Emotion-Aware Ranking

Information Retrieval Final Project Presentation

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Gary Lai, Zelalem Gero, Ali Ahmadvand, Reza Karimi

Motivation

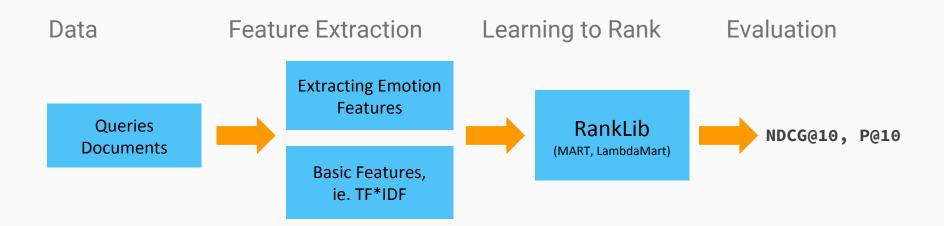
Sadness

What we want What we don't want VS. I missed my flight, that's great **Encouragement**, Anger, Happiness Embarrassment, Sadness I finished my exam, I killed it Power, Pride, Gun, Evil, Happiness, Fear, Death Achievement I broke up with my girlfriend, how awesome is it? Pride, Insecurity,

Happiness,

Peace

Pipeline



Datasets

• Emoji-labeled tweets

1.2 billion tweets (English, Text + Emoji)

Microblog TREC 2011

- o 10 millions tweets
- 50 queries
- o Each query has 1000 tweets with corresponding relevance score

Data Preprocessing

- Stop words, punctuation, numbers, urls removal
- Non-English character and word removal
- Stemming



Emotion Extraction

- For each tweet in the document set:
 - Predict the probability of each emotion
 - Total of 64 emotion classes
- Do the same for each tweet in the query set
- We get two vectors of 64 emotion class probabilities

* Emotion extraction paper [felbo2017]

Features used for LTR (MART, LambdaMART)

Baseline

- Cosine similarity between query and doc
- Retweet counts
- Favorited or not

Our Feature Vector

- Features from baseline
- Emotion from query (64 classes)
- Emotion from tweet/doc (64 classes)

Results

	Baseline		Baseline + Emoji	
	MART	LambdaM ART	MART	LambdaM ART
NDCG@10	0.4049	0.3779	0.4294	0.4737
P@10	0.149	0.1367	0.1633	0.1755

Improvement

	MART	Lambda MART
NDCG	6.05%	25.35%
MAP	9.59%	28.38%

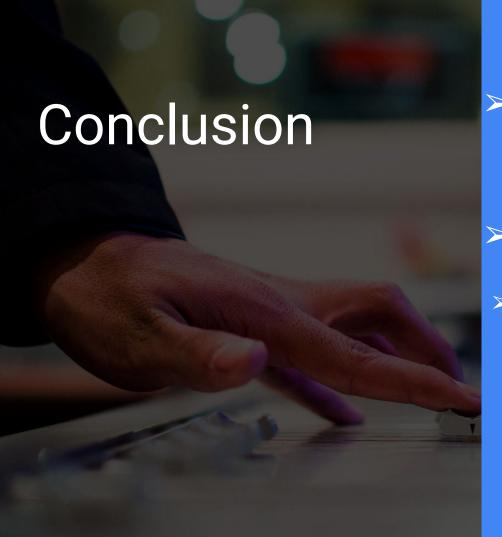


Potential Improvements

 More evaluation data from TREC 2013

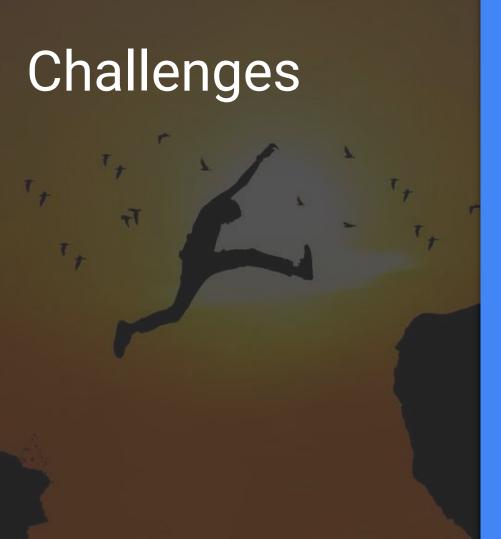
60 additional queries

More documents



- Extracted emotion features from Microblog TREC dataset
- > Fed it into the pipeline
- Got improvements over the baseline

Thank you!



Evaluation Dataset

Are Emotions Effective for other datasets?