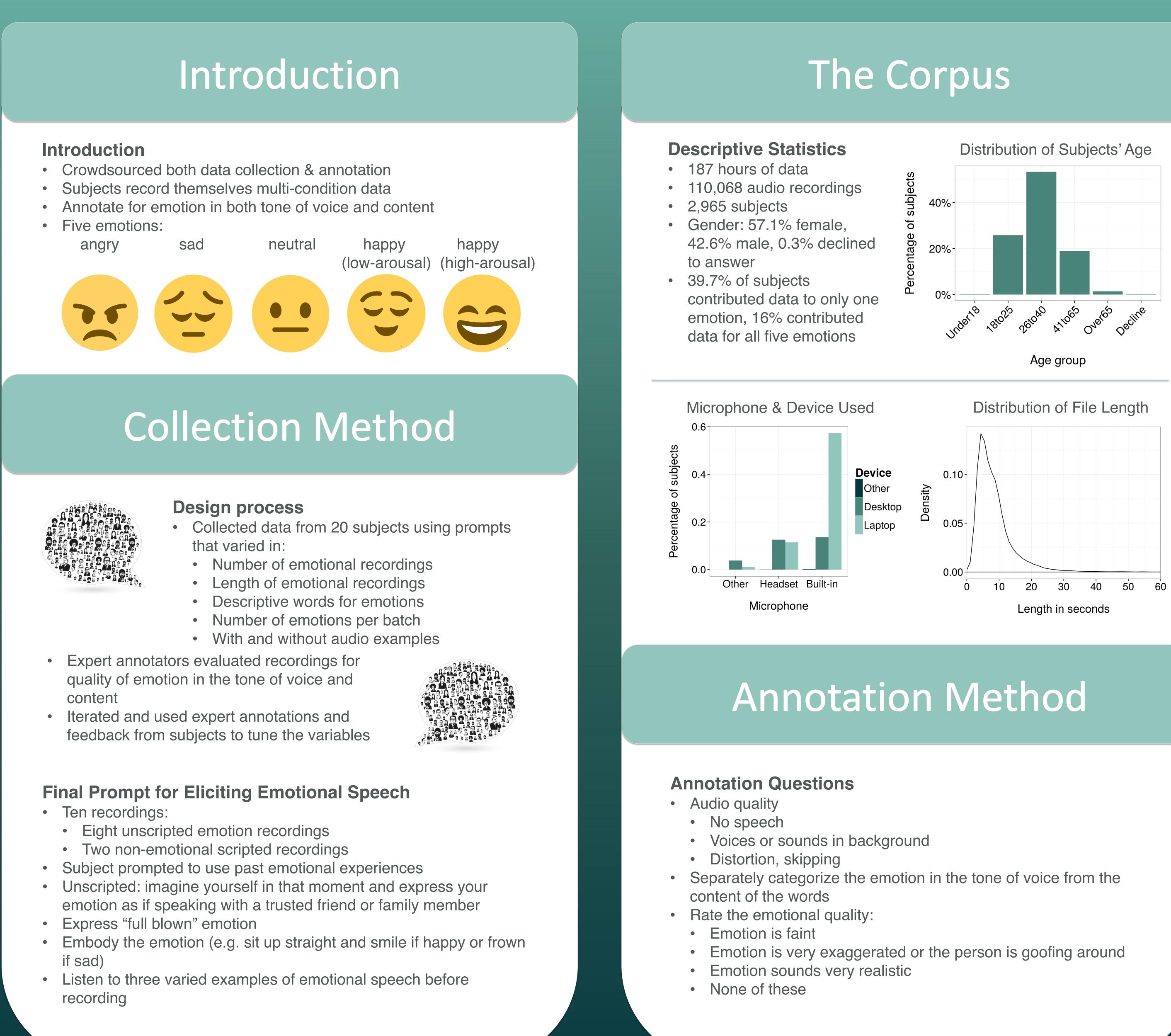
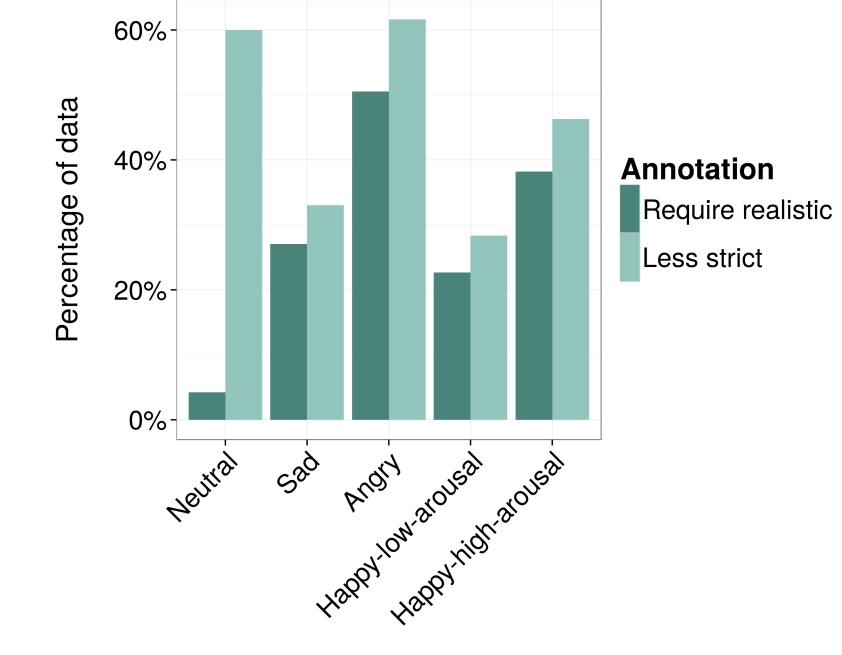
CROWDSOURCING EMOTIONAL SPEECH

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Results & Discussion

Annotation Results



Initial Evaluation of Annotation

- Single-layer Neural Network
- randomly from data

- Balanced classes

Lessons Learned

- large corpus

- emotions
- annotator error in post-processing
- Special attention should be paid to eliciting "neutral" data



• Annotated 5,168 recordings (5% of the data) • Audio quality issues present in 2% of the data Emotion in the tone of voice & content matched for 46% of the data • Emotion rated "very realistic" for 29% of the data Requiring data sound "very realistic" removes about 8-10% of data for all emotions except neutral, for which it removes 60% of data

Emotion

Classification experiment using SenSay[™] platform

• 2 sets: (1) annotated subset; (2) equally sized subset drawn

• Features include spectral, prosodic, articulatory, noise-robust, etc. • 4-way classification: Neutral removed due to lack of data

• 5.3% absolute (13.3% relative) improvement in accuracy

Crowdsourcing is fast and inexpensive way to collect and annotate a

• Multiple short utterances best for eliciting "full-blown" emotion Ten recordings / batch maximizes the number of samples while minimizing emotional-quality loss due to fatigue/boredom Collecting emotions separately minimizes cognitive load of switching

Requiring subjects listen to three varied audio examples increased emotional-quality without narrowing emotional expression Difficult to recruit long-term annotators – better to account for