Information Retrieval & Text Mining

Probabilistic Retrieval Model:
Basic Idea

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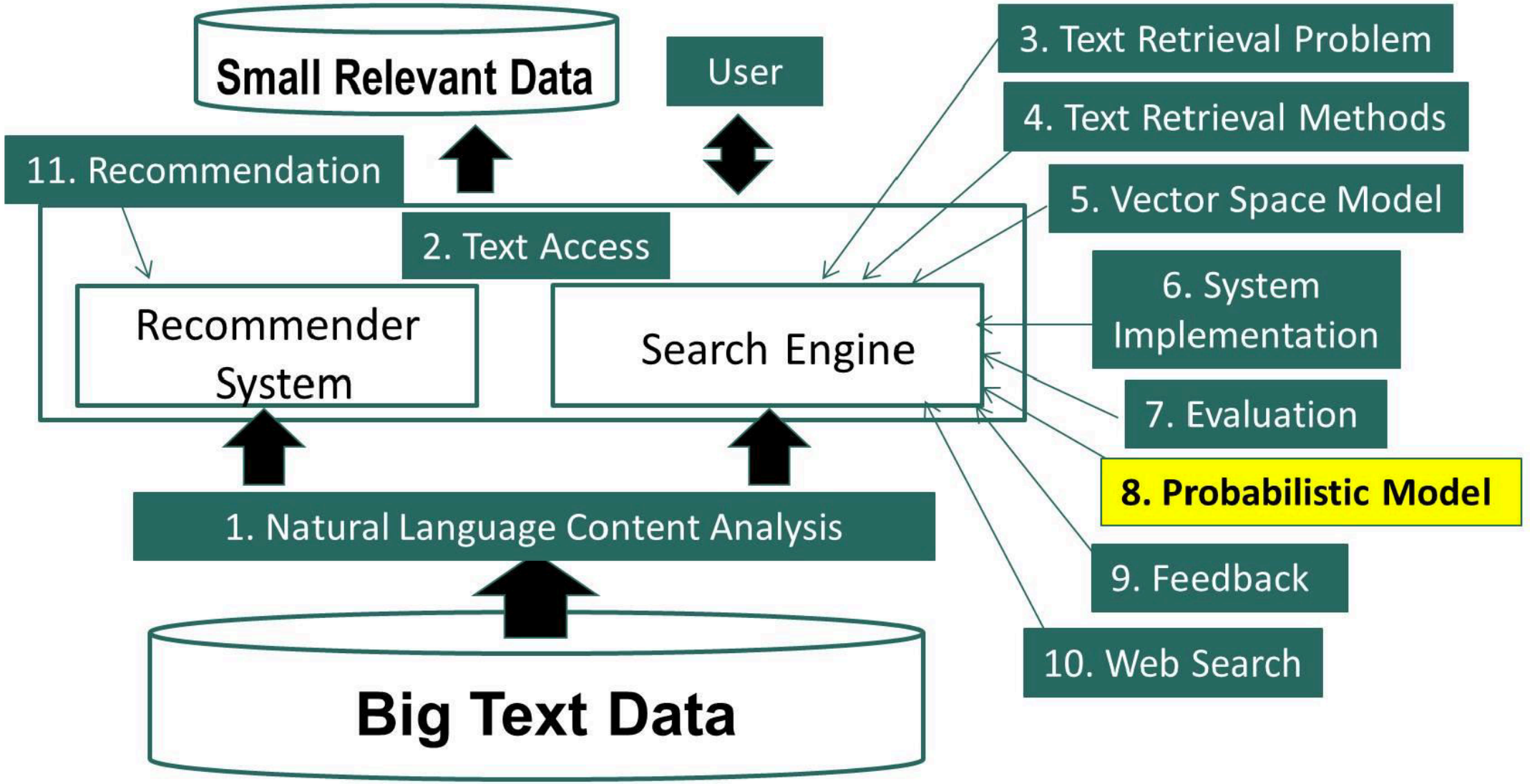

Many Different Retrieval Models

- **Similarity-based models**: f(q,d) = similarity(q,d)
  - Vector space model
- **Probabilistic models**: f(d,q) = p(R=1|d,q), where R $\in$ {0,1}
  - Classic probabilistic model
  - Language model
  - Divergence-from-randomness model
- **Probabilistic inference model**: f(q,d) = p(d→q)
- **Axiomatic model**: f(q,d) must satisfy a set of constraints
- These different models tend to result in similar ranking functions involving similar variables

Probabilistic Model

- We define ranking function that a given document **D** is relevant to a given query **Q**.
- We introduce binary random variable $R \in \{0, 1\}$
- We assume that **Q** and **D** are observations from random variable, in vector space model we assume they are vectors
- Problem of retrieval now becomes the problem to estimate the probability of relevance.

$$f(d,q) = p(R=1 \mid d,q), \quad R \in \{0,1\}$$

Many Different Retrieval Models

- **Probabilistic models**: $f(d,q) = p(R=1|d,q), \quad R \in \{0,1\}$
  - Classic probabilistic model ➔ BM25
  - **Language model ➔Query Likelihood**

$$p(R=1|d,q) \approx p(q|d,R=1)$$

If a user likes document d, how likely would the user enter query q (in order to retrieve d)?

Probabilistic Retrieval Models: Basic Idea

Query	Doc	Rel
q	**d**	**R**
q1	d1	1
q1	d2	1
q1	d3	0
q1	d4	0
q1	d5	1
…		
q1	d1	0
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$$f(q,d)=p(R=1|d,q)=?$$

How can we estimate the probability of relevance?

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$f(q,d)=p(R=1|d,q)=?$ $\frac{count(q,d,R=1)}{count(q,d)}$

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P(R=1|q1,d1) = ?

P(R=1|q1,d2) = ?

P(R=1|q1,d3) = ?

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$P(R=1|q1,d1) = ?$ 1/2

$P(R=1|q1,d2) = ?$ 2/2

$P(R=1|q1,d3) = ?$ 0/2

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P(R=1|q1,d1) = ? 1/2

P(R=1|q1,d2) = ? 2/2

P(R=1|q1,d3) = ? 0/2

What about unseen documents?

Unseen queries?

Query Likelihood Retrieval Model

Query	Doc	Rel
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$$f(q,d)=p(R=1|d,q)\approx \quad \mathbf{p(q|d,R=1)}$$

Approximations

In query likelihood, our assumption is that this probability of relevance can be approximated by the probability of a query given a document and relevance, p(q | d , R = 1). Intuitively, this probability just captures the following probability: if a user likes document d, how likely would the user enter query q in order to retrieve document d? The condition part contains document d and R = 1, which can be interpreted as the condition that the user likes document d.

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User likes d

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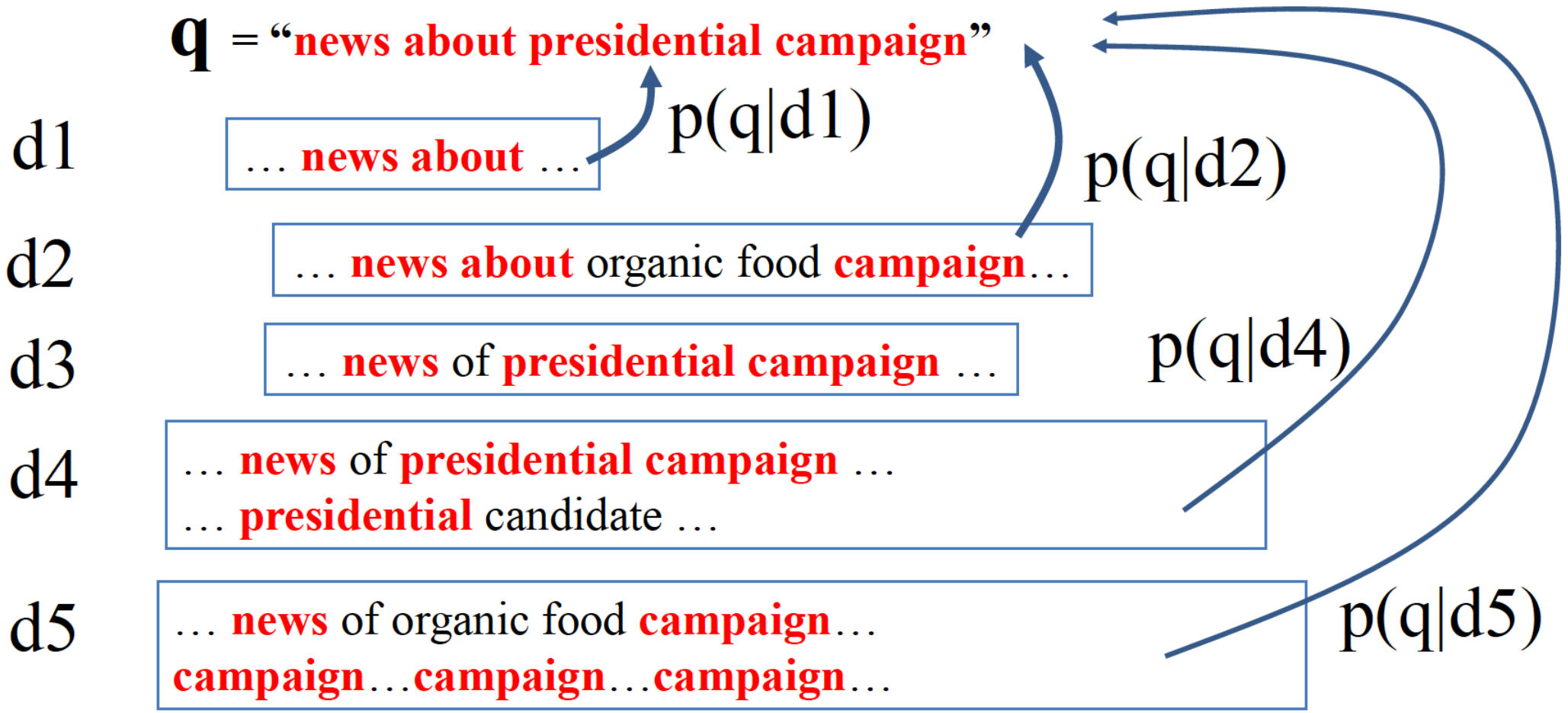
User likes d

$$f(q,d)=p(R=1|d,q)\approx \mathbf{p(q|d,R=1)}$$

How likely the user enters q

Assumption:
A user formulates a query based on an "**imaginary relevant document**"

Which doc is Most Likely the "Imaginary Relevant Doc"?



Summary

- Relevance(q,d) = p(R=1|q,d) ➔p(q|d,R=1)
- **Query likelihood** ranking function: f(q,d)=p(q|d)
  - Probability that a user who likes d would pose query q
- How to compute p(q|d)? How to compute probability of text in general? ➔ Language Model

p(q= "presidential campaign"**|d=** … **news** of **presidential** **campaign** … **presidential** candidate … **)**