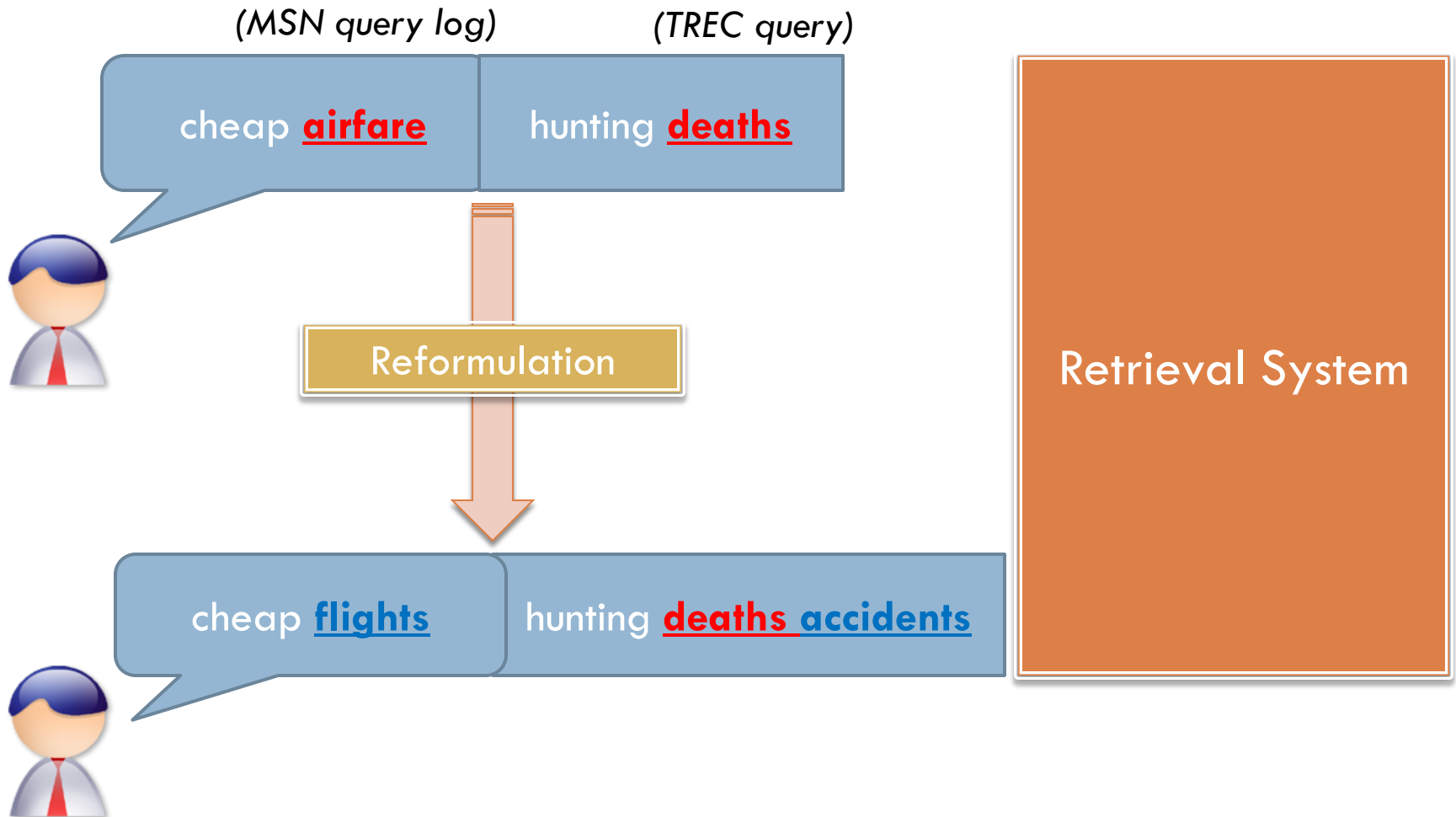




QUERY REFORMULATION USING ANCHOR TEXT

Van Dang and W. Bruce Croft

Query Reformulation



Related work

- Relevance feedback
 - ▣ Well-known, not in the scope of this paper
- Recent reformulation techniques rely on query logs
 - ▣ [Jones et al., 06], [Wang and Zhai, 08]
 - ▣ These techniques have proven effective for real web queries
 - Many of these queries are badly formulated (“cheap airfare”)
 - ▣ What if queries are good? (e.g. “hunting deaths”)
 - Can these techniques still make them better?

Do these methods work with good queries?

Related Work

- Recent reformulation techniques rely on query logs
 - ▣ [Jones, 06], [Wang and Zhai, 08]
- And so do many other tasks
 - ▣ Spelling correction: [Cucerzan et al., 04], [Ahmad et al., 05]
 - ▣ Stemming: [Peng et al., 07]
- Query logs might not be available to research community
 - ▣ Any alternatives?
- `<anchor text, url>` is just like `<query, clicked doc>`.

Can we use anchor text to simulate a query log?

Introduction

Do these methods work with good queries?

- Using TREC collections to evaluate the most recent log-based reformulation technique [Wang and Zhai, 08] on three tasks
 - ▣ Query Substitution
 - ▣ Query Expansion
 - ▣ Query Stemming

Can we use anchor text to simulate a query log?

- Uses anchor text in place of a query log

The Anchor Log

- Extract $\langle \text{anchor}, \text{url} \rangle$ pairs from the Gov-2 collection to create the *anchor log*.

	MSN Log	Anchor Log
# Total Queries	14 million	526 million
# Unique Queries	6 million	20 million
Avg. Query Length	2.68	2.62

- The anchor log is very noisy
 - “click here”, “print version”, ... don’t represent the linked page

Query Substitution

- A context of a word is the unigram preceding it

- Context distribution

$$P(c_i | w) = \frac{\text{count}_w(c_i)}{\sum_{c_j \in C(w)} \text{count}_w(c_j)}$$

The probability that the term c_i appears in w 's context

- The translation model

$$t(s | w) = \frac{e^{-D(P(\cdot|w) \| P(\cdot|s))}}{Z}$$

The KL divergence between the context distributions of w and s

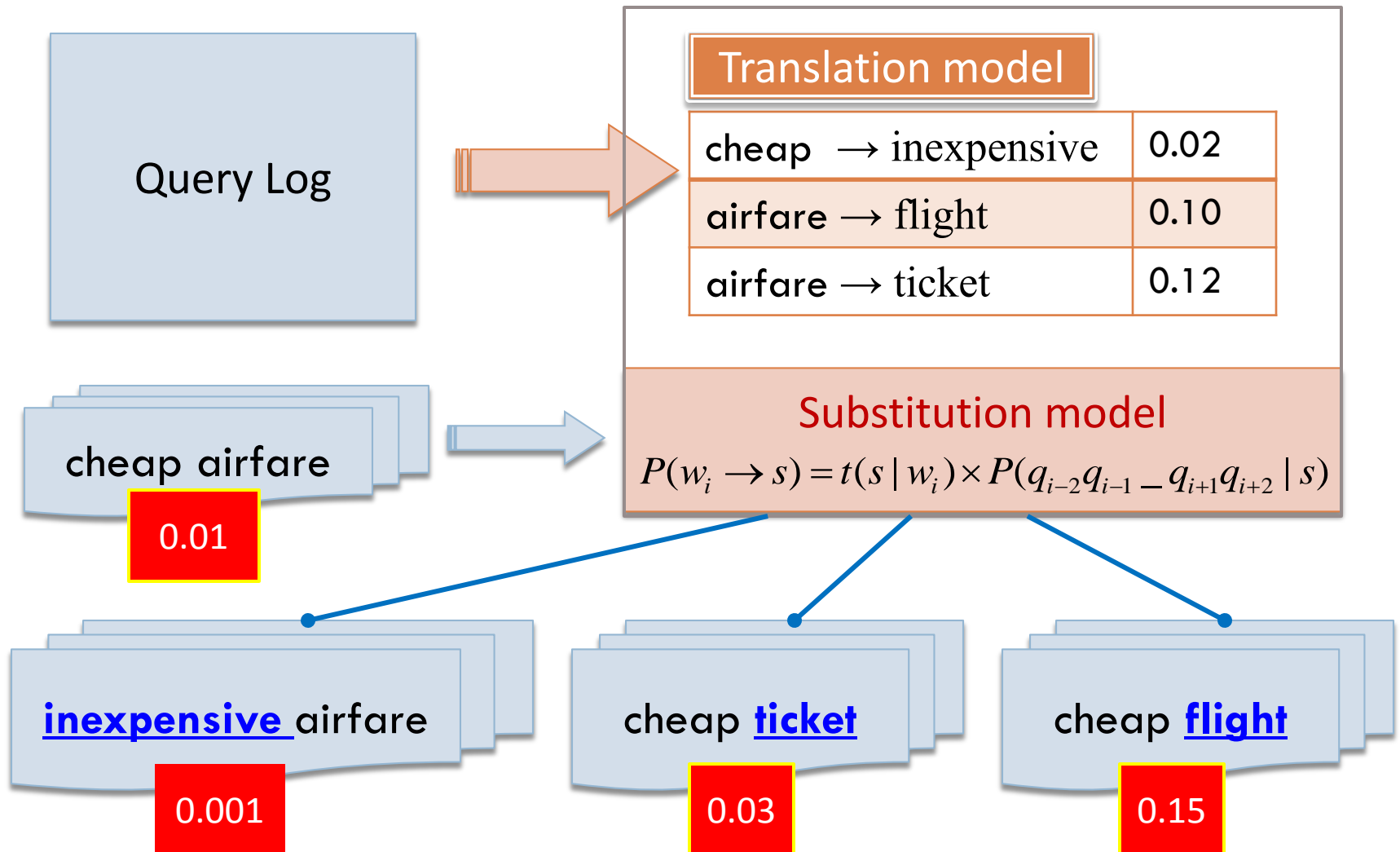
- The substitution model

- $Q = q_1, \dots, q_{i-2}, q_{i-1}, \mathbf{q}_i, q_{i+1}, q_{i+2}, \dots$ candidate = \mathbf{s}

How fit the new term is to the context of the current query

$$P(w_i \rightarrow s) = t(s | w_i) \times P(q_{i-2}q_{i-1} - q_{i+1}q_{i+2} | s)$$

Substitution: An example



Query Expansion and Stemming

- Query Expansion is exactly the same as substitution
 - ▣ We add the new term and keep the original term
- substitution: “cheap *airfare*” → “cheap *flight*”
- expansion: “cheap *airfare*” → “cheap *airfare flight*”
- Stemming
 - ▣ New terms are restricted to Porter-stemmed root terms
- “*drive* direction” → “*drive driving* direction”

Experimental Setup

- Evaluation

- ▣ Conducted on three TREC collections:

- Robust-04 (news)

- WT10G (web)

- Gov-2 (web)

Collection	# Documents	# Queries
Robust-04	0.5 M	250
WT10G	1.5 M	100
Gov-2	25 M	150

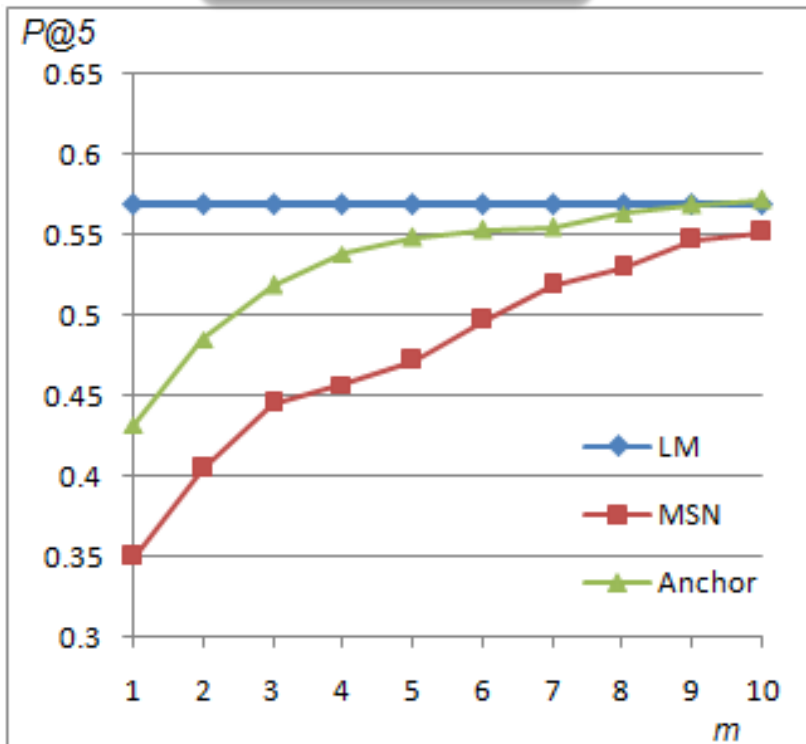
- Title queries vs. Description queries

Evaluation of Reformulated Query

Original Queries	MSN-Log Substitution	Anchor-Log Substitution
Query 1	Substitution 1 Substitution 2 ... Substitution m	Substitution 1 Substitution 2 ... Substitution m
...
Query n	Substitution 1 Substitution 2 ... Substitution m	Substitution 1 Substitution 2 ... Substitution m
P@5	P@5	P@5

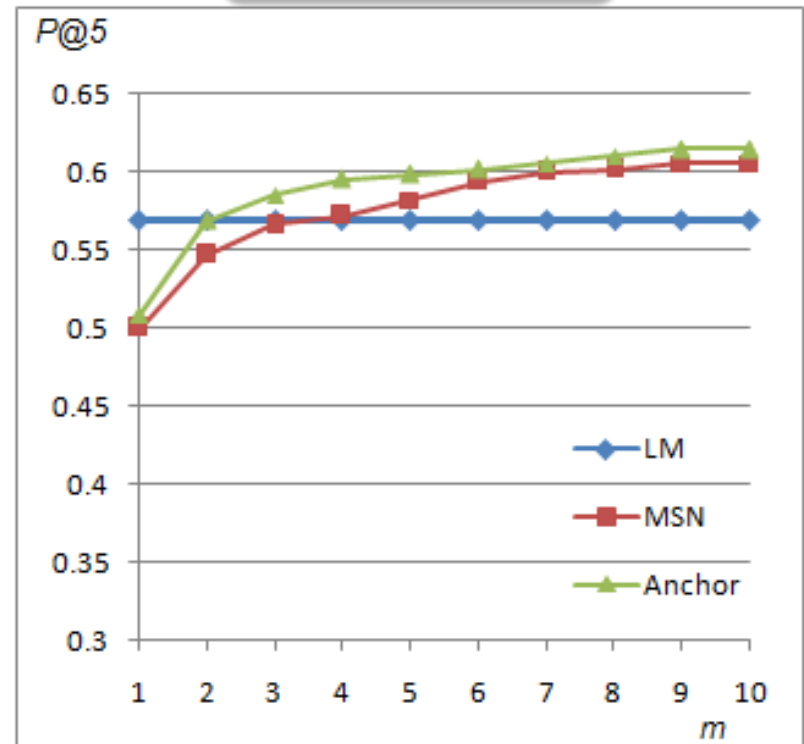
Substitution vs. Expansion (Title Q.)

Substitution



Does NOT help

Expansion



HELPS

The *Anchor log* is comparable to the *MSN Log*

“Chance” vs. “Risk”

- Substitution works for web queries [Wang and Zhai, 08]
 - Does not work here
 - Expansion is much better
 - **Why?**
- Both Substitution and Expansion
 - Introduce a new term to the query
 - “**chance**”: it brings more **relevant** documents
 - “**risk**”: it brings more **non**-relevant documents

“Chance” vs. “Risk”

□ Results

- ▣ Among 99 queries that were reformulated

	# Queries	P@5 change
Substitution helps	34	+110.94%
Expansion helps	32	+88.72%
Substitution hurts	32	-55.29%
Expansion hurts	14	-53.85%

Substitution

Helps
substantially

Hurts
drastically

Does NOT help
in general

Expansion

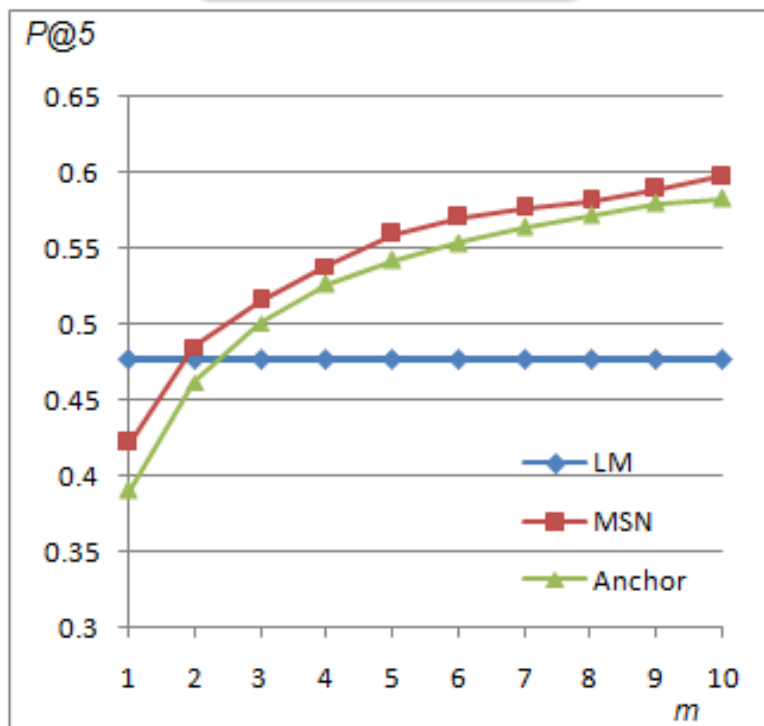
Helps more than it hurts, thus better

“Chance” vs. “Risk”

- Translation model does NOT provide « synonyms »
 - {women, men, children}
 - {diamond, gold, necklace, watches}
- It is undesirable to
 - “diamond smuggling” → “watches smuggling”
- TREC queries have good quality
 - Complete substitution is too risky

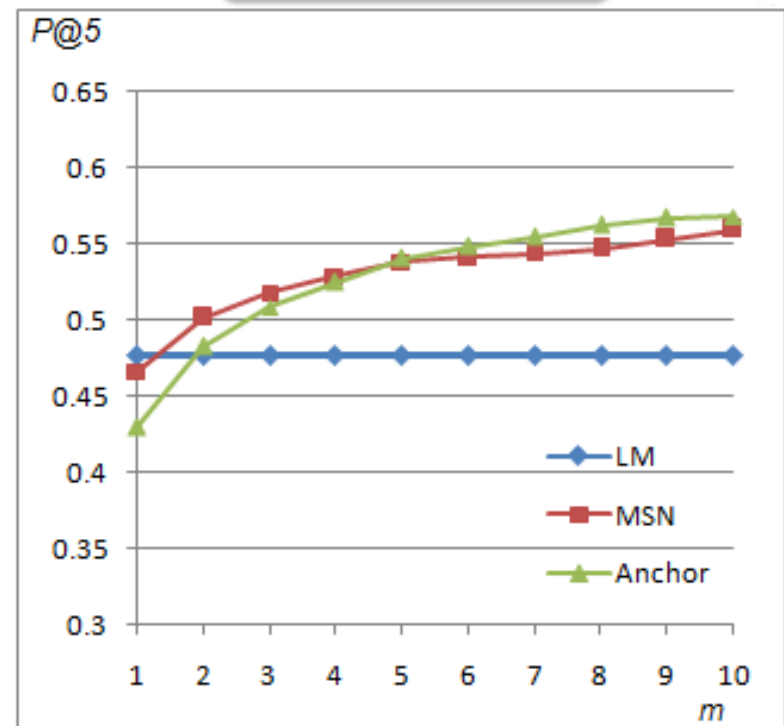
Substitution vs. Expansion (Desc Q.)

Substitution



Helps even more

Expansion



Helps

The *Anchor log* is comparable to the *MSN Log*

Substitution good for Long Query?

- Substitute w for $s = \text{drop } w + \text{add } s$
 - ▣ Q_{org} : original query
 - ▣ Q_{drop} : drop the target word
 - ▣ Q_{add} : add the substitution candidate

			Q_{org}	Q_{drop}	Q_{add}
MSN Log	Short Q.	WT10G	0.3291	0.2734	0.3468
		Robust04	0.4786	0.4009	0.4937
		Gov-2	0.5632	0.4529	0.5515
	Long Q.	WT10G	0.3158	0.3074	0.3768
		Robust04	0.4764	0.5138	0.5976
		Gov-2	0.5238	0.5578	0.6612

Dropping hurts

Dropping helps

[Kumaran et al., 09]

Similar improvement

It is the dropping that helps

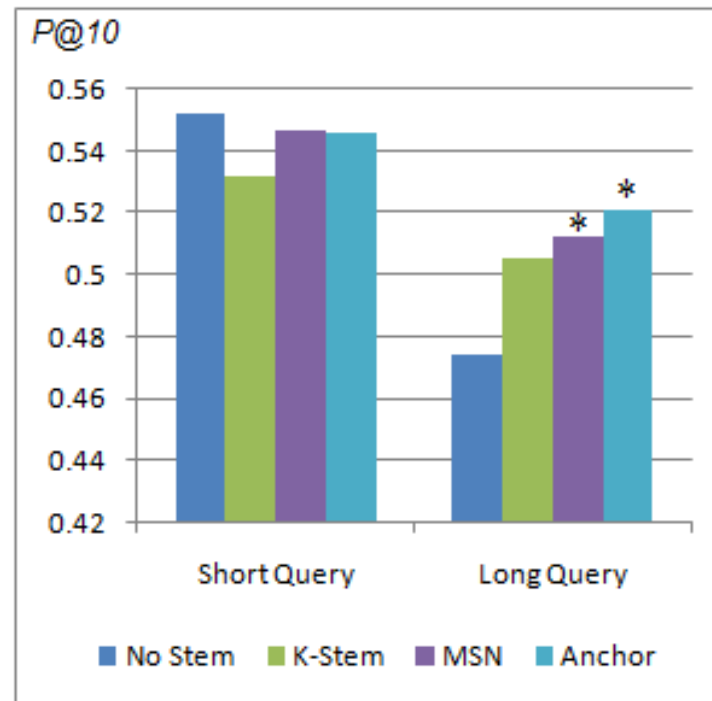
Stemming

□ We compare using P@10 queries

□ Unstemmed

□ Krovetz

□ Log-based (MSN vs. Anchor Log)



The *Anchor log* is comparable to the *MSN Log*

Conclusions

- **Anchor text gives comparable performance to MSN log on**
 - Substitution
 - Expansion
 - Stemming
- Expansion is more reliable than substitution
- Substitution helps with long (desc) queries
 - It is the dropping that helps
- Log-based stemming is promising