

#### A Framework for Political Portmanteau Decomposition



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    - libtard = liberal + retard
      repugnican = repugnant + republican

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    - libtard = liberal + retard
  - offensive; political framing
  - creative, humorous, slang, sticky
  - can be used in hate speech



repugnican = **repugn**ant + republ**ican** 

#### Contributions

- Framework for identifying political portmanteau from the web
- Algorithm for PP detection and decomposition into root words
- First shared dataset of PP

# Method



- Extract words from Reddit news comments
- Apply slang detection algorithm
- Classify the detected words into PP vs not-PP

Hossain, Nabil, Thanh Thuy Trang Tran, and Henry Kautz. "Discovering Political Slang in Readers' Comments." In ICWSM 2018.

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  - $E + C \rightarrow PP$  or  $C + E \rightarrow PP$



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# **Model Details**

- $\beta$  distribution Model no contextual features
  - Edit distances, word length, usage frequency
  - capture sound blending and word popularity
- XGBoost uses pre-trained GloVe word vector features from comments
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Model	Top1	Top3	Top5	Top10
Random	47.48	57.75	62.57	67.61
Beta Model (no context)	62.17	72.43	75.45	79.07
XGBoost (with context)	76.23	86.72	90.34	93.36

**PP Decomposition Accuracy** 



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Model	Accuracy
Chance (always predict "not PP")	66.2
Bi-LSTM	69.5
Bi-LSTM + GloVe	75.1
BERT	78.8
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