

Sentiment Analysis of Term in Context: Task A

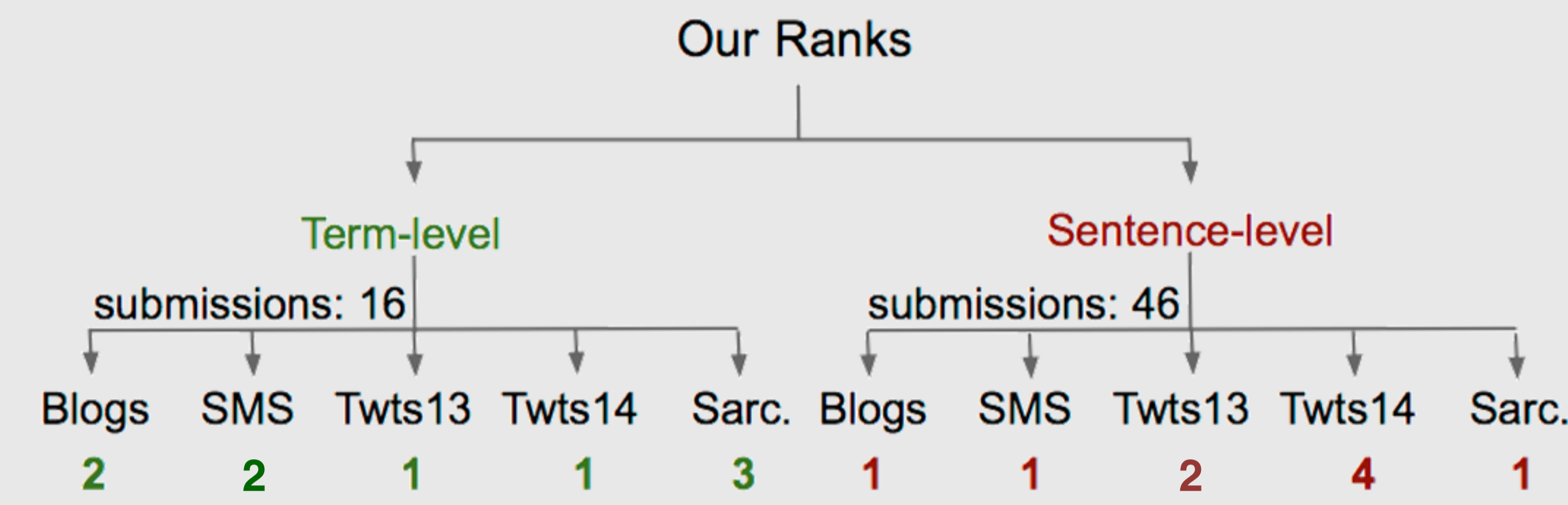
Polarity of the target: positive, negative, neutral?

Tweet: The movie is visually spectacular.
target is positive

Tweet: The movie felt like a slow documentary.
target is negative

Tweet: The NatGeo documentary was fascinating.
target is neutral

We present a sentiment analysis system which determines both term- and message-level polarity of a message.



The NRC-Canada-2014 System

Sentiment Analysis of Message: Task B

Polarity of the message: positive, negative, neutral?

Tweet: The movie is visually spectacular.
target is positive

Tweet: The movie felt like a slow documentary.
target is negative

Tweet: The NatGeo documentary is at 7pm.
target is neutral

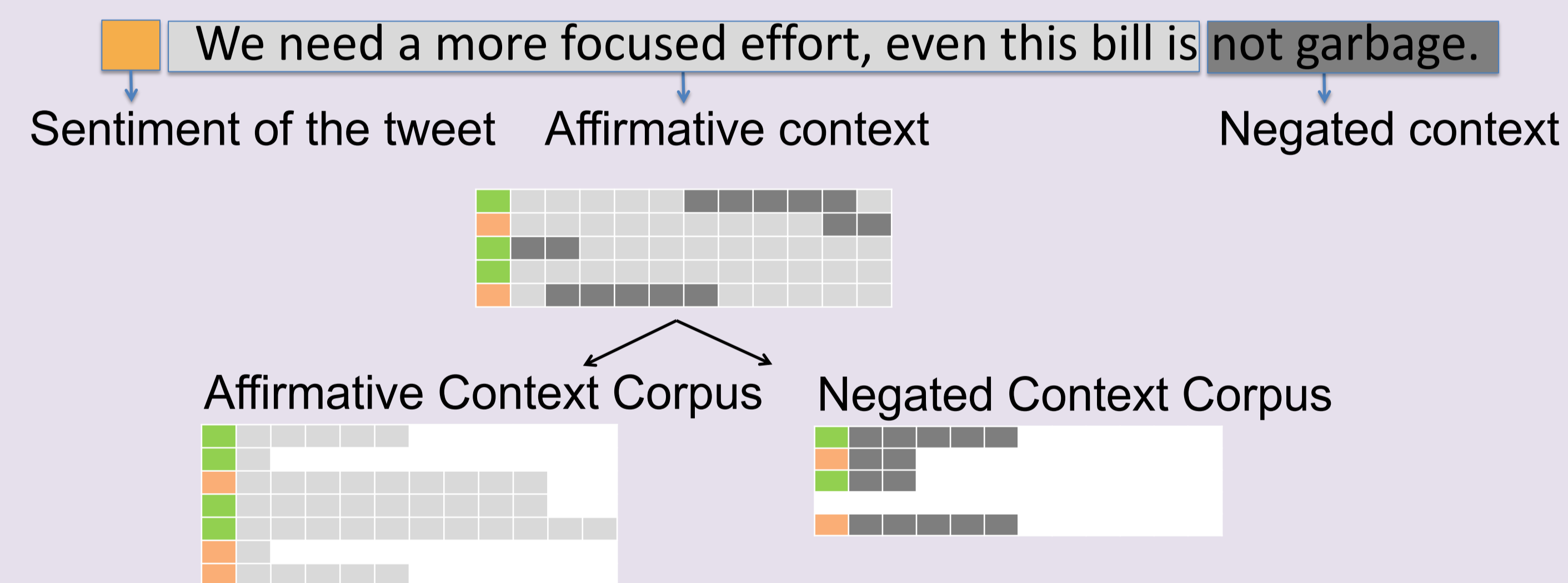
List of Features

sentiment lexicon	# sentiment words, score
word n-grams	spectacular, like documentary
char n-grams	spect, docu, visua
part of speech	N: 5, V: 2, A:1
negation	Neg: 1;
word clusters	probably, definitely, def
all-caps	YES, COOL
punctuation	!+: 1, ?+: 0, !?+: 0
emoticons	:D, >:(
hashtags	#excited, #NowPlaying
elongated words	soooo, yaayyy

Improving Sentiment Lexicons

We proposed a lexicon-based approach (Kiritchenko et al., 2014) to determine the sentiment of words in affirmative and negated context.

- Hashtags or emoticons are used as the gold labels of sentiment in tweets (Mohammad, 2012).
- A tweet corpus is split into two parts: Affirmative Context Corpus and Negated Context Corpus.



- For each word w in a tweet corpus, an association score is generated using the Affirmative and Negated Context, separately.

$$\text{score}(w) = \text{PMI}(w, \text{positive}) - \text{PMI}(w, \text{negative})$$

- If $\text{score}(w) > 0$, then word w is positive
- If $\text{score}(w) < 0$, then word w is negative

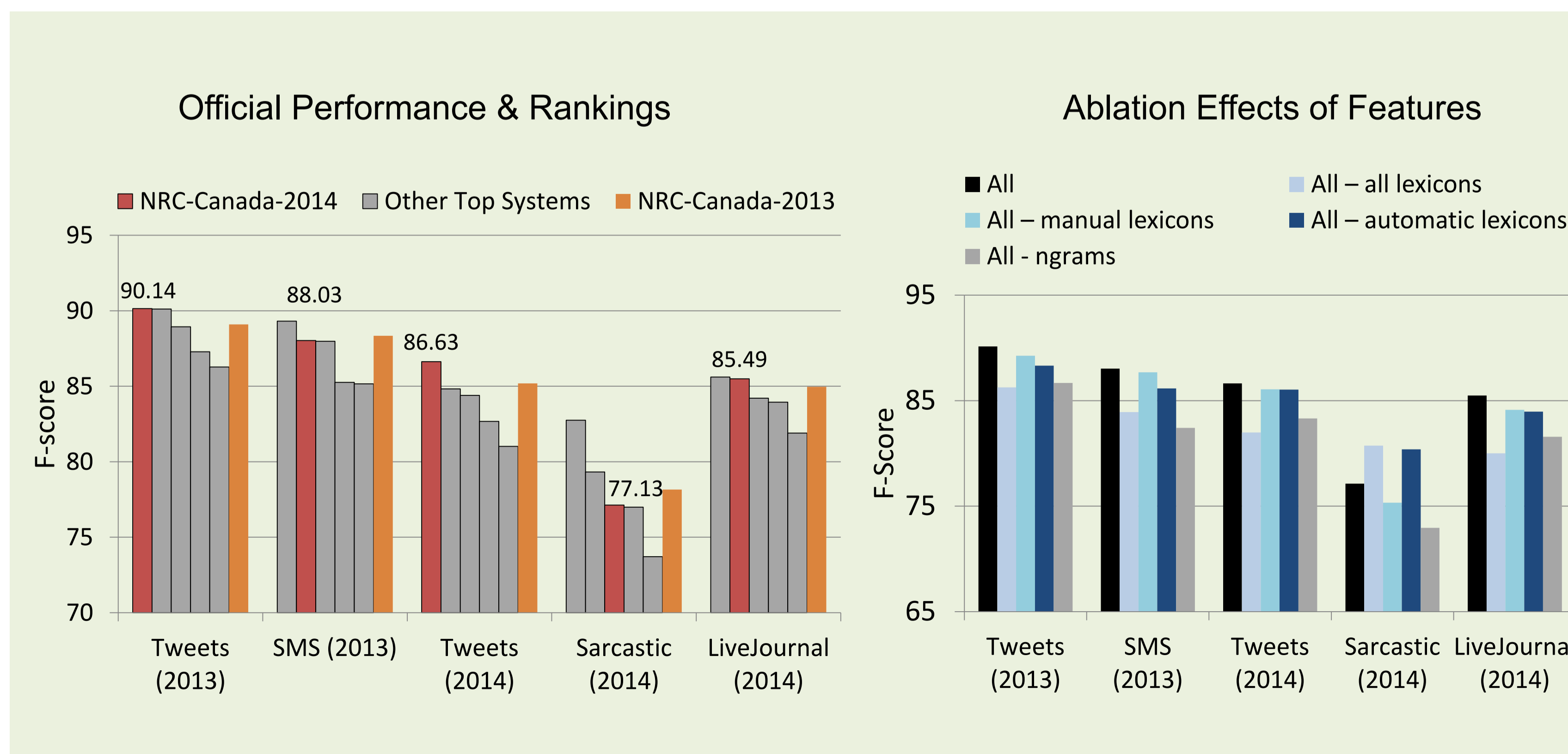
Examples: garbage **-0.65** garbage_negatedContext **0.10**

- Using these new lexicons, we generate sentiment-lexicon features and add them to our systems (e.g., # sentiment words and sentiment scores.)

Discriminating Negation Words

- Different negation words, e.g., *never* and *didn't*, often have different effects on sentiment (Zhu et al., 2014; Taboada et al., 2011).
- In our term-level system, we take this into consideration. For example, a word *acceptable* is treated as *acceptable_beNever* and *acceptable_cannot* when it appears after word *never* and *cannot*, respectively.

Results: Task A



Conclusions

- Best micro-averaged results over all 5 datasets in both tasks.
- Message-level task: best results on 3 out of 5 datasets; term-level task: best results on 2 out of 5 datasets.
- Most useful features: sentiment lexicons, especially automatic tweet-specific lexicons.

NRC Sentiment Lexicons are available for download:

www.purl.com/net/sentimentoftweets

Results: Task B

