## Now that you've bought into Model 2, we'll tell you why to get Model I

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# Two principal approaches to Expert Finding

- Model I: Candidate models
  - Create a textual model ("profile" document) of candidates' knowledge according to the documents with which they are associated
  - Rank profiles using document retrieval techniques
- Model 2: Document models
  - Find documents that are relevant to the topic
  - Find out who is most strongly associated with these documents

#### Results

	TREC 2005		TREC 2006		TREC 2007	
	MAP	MRR	MAP	MRR	MAP	MRR
Model I	.1883	.4692	.3206	.7264	.3700	.5303
Model 2	.2053	.6088	.4660	.9354	.4137	.5666

K. Balog. People Search in the Enterprise. PhD thesis, University of Amsterdam, 2008.

- Baseline settings
  - Automatic smoothing parameter estimation
  - Baseline document-candidate associations
  - No query expansion

#### Components / extensions

- Proximity-based versions
- Smoothing parameter estimation
- Document-candidate associations
- Query models

#### Proximity-based versions

	TREC	2005	TREC 2006		TREC 2007	
	MAP	MRR	MAP	MRR	MAP	MRR
Model I	.1883	.4692	.3206	.7264	.3700	.5303
Model IB	.2020	.5928	.4254	.9048	.3608	.5003
Model 2	.2053	.6088	.4660	.9354	.4137	.5666
Model 2B	.2194	.6096	.4544	.9235	.4303	.5656

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# Smoothing parameter estimation

		TREC	2005	TREC 2006		TREC 2007	
		MAP	MRR	MAP MRR		MAP	MRR
МІ	auto	.1883	.4692	.3206	.7264	.3700	.5303
	opt.	.1912	.5747	.3834	.8647	.3801	.5571
M2	auto	.2053	.6088	.4660	.9354	.4137	.5666
	opt.	.2211	.6302	.4697	.9558	.4142	.5671

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## Smoothing parameter estimation

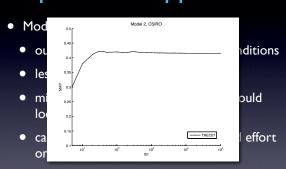
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### The preferred approach is...

- Model 2, because
  - outperforms Model I in nearly all conditions
  - less sensitive to smoothing
  - mimics the process how real users would look for experts
  - can be implemented with very limited effort on top of a document search engine
  - effective (enough to examine the top ranked documents)

#### The preferred approach is...



effective (enough to examine the top ranked documents)

#### But...

- Better document retrieval ⇒ better expert finding?
  - No, the relation is not that simple

	Baseline	QM_BFB	QM_EX
Document search	.3576	.3797	.4449
Expert search (M2)	.4137	.3795	.4652

MAP scores on the TRECENT 2007 document and expert search tasks.

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#### Moreover ...

- So far: changing one component at a time
  - What if we start combining and stacking extensions?
- How to integrate external (e.g., Web) evidence?
- How to visualize?

## Stacking things

	estim	nated	optimal		
	MAP	MRR	MAP	MRR	
Model I	.3700	.5303	.3801	.5571	
Model IB	.3608	.5003	.4633	.6236	
Model 2	.4137	.5666	.4142	.5671	
Model 2B	.4303	.5656	.4323	.5790	

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#### Visualizing the model

Model I

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- Model 2
  - List of documents? Snippets?

### ILPS@TRECENT 2008

#### **Approach**

- Proximity-based version of Model I
- Use global evidence to estimate the importance of a text snippet [Balog & de Rijke, CIKM 2008]
- Expansion
  - Profile side: evidence from the Web
  - Query side: terms from candidate profiles

#### Setup

- Minimal corpus preprocessing (stopword removal, no stemming)
- Use candidate list from [Balog & de Rijke, CIKM 2008]
  - Made available for TREC participants http://es.csiro.au/cerc/data/balog/
- Science communicators are filtered out
- Results reported on 2007 topics

# Query expansion using candidate profiles

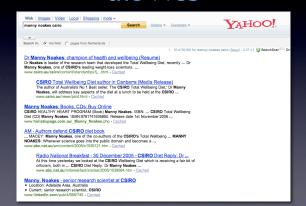
 Add top K (=10) terms from the profiles of top M (=5) experts

	Query model	MAP	MRR
Model IB	QM_BL	.4838	.6280
	QM_PROFILES	.4933	.6422
Model 2	QM_BL	.4559	.6172
	QM_PROFILES	.4799	.6268

## Profile expansion using the Web

- Use candidate name + CSIRO as a query
- Use Yahoo! API to get documents
- Build candidate profile from result snippets

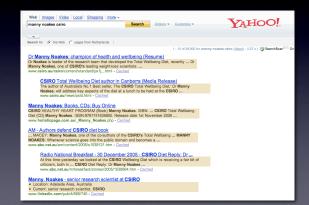
# Profile expansion using the Web



# Profile expansion using the Web

	Query model	MAP	MRR
Model IB	QM_BL	.4838	.6280
	QM_PROFILES	.4933	.6422
Web	QM_BL	.3840	.5217
	QM_PROFILES	.4153	.5478
Model IB (I) + Web (4)		.5323	.6371

# Profile expansion using the Web



#### Now what?

Model I



Model 2

#### What about...

Model I



Model 2

#### Model I + Model 2

	MAP	MRR
Model IB	.4838	.6334
Model 2	.4799	.6268
MIB + M2	.5267	.6828
MIB + M2 + Web	.5405	.6468

#### Wrap-up

- Model I: better candidate representations result in better performance
- Model 2: better document retrieval doesn't necessarily result in better ranking of experts
- Which of Model 1 or Model 2 should be preferred then?
  - It depends on...

#### Conclusions

Now that you've bought into Model 2, we'll tell you why to get Model 1 too

#### Future?

#### Questions?

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