



PRUM

Precision Recall with User Modelling

Application to XML

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Outline

- Why measuring?
- XML Information Retrieval
- Current metrics
 - GPR, PRng, XCG
- PRUM
 - A new user model
 - A new metric
- Comparison

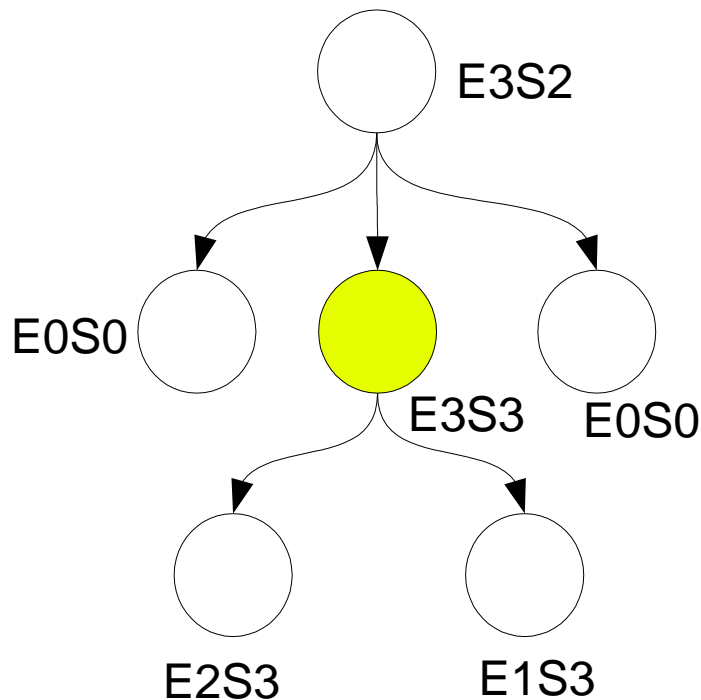
Why measuring?

- The evaluation methodology started early (Cleverdon, 1967)
- Justification of pragmatic/theoretic developments
- In new IR paradigms (XML, video, etc.), measures have the same role to play... but none of the proposed metrics are satisfying.

XML Retrieval (and INEX...)

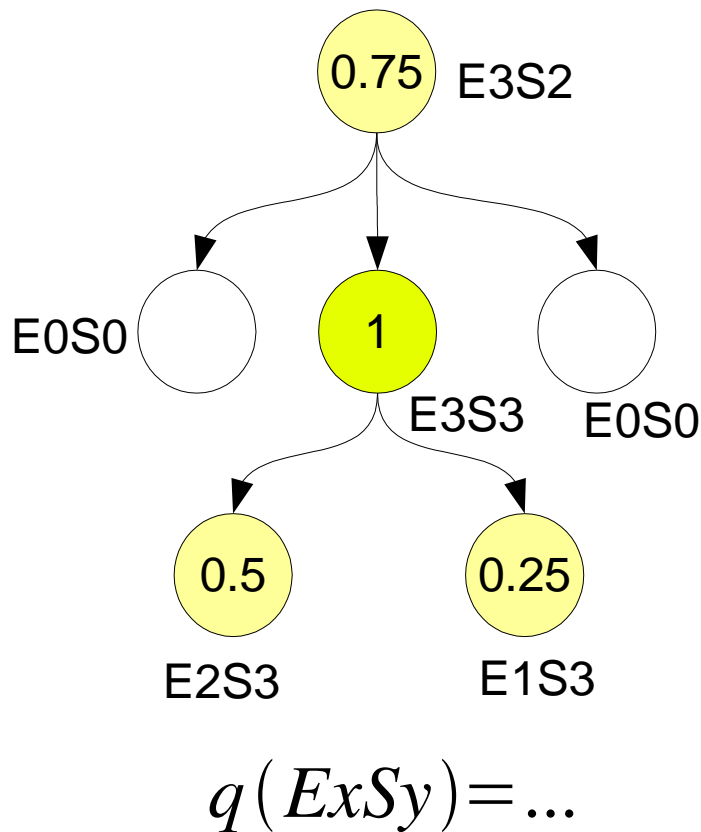
- “...more precise access by giving more specific answers...” (ie. XML nodes)
- Assessments: a new scale
 - Specificity (S): The extent to which a document component is focused on the information (4 values: 0 to 3)
 - Exhaustivity (E): The extent to which the information contained in a document component satisfies the user's query need (4 values: 0 to 3)

Precision Recall



- Only exact answers are rewarded
- ▶ Too strict!
- ▶ Overlap problem in recall base

Generalised Precision Recall

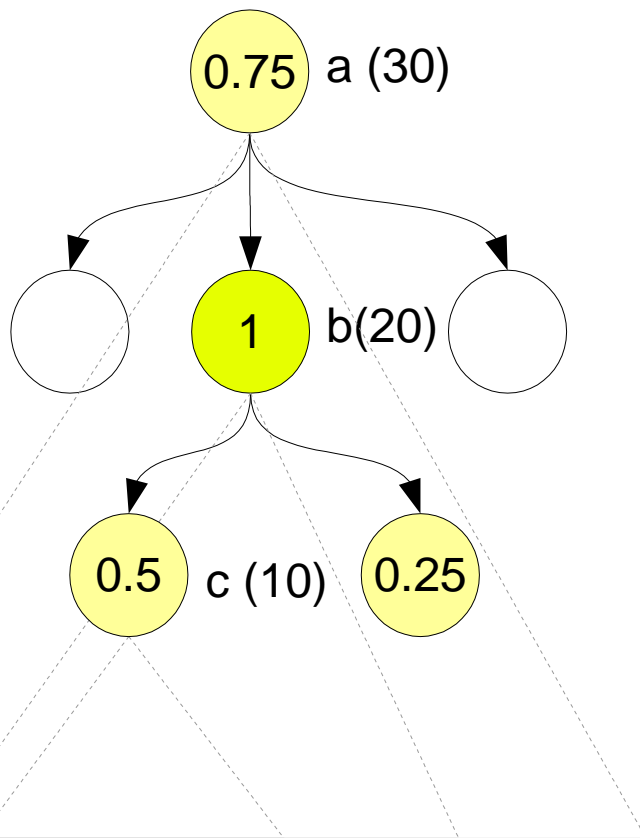


- Assign a relevance value to near misses (quantisation)
- ▶ Recall base is expanded
- ▶ Overlap is a problem

PRng (N. Gövert)

- Exhaustivity and specificity based on the notion of an ideal concept space upon which precision and recall are defined.
- Considers overlap in retrieval results
 - Very sensitive to list order
- Does not consider overlap in recall-base (not completely true)

PRng (N. Gövert)



$$\begin{aligned}\text{Recall}(a,b,c) \\ &= 0.75 + 0 \times 1 + 0 \times 0.5 \\ &= 0.75\end{aligned}$$

$$\begin{aligned}\text{Recall}(c,b,a) \\ &= 0.5 + .5 \times 1 + 1/3 \times 0.75 \\ &= 1.25\end{aligned}$$

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XCG (G. Kazai)

- Based on cumulated gain measure for IR
- Accumulates gain obtained by retrieving elements at fixed ranks
- Based on the construction of an **ideal recall-base**
- Consider overlap in both retrieval results and recall-base

XCG (G. Kazai)

- The most consistent metric
 - Normalisation maximises the gain of near misses
 - Normalisation to ensure that performance is bounded by an “ideal” system
- ... but
 - No precise user model
 - No precision dimension

Towards user models

- Web and others
 - Quintana
 - Dunlop... leads to:
- Tolerance To Irrelevance (A. de Vries)
 - The first real user model in XML IR
 - Some theoretic problems

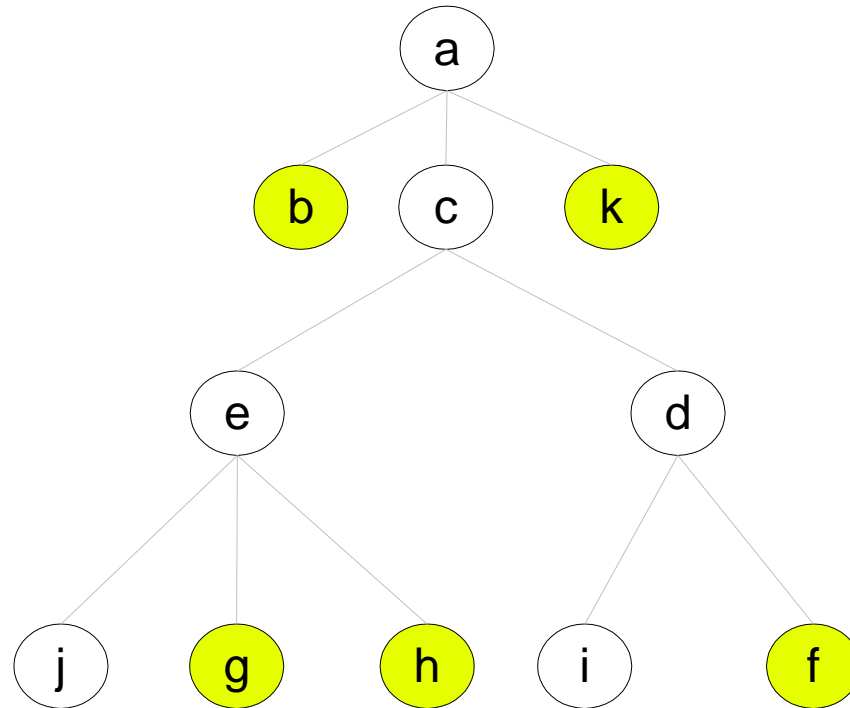
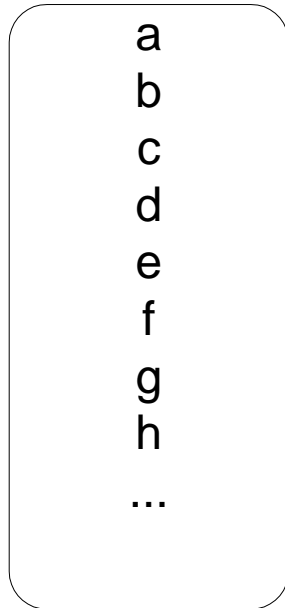


PRUM: main ideas

- Construction of an ideal recall base without overlap (like XCG)
- Definition of a precise user model (which includes T2I as a special case)
- Sound formal grounds (based on Raghavan's probabilistic precision-recall)

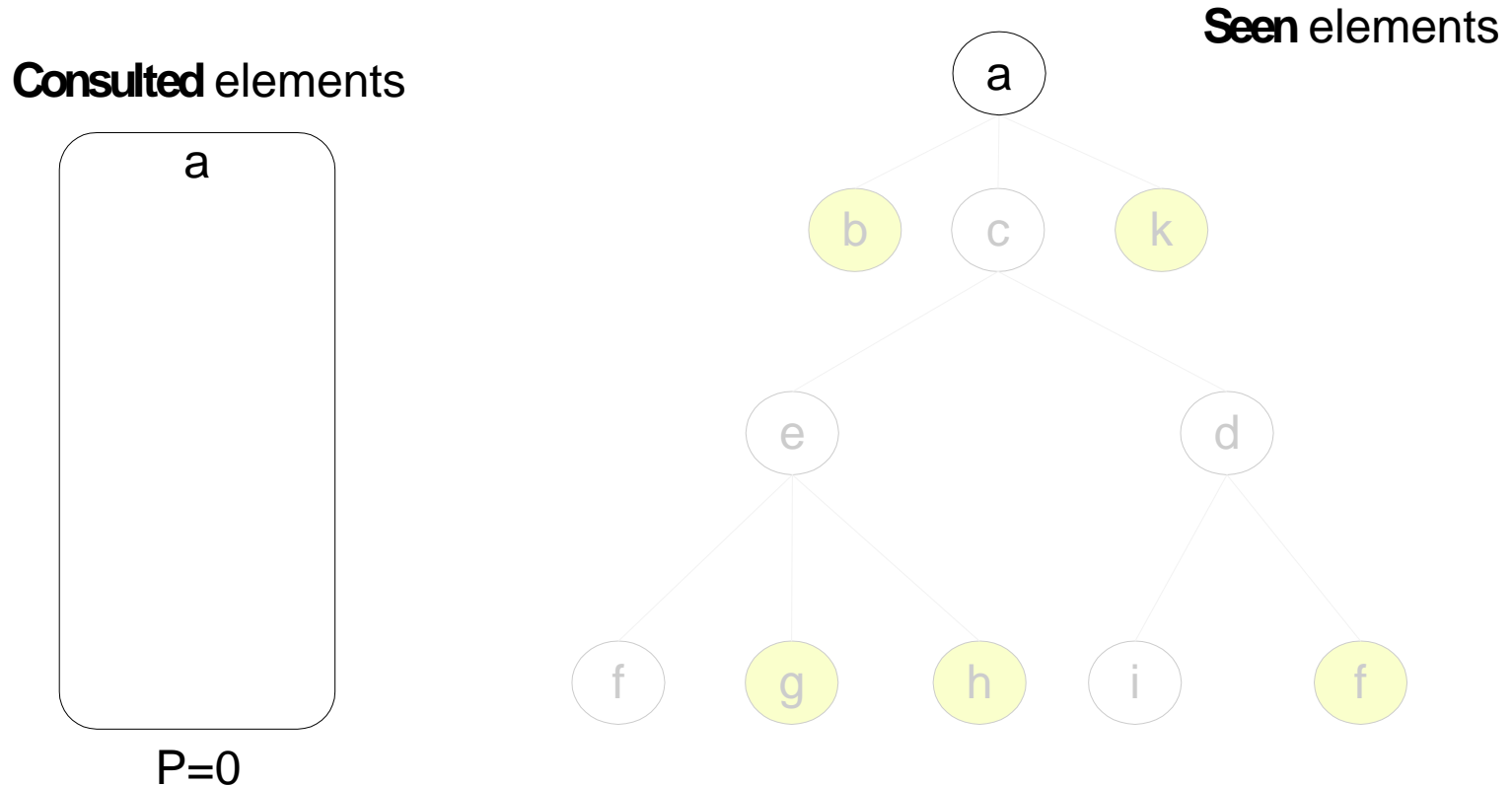
The PRUM user model

Retrieved list



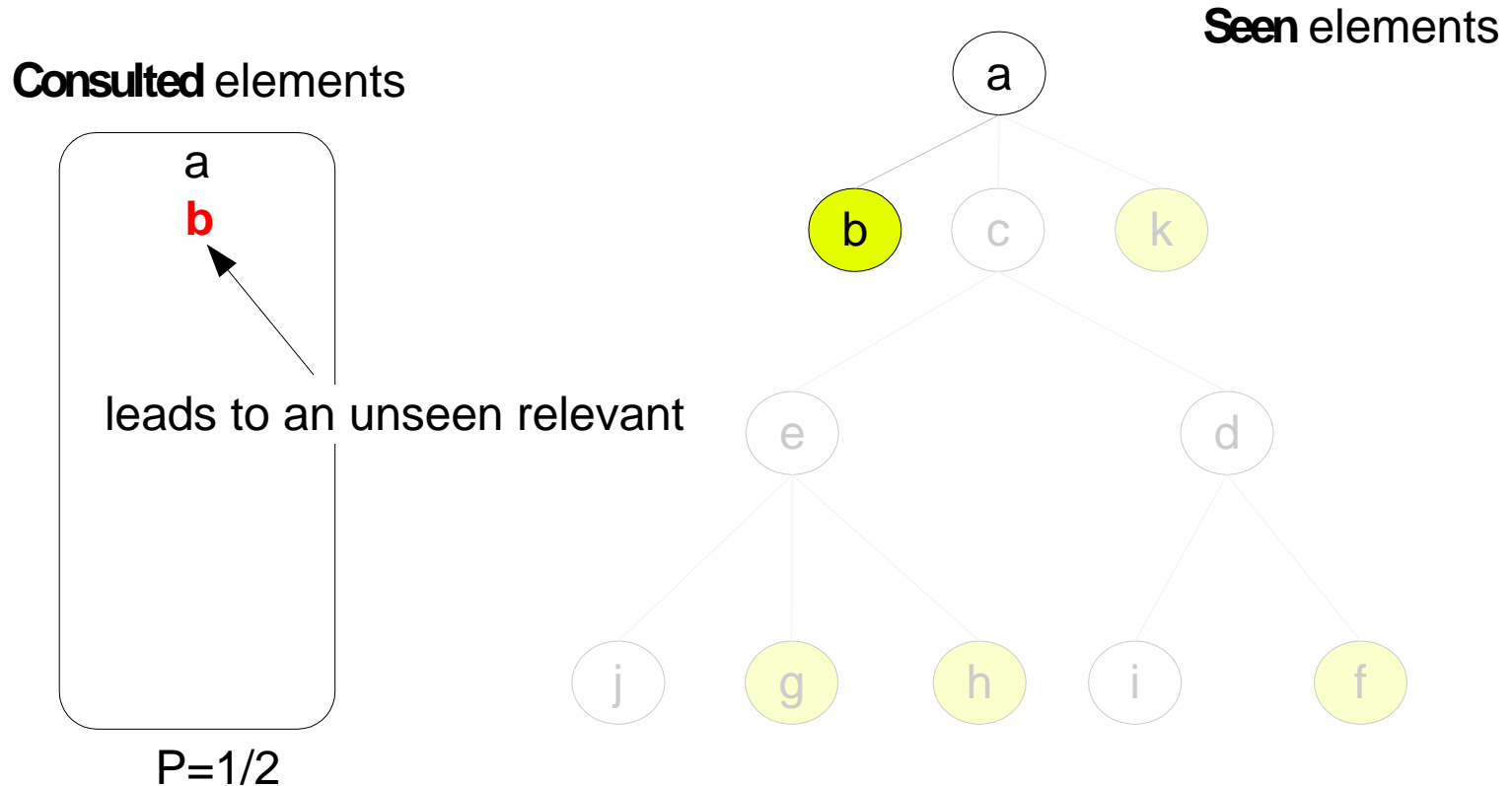
The user wants to see 4 relevant elements

The PRUM user model



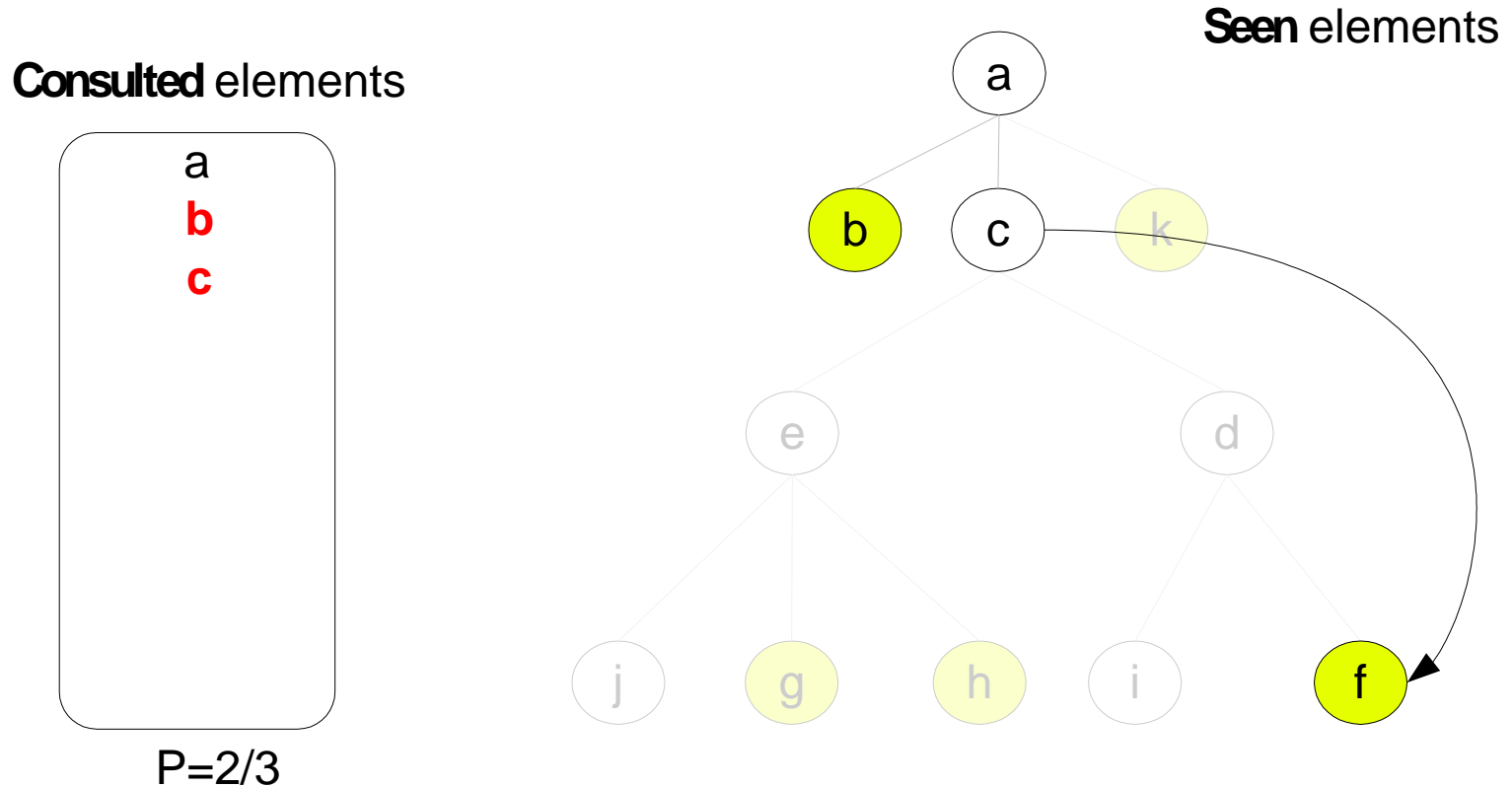
The user has seen 0 relevant element

The PRUM user model



The user has **seen 1** relevant element

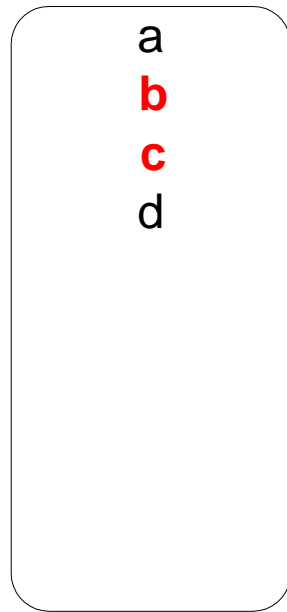
The PRUM user model



The user has **seen 2** relevant elements

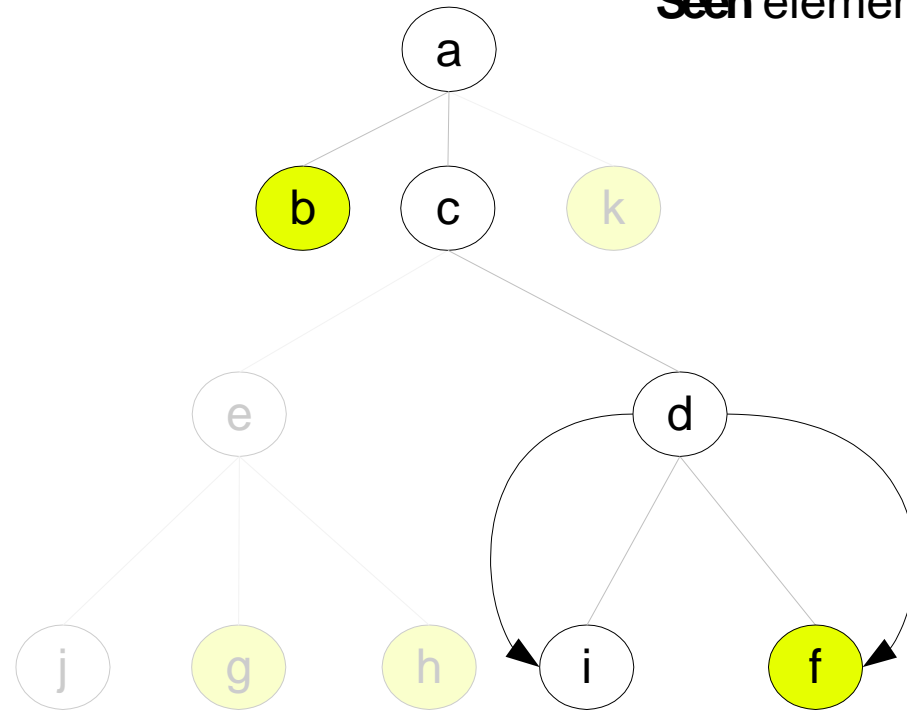
The PRUM user model

Consulted elements



$P=2/4$

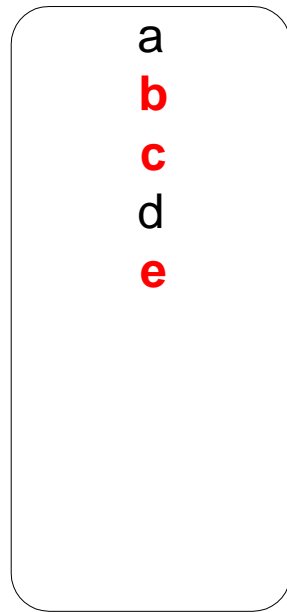
Seen elements



The user has **seen 2** relevant elements

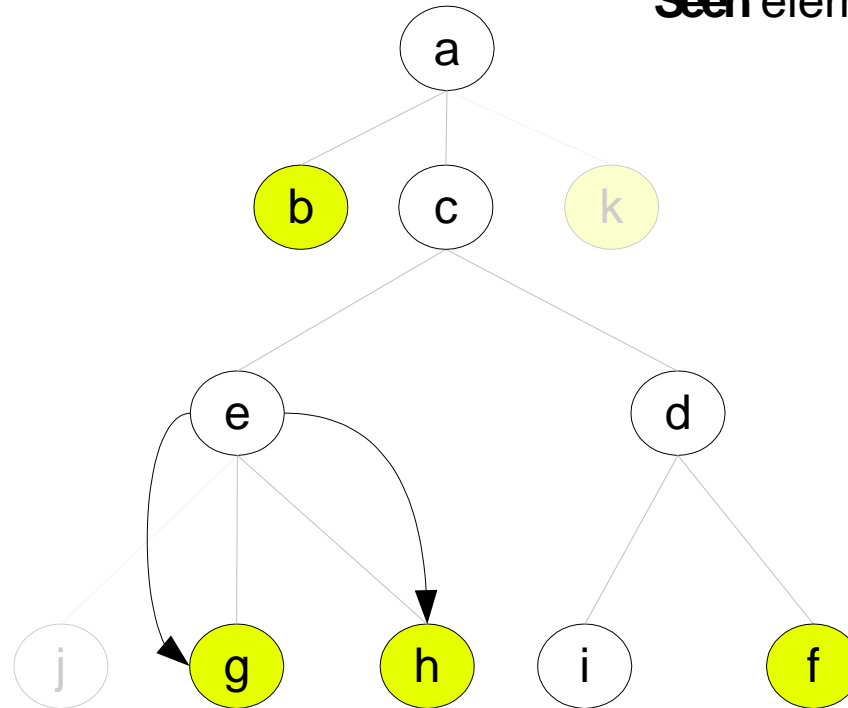
The PRUM user model

Consulted elements



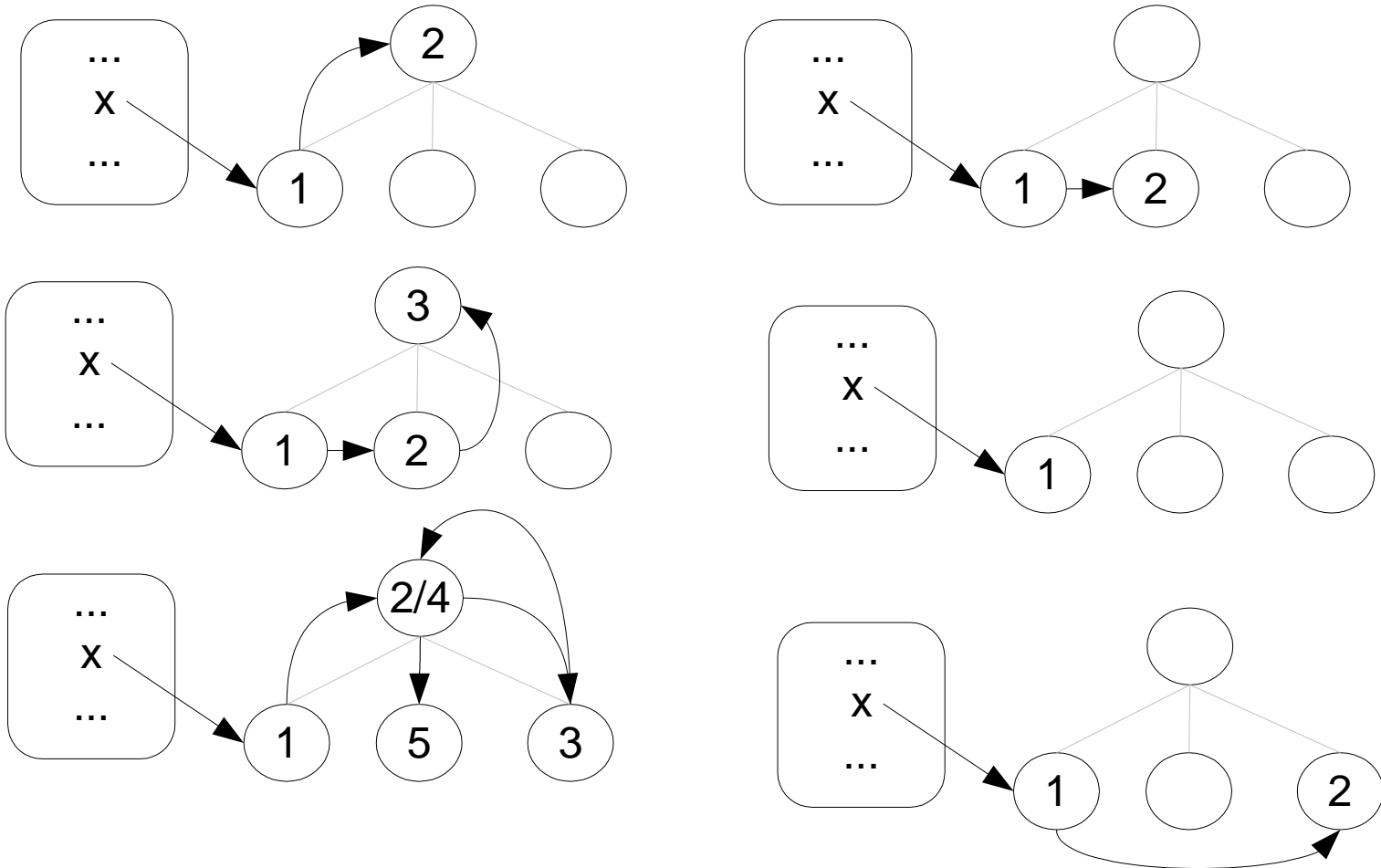
$P=3/5$

Seen elements

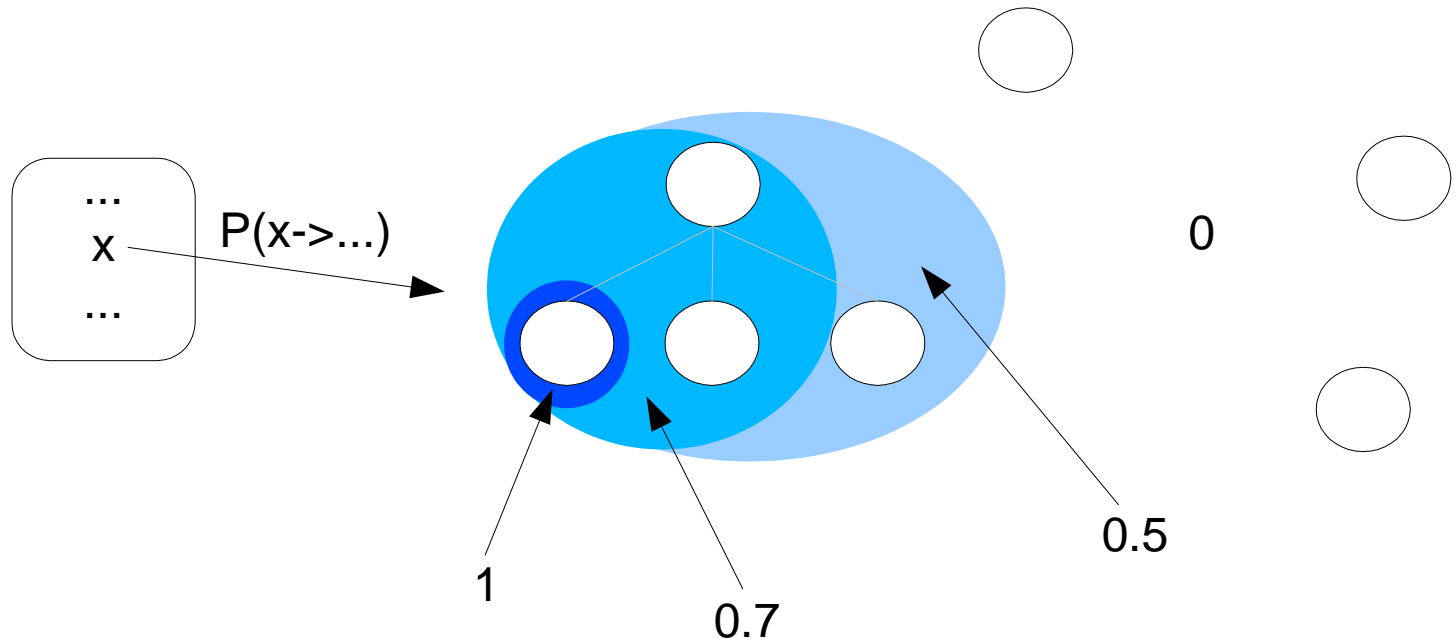


The user has **seen** 4 relevant elements

PRUM stochastic user model



PRUM stochastic user model



PRUM definition

- Extends Raghavan's definition

$$P(L_{ur} | C_s, L=l, Q=q)$$

The element
Leads to an
Unseen
Relevant

The element is
Consulted by
the user

The user wants
to see l % of
the relevant
elements

Computing PRUM

- Two probabilities have to be computed at each rank in the consulted list:
 - The probability that the user has seen s distinct relevant elements

$$P(F_i = s)$$

- The probability that the user sees a new relevant element (knowing that ...)

$$P(F_i > F_{i-1} | F_{i-1} = s)$$

Experiments (INEX 2004)

- 5 hypothetic systems

- 1) ideal (I) = ancestors (A)

- 2) parent (P) ~ biggest child (B)

- 3) document (D)

- Orders

- GRP: $A \gg I > P > D \gg B$

- RPng: $A > I \sim P > D \gg B$

- PRUM: $I = A \gg B \gg P \gg D$

Conclusion

■ PRUM

- Extension of Precision-Recall (probabilistic)
- Explicit parameters (\neq GPR, PRng, XCG)
- Good behaviour of the metric
- Adapted to several IR formalisms

■ Work to be done...

- More realistic user model parameters

■ Shortcomings

- Graded scale
- Complexity