LambdaMerge: Merging the Results of Query Reformulations

Daniel Sheldon
Oregon State University
sheldon@eecs.oregonstate.edu

Milad Shokouhi, Martin Szummer, and Nick Craswell
Microsoft Research
{milads,szummer,nickcr}@microsoft.com

Merging Query Reformulations

Query reformulation algorithms

- Improve retrieval by alleviating Q-D mismatch bill gates bio → bill gates biography
- Or can make things worse e.g. bio → biog
- Or even drift off-topic e.g. bill → melinda

Approach: Run multiple queries, merge results

- Post-retrieval it is easier to detect quality/drift
- Multiple queries give diverse relevance evidence

Contribution: New merging methods

- 1. CombRW: Weighted CombSUM (unsupervised)
- 2. LambdaMerge: Supervised merging
 - Trained to maximize target such as NDCG
 - Incorporating quality and drift features
 - Robust to bad reformulations

LambdaMerge

Issue *k* formulations to search engine:

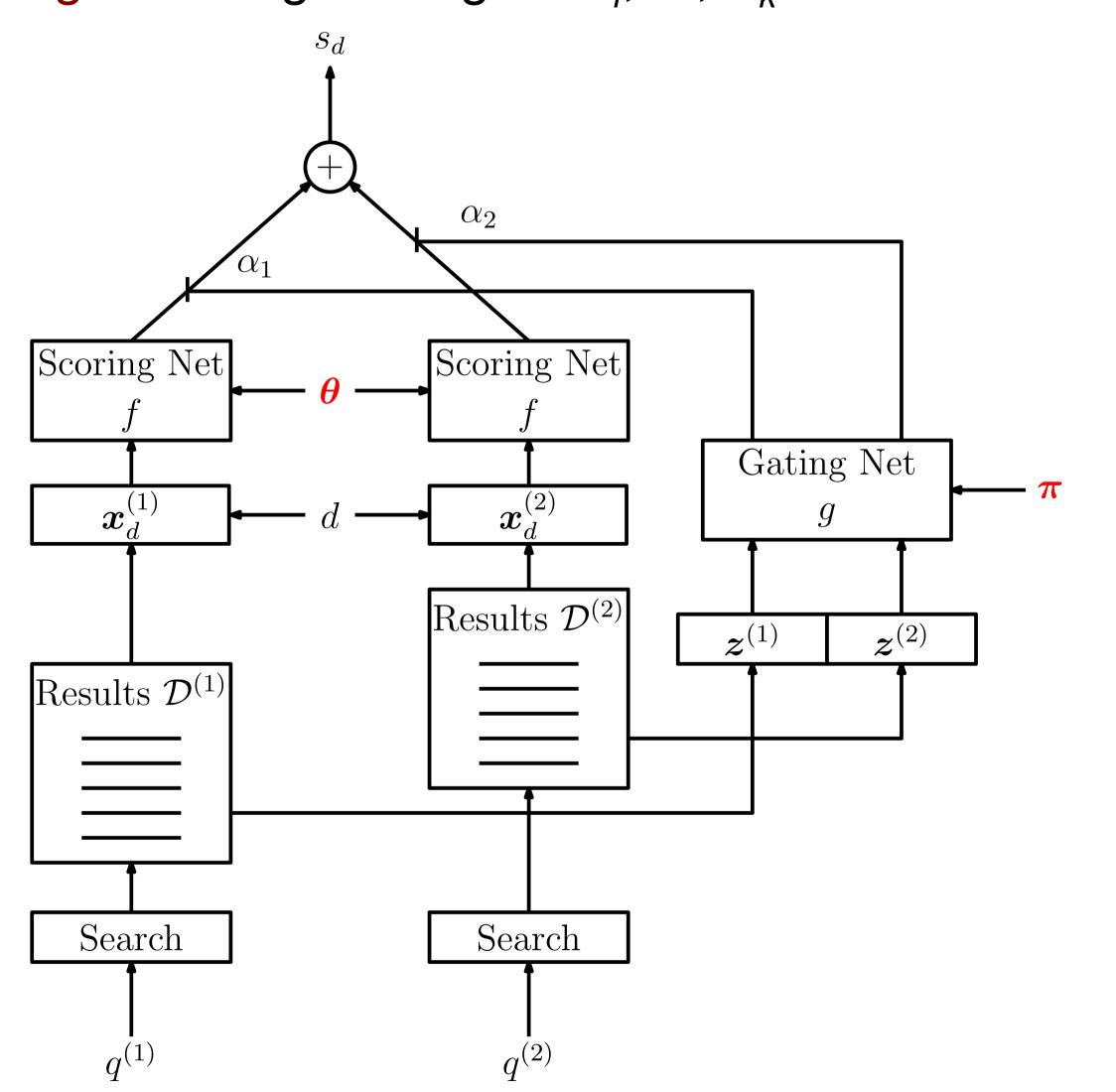
- Original query $q^{(1)}$ plus reformulations $q^{(2)},...,q^{(k)}$
- Get top-*N* lists *D*⁽¹⁾, ..., *D*^(k)

Generate features:

- Query-document features $\mathbf{x}_d^{(k)}$: relevance of document d specific to $D^{(k)}$
- Gating features $z^{(k)}$: drift + overall quality of $D^{(k)}$

Scoring net assigns score $f(\mathbf{x}_d^{(k)}; \boldsymbol{\theta})$ to each formulation-document pair

Gating net assigns weights $\alpha_1, ..., \alpha_k$ to formulations



Overall document score: $s_d = \sum_k \alpha_k \cdot f(\boldsymbol{x}_d^{(k)}; \boldsymbol{\theta})$

Parameters θ (scoring) and π (gating) trained by backprop with LambdaRank gradients to optimize NDCG [1]

Experiments

Test collections: Bing data and GOV2

Reformulations: Click graph random walk [2]

Single-query methods:

- ORG: Original query
- RW1: Most likely alternative query from RW
- RAPP-L: Predict best query (lin. regression) [3]
- RAPP(Ω): Choose query via NDCG@5 (oracle)

Merging methods:

- CombSUM: Sum scores
- CombRW: CombSUM with random walk weight
- LambdaMerge: Using these features

Table 1: List of features.		
Query-document	Score, Rank, $NormScore_{[0,1]}$,	
features	$NormScore_{\mathcal{N}(0,1)}$, $IsTopN$	
Gating features	ListMean, ListStd, ListSkew,	
(difficulty)	Clarity, RewriteLen, RAPP	
Gating features	IsRewrite, RewriteRank,	
(drift)	RewriteScore, Overlap@N	

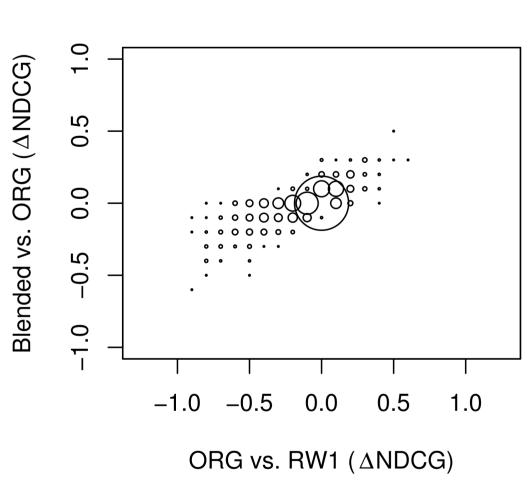
Results:

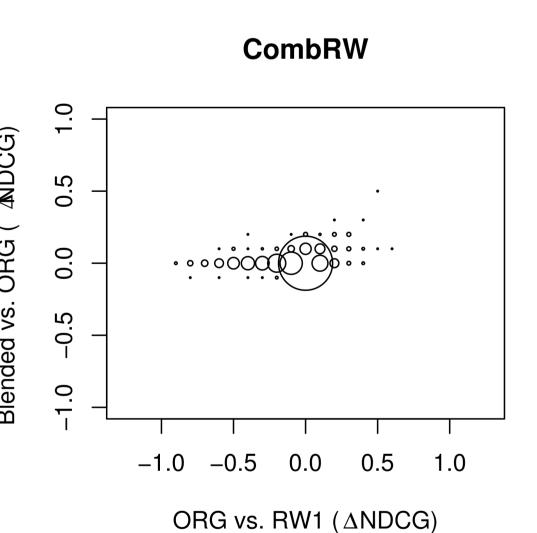
	NDCG@5	NDCG@10
ORG	0.538	0.524
RW1	0.422	0.387
CombSUM	0.510	0.486
CombRW	0.542	0.516
RAPP-L	0.534	0.524
$\lambda ext{-Merge}$	0.555	0.539
$\mathrm{RAPP}(\Omega)$	0.556	0.530

GOV2

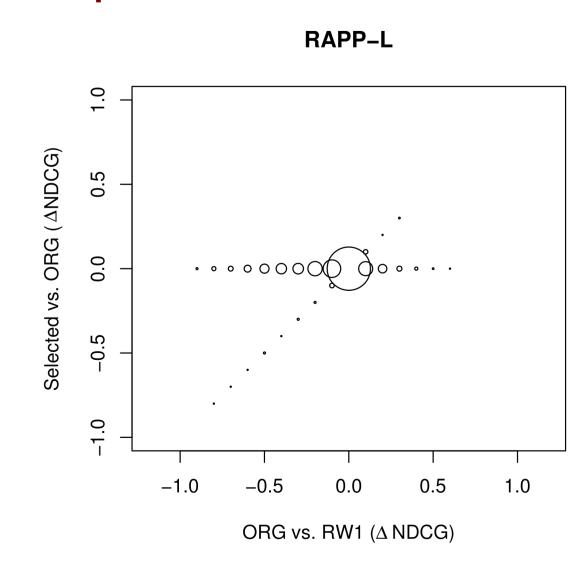
Bing

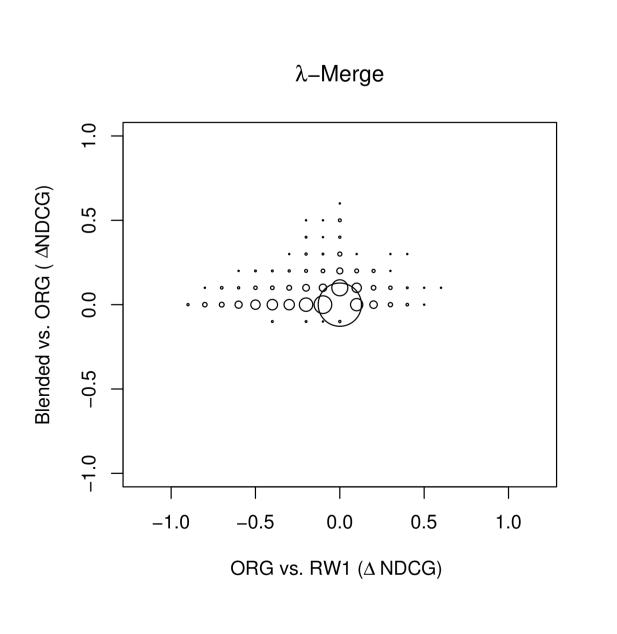
Robustness Analysis Unsupervised methods: Combsum





Supervised methods:





- [1] Burges, Ragno, and Le. Learning to rank with nonsmooth cost functions. NIPS 2006
- [2] Craswell and Szummer. Random Walks on the Click Graph. SIGIR 2007
- [3] Balasubramanian, Kumaran, and Carvalho. Predicting query performance on the web. SIGIR 2010 poster