

A Probability Ranking Principle for Interactive IR

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Outline

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Motivation

The classical PRP

Questioning the PRP assumptions

Interactive Retrieval

The classical PRP

- ▶ Task: Retrieve relevant documents
- ▶ Relevance of a document to a query is independent of other documents
- ▶ Scanning through the ranked list is the major task of the user (and the only one considered)

Questioning the PRP assumptions

- ▶ Relevance depends on documents the user has seen before
- ▶ Relevance judgment is not the most expensive task for a user

Interactive Retrieval

- ▶ User has a rich set of interaction possibilities
 - ▶ (re)formulate query
 - ▶ selection based on summaries of various granularity
 - ▶ select related terms from list
 - ▶ follow document link
 - ▶ relevance judgment
- ▶ Information need changes during a search
- ▶ No theoretic foundation for constructing IIR systems

Approach

Requirements for an IIR-PRP

Basic Assumptions

Abstraction: Situations with Lists of Choices

Requirements for an IIR-PRP

- ▶ Consider the complete interaction process
- ▶ Allow for different costs for different activities
- ▶ Allow for changes of the information need

Basic Assumptions

- ▶ Focus on a functional level of interaction (usability issues disregarded here)
- ▶ System presents list of choices to the user
- ▶ Users evaluate choices in linear order
- ▶ Only positive decisions/choices are of benefit for a user

Examples of decision lists

- ▶ ranked list of documents
- ▶ list of summaries
- ▶ list of document cluster
- ▶ KWIC list
- ▶ list of expansion terms
- ▶ links to related documents
- ▶ ...

Example: Non-linear decision list

Australia is the only place in the world with 25 types of fleas - Eva Glass

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the INQUIRER

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Headlines week ending - Thursday 09 October 2008

Real DVD ban extended

While judge has a ponder
Thu, 09 October 2008, 12:28

Brits are losing their memories

Sticks, not minds, that is
Thu, 09 October 2008, 10:57

Nvidia 270, 290 and GX2 roll out in November

Why you shouldn't get too excited
Thu, 09 October 2008, 10:01

Acer claims top laptop spot

Number one
Thu, 09 October 2008, 9:30

6TB NAS tested

Hardware Rounds
Synology's DS408 leaves an impression
Thu, 09 October 2008, 8:48

PowerVR 531 is out and about

Imagination reckons UMPs are hot
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Symantec acquires MessageLabs

\$695 million cash deal
Wed, 08 October 2008, 14:40

Baltimore gets mobile WiMAX

Sprints to the finish line
Thu, 09 October 2008, 11:17

PPP cleanup attempt has familiar Eckoh

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Microsoft gets off of its Cloud

Volish cloud becomes Strata
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Youtube adds vocals to comment

All you ever wanted
Thu, 09 October 2008, 9:07

Sun terrible place to work

MySQL co-founder walks
Thu, 09 October 2008, 6:57

Universal Music Group joins Kiwibox

Music match made in heaven
Wed, 08 October 2008, 15:12

Aussie TV network guilty of subliminal ads

Admonished but not penalised

Adobe warns Clickjackers could take over your web cam

Put some clothes on, at least
Thu, 09 October 2008, 11:12

China cracks down on mobile spam

Shutters 18 service providers
Thu, 09 October 2008, 10:28

HP cuts over 3000 EDS jobs in UK

Scrapheaped
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Create your own game as you drive

More augmented reality than Top Gear
Thu, 09 October 2008, 8:57

Spammers prefer Obama to McCain

Bigger interest makes a better spam hook
Wed, 08 October 2008, 16:12

Opera maverick is still making waves

Speaker's Corner Jon Tetzchner, Opera Software
Wed, 09 October 2008, 14:40

Bust hackers

IThound

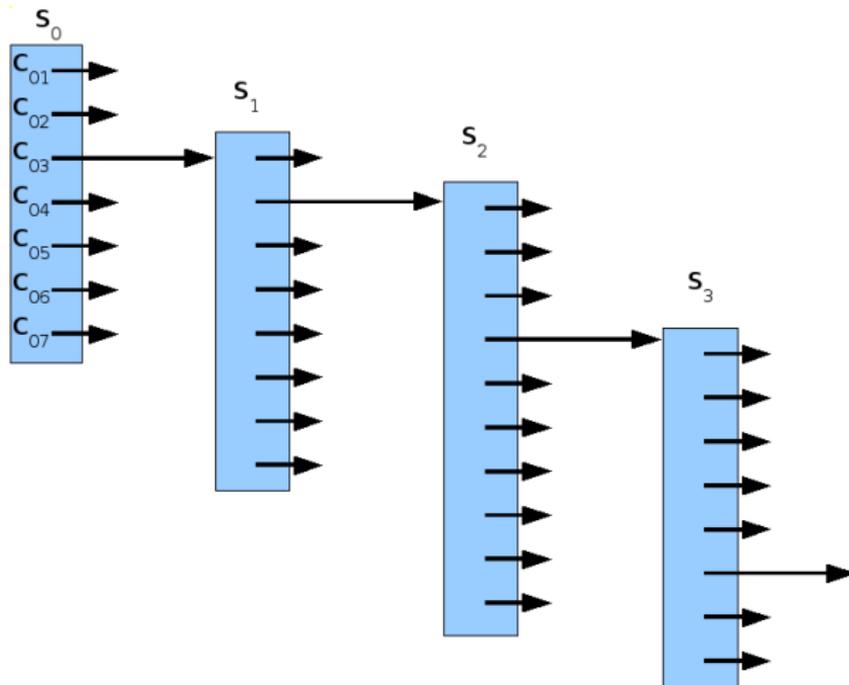
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1. Nvidia 270, 290 and GX2 roll out in November
2. Ipod doomed says Wozniak
3. Adobe warns Clickjackers could take over your web cam
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5. Brits are losing their memories

Abstraction: Situations with Lists of Choices



The Model

Choices

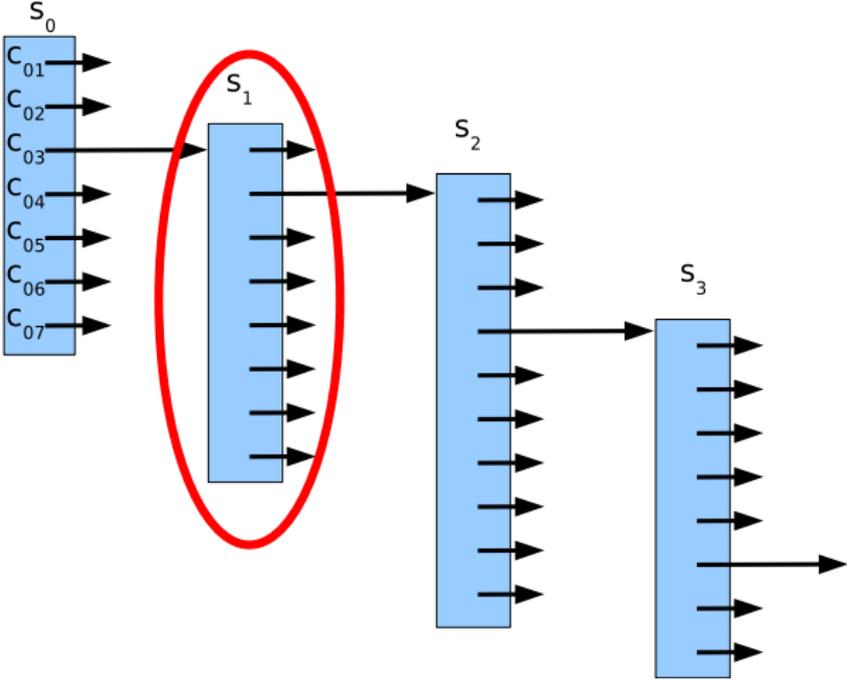
Selection lists

Ranking of choices

Basic ideas

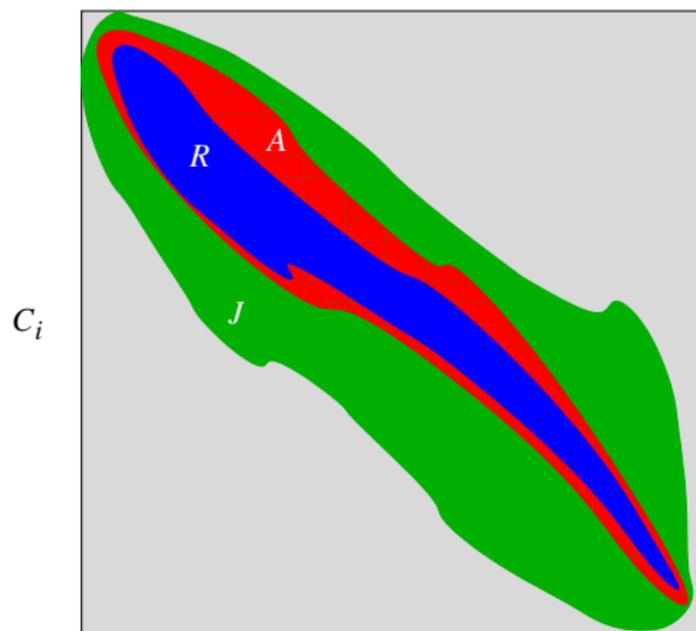
- ▶ A user moves from situation to situation
- ▶ In each situation s_i , the user is presented a list of (binary) choices $\langle c_{i1}, c_{i2}, \dots, c_{i,n_i} \rangle$
- ▶ The user decides about each of these choices sequentially
- ▶ The first positive decision moves the user to a new situation s_j
- ▶ A decision may be wrong, requiring backtracking

Probabilistic model focusing on single situation



Probabilistic Event space

U_i



U_i : Uses in situation s_i
 C_i : choices in situation s_i

$J \subset U_i \times C_i$:
judged choices

$A \subset J$:
accepted choices

$R \subseteq A$: 'right' choices

Expected Benefit of a choice

p_{ij} probability that the user will accept choice c_{ij}

q_{ij} probability that this decision was right

$e_{ij} < 0$: effort for evaluating the choice c_{ij}

$b_{ij} > 0$: resulting benefit from positive, correct decision

$g_{ij} \leq 0$: cost for correcting a wrong decision

Expected benefit of choice c_{ij}

$$E(c_{ij}) = e_{ij} + p_{ij} (q_{ij}b_{ij} + (1 - q_{ij})g_{ij})$$

Example

Web search: 'Java' $\rightarrow n_0=290$ mio. hits

System proposes extension terms:

term	n_i	p_{ij}	b_{ij}	$p_{ij}b_{ij}$
program	195 mio	0.67	0.4	0.268
blend	5 mio	0.02	4.0	0.08
island	2 mio	0.01	4.9	0.049

benefit $b_{ij} = \log \frac{n_0}{n_i}$

Strategies for maximizing expected benefit

$$E(c_{ij}) = e_{ij} + p_{ij} (q_{ij}b_{ij} + (1 - q_{ij})g_{ij})$$

(assume that benefit b_{ij} and corr. effort g_{ij} are given)

1. minimize effort $|e_{ij}|$ —
but keep p_{ij} (selection prob.) and q_{ij} (success prob.) high
2. maximize p_{ij} : user should choose c_{ij} whenever it is appropriate —
but keep success probability q_{ij} high
 \rightsquigarrow increased effort e_{ij}
3. maximize q_{ij} by avoiding erroneous positive decisions
 \rightsquigarrow increased effort e_{ij}

Further remarks

$$E(c_{ij}) = e_{ij} + p_{ij} (q_{ij}b_{ij} + (1 - q_{ij})g_{ij})$$

- ▶ Expected benefit should be positive
choices with negative values should not be presented to a user.
- ▶ Methods for estimating parameters p_{ij} , q_{ij} , b_{ij} , e_{ij} , g_{ij} :
Issue of further research
- ▶ In the following, let $a_{ij} = q_{ij}b_{ij} - (1 - q_{ij})g_{ij}$
("average benefit")

$$E(c_{ij}) = e_{ij} + p_{ij}a_{ij}$$

Selection list

situation s_i with list of choices $r_i = \langle c_{i1}, c_{i2}, \dots, c_{i,n_i} \rangle$

expected benefit of choice list:

$$\begin{aligned} E(r_i) &= e_{i1} + p_{i1} a_{i1} + \\ &\quad (1 - p_{i1}) (e_{i2} + p_{i2} a_{i2} + \\ &\quad (1 - p_{i2}) (e_{i3} + p_{i3} a_{i3} + \\ &\quad \dots \\ &\quad (1 - p_{i,n-1}) (e_{in} + p_{in} a_{in}))) \\ &= \sum_{j=1}^n \left(\prod_{k=1}^{j-1} (1 - p_{ik}) \right) (e_{ij} + p_{ij} a_{ij}) \end{aligned}$$

Expected benefit of a choice list

$$E(r_i) = \sum_{j=1}^n \left(\prod_{k=1}^{j-1} (1 - p_{ik}) \right) (e_{ij} + p_{ij} a_{ij})$$

Ranking of choices

Consider two subsequent choices c_{il} and $c_{i,l+1}$

$$E(r_i) = \sum_{\substack{j=1 \\ l \neq j \neq l+1}}^n \left(\prod_{k=1}^{j-1} (1 - p_{ik}) \right) (e_{ij} + p_{ij} a_{ij}) + t_i^{l,l+1}$$

where

$$t_i^{l,l+1} = (e_{il} + p_{il} a_{il}) \prod_{k=1}^{l-1} (1 - p_{ik}) + (e_{i,l+1} + p_{i,l+1} a_{i,l+1}) \prod_{k=1}^l (1 - p_{ik})$$

analogously $t_i^{l+1,l}$ for $\langle \dots, c_{i,l+1}, c_{il}, \dots \rangle$

Difference between alternative rankings

$$\begin{aligned}d_i^{l,l+1} &= \frac{t_i^{l,l+1} - t_i^{l+1,l}}{\prod_{k=1}^{l-1} (1 - p_{ik})} \\&= \mathbf{e}_{il} + p_{il} \mathbf{a}_{il} + (1 - p_{il})(\mathbf{e}_{i,l+1} + p_{i,l+1} \mathbf{a}_{i,l+1}) - \\&\quad (\mathbf{e}_{i,l+1} + p_{i,l+1} \mathbf{a}_{i,l+1} + (1 - p_{i,l+1})(\mathbf{e}_{il} + p_{il} \mathbf{a}_{il})) \\&= p_{i,l+1}(\mathbf{e}_{il} + p_{il} \mathbf{a}_{il}) - p_{il}(\mathbf{e}_{i,l+1} + p_{i,l+1} \mathbf{a}_{i,l+1})\end{aligned}$$

For $d_i^{l,l+1} \stackrel{!}{\geq} 0$, we get

$$\mathbf{a}_{il} + \frac{\mathbf{e}_{il}}{p_{il}} \geq \mathbf{a}_{i,l+1} + \frac{\mathbf{e}_{i,l+1}}{p_{i,l+1}}$$

PRP for Interactive IR

$$a_{ij} + \frac{e_{ij}}{p_{ij}} \geq a_{i,l+1} + \frac{e_{i,l+1}}{p_{i,l+1}}$$

↪ Rank choices by decreasing values of

$$\varrho(c_{ij}) = a_{ij} + \frac{e_{ij}}{p_{ij}}$$

Expected benefit: single choices vs. list

$$\text{expected benefit: } E(c_{ij}) = p_{ij}a_{ij} + e_{ij}$$

$$\text{ranking criterion: } \varrho(c_{ij}) = a_{ij} + \frac{e_{ij}}{p_{ij}}$$

Example:

choice	p_{ij}	a_{ij}	e_{ij}	$E(c_{ij})$	$\varrho(c_{ij})$
c_1	0.5	10	-1	4	8
c_2	0.25	16	-1	3	12

$$E(\langle c_1, c_2 \rangle) = 4 + 0.5 \cdot 3 = 5.5$$

$$E(\langle c_2, c_1 \rangle) = 3 + 0.75 \cdot 4 = 6$$

IIR-PRP vs. PRP

$$a_{ij} + \frac{e_{ij}}{p_{ij}} \geq a_{i,l+1} + \frac{e_{i,l+1}}{p_{i,l+1}}$$

Let $e_{ij} = -\bar{C}$, $\bar{C} > 0$ and $a_{ij} = C$:

$$\begin{aligned} C - \frac{\bar{C}}{p_{ij}} &\geq C - \frac{\bar{C}}{p_{i,l+1}} \\ \Rightarrow p_{ij} &\geq p_{i,l+1} \end{aligned}$$

↪ Classic PRP still holds!

IIR-PRP: Observations

Rank choices by $a_{ij} + \frac{e_{ij}}{p_{ij}}$

- ▶ p_{ij} 'probability of relevance' still involved
- ▶ tradeoff between effort e_{ij} and benefit a_{ij}
- ▶ difference between PRP and IIR-PRP due to variable values for e_{ij} and a_{ij}
- ▶ IIR-PRP looks only for the first positive decision

Towards application

Parameter estimation

Saved effort

Parameter estimation

1. Selection probability p_{ij} :
focus of many IR models,
but models for dynamic info needs required
2. Effort parameters e_{ij} , g_{ij} + success probability q_{ij} :
most research needed
3. Benefit b_{ij} :
 - ▶ information value ?
 - ▶ saved effort (see below)

Saved effort

- ▶ methods for estimating number r_q of relevant documents for query q
- ▶ linear recall-precision curve: $P(R) := P^0 \cdot (1 - R)$
- ▶ position of the first relevant document: $n_q = \frac{r_q}{P^0(r_q-1)}$
- ▶ user's choice transforms current query q' to optimum query q
- ▶ $P(q|q')$: probability that a random document from the result list of q also occurs in the result list of q'
- ▶ $n_{q'} = \frac{r_q}{P(q|q')P^0(r_q-1)}$
- ▶ benefit for moving from q' to q : $n_{q'} - n_q$.

Saved effort: Example

term	n_i	p_{ij}	b_{ij}	$p_{ij}b_{ij}$	$n_{q'}$	q_{ij}
program	195m	0.67	0.4	0.268	3	-0.5
blend	5m	0.02	4.0	0.08	116	56
island	2m	0.01	4.9	0.049	290	145

Conclusion and Outlook

Conclusion and Outlook

- ▶ Current IIR systems lack theoretic foundation
- ▶ Interactive IR as decision making
- ▶ user works on linear list of choices
- ▶ positive choices move user to new situation, with (possibly) new choice list
- ▶ IIR-PRP is generalization of classical PRP
- ▶ introduced new parameters
- ▶ parameter estimation is issue of further research