A Methodology for Using Crowdsourced Data to Measure Uncertainty in Natural Speech

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Introduction

Prosody: the melodic structure of speech → meaning in speech
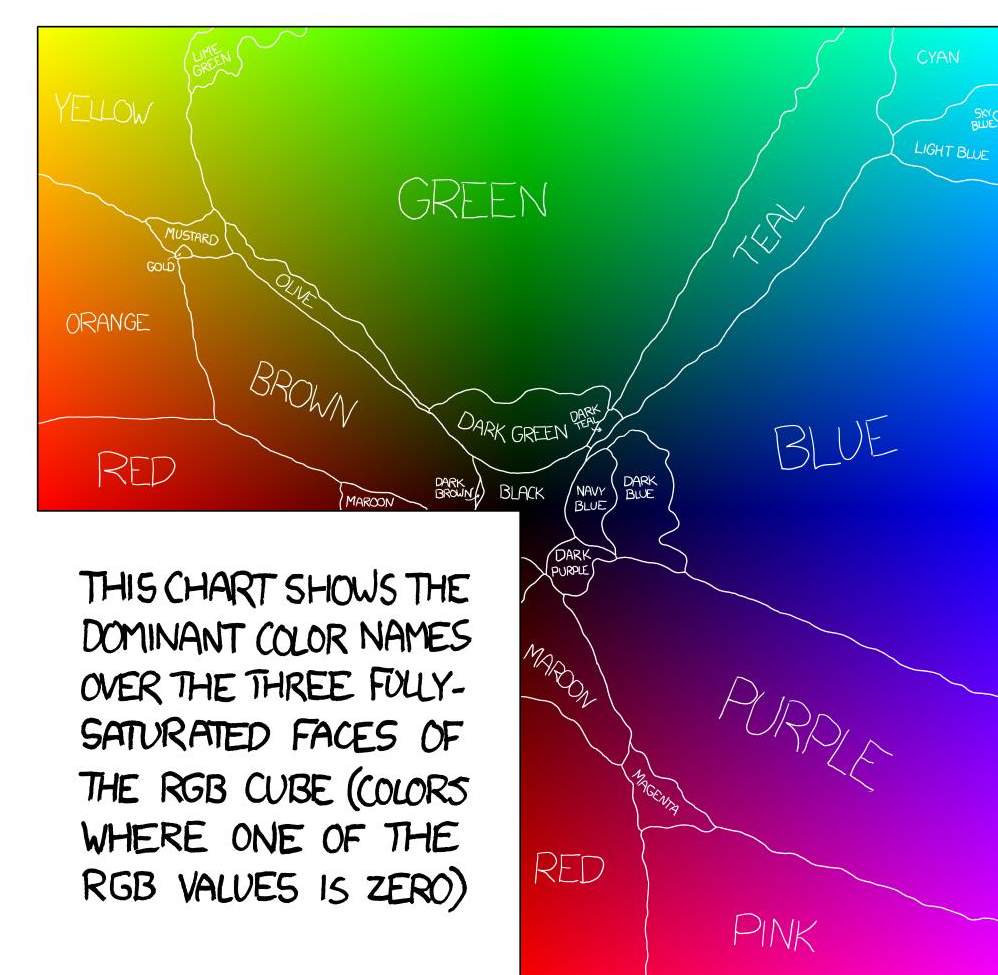**Our focus:** uncertainty in prosody → entropy
Question: Does the amount of uncertainty a person expresses correlate with the ambiguity of the color one is trying to describe?
Hypothesis: If so, there will be prosodic information that distinguishes utterances in low-entropy contexts from utterances in high-entropy contexts.

Methods

Data:
- XKCD color internet survey
- 222,500 participants typed color names

THIS CHART SHOWS THE DOMINANT COLOR NAMES OVER THE THREE FULLY-SATURATED FACES OF THE RGB CUBE (COLORS WHERE ONE OF THE RGB VALUES IS ZERO)

http://imgs.xkcd.com/blag/satfaces_map_1024.png

- 60 RGB colors (30 high and 30 low entropy)
- Presented one at a time in random order in a 2x2-inch square

Low Entropy   High Entropy

selection → collection →
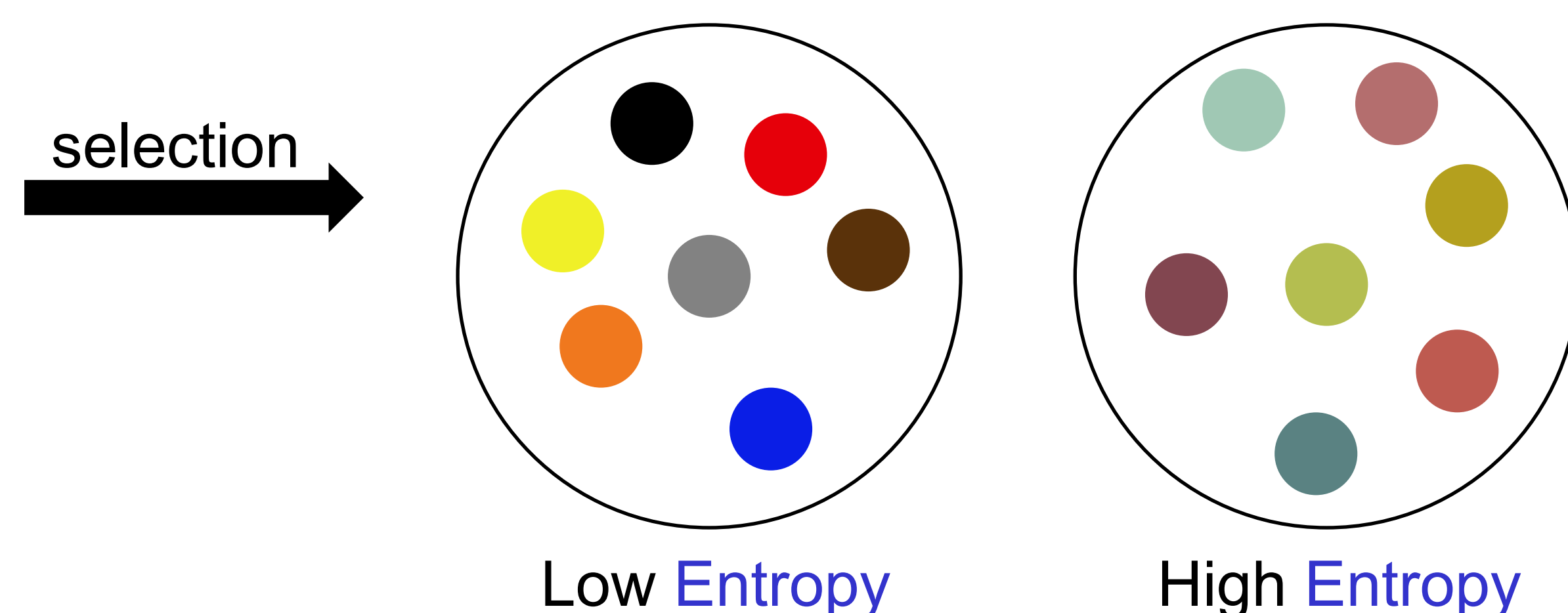
- 18 American native-English speakers (10 female, 8 male)
- Not colorblind
- Given 6 seconds to verbally state the name of the color (while being recorded)
- 488 low-entropy recordings
- 468 high-entropy recordings

Classification:

Features of Uncertainty

| # of syllables, # of pauses (>= 0.3 s), total duration, length of phonation, speech rate, articulation rate, ASD (phonation length/ # syllables) (Wempe & de Jong 2008) | # filled-pauses (words like "um" and "uh") & Gender of participant |

automatic / by hand — Praat

ToBI final tones (L-L%, H-L%, L-H%, H-H%)

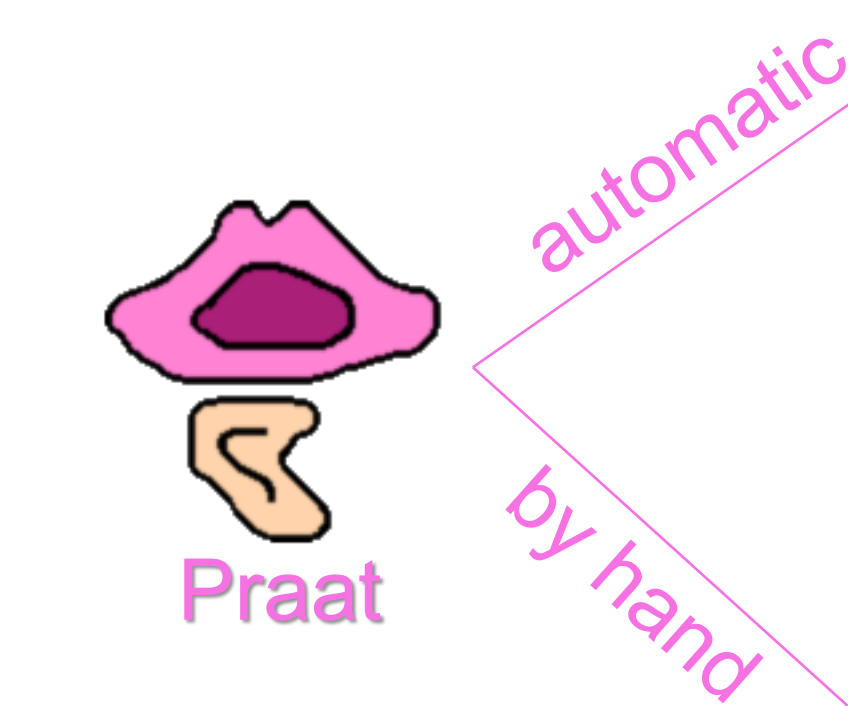### Classes

Low Entropy Color

High Entropy Color

Details

- Binary Classification
- 10-fold CV using Weka
- Naïve Bayes classification tree

Results

65.27% accuracy with gender
62.34% accuracy without gender
P < 0.0001

	Hyp-Low	Hyp-High	Total
Ref-Low	**346**	142	488
Ref-High	190	**278**	468
Total	536	420	

Confusion matrix of high & low entropy classifications produced from Naïve Bayes Tree

Naïve Bayes Tree, where the leaves are different NB classifiers which use all the features

Example:

2x2-inch square
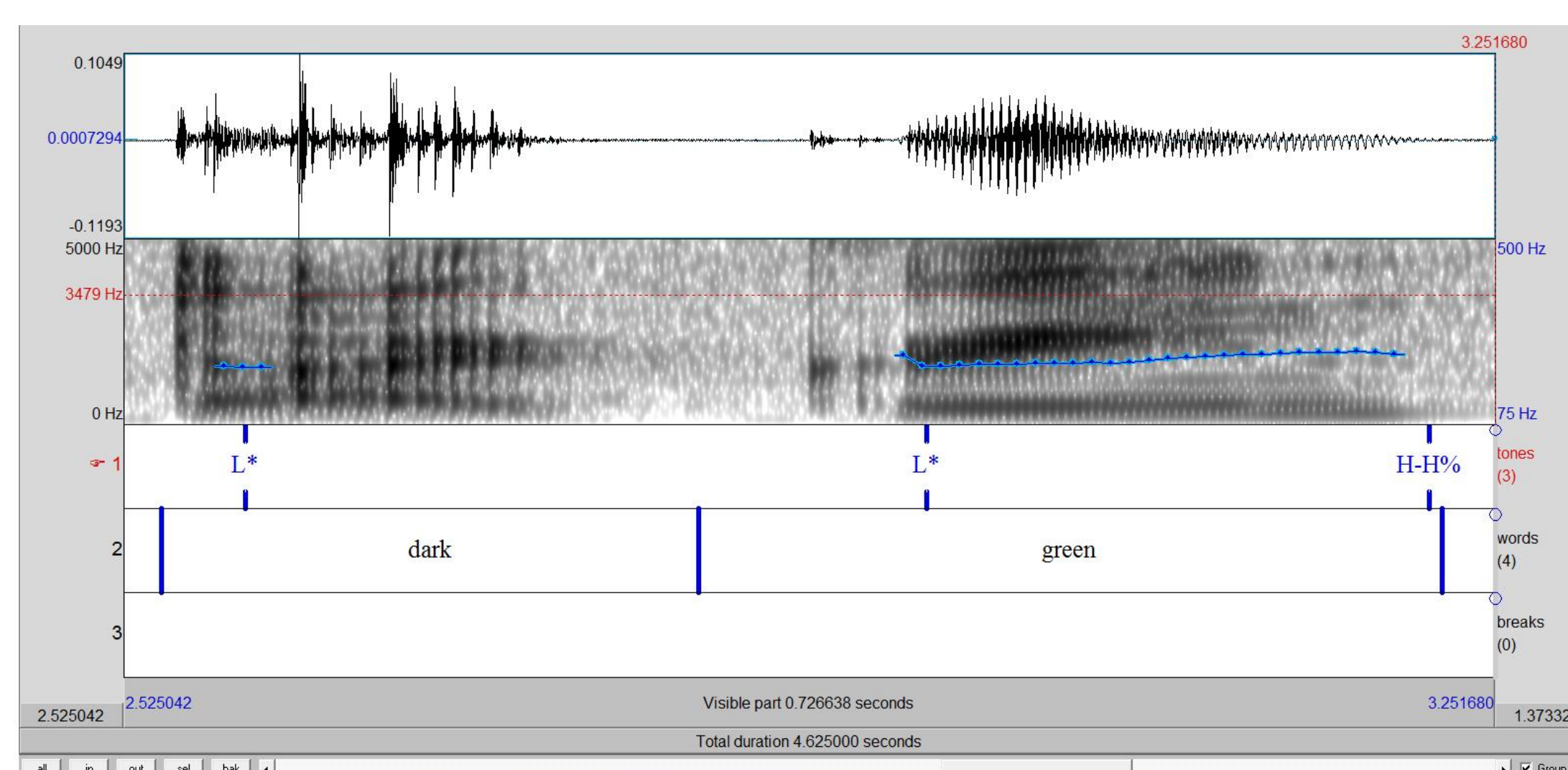
Participant saying "dark green" with try tone (H-H%).

Questions for Future Work

- Were these prosodic cues really measuring uncertainty or just the difficulty of the task?
- Why didn't intonation play a bigger role? Why were people mostly using H-L% monotone or L-H%? Could we make the task more like a conversation?
- Why was there a gender difference? Cultural? Biological?
- Can this be used in applications? For example, if people used more uncertainty while using voice search, would it improve precision and/or recall?
- Will this work with other media besides color?
- Are these features cross-lingual?

Conclusions

- Entropy **can** correlate with prosodic cues thought to predict uncertainty in speech
- We **can** use crowdsourced data to calculate entropy values, despite being from a non-verbal task

For references, please see paper.