## Crowdsourcing the Creation of an Emotion Lexicon



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#### Outline

- Emotions
- Crowdsourcing
- Annotation analysis

## •° EMOTIONS

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#### Examples



joy sadness surprise hate love relief shame fear anxiety And many many more.



### Pervasiveness

- Cross-cultural
  - Paul Ekman's work on facial expressions
- Animals express/feel emotions too
  - Charles Darwin's
    The Expressions of the
    Emotions in Man and Animals











#### CHARLES DARWIN

The Expression of the Emotions in Man and Animals

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## Definition



A mental reaction subjectively experienced as strong feeling usually directed toward a specific object and typically accompanied by physiological and behavioral changes in the body.

- subjectively experienced
- strong feeling
- directed toward (or evoked by) a specific object



# Painting



#### The Destroyer - Frank Frazetta



#### Sentence



(Phil, from the San Francisco Chronicle) speaker/writer



Death threats over South Park episode Event

When your cartoon can get you killed



listener/reader



Extremists Participants



Trey Parker, Matt Stone Participants

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#### Our focus: words



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# Our goal

- Create and analyze an emotion lexicon through input from a large number of people.
- Examples:
  - vampire typically evokes fear
  - vacation typically evokes joy
  - death typically evokes sadness
  - result typically evokes anticipation

# Motivation for emotion detection

- Devising automatic dialogue systems that respond appropriately to different emotional states of the user.
  - customer relation models
  - intelligent tutoring systems
  - emotion-aware games
- Tracking sentiment towards politicians, movies, products.
- Determining emotional intelligence.
- Assisting in writing e-mails, documents, and other text to convey desired emotion (and avoiding misinterpretation).

# Motivation for emotion detection (continued)

- Depicting flow of emotions in novels and other books
- Identifying what emotion a newspaper headline is trying to evoke
- Re-ranking and categorizing information/answers in a web 2.0 world
- Detecting how people use emotion-bearing-words and metaphors to persuade and coerce others
- Developing humanoid robots



### Which emotions?





# Kinds of emotions



- Instinctual versus cognitive emotions
  - sensing and perceiving from the amygdala versus thinking and reasoning through the prefrontal cortex
- Basic versus complex
  - base emotions lead to more complex ones
- Categorization based on duration
  - some emotions occur over a period of seconds (e.g. surprise), whereas others can last years (e.g. love)



### **Base emotions**

- Ekman: 6
  - joy, sadness, fear, anger, surprise, disgust
- Plutchik: 8
  - Ekman's 6 + anticipation + trust
  - 4 pairs of antonymous emotions
- More proposals by Parrot, Loyban, and others

# Robert Plutchik's psychoevolutionary theory

- Biologically primitive
- Evolved to increase "reproductive fitness"
- Triggers for behavior with high survival value
  - e.g., fear inspires fight-or-flight response



# Plutchik's wheel of emotions



## Using Mechanical Turk for LARGE-SCALE EMOTION ANNOTATION

# Amazon's Mechanical Turk

- Requester
  - breaks task into small independent units HITs
  - specifies:
    - compensation for solving each HIT
    - # of independent annotations required for each HIT a.k.a. # of assignments/HIT
  - uploads HITs
- Turkers
  - attempt as many HITs as they wish
- Requester
  - inspects each assignment: approves or rejects

# Amazon's Mechanical Turk: Features

- Inexpensive
  - \$1/hour is not uncommon
- Convenient
  - Web-based
  - Scripts to upload HITs and review assignments
- Takes care of certain ethics issues
  - Anonymity
  - No pressure on workers to solve HITs

# Amazon's Mechanical Turk: Challenges

- Annotator time is precious
  - Minimum reading, minimum writing
  - Maximum information throughput
- Requestor time is precious
  - Automatic review and assimilation of annotations

**One solution**: Multiple choice questions, with examples instead of explanations.



# Example question

How much does vampire evoke/produce the emotion fear? (For example, horror and scary may strongly evoke fear.)

- vampire does not evoke fear
- $_{\rm C}\,$  vampire weakly evokes fear
- $_{\rm C}$  vampire moderately evokes fear
- o vampire strongly evokes fear

# Amazon's Mechanical Turk: Challenges

- Malicious annotations
  - Random selection or garbage data entry
  - Deliberate incorrect annotation
- Inadvertent and infrequent errors
  - Turker attempts HITs for unfamiliar words too

# **Emotion annotation: Challenges**

 Words used in different senses and in different contexts can evoke different emotions.

High aspect ratio wings allow low speed <u>flight</u>.

The fight or <u>flight</u> response is crucial for survival.

- How to convey the target sense to the annotator?
  - definitions are long
  - need to discourage annotation for unfamiliar words





# Our solution

**Directions:** Attempt HIT only if you are familiar with the word. Words in different senses may have different emotion associations. Question 1 will guide you to the intended sense. Q1. Which word is closest in meaning (most related) to flight?

- buying
- avoidance
- o doubt
- $\circ$  boredom
- Near-synonym is taken from a thesaurus.
  - Categories in a thesaurus act as coarse senses
- Three distracters are chosen at random

# **Emotion annotation: Challenges**

Words used in different senses and in different contexts can evoke different emotions.

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# Identifying bad assignments

- If the word choice question is answered wrongly, then the whole assignment is discarded (answers to all questions in the HIT by the Turker are discarded)
- If an annotator gets more than 1 in 3 questions wrong, then we assume they are not following instructions.
  - We reject all their assignments.

# Amazon's Mechanical Turk: Challenges

Malicious annotations

- Random selection or garbage data entry
- Deliberate incorrect annotation
- Inadvertent and infrequent errors
  - Turker attempts HITs for unfamiliar words too

Will detect it 75% of the time.



# Target n-grams

- Conditions:
  - Most frequent terms in the Google n-gram corpus
  - Must be in the thesaurus in just one or two categories
- Most frequent monosemous n-grams in each of the following categories:
  - noun unigrams (200)
  - noun bigrams (200)
  - verb unigrams (200)
  - verb bigrams (200)
  - adjective unigrams (200)
  - adjective bigrams (200)
  - adverb unigrams (200)
  - adverb bigrams (200)



### Target n-grams (continued)

- Most frequent monosemous terms in the General Inquirer (GI) that are:
  - marked as positive (200)
  - marked as negative (200)
- Terms in WordNet Affect Lexicon (WAL) that have one or two senses and are:
  - marked as anger terms (107)
  - marked as disgust terms (25)
  - marked as fear terms (58)
  - marked as joy terms (109)
  - marked as sadness terms (86)
  - marked as surprise terms (39)

#### 2176 terms in all.



#### **Questions:**

- 1. Which word is closest in meaning (most related) to flight?
  - $\circ$  buying
  - $\circ$  avoidance
  - o doubt
  - $\circ$  boredom
- 2. How positive (good, praising) is flight (for example, nice and **excellent** are strongly positive):
  - flight is not positive
  - flight is weakly positive
  - flight is moderately positive
  - flight is strongly positive



#### **Questions (continued):**

- 3. How negative (bad, criticizing) is flight (for example, poor and pathetic are strongly negative):
  - flight is not negative
  - c flight is weakly negative
  - c flight is moderately negative
  - flight is strongly negative
- 4. How much does flight evoke/produce the emotion joy (for example, happy and fun may strongly evoke joy):
  - flight does not evoke joy
  - c flight weakly evokes joy
  - flight moderately evokes joy
  - ⊂ flight strongly evokes joy

## **O** ANNOTATION ANALYSIS

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### Numbers

- 2176 (HITs) x 5 (assignments per HIT) = 10,880 assignments
- Annotators: 1012
- Turkers spent on average about 1 minute per HIT
- Hourly wage was about \$2.40 (about 4 cents per HIT)
- Total cost: US \$470 (cost per term: about 22 cents)
- More than 95% of the assignments had the correct answer for the word choice question.
  - The rest were discarded.
- 2081 terms had 3 or more valid assignments
  - on average 4.75 assignments per HIT



### **Emotive and non-emotive**

- Practical NLP applications may care for only two levels of intensity
- Example: vampire-fear



# Percent of most frequent terms that are emotive



# % of WAL joy terms evocative of different emotions as per the Turkers



# % of WAL surprise terms evocative of different emotions as per the Turkers



# % of WAL anger terms evocative of different emotions as per the Turkers





### What was missed?

baffled covetousness exacerbate gravel pesky pestering

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# Anger and Joy!

adjourn credit card find out gloat spontaneously surprised

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Agreement at two intensity levels: Majority class (m) = 3, 4, 5

= m = three = m = four = m = five





# Conclusions

- Regular folks can produce high quality emotion annotations with proper guidelines and checks:
  - Annotations match those in GI and WAL
  - Reasonable degree of agreement
    - Anticipation and trust are sources of more disagreement
- A large number of commonly used terms are emotive:
  - About 61%
    (evoke one or the other base emotion)



## Current work



# % of terms where all 5 agree





## Current and future work

- Determining which terms have strong color associations and if there is a correlation with emotions.
- Determine how much near synonyms vary in emotional content.
- Empirically verify if complex emotions are indeed combinations of basic emotions.
- Create a much larger lexicon (40,000 terms, say).
  - Make lexicon publicly available.
- Incorporate a game and scoring in the annotation.
- Use lexicon in applications.

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# Example HIT

#### Answer questions pertaining to "vampire".

#### **Directions:**

- Return/skip HIT if you do not know the meaning of the word.
- Attempt HITS only if you are a native speaker of English, or fluent in English.
- Certain "check questions" will be used to make sure the annotation is responsible and reasonable. Assignments that fail these tests will be rejected. If you fail more than 33% of these check questions, then it will be assumed that you are not following instructions 1 and/or 2 above, and ALL your assignments will be rejected.
- Only those HITs with ALL questions answered reasonably and responsibly will be approved.
- Expected date the assignments will be approved: April 20, 2010.
- Confidentiality notice: Your responses are confidential...