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# Query Operations

Relevance Feedback

Query Expansion

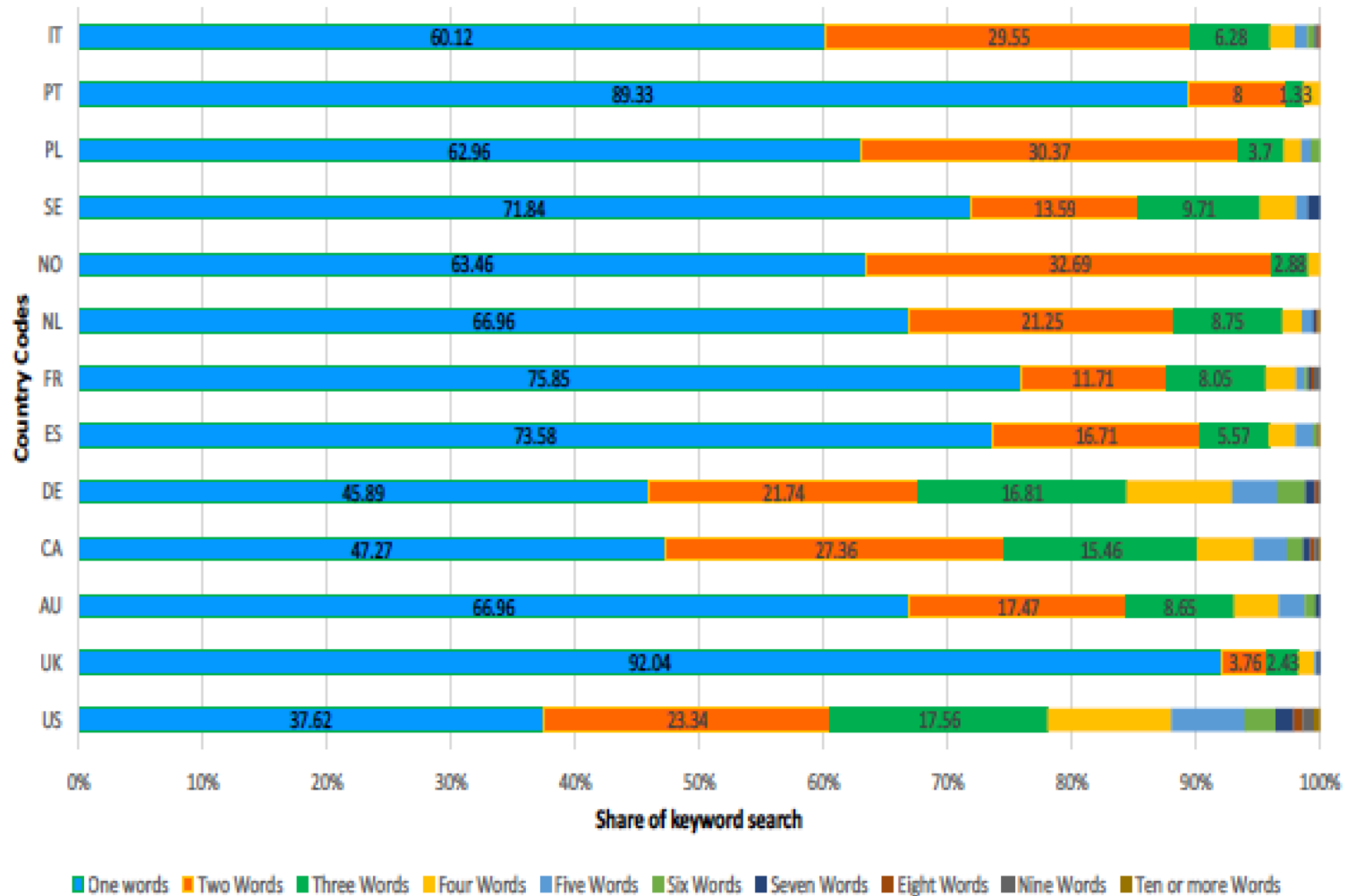
Query interpretation

# Relevance Feedback

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- After initial retrieval results are presented, allow the user to provide feedback on the relevance of one or more of the retrieved documents.
- Use this feedback information to reformulate the query.
- Produce new results based on reformulated query.
- Allows more interactive, multi-pass process.

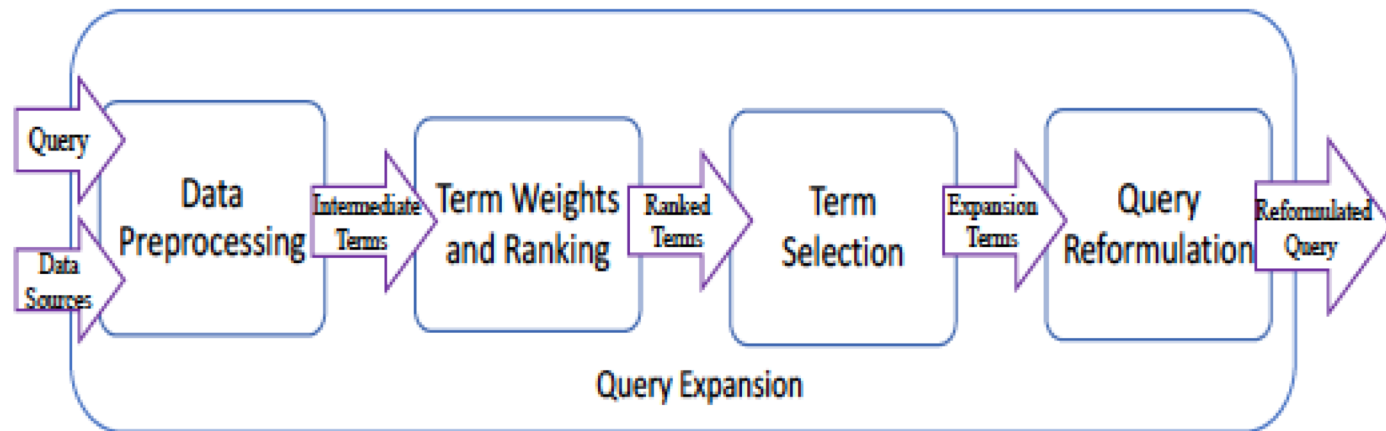
# Statistics on query dimension on web search engines by country



From: Azad Deepak 2017

# Workflow of query expansion

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From: Azad Deepak 2017



# Query expansion methods

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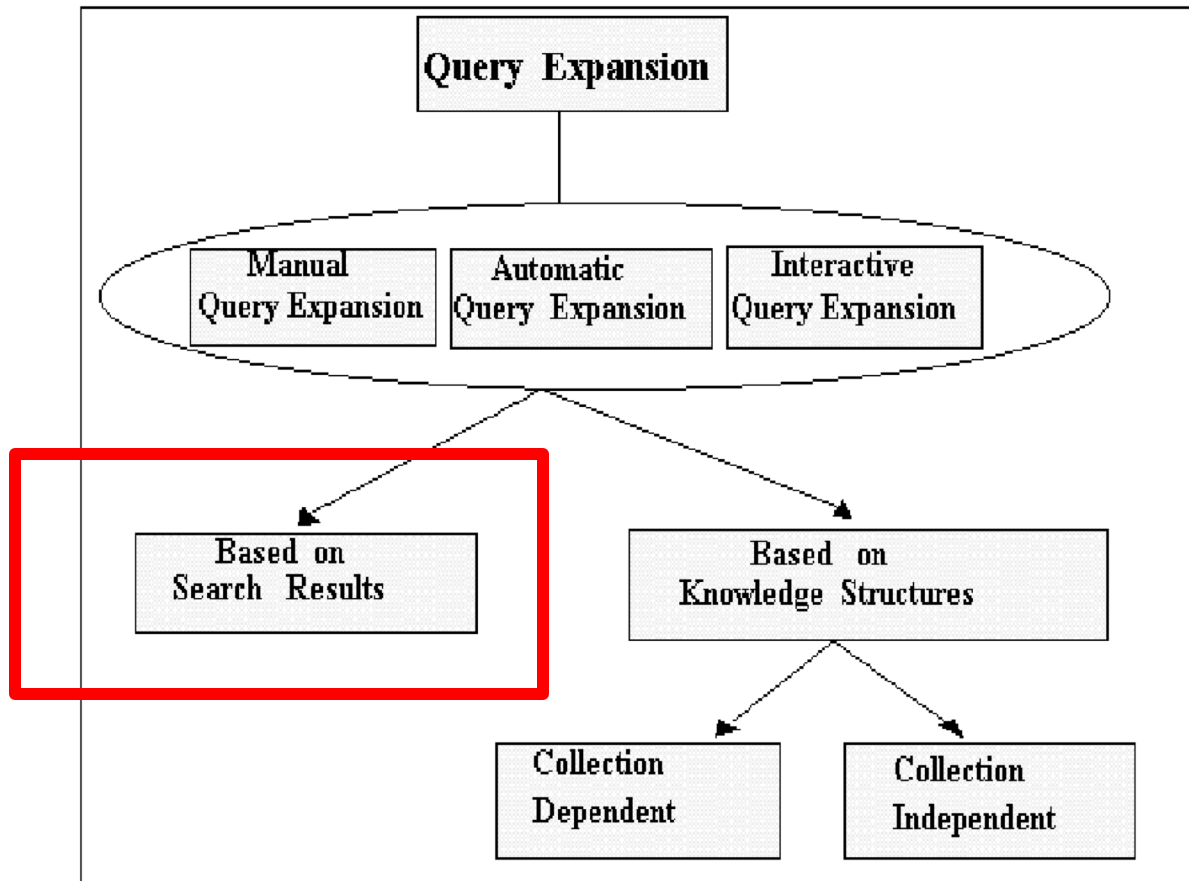
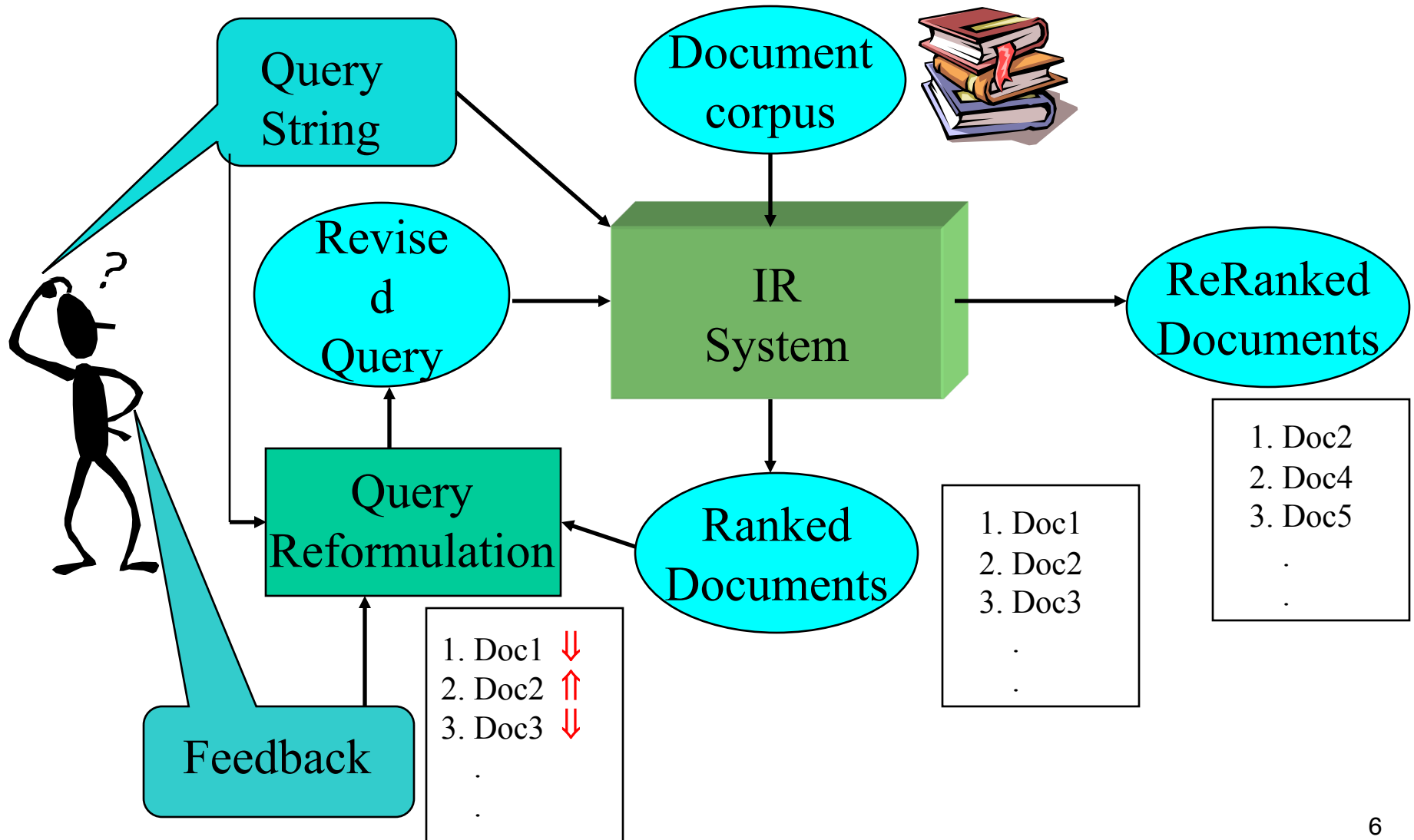


Figure 1: Query Expansion: Methods and Sources

# Relevance Feedback Architecture



# Query Reformulation

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- Revise query to account for feedback:
  - **Query Expansion**: Add new terms to query extracted from relevant documents.
  - **Term Re-weighting**: Increase weight of terms in relevant documents and decrease weight of terms in irrelevant documents.
- Several algorithms for query reformulation.

# Query Reformulation for VSR (vector space retrieval)

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- General idea: change query vector using vector algebra:
  - **Add** the vectors for the **relevant** documents to the query vector.
  - **Subtract** the vectors for the **irrelevant** docs from the query vector.
- This adds both positively and negatively weighted terms to the query as well as reweighting the initial terms.

# Optimal Query

- Assume that the relevant (to the user's query) set of documents  $C_r$  is known.
- Then the best query that ranks all and only the relevant documents at the top is:

$$\vec{q}_{opt} = \frac{1}{|C_r|} \sum_{\forall \vec{d}_j \in C_r} \vec{d}_j - \frac{1}{N - |C_r|} \sum_{\forall \vec{d}_j \notin C_r} \vec{d}_j$$

Where  $N$  is the total number of documents. The query vector **sum** the weights  $w_{ij}$  for all  $d_j$  in  $C_r$  and **subtracts** all the weights  $w'_{ik}$  for all  $d_k$  not in  $C_r$ .

# Example (query is «information retrieval»)

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- Vocabulary: (*information, method, performance, retrieval, system*)
- **D1: (1,0,1,1,0)** “*information retrieval performances*”
- **D2: (1,0,1,1,1)** “*performance of information retrieval systems*”
- **D3: (0,1,0,0,1)** “*system’s method*”
- **C<sub>r</sub>: D1, D2; N-C<sub>r</sub> = D3**

$$q_{opt} = \frac{1}{2} \{ (1,0,1,1,0) + (1,0,1,1,1) \} - \frac{1}{3-2} (0,1,0,0,1) =$$

$$\frac{1}{2} (2,0,2,2,1) - (0,1,0,0,1) = (1,0,1,1,0.5) - (0,1,0,0,1) =$$

$$(1, -1, 1, 1, -0.5)$$

# Standard Rocchio Method

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- Previous method is not realistic since all relevant documents are unknown. Rocchio method uses the **known** relevant ( $D_r$ ) and irrelevant ( $D_n$ ) sets (among the first k ranked) of documents and include them in initial query  $q$ .

$$\vec{q}_m = \alpha \vec{q} + \frac{\beta}{|D_r|} \sum_{\forall \vec{d}_j \in D_r} \vec{d}_j - \frac{\gamma}{|D_n|} \sum_{\forall \vec{d}_j \in D_n} \vec{d}_j$$

$\alpha$ : Tunable weight for initial query.

$\beta$ : Tunable weight for relevant documents.

$\gamma$ : Tunable weight for irrelevant documents.

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### Eclipse - The Eclipse Foundation open source community w...

<https://www.eclipse.org/> Traduci questa pagina

A project aiming to provide a universal toolset for development. Open Source IDE, mostly provided in Java, but the development language is independent and ...

#### Downloads

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#### About the Eclipse Foundation

The Eclipse Foundation is a not-for-profit, member supported ...

#### Eclipse Marketplace

Eclipse Marketplace is the source for Eclipse-based solutions ...

#### Documentation

Current releases. Eclipse Kepler (4.3) Documentation (HTML Help ...

#### Projects

List of Projects - Simultaneous Releases - Eclipse Project Tools

[Altri risultati in eclipse.org »](#)

### Eclipse (informatica) - Wikipedia

[it.wikipedia.org/wiki/Eclipse\\_\(informatica\)](it.wikipedia.org/wiki/Eclipse_(informatica))

**Eclipse** è un ambiente di sviluppo integrato multi-linguaggio e multiplatforma. Ideato da un consorzio di grandi società quali Ericsson, HP, IBM, Intel, ...

### The Twilight Saga: Eclipse - Wikipedia

[it.wikipedia.org/wiki/The\\_Twilight\\_Saga:\\_Eclipse](it.wikipedia.org/wiki/The_Twilight_Saga:_Eclipse)

The Twilight Saga: **Eclipse** è un film del 2010 diretto da David Slade. Sceneggiato da Melissa Rosenberg, è il terzo film tratto dalla serie di Twilight. La pellicola ...

### Eclipse - Wikipedia

<it.wikipedia.org/wiki/Eclipse>

Questa è una pagina di disambiguazione; se sei giunto qui cliccando un collegamento, puoi tornare indietro e correggerlo, indirizzandolo direttamente alla voce ...

### Eclipse (software) - Wikipedia, the free encyclopedia

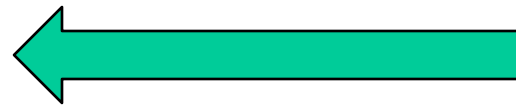
[en.wikipedia.org/wiki/Eclipse\\_\(software\)](en.wikipedia.org/wiki/Eclipse_(software)) Traduci questa pagina

In computer programming, **Eclipse** is an integrated development environment (IDE). It contains a base workspace and an extensible plug-in system for ...

Query is: *eclipse saga*

**D<sub>r</sub>**: *saga, movie, director, david slade, licanthropus, melissa rosenberg..*

**D<sub>n</sub>**: *foundation, software, development, tool, environment....*





# Ide Regular Method

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- Since more feedback should perhaps increase the degree of reformulation, do not normalize :

$$\vec{q}_m = \alpha \vec{q} + \beta \sum_{\forall \vec{d}_j \in D_r} \vec{d}_j - \gamma \sum_{\forall \vec{d}_j \in D_n} \vec{d}_j$$

$\alpha$ : Tunable weight for initial query.

$\beta$ : Tunable weight for relevant documents.

$\gamma$ : Tunable weight for irrelevant documents.

# Example

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- $q (w_{1q}, w_{2q}, w_{3q}, w_{4q})$
- Dr:  $[d_1(w_{11}, w_{21}, w_{31}, w_{41}), d_2(w_{12}, w_{22}, w_{32}, w_{42})]$
- Dn:  $d_3(w_{13}, w_{23}, w_{33}, w_{43})$
- $q_{\text{exp}}: ((\alpha w_{1q} + \beta(w_{11} + w_{12}) - \gamma w_{13}), (\alpha w_{2q} + \beta(w_{21} + w_{22}) - \gamma w_{23}), (\alpha w_{3q} + \beta(w_{31} + w_{32}) - \gamma w_{33}), (\alpha w_{4q} + \beta(w_{41} + w_{42}) - \gamma w_{43}))$

# Ide “Dec Hi” Method

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- Bias towards rejecting **just** the highest ranked of the irrelevant documents:

$$\vec{q}_m = \alpha \vec{q} + \beta \sum_{\forall \vec{d}_j \in D_r} \vec{d}_j - \gamma \max_{non-relevant} (\vec{d}_j)$$

$\alpha$ : Tunable weight for initial query.

$\beta$ : Tunable weight for relevant documents.

$\gamma$ : Tunable weight for irrelevant document.

# Comparison of Methods

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- Overall, experimental results indicate no clear preference for any one of the specific methods.
- All methods generally improve retrieval performance (recall & precision) with feedback.
- Generally tunable constants  $\alpha$ ,  $\beta$ ,  $\gamma$  equal 1.

# Example

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- Initial query: “*New space satellite applications*”
  - + 1. 0.539, 08/13/91, NASA Hasn’ t Scrapped Imaging Spectrometer
  - + 2. 0.533, 07/09/91, NASA Scratches Environment Gear From Satellite Plan
  - 3. 0.528, 04/04/90, Science Panel Backs NASA Satellite Plan, But Urges Launches of Smaller Probes
  - 4. 0.526, 09/09/91, A NASA Satellite Project Accomplishes Incredible Feat: Staying Within Budget
  - 5. 0.525, 07/24/90, Scientist Who Exposed Global Warming Proposes Satellites for Climate Research
  - 6. 0.524, 08/22/90, Report Provides Support for the Critics Of Using Big Satellites to Study Climate
  - 7. 0.516, 04/13/87, Arianespace Receives Satellite Launch Pact From Telesat Canada
  - + 8. 0.509, 12/02/87, Telecommunications Tale of Two Companies
- User marks relevant documents with “+”.

## Expanded query after relevance feedback

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- 2.074 new
- 30.816 satellite
- 5.991 nasa
- 4.196 launch
- 3.516 instrument
- 3.004 bundespost
- 2.790 rocket
- 2.003 broadcast
- 0.836 oil
- 15.106 space
- 5.660 application
- 5.196 eos
- 3.972 aster
- 3.446 arianespace
- 2.806 ss
- 2.053 scientist
- 1.172 earth
- 0.646 measure

# Results for expanded query

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- 2 1. 0.513, 07/09/91, [NASA Scratches Environment Gear From Satellite Plan](#)
- 1 2. 0.500, 08/13/91, [NASA Hasn't Scrapped Imaging Spectrometer](#)
3. 0.493, 08/07/89, [When the Pentagon Launches a Secret Satellite, Space Sleuths Do Some Spy Work of Their Own](#)
4. 0.493, 07/31/89, [NASA Uses 'Warm' Superconductors For Fast Circuit](#)
- 8 5. 0.492, 12/02/87, [Telecommunications Tale of Two Companies](#)
6. 0.491, 07/09/91, [Soviets May Adapt Parts of SS-20 Missile For Commercial Use](#)
7. 0.490, 07/12/88, [Gaping Gap: Pentagon Lags in Race To Match the Soviets In Rocket Launchers](#)
8. 0.490, 06/14/90, [Rescue of Satellite By Space Agency To Cost \\$90 Million](#)

# Relevance Feedback on the Web

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- Some search engines offer a *similar/related* pages feature (this is a trivial form of relevance feedback)
  - Google (link-based, but is now hidden). It rather shows “related search”
  - But some don’t because it’s hard to explain to average user why a page is suggested
- Specialized search engines are those who more often use feedback



# Why is Feedback Not Widely Used

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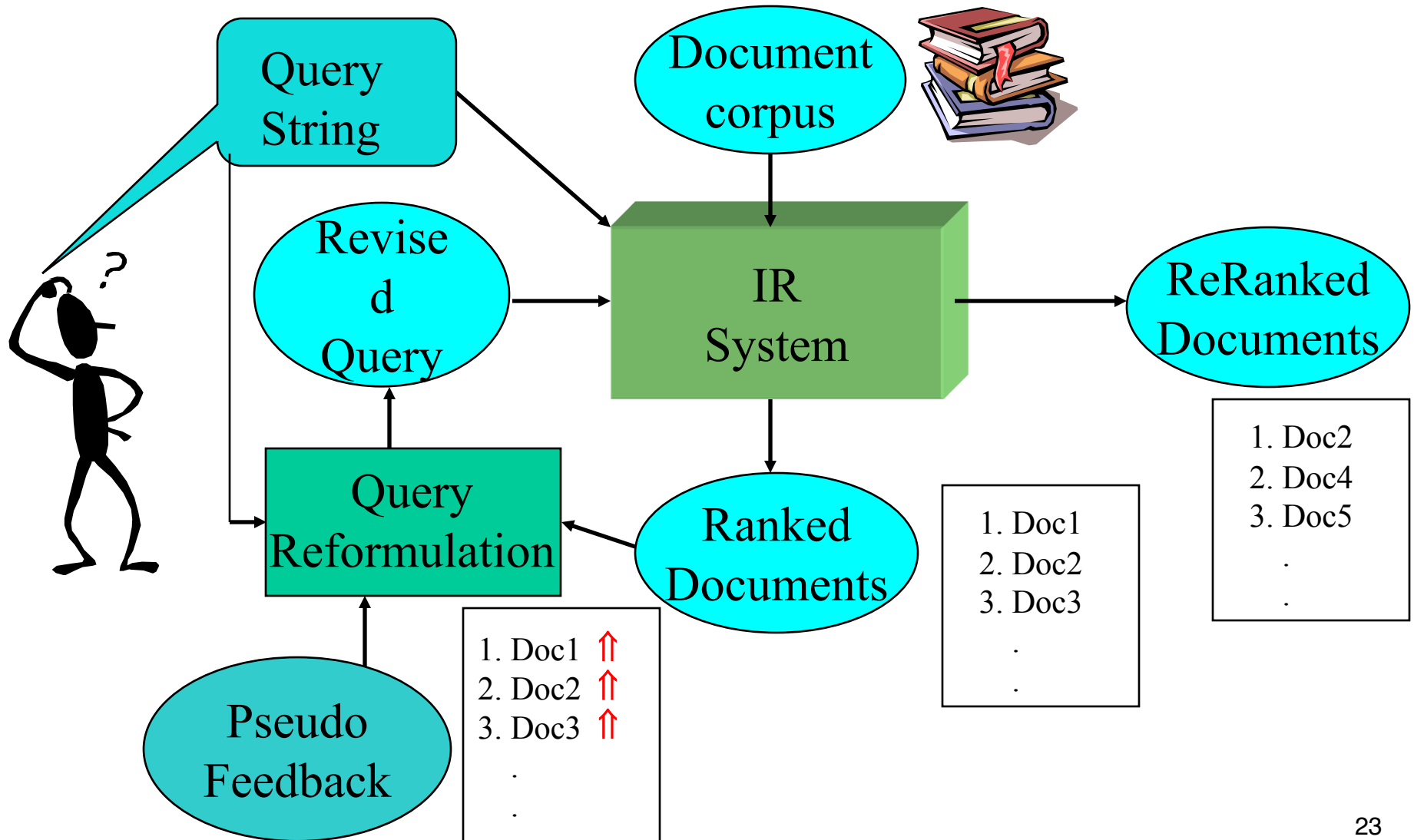
- Users sometimes reluctant to provide explicit feedback.
- Results in long queries that require more computation to retrieve documents: search engines process lots of queries and allow little time for each one.
- Makes it harder to understand why a particular document was retrieved.

# Pseudo Feedback

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- Use relevance feedback methods **without explicit user input**.
- Just **assume** the top  $m$  retrieved documents are relevant, and use them to reformulate the query.
- Allows for query expansion that includes terms that are correlated with the query terms.
- Would not work well for previous “Eclypse” example but common queries are less ambiguous,
- E.g. Eclypse licanthropous, Eclypse moon

# Pseudo Feedback Architecture



# PseudoFeedback Results

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- Found to improve performance on public IR competitions ( ad-hoc retrieval task).
- Works even better if top documents must also satisfy additional boolean constraints in order to be used in feedback (especially negative constraints like *eclipse AND (licantropus OR not moon)*).

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# **OTHER METHODS FOR QUERY EXPANSION**

# Query expansion methods

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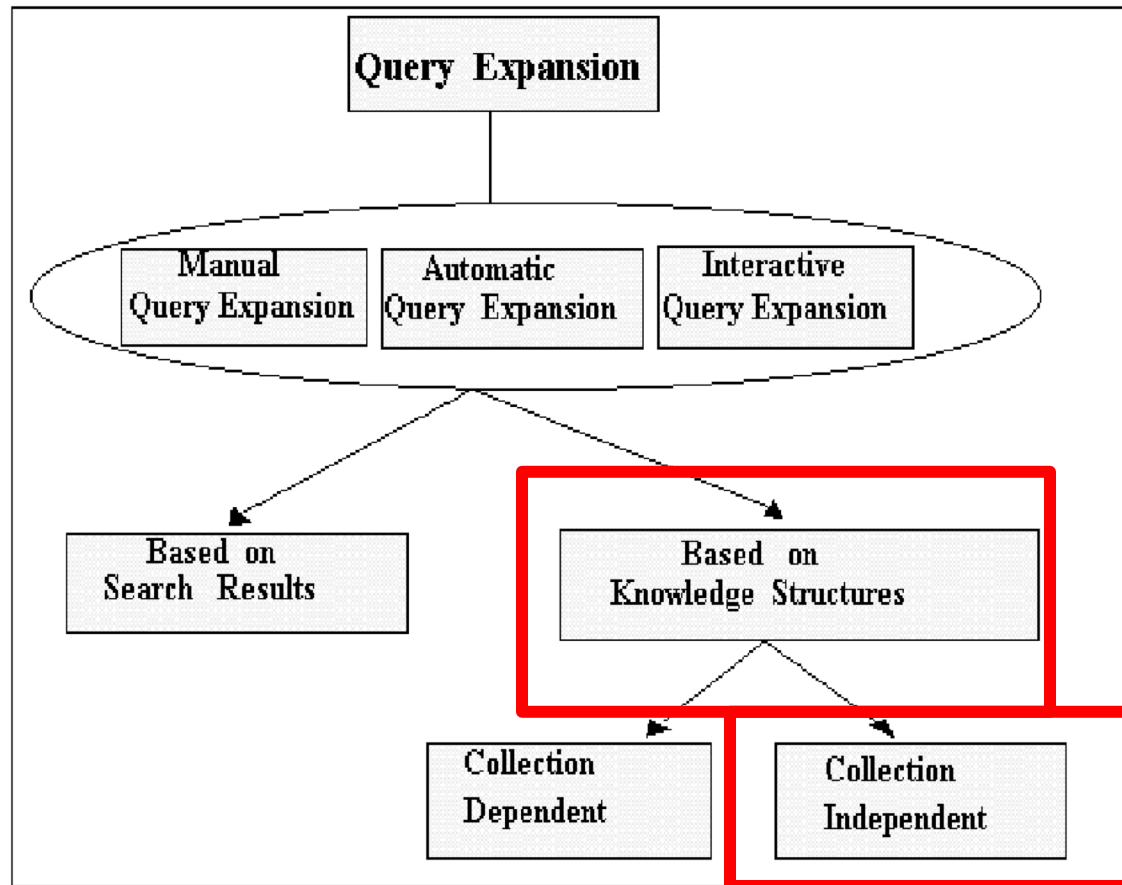


Figure 1: Query Expansion: Methods and Sources

# Thesaurus

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- A thesaurus provides information on synonyms and semantically related words and phrases.
- Example:

physician

syn: ||croaker, doc, doctor, MD,  
medical, mediciner, medico, ||sawbones

rel: medic, general practitioner,  
surgeon,

# Thesaurus-based Query Expansion

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- For each term,  $t$ , in a query, expand the query with synonyms and related words of  $t$  from the thesaurus.
- Can weight added terms less than original query terms (= *discount factor* for terms not in original query).
- Generally increases recall.
- May significantly decrease precision, particularly with ambiguous terms.
  - “interest rate” → “interest rate fascinate evaluate”



# WordNet

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- A more detailed database of semantic relationships between English words.
- Developed by famous cognitive psychologist George Miller and a team at Princeton University.
- About 144,000 English words.
- Nouns, adjectives, verbs, and adverbs grouped into about 109,000 synonym sets called *synsets*.

# Wordnet

## WordNet Search - 3.1

- [WordNet home page](#) - [Glossary](#) - [Help](#)

Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations  
Display options for sense: (gloss) "an example sentence"

### Noun

- [S:](#) (n) **Moon**, **moon** (the natural satellite of the Earth) "*the average distance to the Moon is 384,400 kilometers*"; "*men first stepped on the moon in 1969*"
- [S:](#) (n) **moon** (any object resembling a moon) "*he made a moon lamp that he used as a night light*"; "*the clock had a moon that showed various phases*"
- [S:](#) (n) [lunar month](#), **moon**, [lunation](#), [synodic month](#) (the period between successive new moons (29.531 days))
- [S:](#) (n) [moonlight](#), [moonshine](#), **Moon** (the light of the Moon) "*moonlight is the smuggler's enemy*"; "*the Moon was bright enough to read by*"
- [S:](#) (n) **Moon**, [Sun Myung Moon](#) (United States religious leader (born in Korea) who founded the Unification Church in 1954; was found guilty of conspiracy to evade taxes (born in 1920))
- [S:](#) (n) **moon** (any natural satellite of a planet) "*Jupiter has sixteen moons*"

### Verb

- [S:](#) (v) [daydream](#), **moon** (have dreamlike musings or fantasies while awake) "*She looked out the window, daydreaming*"
- [S:](#) (v) **moon**, [moon around](#), [moon on](#) (be idle in a listless or dreamy way)
- [S:](#) (v) **moon** (expose one's buttocks to) "*moon the audience*"

Word to search for:  Search WordNet

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

## Noun

- **S: (n) car**, [auto](#), [automobile](#), [machine](#), [motorcar](#) (a motor vehicle with four wheels; usually propelled by an internal combustion engine) *"he needs a car to get to work"*
  - [direct hyponym](#) / [full hyponym](#)
  - [part meronym](#)
  - [domain term category](#)
  - [direct hypernym](#) / [inherited hypernym](#) / [sister term](#)
    - **S: (n) motor vehicle**, [automotive vehicle](#) (a self-propelled wheeled vehicle that does not run on rails)
      - **S: (n) self-propelled vehicle** (a wheeled vehicle that carries in itself a means of propulsion)
        - **S: (n) wheeled vehicle** (a vehicle that moves on wheels and usually has a container for transporting things or people) *"the oldest known wheeled vehicles were found in Sumer and Syria and date from around 3500 BC"*
          - **S: (n) vehicle** (a conveyance that transports people or objects)
            - **S: (n) conveyance**, [transport](#) (something that serves as a means of transportation)
              - **S: (n) instrumentality**, [instrumentation](#) (an artifact (or system of artifacts) that is instrumental in accomplishing some end)
                - **S: (n) artifact**, [artefact](#) (a man-made object taken as a whole)
                  - **S: (n) whole**, [unit](#) (an assemblage of parts that is regarded as a single entity) *"how big is that part compared to the whole?"; "the team is a unit"*
                    - **S: (n) object**, [physical object](#) (a tangible and visible entity; an entity that can cast a shadow) *"it was full of rackets, balls and other objects"*

# Wordnet hierarchy

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# WordNet Synset Relationships

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- **Antonym**: front → back
- **Attribute**: benevolence → good (noun to adjective)
- **Pertainym**: alphabetical → alphabet (adjective to noun)
- **Similar**: unquestioning → absolute
- **Cause**: kill → die
- **Entailment**: breathe → inhale
- **Holonym**: chapter → text (part-of)
- **Meronym**: computer → cpu (whole-of)
- **Hyponym**: plant → tree (specialization)
- **Hypernym**: apple → fruit (generalization)

# WordNet Query Expansion

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- Add synonyms in the same synset.
- Add hyponyms to add specialized terms.
- Add hypernyms to generalize a query.
- Add other related terms to expand query.
- In case of ambiguity, which synset?

Example query: **car rental**

Expanded query: (car OR automonile OR machine OR ..)AND  
(rental OR leasing OR..)

# Not all senses available

## WordNet Search - 3.1

- [WordNet home page](#) - [Glossary](#) - [Help](#)

Word to search for:

Display Options:

Key: "S:" = Show Synset (semantic) relations, "W:" = Show Word (lexical) relations

Display options for sense: (gloss) "an example sentence"

### Noun

Computer sense of apple is missing

- **S: (n) apple** (fruit with red or yellow or green skin and sweet to tart crisp whitish flesh)
  - **direct hyponym / full hyponym**
    - **S: (n) crab apple, crabapple** (small sour apple; suitable for preserving) *"crabapples make a tangy jelly"*
    - **S: (n) eating apple, dessert apple** (an apple used primarily for eating raw without cooking)
    - **S: (n) cooking apple** (an apple used primarily in cooking for pies and applesauce etc)
  - **direct hypernym / inherited hypernym / sister term**
  - **part holonym**
- **S: (n) apple, orchard apple tree, Malus pumila** (native Eurasian tree widely cultivated in many varieties for its firm rounded edible fruits)

# A better source: Wikipedia (disambiguation page)

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## Apple (disambiguation)

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From Wikipedia, the free encyclopedia

The **apple** is the pomaceous edible fruit of a temperate-zone deciduous tree.

**Apple** or **apples** may also refer to:

### Plants and plant parts

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- *Malus*, the genus of all apples and crabapples
- *Cashew apple*, the fruit that grows with the cashew nut
- Several fruits called *Custard apple*
- Love apple
  - Tomato
  - *Syzygium samarangense*
- Plants called *Mammee apple*
- May apple, *Podophyllum peltatum*
- *Oak apple*, a type of gall that grows on oak trees
- Several fruits called *rose apple*
- Thorn apple:
  - *Crataegus* species
  - *Datura* species
- Wax apple, *Syzygium samarangense*

### Companies

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- *Apple Corps*, a multimedia corporation founded in the 1960s by The Beatles
- *Apple Inc.*, a consumer electronics and software company founded in the 1970s
- *Apple Bank*, an American bank in the New York City area

### Films

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- *The Apple (1980 film)*, a 1980 musical science fiction film
- *The Apple (1998 film)*, by Samira Makhmalbaf

### Television

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- "The Apple" (*Star Trek: The Original Series*), a 1967 second season episode

### Music

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- *Apple (album)*, a 1990 album by Mother Love Bone

# An even better source: BabelNet



BabelNet

• Noun

apple ENGLISH TRANSLATE INTO... SEARCH

All Concepts Named Entities    + more domains 17 results

## Noun



EN apple • apple blossom • apple peel

Fruit with red or yellow or green skin and sweet to tart crisp whitish flesh

bn:00005054n | Concept



EN apple • Malus pumila • orchard apple tree

Native Eurasian tree widely cultivated in many varieties for its firm rounded edible fruits

bn:00005055n | Concept



EN Apple Inc. • Apple • Apple Computer

Apple Inc. is an American multinational technology company headquartered in Cupertino, California that design develops, and sells consumer electronics, computer software, and online services.

bn:03739345n | Named Entity



EN Apple Store • Apple Retail Store • Apple (store)

Apple Store is a chain of retail stores owned and operated by Apple Inc.

bn:03283215n | Named Entity

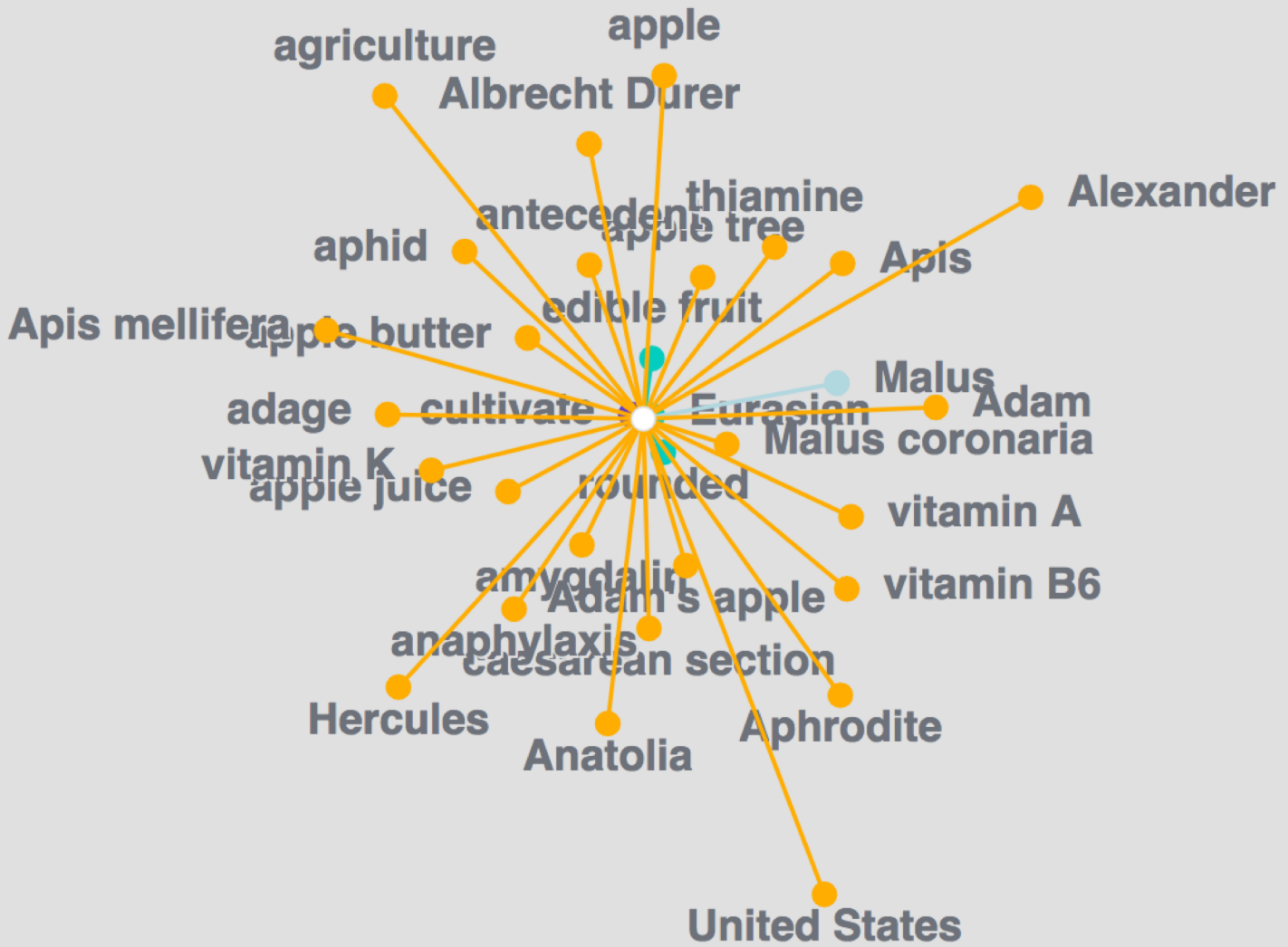


EN Yabloko • Russian United Democratic Party • Apple (Russia)

The Russian United Democratic Party "Yabloko" is a Russian social-liberal political party founded by Grigory Yavlinsky and currently led by Emilia Slabunova.

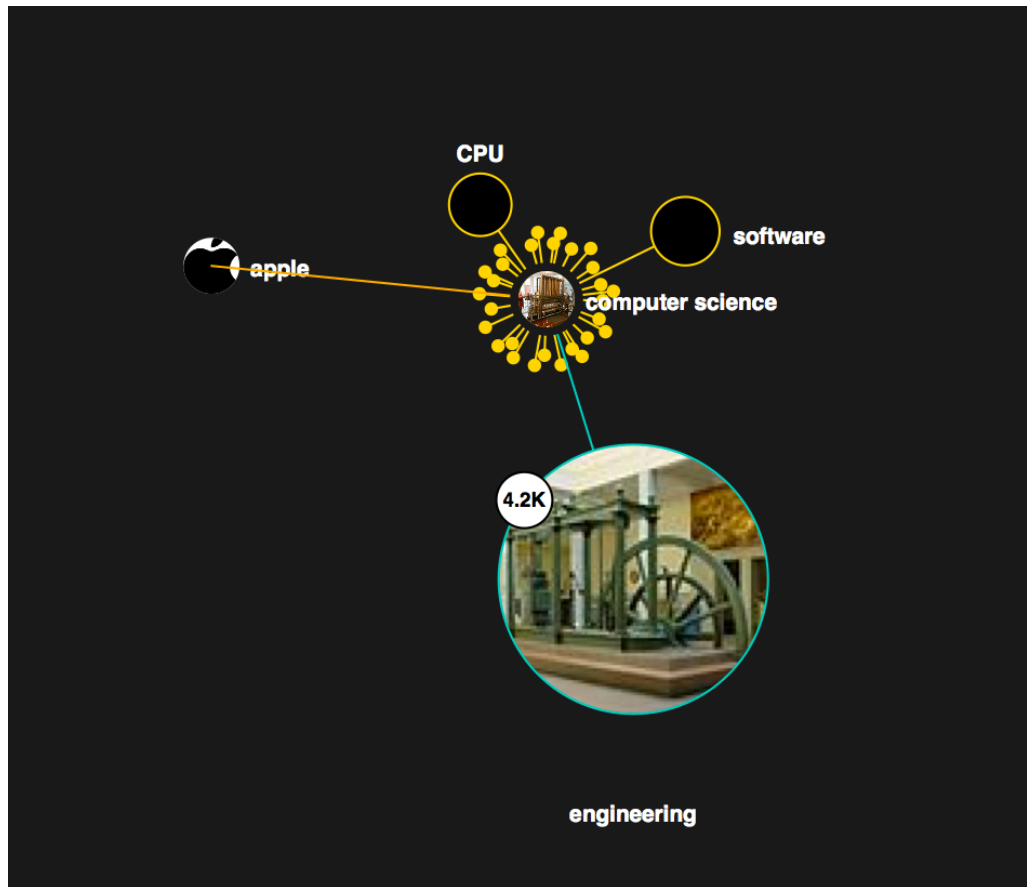






# Use context in query to disambiguate (e.g. “apple computer”) or click on right sense

**Apple** Inc. is an American multinational corporation that designs and sells consumer electronics, **computer** software, and personal computers.



Explore: apple computer

apple, CPU, engineering , computer science, software,

# Query expansion methods

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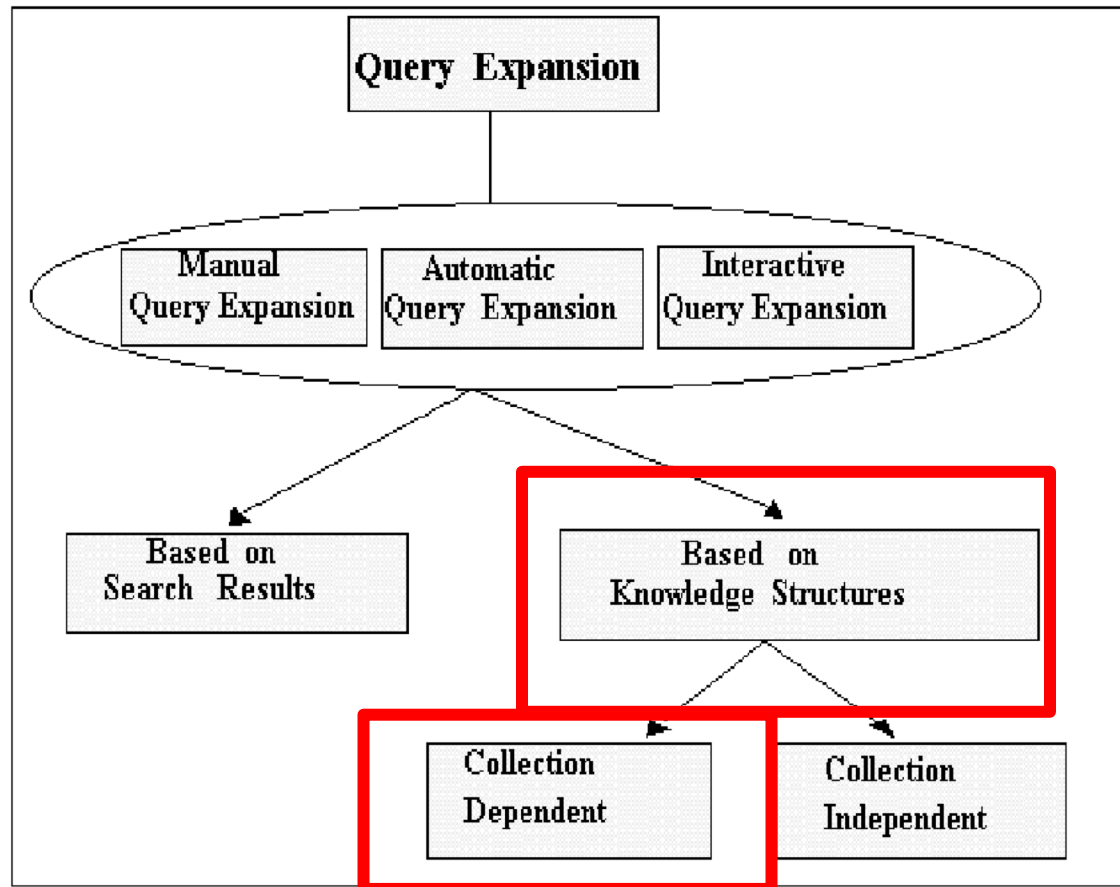


Figure 1: Query Expansion: Methods and Sources

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# STATISTICAL QUERY EXPANSION

# Statistical Expansion

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- Existing human-developed thesauri are not easily available in all languages (even though now BabelNet has 100 languages).
- More importantly: *Semantically related terms can be more easily discovered from statistical analysis of corpora.*
- E.g. “licantrope” and “eclipse” **may not co-occur in thesauri**: “*The Twilight Saga: Eclipse, commonly referred to as Eclipse, is a 2010 American romantic fantasy film based on Stephenie Meyer's 2007 novel, Eclipse*” but they **do co-occur in texts (more free texts available than thesauri..)**

# Automatic Global Analysis

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- Determine term similarity through a **pre-computed statistical analysis** of the complete corpus.
- Compute association matrices which quantify term correlations in terms of how frequently they co-occur.
- Expand queries with statistically most similar terms.

# Association Matrix

|       | $W_1$    | $W_2$    | $W_3$    | ..... | $W_n$    |
|-------|----------|----------|----------|-------|----------|
| $W_1$ | $c_{11}$ | $c_{12}$ | $c_{13}$ | ..... | $c_{1n}$ |
| $W_2$ | $c_{21}$ |          |          |       |          |
| $W_3$ | $c_{31}$ |          |          |       |          |
| .     | .        |          |          |       |          |
| .     | .        |          |          |       |          |
| $W_n$ | $c_{n1}$ |          |          |       |          |

$c_{ij}$ : Correlation factor between term  $i$  and term  $j$

$$c_{ij} = \sum_{d_k \in D} f_{ik} \times f_{jk}$$

$c_{ij}=0$  if either  $i$  or  $j$  do not occur in  $d_k$   
 $c_{ii}$ = sum of quadratic frequencies

$f_{ik}$ : Frequency of term  $i$  in document  $k$

# Normalized Association Matrix

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- Frequency based correlation factor favors more frequent terms: need to discriminate chance (they co-occur because they are very frequent) from genuine relatedness
- **Normalize** association scores:

$$s_{ij} = \frac{c_{ij}}{c_{ii} + c_{jj} - c_{ij}}$$

**Numerator:** SUM(product of  $i$ - $j$  frequencies)

**Denominator:** SUM(frequency of  $i$ )<sup>2</sup> + SUM(frequency of  $j$ )<sup>2</sup> – numerator

- Normalized score is 1 if two terms have the same frequency in all documents.



## Example (assuming freq=1 or 0 in all docs)

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- Documents with “information” : 5500
- Documents with “retrieval” : 2600
- Documents with both: 2500

$$\frac{2500}{5500 + 2600 - 2500} = \frac{2500}{5600} = 0,45$$

# Metric Correlation Matrix

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- Association correlation does not account for the proximity of terms in documents, just co-occurrence **frequencies** within documents.
- Metric correlations account for term **proximity**.

$$c_{ij} = \sum_{k_u \in V_i} \sum_{k_v \in V_j} \frac{1}{r(k_u, k_v)}$$

$V_i$ : Set of all occurrences of term  $i$  in any document.

$r(k_u, k_v)$ : Distance in words between word occurrences  $k_u$  and  $k_v$   
( $\infty$  if  $k_u$  and  $k_v$  are occurrences in different documents).

# Normalized Metric Correlation Matrix

---

- Normalize scores to account for term frequencies:

$$s_{ij} = \frac{c_{ij}}{|V_i| \times |V_j|} = \sum_{k_u \in V_i} \sum_{k_v \in V_j} \frac{1}{r(k_u, k_v)} / (|V_i| \times |V_j|)$$

$V_i$ ,  $V_j$  are the subset of documents in the collection including term  $i$  or term  $j$

# Query Expansion with Correlation Matrix

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- For each term  $i$  in query, expand query with the  $n$  terms,  $j$ , with the highest value of  $c_{ij}$  ( $s_{ij}$ ).
- **This adds related terms found in the “neighborhood” of the query terms.**

# Co-occurrence table

## Example

| word        | ten nearest neighbors                         |
|-------------|---|
| absolutely  | absurd whatsoever totally exactly nothing     |
| bottomed    | dip copper drops topped slide trimmed slight  |
| captivating | shimmer stunningly superbly plucky witty      |
| doghouse    | dog porch crawling beside downstairs gazed    |
| Makeup      | repellent lotion glossy sunscreen Skin gel p  |
| mediating   | reconciliation negotiate cease conciliation p |
| keeping     | hoping bring wiping could some would othe     |
| lithographs | drawings Picasso Dali sculptures Gauguin l    |
| pathogens   | toxins bacteria organisms bacterial parasite  |
| senses      | grasp psyche truly clumsy naive innate awl    |

# Problems with Global Analysis

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- Term ambiguity may introduce irrelevant statistically correlated terms.
  - “Apple computer” → “Apple red fruit computer”
- Since terms are highly correlated anyway, expansion may not retrieve many additional documents.

# Automatic Local Analysis

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- At query time, dynamically determine similar terms based on analysis of top-ranked retrieved documents.
- Base correlation analysis on only the “local” set of retrieved documents for a specific query.
- Avoids ambiguity by determining similar (correlated) terms only within relevant documents.
  - “Apple computer” →  
“Apple computer Powerbook laptop”

# Example (apple computer)

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[Apple Computer - Get great deals for Apple Computer on eBay!](#)

[popular.ebay.com/computers.../apple...](#) - [Stati Uniti](#) - Traduci questa pagina

The **Apple Computer** Co. began in the 1970s with the production of the behemoth Apple II microcomputer. Based in Cupertino, CA, in the heart of Silicon Valley, ...

[AAPL Stock Price Today - Apple Inc. Stock Quote - WSJ.com](#)

[quotes.wsj.com/AAPL](#) - Traduci questa pagina

**Apple Inc.** AAPL (U.S.: Nasdaq). Help. Real-time prices for U.S.-listed stocks, including premarket and after hours, reflect trading through Nasdaq only.

[AAPL: Summary for Apple Inc.- Yahoo! Finance](#)

[finance.yahoo.com/q?s=AAPL](#) - [Stati Uniti](#) - Traduci questa pagina

View the basic AAPL stock chart on Yahoo! Finance. Change the date range, chart type and compare **Apple Inc.** against other companies.

[Apple Inc.: NASDAQ:AAPL quotes & news - Google Finance](#)

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Get detailed financial information on **Apple Inc.** (NASDAQ:AAPL) including real-time stock quotes, historical charts & financial news, all for free!

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# Global vs. Local Analysis

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- Global analysis requires intensive term correlation computation **only occasionally**.
- Local analysis requires intensive term correlation computation for every query at **run time** (although number of terms and documents is less than in global analysis).
- But local analysis gives better results.

# Global Analysis Refinements

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- Only expand query with terms that are similar to ***all terms in the query.***

$$\text{sim}(k_i, Q) = \sum_{k_j \in Q} c_{ij}$$

- “fruit” not added to “Apple computer” since it is far from “computer.”
- “fruit” added to “apple pie” since “fruit” close to both “apple” and “pie.”
- Use more sophisticated term weights (instead of just frequency) when computing term correlations.

# Query Expansion with co-occurrences:

## Conclusions

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- Expansion of queries with related terms can improve performance, **particularly recall** (more terms=more documents with same rank threshold).
- However, must select similar terms very carefully to avoid problems, such as loss of **precision** (e.g. if unrelated terms are added, precision might considerably decrease).

# Query expansion methods

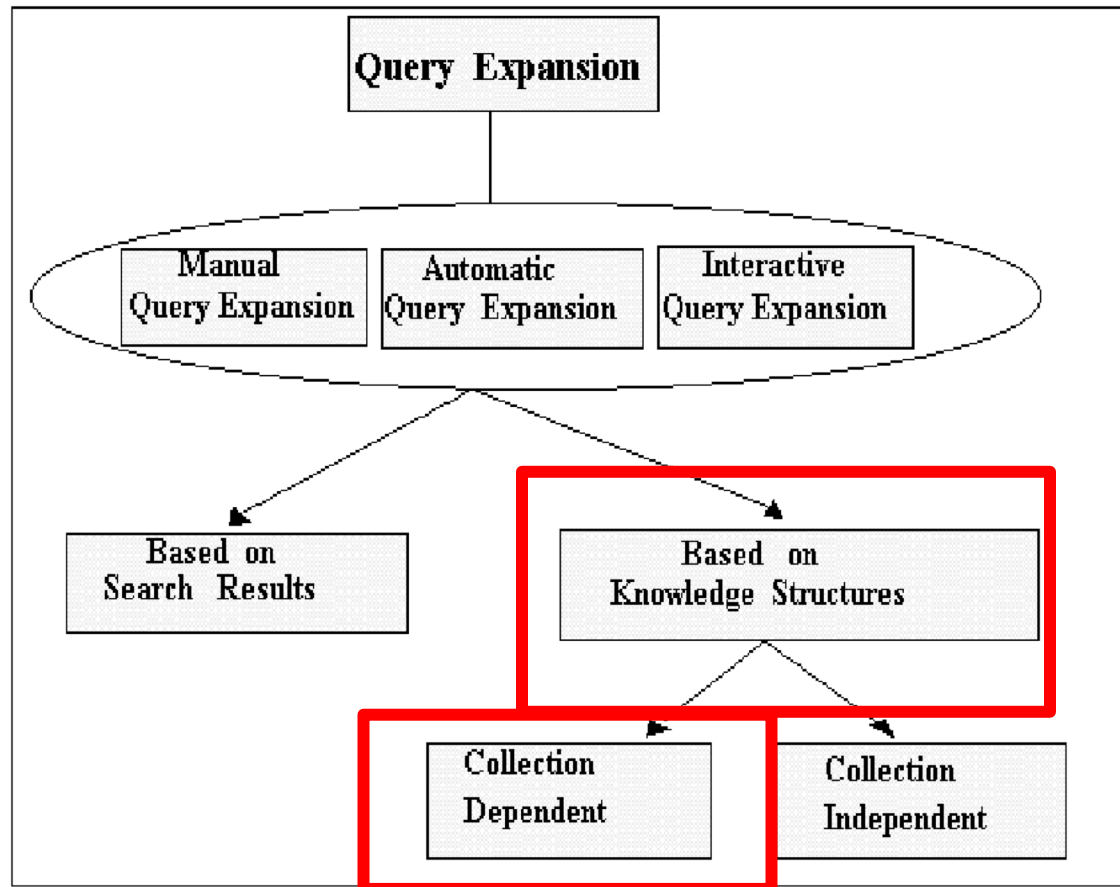
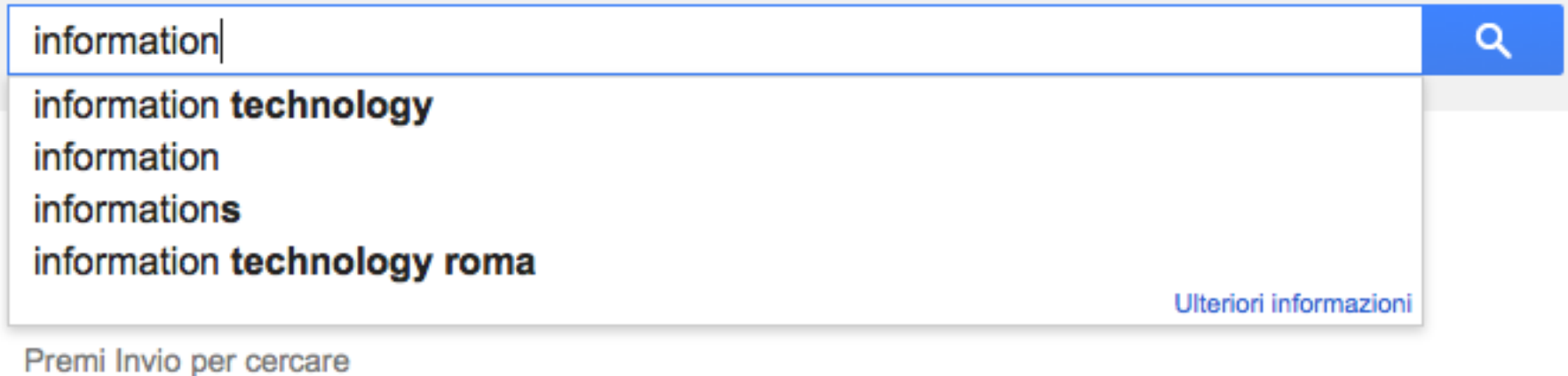


Figure 1: Query Expansion: Methods and Sources

# Expansion with query logs

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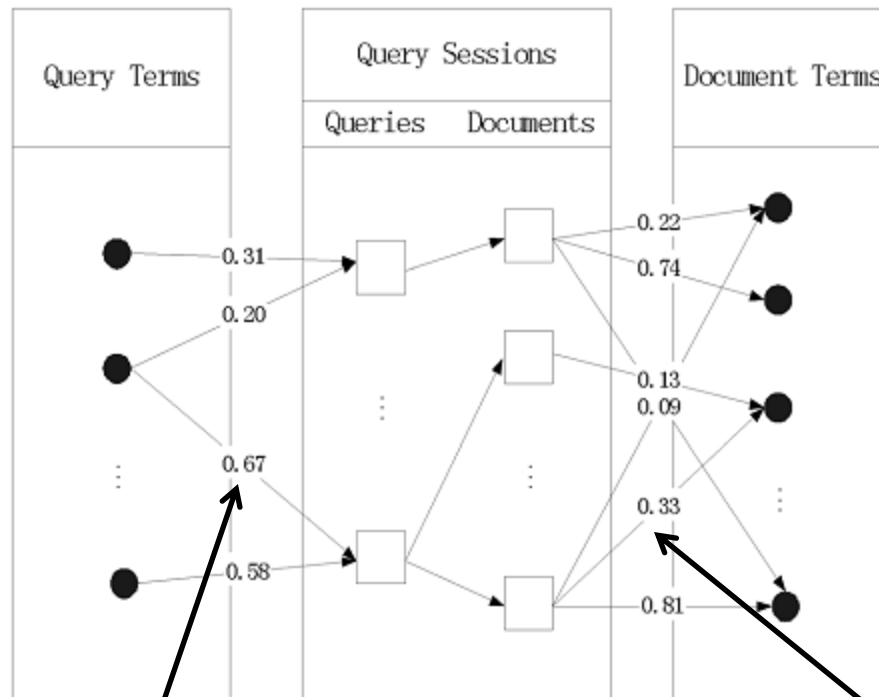
- Google use query logs:



Query is expanded with “hints” as you type words into the query window

# Query expansion with query logs

Query logs: given a query, which documents have been accessed?



Prob of term  $j$  in query  $k$

Relevance of term  $i$  in doc  $j$

# Learning from querylogs is important, however

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- Google process 3 billion queries per day
- Lots of data, however, 20-25% of these queries are NEW, have never been done before
- 450ml previously “unseen” queries
- Therefore “brute force” is not enough, need to LEARN from previous query and GENERALIZE
- GENERALIZE= learn word meaning and word correlations

# Is there anything more advanced than co-occurrences to learn correlations?

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- To detect these similarities (next lessons):
  - Latent Semantic Indexing
  - Word embeddings (a.k.o. deep method)