

Does
Optical Character Recognition
and Caption Generation
Improve Emotion Detection in
Microblog Posts?

June 22<sup>nd</sup>, 2017

Roman Klinger



#### **Motivation**



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Happy Birthday to my incredible father @realDonaldTrump. I hope that the year to come is your best yet!







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My wife says to let my #anger toward my #Lupus #FuelMyFight. It works.

# I have an attitude

AND IT'S NOT BECAUSE I'M A BAD PERSON, BECAUSE I Wasn't raised right or because I'm Just Mean.

#### DAILY BASIS

J'm tired, J'm grumpy, and all J want to do is sleep. But, of course, my body doesn't want me to SOIHAVE TO BE HERE, BE AWAKE, AND BE IN PAIN.
I'M SO IRRITATED I DON'T EVEN WANT TO DEAL WITH MYSELF SOMETIMES.

SO, PLEASE EXCUSE ME ON MY HARD DAYS. IT'S NOT YOU, IT'S NOT ME, IT'S MY BODY.

# **Outline**

- Motivation
- 2 Emotion Analysis on Tweets
- Methods & Experimental Setting
- 4 Results
- 5 Conclusion & Discussion

# What is Emotion Analysis?

#### Sentiment analysis

positive vs. negative (neutral, mixed)

Emotion analysis discrete (Ekman/Plutchik)

discrete emotion classes

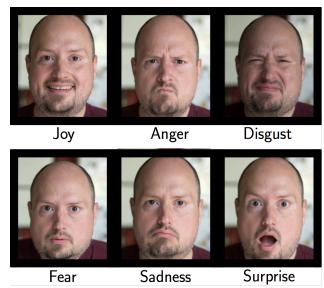
Subjectivity analysis

subjective vs. objective

**Emotion analysis** continuous (Posner/Russell/Peterson)

valence and arousal

# **Emotion Models: Ekman**



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# Task Description and Research Question

#### Task

Given a Tweet, assign an emotion from the set: anger, fear, joy, sadness, surprise, disgust, love, shame, trust

#### Research Questions

- Which feature sets contribute to the task?
  - Tweet Text
  - Text extracted from attached image
  - · Automatically generated caption for attached image
- Can we operationalize this with off-the-shelf tools?

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#### **Feature Extraction & Classifier**

#### **Tweet Text**

- Bag of words, keep [^A-Za-z0-9# ]
- Username blinding, emotion hashtags are ignored

#### **Optical Character Recognition**

- Tesseract 3.04.01
- Ignore output with less than 6 bytes

#### **Caption Generation**

- NeuralTalk 2
- · Pretrained COCO data set model
- Classifier: linear maximum entropy classifier with L2 regularization

# **Example for Caption Generation**



@realDonaldTrump. I hope that the year to come is your best yet!



"a man and a woman standing next to each other"

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My wife says to let my #anger toward my #Lupus #FuelMvFight. It works.

## I have an attitude

I'm tired, I'm grumpy, and all I want to do is sleep. But, of course, my body doesn't want me to SO I HAVE TO BE HERE, BE AWAKE, AND BE IN PAIN. WANT TO DEAL WITH MYSELF

SOMETIMES . SO. PLEASE EXCUSE ME ON MY HARD DAYS, IT'S NOT YOU, IT'S NOT ME, IT'S MY BODY.

9 have an attitude

AND IT'S NDT BECAUSE I'M A BAD PERSON, BECAUSE I WASN'T RAISED RIGHT CR BECAUSE I'M JUST MEAN. OSOEENOOE 'EEEE OENO NN

J'm timd, J'm grumpy, and all 2 want tvdoiz sleep. 3m. of course, my body doesn't want me to SIT I HAVE TIT BE HERE, BE AWAKE, AND BE IN PAIN. I'M so IRRITATED I DON'T EVEN

WANT TO DEAL WITH MYSELF SOMETIMES.

SD. PLEASE EXCUSE ME ON MY HARD DAYS. IT'S NOT YDU, IT'S NOT ME, IT'S MY BODY.

"a picture of a man in a suit"

### Data collection: self labeling

Crawl corpus from Twitter with hashtags for emotions

#### Sample three corpora of 200,000 Tweets

- D<sub>all</sub> without constraints
- D<sub>OCR</sub> from all instances with image and with OCR output
  - ⇒ Tweets with images with text
- D<sub>Vis</sub> from all instances with image and without OCR output
  - ⇒ Tweets with images without text

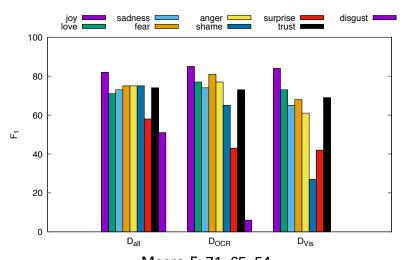
(captions are generated on all images)

# **Corpus Statistics**

Emotion	D <sub>all</sub>	$\mathbf{w}/\Phi_{OCR}^d$	$\mathbf{w}/\Phi^d_{Vis}$	D <sub>OCR</sub>	$D_{Vis}$
joy	91,836	6,927 (8%)	18,672 (20 %)	92,066	111,604
love	41,470	3,477 (8%)	8,963 (22%)	46,290	50,974
sadness	26,521	1,495 (6%)	2,952 (11 %)	19,707	14,370
fear	12,721	1,490 (12%)	2,299 (18%)	19,400	7,925
anger	11,902	831 (7%)	1,384 (12%)	10,379	5,317
surprise	5,492	195 (4%)	792 (14%)	2,790	5,681
trust	5,170	561 (11 %)	886 (17 %)	7,274	3,151
shame	4,562	138 (3%)	193 (4%)	1,988	862
disgust	326	8 (2%)	22 (7%)	106	116
All	200,000	15,122 (8 %)	36,163 (18%)	200,000	200,000

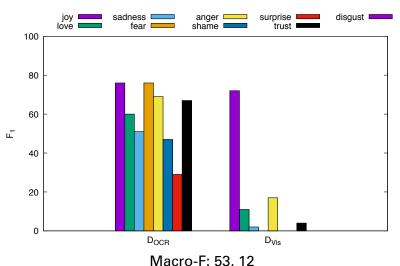
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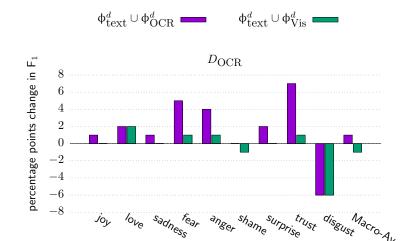
# Experiment 1: Tweet text features only on $D_{all}$ , $D_{OCR}$ , $D_{Vis}$



Macro-F: 71, 65, 54
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# Experiment 2: OCR Features on $D_{OCR}$ , Caption Features on $D_{Vis}$

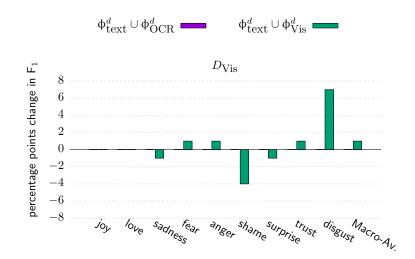




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# Feature Impact of OCR and Captions on $D_{Vis}$



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#### Conclusion & Discussion

- Optical character recognition improves emotion recognition
- Caption generation does not help
- Classification with images works surprisingly well based on Tweet text only (at least on frequent emotion classes)

#### Multiple classifiers for different object types (e.g. Caffe2 model zoo)

- Facial emotion recognition
- Age, gender
- Landmarks
- Include deep learning emotion classifier taking the image as input directly
- Tweet generator instead of caption generator?

