

# MEASURING SENTIMENT DURING THE 2016 U.S. PRESIDENTIAL ELECTION USING EMOJI ANALYSIS

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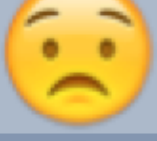
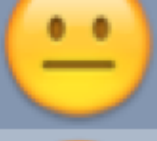



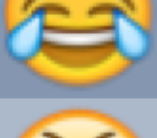
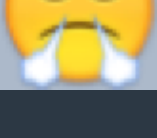
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## INTRODUCTION

- Sentiment analysis of Twitter data has been used to predict election results<sup>1</sup>
- However, sentiment analysis and results can be confusing to the general public
- Pairing sentiment analysis with emoji analysis might offer a way to quickly disseminate findings to the general public
- Emoji analysis needs to be tested for efficacy

## METHODOLOGY

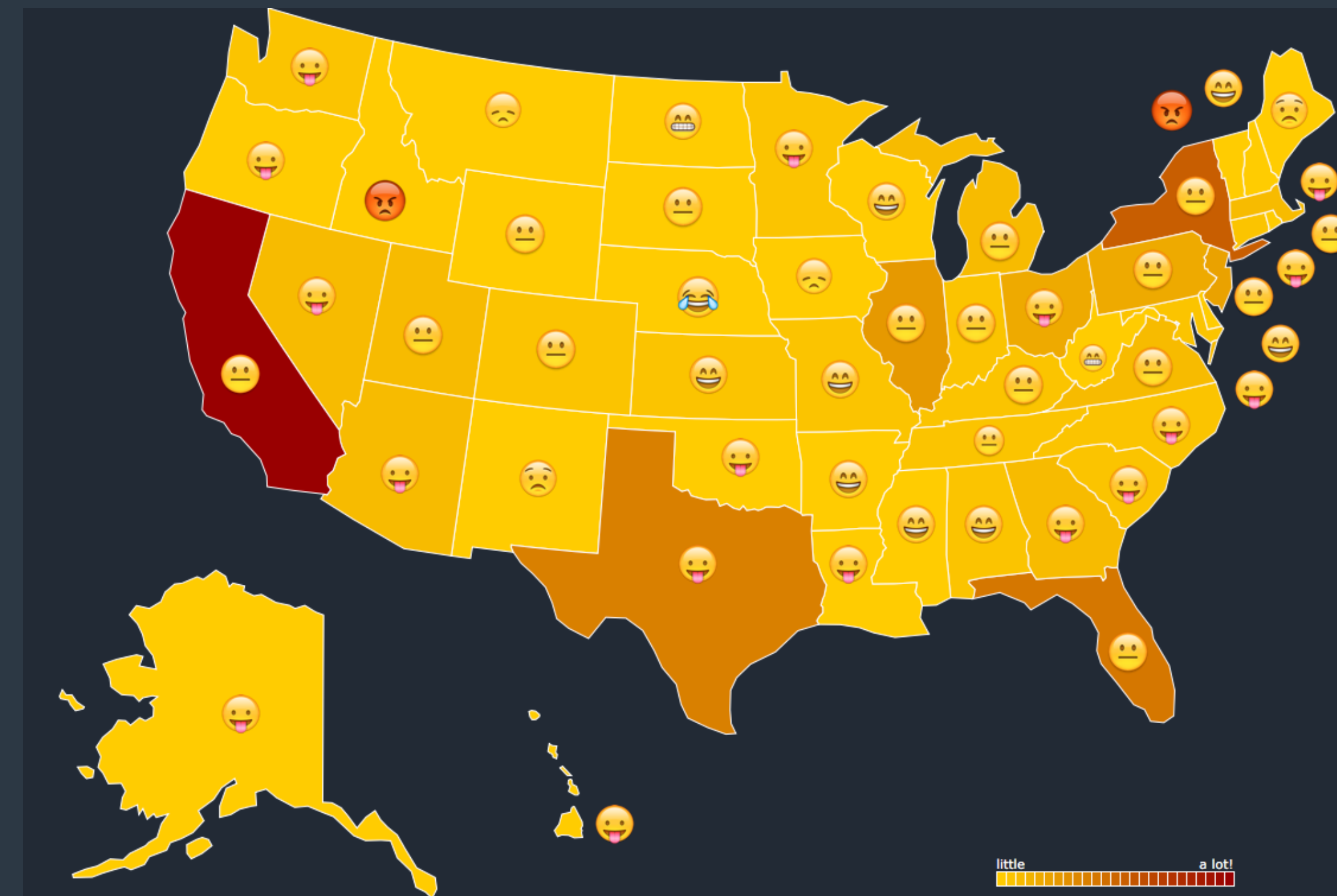
- Collected tweets containing **#clinton**, **#trump**, or **#election2016**
  - Time range: September 6 – November 14
- Created a smaller dataset of tweets containing **geo-location data** and one of **67 different emojis** (**n=8,293**)
- Separated tweets containing each hashtag, resulting in three sub-datasets:
  - #clinton: 680 tweets
  - #trump: 4,305 tweets
  - #election2016: 3,407 tweets
- Emojis were rated on a -5 to 5 polarity (emotion) scale in prior work<sup>2</sup>
- Generated state-by-state heatmaps of aggregate emotion score and tweet volume
- Compared these heatmaps to general election results to draw conclusions about value of emoji analysis

Emoji	Name(s)	Score
	rage; pout	-5
	angry	-4
	worried	-3
	disappointment	-2
	frowning	-1
	neutral_face	0
	stuck_out_tongue	1
	smile	2
	grinning	3
	joy	4
	triumph	5

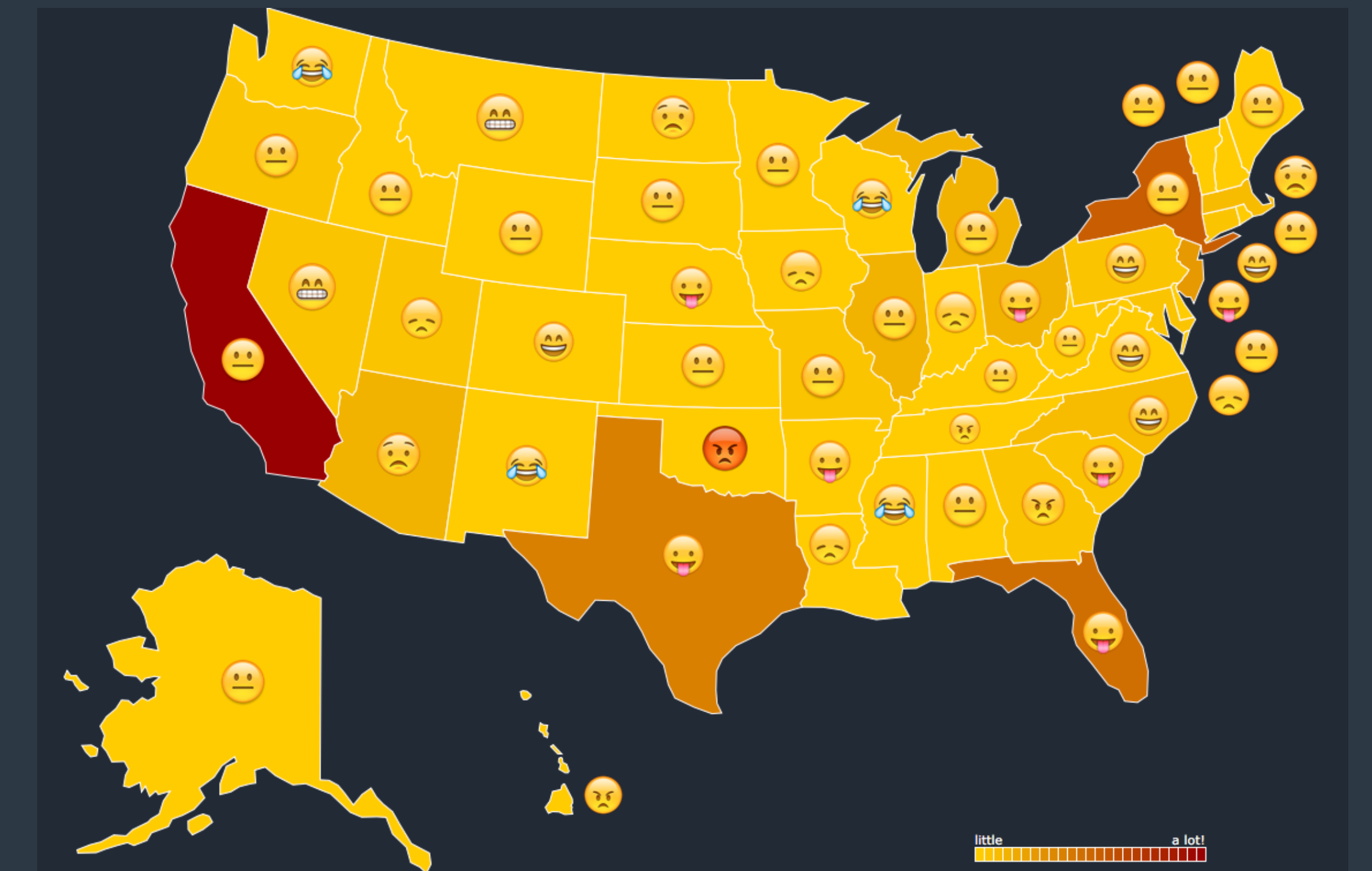
*Examples of emojis, name(s), and polarity score*

## RESULTS

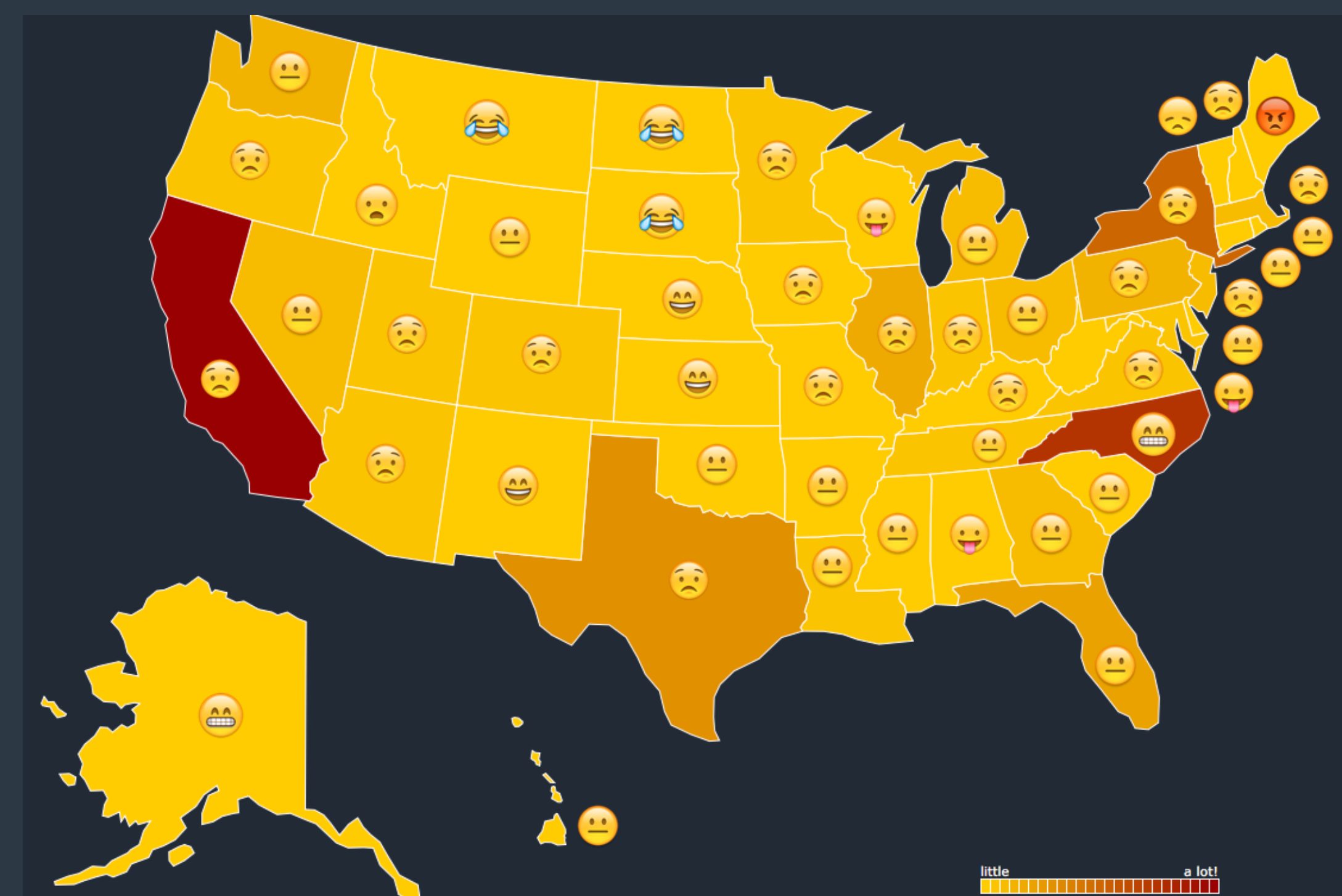
#trump heatmap



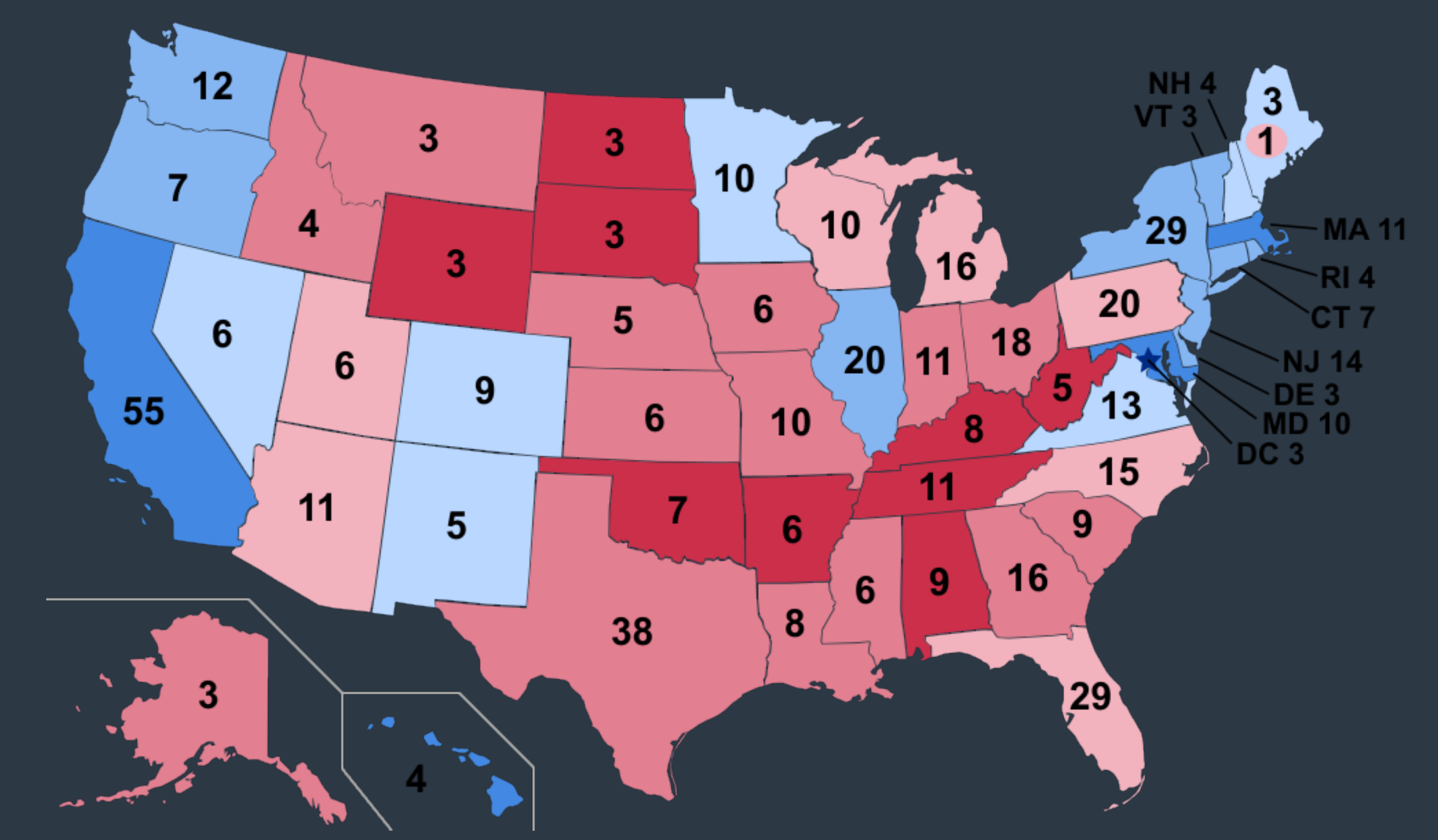
#clinton heatmap



#election2016 heatmap



2016 Presidential election results<sup>3</sup>



## CONCLUSION

- Emoji analysis can act as an easily decipherable complement to traditional sentiment analysis
- In “swing states”, the emoji analysis frequently matched the outcome of the general election
  - States with a higher score in the #trump set voted for Donald Trump
  - States with a higher score in the #clinton set voted for Hillary Clinton
- States that traditionally voted for specific parties were harder to decipher
- This type of complimentary analysis might be more effective in less contentious situations (i.e. an election with better liked candidates)
  - Future work can investigate this situational effect

<sup>1</sup>Bermingham, A., & Smeaton, A. F. (2011). On using Twitter to monitor political sentiment and predict election results.

<sup>2</sup><https://github.com/woorm/emoji-emotion/blob/master/support.md>

<sup>3</sup>By Ali Zifan - File:Electoral College 2016.svg, CC BY-SA 4.0, <https://commons.wikimedia.org/w/index.php?curid=53004568>