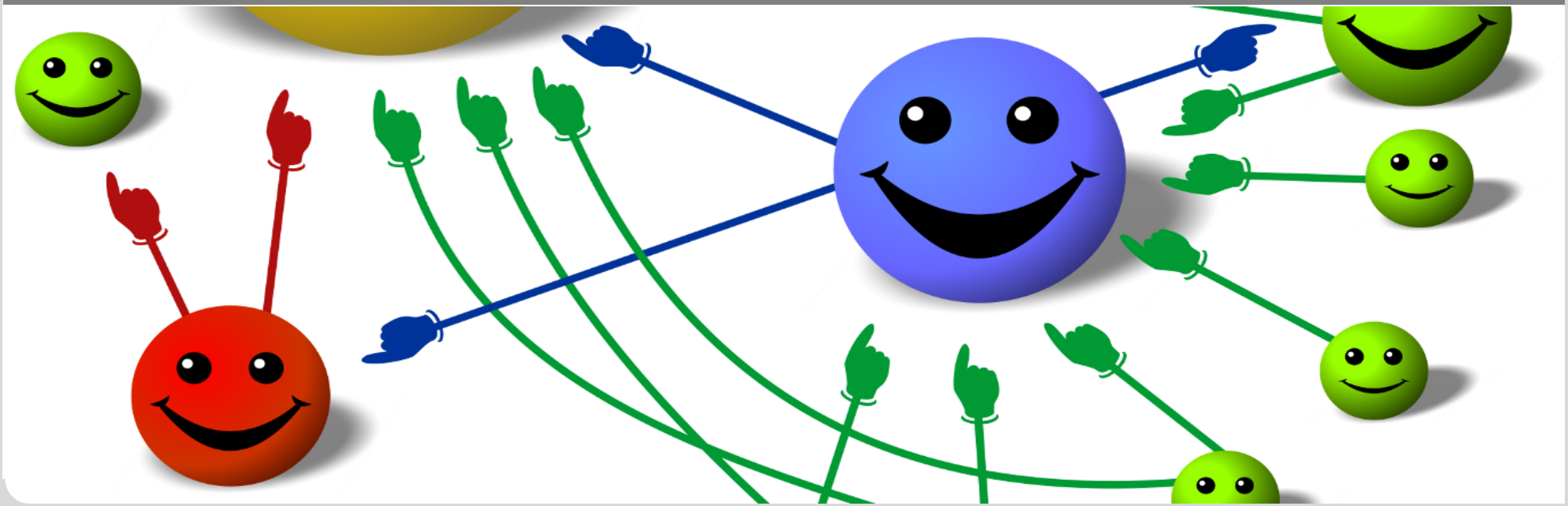
Karlsruhe Institute of Technology

How Google Searches Work

History of searching and the PageRank algorithm

Martin Thoma, Benjamin Lipp | 7th of February, 2013

SPRACHENZENTRUM



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2. PageRank
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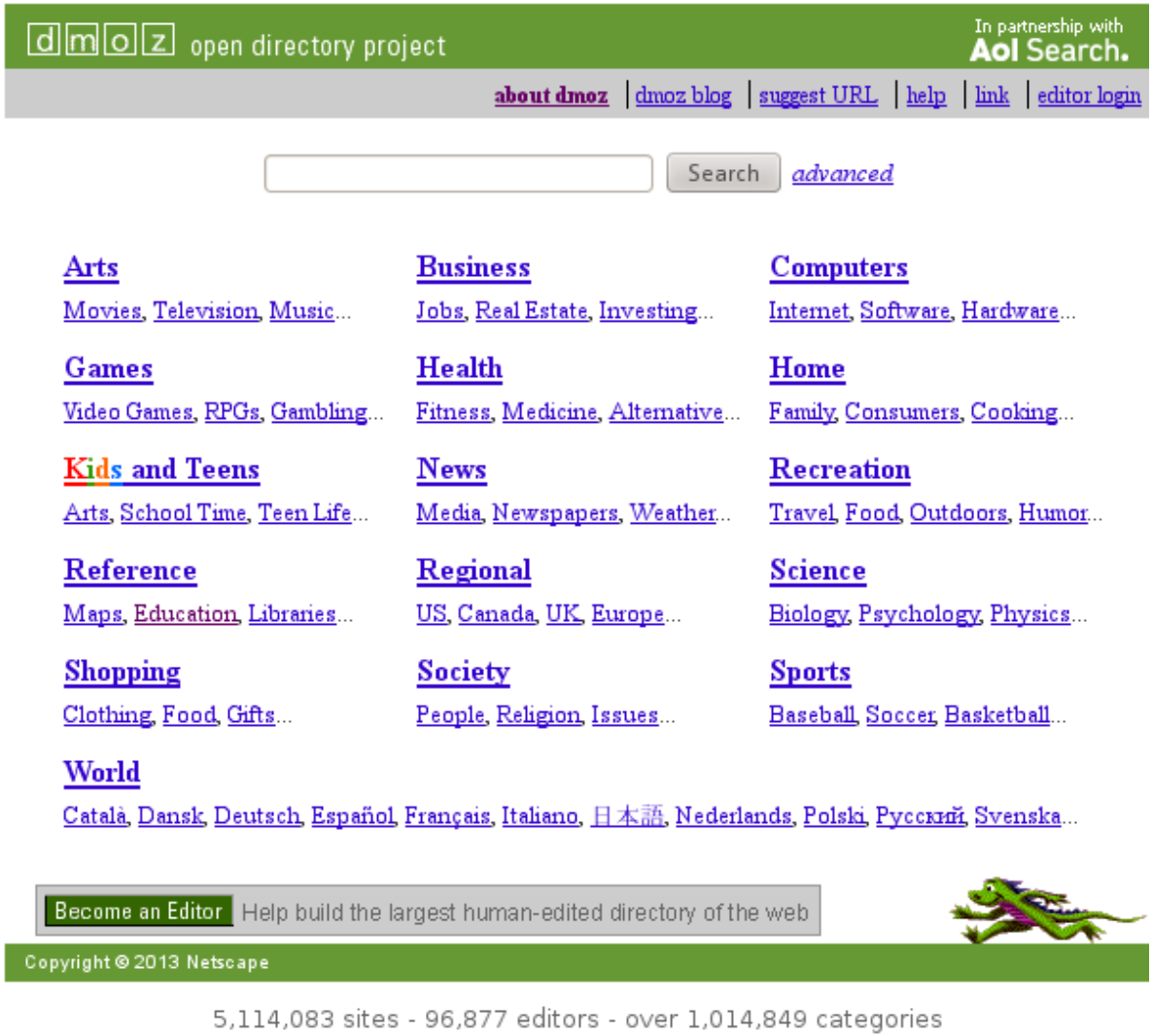
The Early Days: Web Directories



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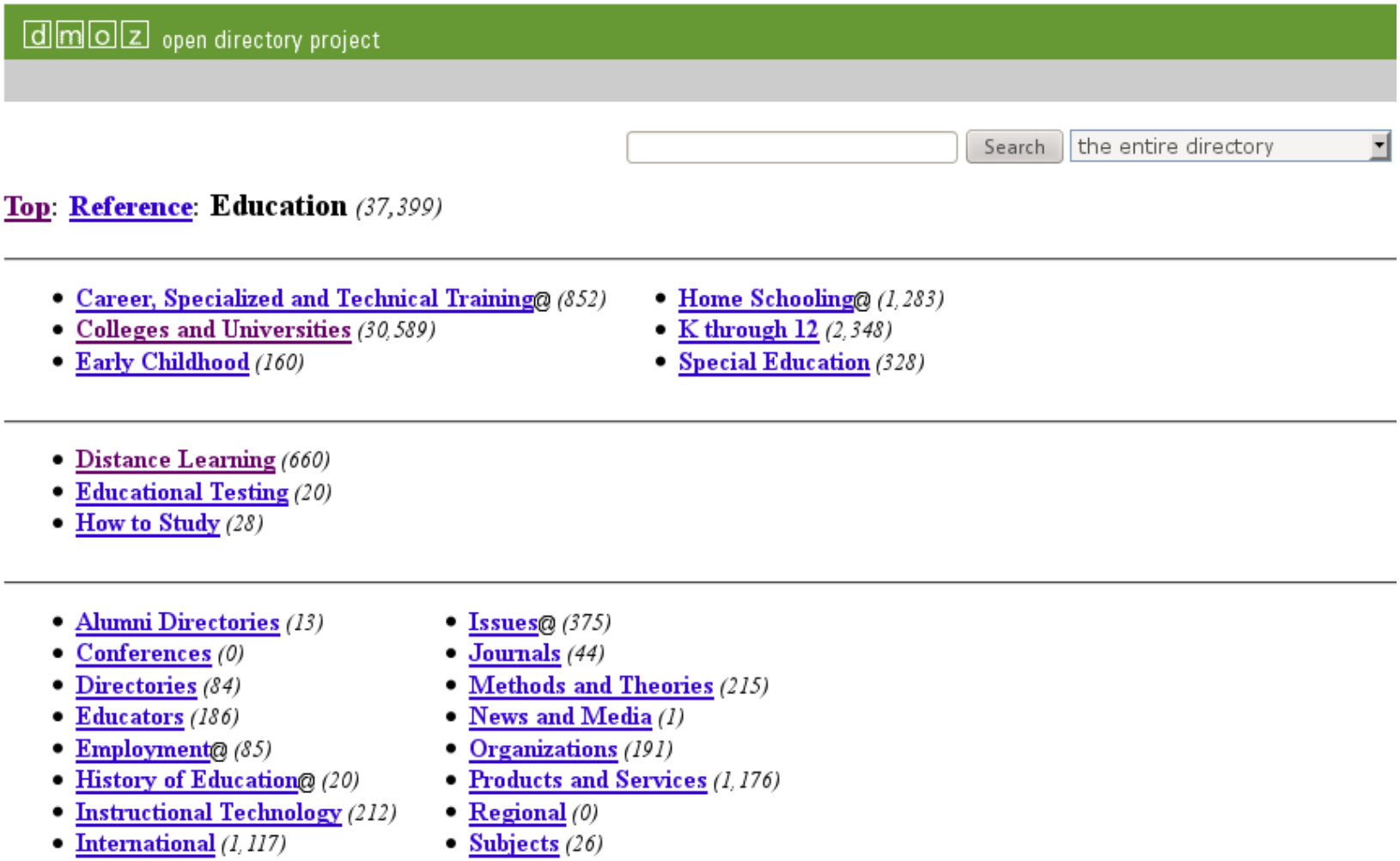

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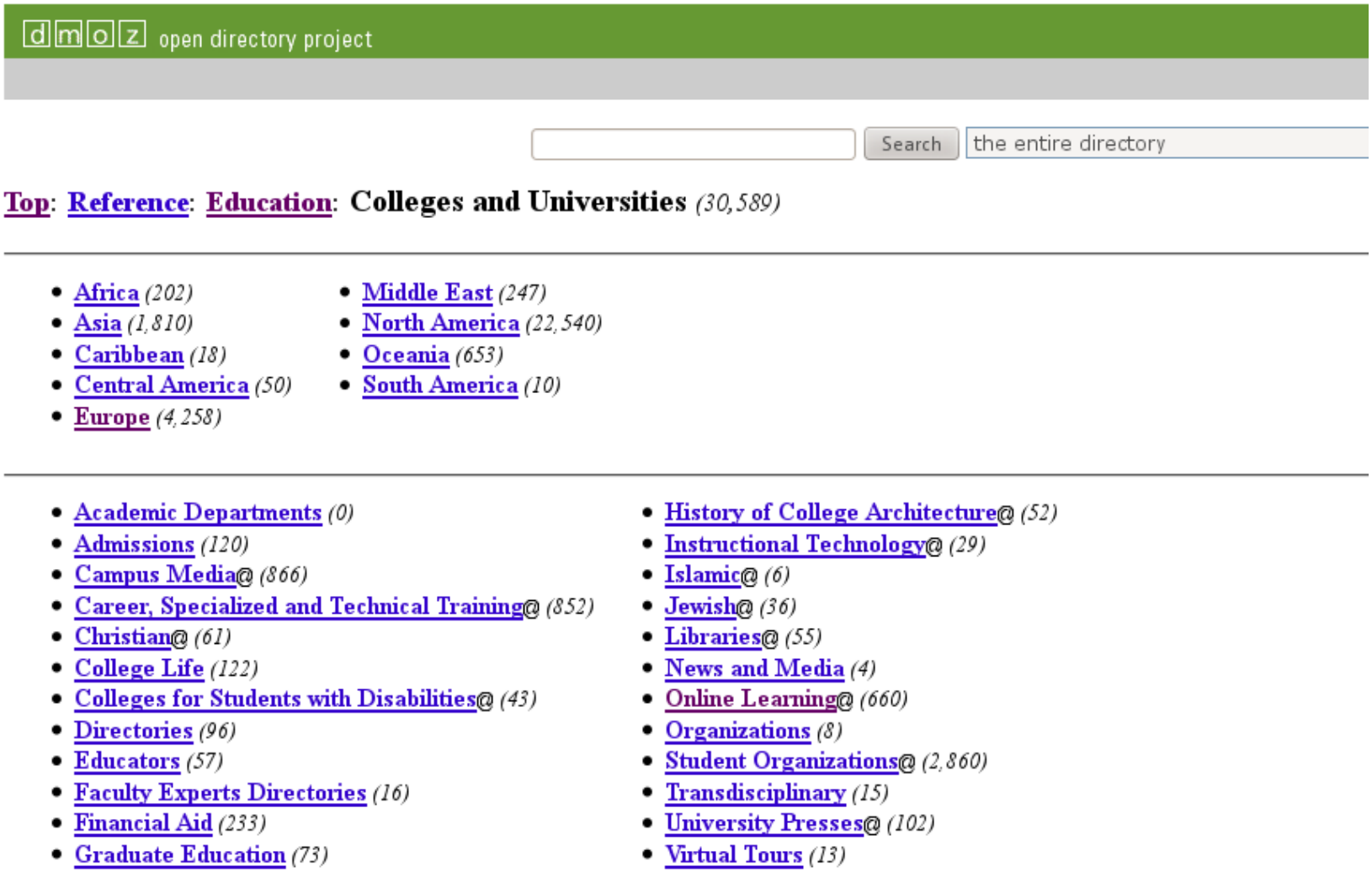

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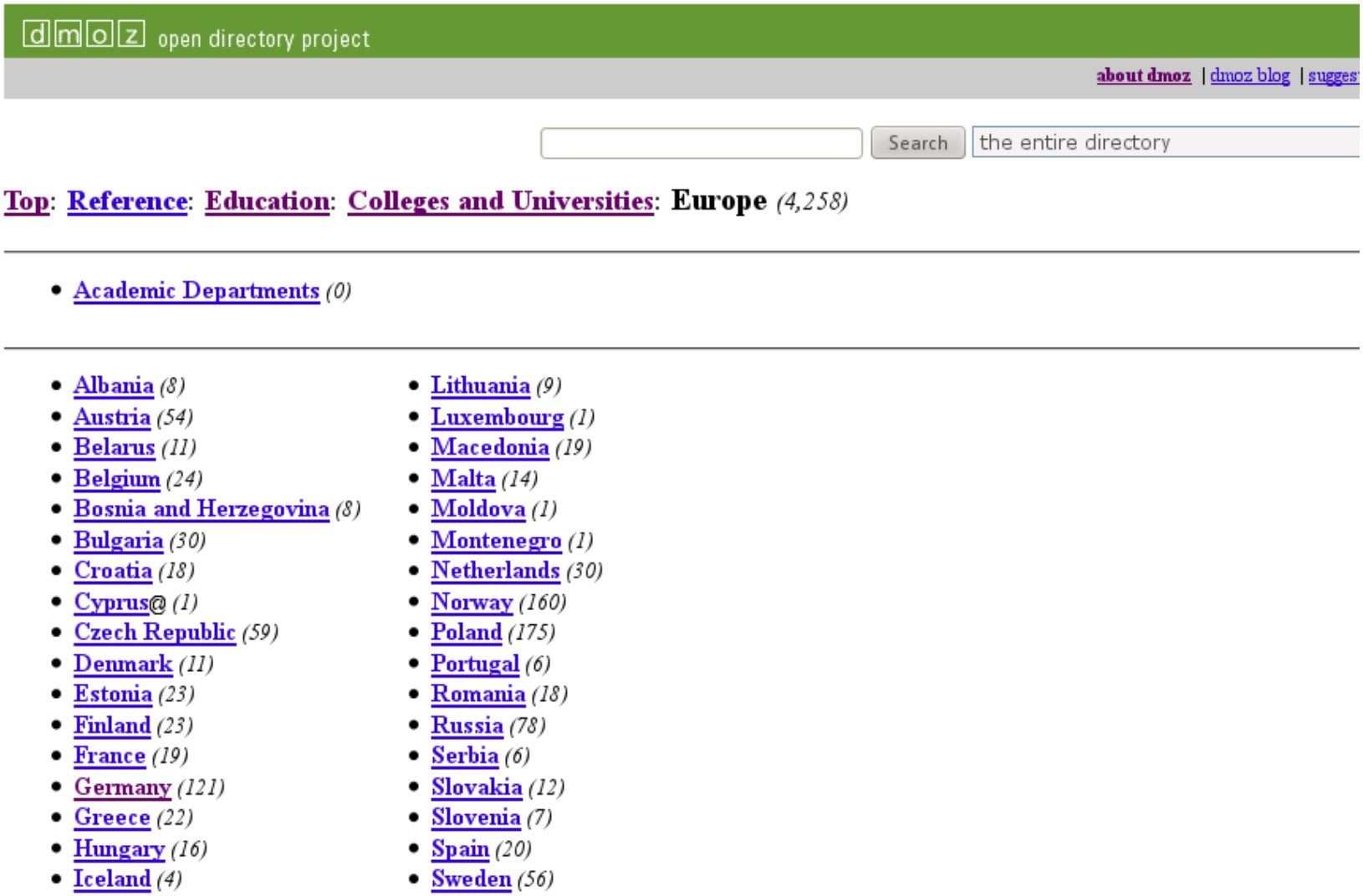

The Early Days: Web Directories

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Search | the entire directory

Top: **Reference**: **Education**: **Colleges and Universities**: **Europe**: **Germany** *(121)*

[A | B | C | D | E | F | G | H | I | J | K | L | M | N | O | P | Q | R | S | T | U | V | W | X | Y

- Baden-Württemberg *(15)*
- Bavaria *(10)*
- Berlin *(35)*
- Brandenburg *(5)*
- Bremen *(2)*
- Hamburg *(3)*
- Hesse *(7)*
- Lower Saxony *(3)*
- Mecklenburg-Western Pomerania *(0)*
- North Rhine-Westphalia *(23)*
- Rhineland-Palatinate *(4)*
- Saarland *(2)*
- Saxony *(1)*
- Saxony-Anhalt *(0)*
- Schleswig-Holstein *(2)*
- Thuringia *(5)*

See also:

- Regional: Europe: Germany *(3,998)*
- Regional: Europe: Germany: Education *(11)*

This category in other languages:

French *(6)* German *(2,229)*

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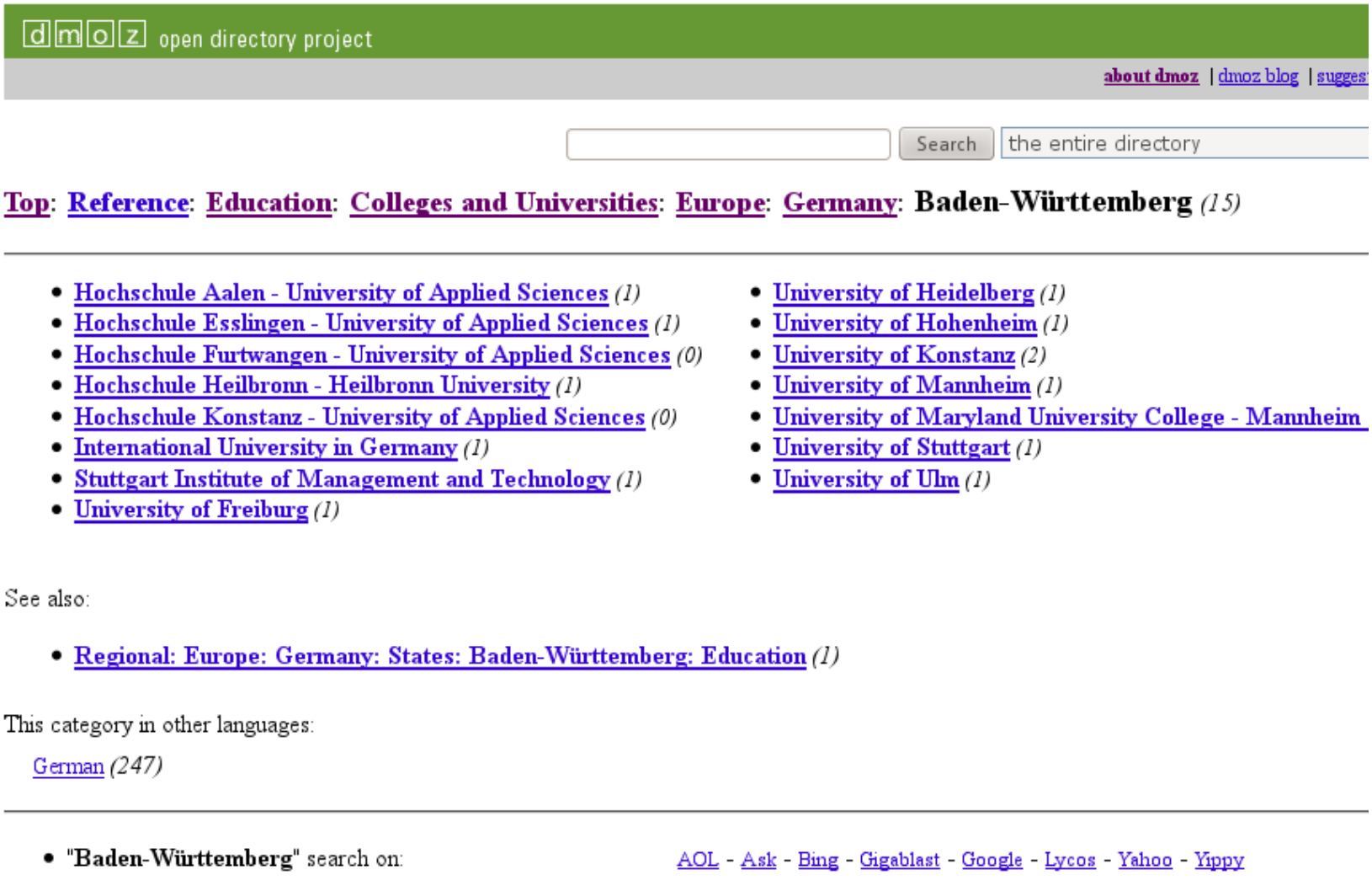

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Top: Reference: Education: Colleges and Universities: Europe: Germany: **Baden-Württemberg** *(15)*

- Hochschule Aalen - University of Applied Sciences *(1)*
- Hochschule Esslingen - University of Applied Sciences *(1)*
- Hochschule Furtwangen - University of Applied Sciences *(0)*
- Hochschule Heilbronn - Heilbronn University *(1)*
- Hochschule Konstanz - University of Applied Sciences *(0)*
- International University in Germany *(1)*
- Stuttgart Institute of Management and Technology *(1)*
- University of Freiburg *(1)*
- University of Heidelberg *(1)*
- University of Hohenheim *(1)*
- University of Konstanz *(2)*
- University of Mannheim *(1)*
- University of Maryland University College - Mannheim
- University of Stuttgart *(1)*
- University of Ulm *(1)*

See also:

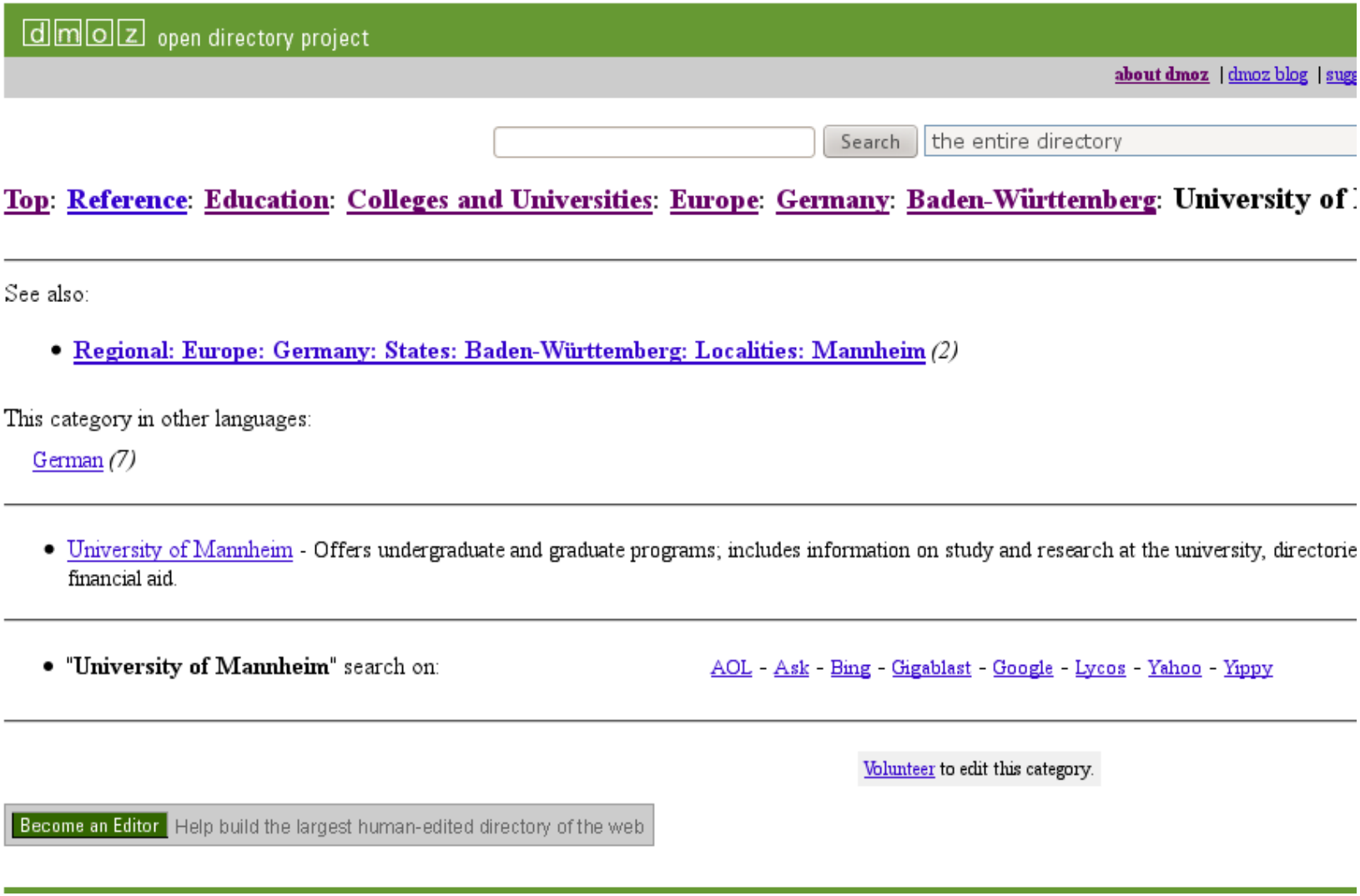
- Regional: Europe: Germany: States: Baden-Württemberg: Education *(1)*

This category in other languages:

German *(247)*

- "**Baden-Württemberg**" search on: AOL - Ask - Bing - Gigablast - Google - Lycos - Yahoo - Yippy

The Early Days: Web Directories



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Search | the entire directory

Top: Reference: Education: Colleges and Universities: Europe: Germany: Baden-Württemberg: University of

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See also:

- Regional: Europe: Germany: States: Baden-Württemberg: Localities: Mannheim *(2)*

This category in other languages:

German *(7)*

- University of Mannheim - Offers undergraduate and graduate programs; includes information on study and research at the university, directorie financial aid.

- "**University of Mannheim**" search on: AOL - Ask - Bing - Gigablast - Google - Lycos - Yahoo - Yippy

---

Volunteer to edit this category.

Become an Editor Help build the largest human-edited directory of the web

The Early Days: Web Directories

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Search: **uni mannheim**

Open Directory Categories (1-5 of 100)

1. **World: Français: Régional: Amérique: Etats-Unis: Etat et politique: Ambassades et consulats: Pays représentés** (39)
2. **World: Deutsch: Wissen: Bibliotheken: Universitäts- und Hochschulbibliotheken: Deutschland** (33)
3. **World: Français: Régional: Amérique: Etats-Unis: Etat et politique: Ambassades et consulats: Représentations à l'étranger** (24)
4. **World: Nederlands: Maatschappij: Overheid: Europese Unie** (22)
5. **World: Deutsch: Gesundheit: Krankenhäuser und Kliniken: Universitätskliniken: Deutschland** (21)

more...

Open Directory Sites (1-20 of 15698)

1. amnesty international Hochschulgruppe Mannheim - Die amnesty-Gruppe an der Uni Mannheim setzt sich unter anderem mit Petitionen, Appellbrief ein und stellt aktuelle Aktivitäten vor.
   -- *http://www.amnesty-uni-mannheim.de/* *World: Deutsch: Regional: Europa: Deutschland: Baden-Württemberg: Städte und Gemeinden: M: Mannheim: Gesellschaft (23)*

2. LHG - Liberale Hochschulgruppe Mannheim - Die liberale Studierenden an der Uni Mannheim stellen sich vor.
   -- *http://www.uni-mannheim.de/studorg/liberale/* *World: Deutsch: Wissen: Bildung: Hochschulen: Europa: Deutschland: Baden-Württemberg: Universität Mannheim ()*

3. Uni Mannheim - Marktübersicht deutscher Anbieter für Online-Preisvergleiche.
   -- *http://projekt.wifo.uni-mannheim.de/preisvergleich/* *World: Deutsch: Zuhause: Verbraucherinformationen: Preisagenturen: Online-Preisvergleiche ()*

4. University of Mannheim - Offers undergraduate and graduate programs; includes information on study and research at the university, directories of financial aid.
   -- *http://www.uni-mannheim.de/* *Reference: Education: Colleges and Universities: Europe: Germany: Baden-Württemberg: University of Mannheim ()*

The Early Days: Web Crawlers

- crawls through the web using hyperlinks
- makes an index of the words contained in a page
- ranks pages for a search query according to number of occurences of keywords

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Basics of PageRank

- Humans know what is good for them
- Humans create Websites
- Humans will only link to Websites they like

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How could we use that?

- Simply count number of links to a Website
- ✗ 10,000 links from only one page
- Count numbers of Websites that link to a Website
- ✗ Quality of the page matters
- ✗ Total number of links on the source page matters

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A brilliant idea

Sergey Brin

Larry Page

Ideas of PageRank

- Decisions of humans are complicated
- A lot of webpages get visited

$\Rightarrow$ modellize clicks on links as random behaviour

- Links are important
  - Links of page A get less important, if A has many links
  - Links of page A get more important, if many link to A

$\Rightarrow$ if B has a link from A, the rank of B increases by $\frac{Rank(A)}{Links(A)}$

if A links to B **then**

$$Rank(B) \mathrel{+}= \frac{Rank(A)}{Links(A)}$$

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Ants

- Websites = nodes = anthill
- Links = edges = paths
- You place ants on each node
- They walk over the paths
  (at random, they are ants!)
- After some time, some anthills will have more ants than others
- Those hills are more attractive than others
- # ants is probability that a random user would end on a website

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Mathematics

Let x be a web page. Then

- $L(x)$ is the set of Websites that link to x
- $C(y)$ is the out-degree of page y
- α is probability of random jump
- N is the total number of websites

$$PR(x) := \alpha \left(\frac{1}{N}\right) + (1-\alpha) \sum_{y \in L(x)} \frac{PR(y)}{C(y)}$$

Pseudocode

function PageRank(Graph web, double $q = 0.15$, int $iterations$)

```
for all page ∈ web do
    page.pageRank = 1/|web|                    ▷ intial probability
end for
while iterations > 0 do
    for all page ∈ web do                      ▷ calculate pageRank of page
        page.pageRank = q
        for all y ∈ L(page) do
            page.pageRank += y.pageRank / C(y)
        end for
    end for
    iterations -= 1
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Factors of Influence

- language
- place
- social information

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The Filter Bubble

dontbubble.us
www.thefilterbubble.com

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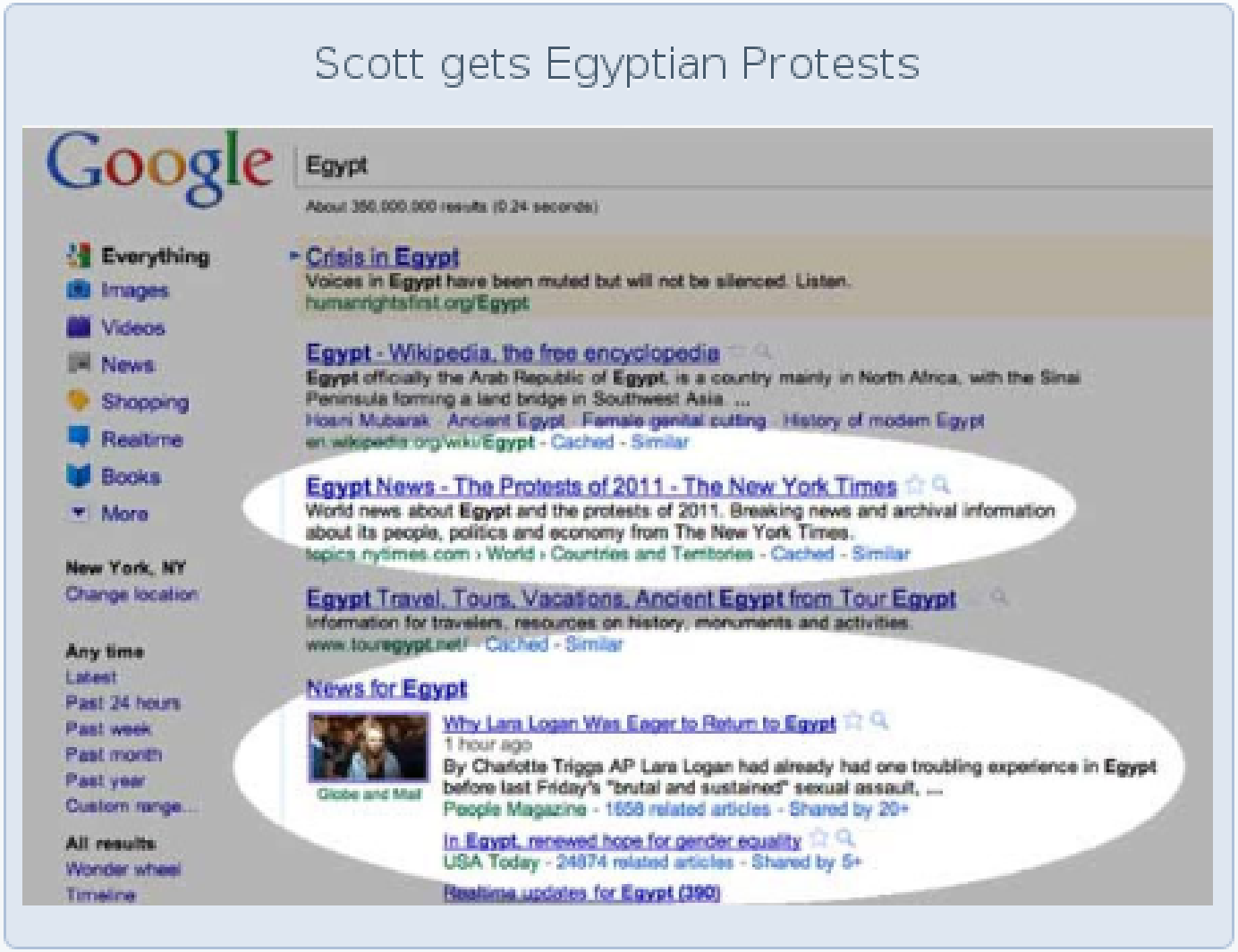

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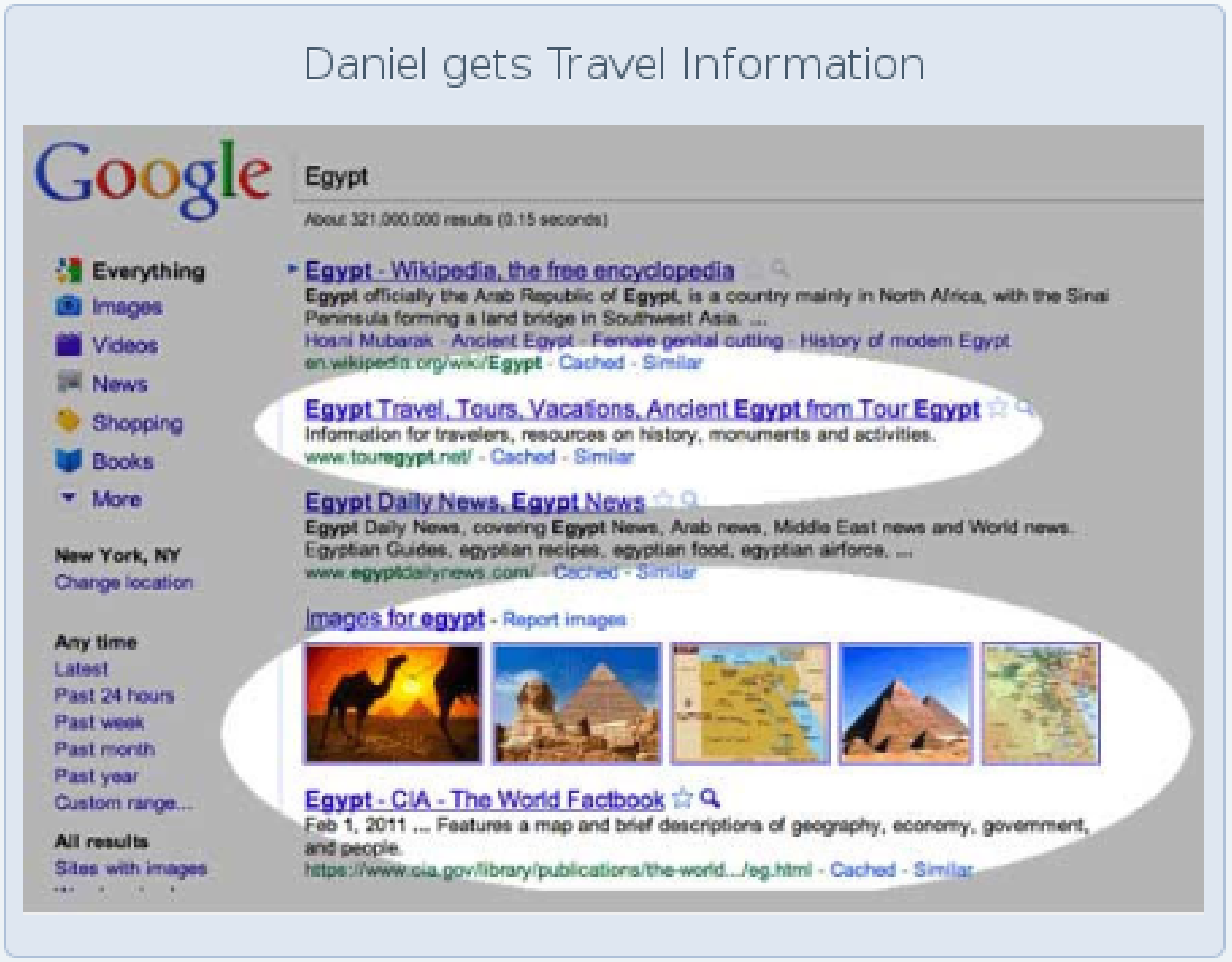

What You've Learned

- web directories
- web crawler
- graph (nodes, eges)
- random walk (ants)
- PageRank
- read pseudocode
- filter bubble

Image Sources

- PageRank by Felipe Micaroni Lalli
- screenshots of www.dmoz.org
- Hyperlink by Bernard Ladenthin
- screenshots of dontbubble.us
- Sergey Brin by enlewof
- Larry Page by aweigend

Thanks for Your Attention!