SURVEY OF AFFECT REPRESENTATION SCHEMES FOR SPEECH EMOTION RECOGNITION

ADITI RAWAT

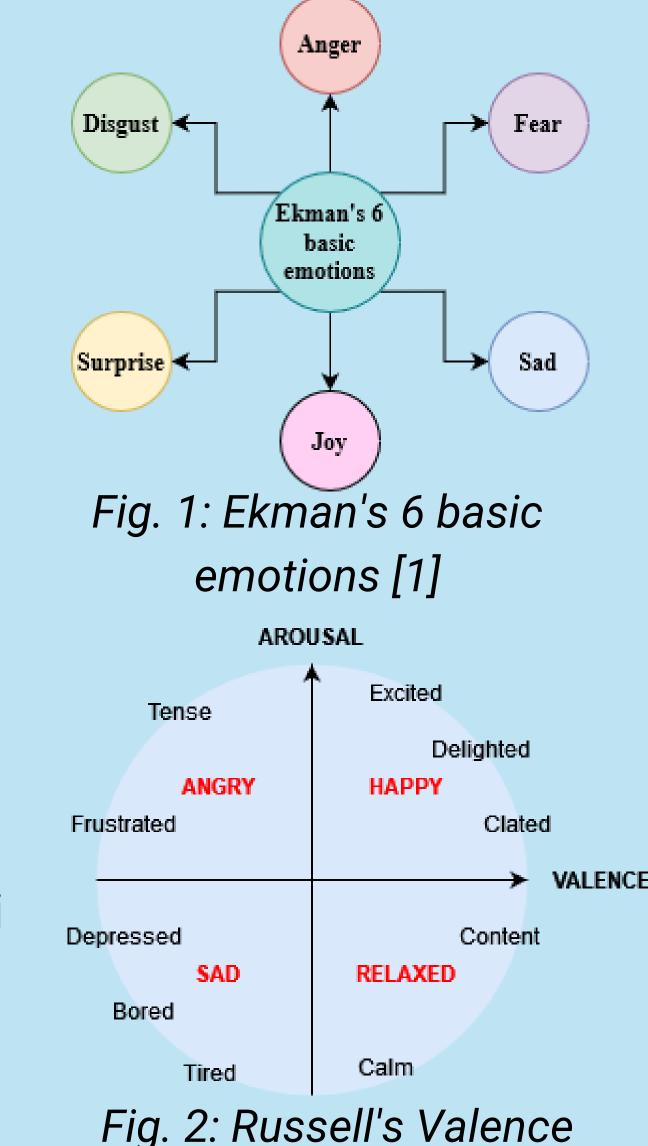
a.rawat-4@student.tudelft.nl

. BACKGROUND

- Affect refers to feelings experienced; subjective in nature
- Automatic affect prediction systems have underlying dimensional or categorical affect representation schemes (ARS)

APPLICATIONS

- Emotionally-receptive Alexa or Siri
- Human-Robot Interaction
- Market research
- Monitoring mental health



and Arousal ARS [2]

2. RESEARCH QUESTION

How are different affect representation schemes utilised in the context of speech emotion recognition?

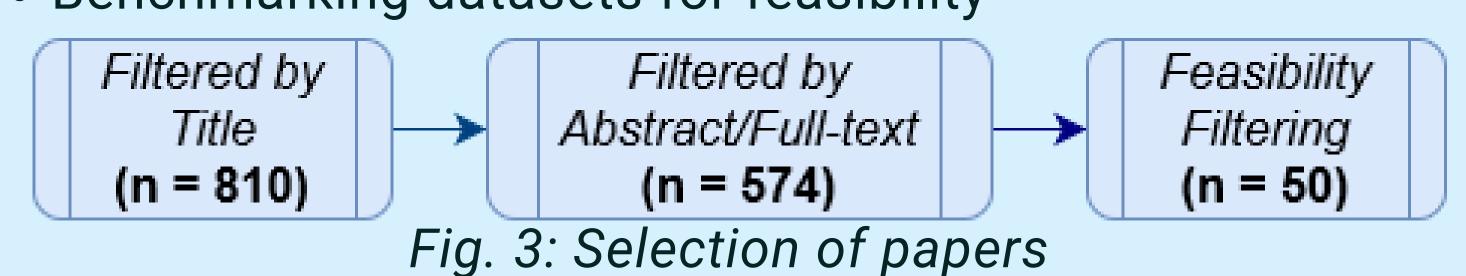
- ARS used in combinations
- Popularity of ARS overall and over time

Targeted types of affective states Motivations for different ARS

Psychological basis for ARS

3. METHODOLOGY

- PRISMA guidelines for systematic review [3]
- Databases: Scopus & Web of Science
- Included: Automatic affect prediction via speech
- Excluded: Multimodal, Non-English papers, Reviews
- Benchmarking datasets for feasibility



- 50 papers selected (10 each from years 2019-2023)
- 15 supplementary dataset papers

ARS IDENTIFIERS EXAMPLES **D**: Dimensional C: Categorical vad: valence, arousal, Type of Affective State Targeted **Corresponding ARS** All except *D.5f*, *D.5a* dominance Emotion Emotion-related state / Affect / Behaviour Behaviour D.7r, D.6t Attitude Table. 1: Types of affective states targeted **BOREDOM** Behaviors: aggressive, cheerful, intoxicate, nervous, tired Emotion-related states: anger, emphatic, positive, rest MOTIVATION Context-specific: Labels based on interaction with robot drop happiness (COVID-related) Fig..3: Emotions targeted by Database-specific: to be consistent with other datasets categorical ARS convenient annotation of dataset (motivation for COMBINATIONS OF ARS categorical schemes) 56% of the systems use multiple imbalance in dataset Better discriminability categorical schemes together Categories being too limited 1 paper uses coarse to fine approach • Not enough positively-valenced emotions: use of . division of valence-arousal calm as baseline space into quadrants • Time-specific: dimensional ARS can recognize rise 2. quadrants mapped to and fall in emotions categories POPULARITY Categorical ARS much more popular Most popular ARS are also from _ C.4e most popular datasets C.5a • C.6e: Ekman's basic 6 C.2i - C.6i • C.6ed: EMO-DB (anger, boredom, D.vad disgust, fear, happiness, sadness) • C.3i: IEMOCAP (anger, happiness, sadness) Fig. 4: ARS Popularity Overall • C.6e: Ekman's basic 6 (most D.va popular in 2023) show steady growth • C.3i: IEMOCAP and C.6ed: EMO-DB grow in popularity with a slight fall in 2023 • C.7r: RAVDESS dataset Fig. 5: ARS Popularity 2019-2023 (Ekman's 6 + calm) rose in ARS AND PSYCHOLOGY popularity significantly • Ekman's basic 6 emotion models and modifications Dimensional ARS only in 2020 Russell's Valence and Arousal and 2021 (mostly valence &

5. DISCUSSION

- Papers themselves do not motivate ARS
- Supplementary papers are more informative
- Neutral as an emotion/behavior is controversial
- Confusion between types of affective states
- Too many ARS relative to number of papers
- Datasets provide opportunity to use dimensional with categorical, but systems do not use this

6. CONCLUSIONS

- Categorical schemes overwhelmingly more popular than dimensional (in combinations as well)
- Happiness, Anger, Sadness, Neutral most common
- Most ARS based on Ekman's but varying labels
- Popularity of ARS highly linked with that of dataset
- Practicality prioritized over psychological research

7. FUTURE EXTENSIONS

- More researchers; more representative sample
- Exploration into supplementary dataset papers
- Multimodal input (for example, audio-visual)
- Motivation behind lack of dimensional representation

REFERENCES

Paul Ekman and Wallace V. Friesen. Constants across cultures in the face and emotion. Journal of Personality and Social Psychology. 17(2):124-129, 1971 2] Pekka Siirtola, Satu Tamminen, Gunjan Chandra, Anusha Ihalapathirana, and Juha R¨oning. Predicting emotion with biosignals: A comparison of classification and regression models for estimating valence and arousal level using wearable sensors. Sensors, 23(3):1598, 2023. 3] Matthew J Page, Joanne E McKenzie, Patrick M Bossuyt, Isabelle Boutron, Tammy C Hoffmann, Cynthia D Mulrow, Larissa Shamseer, Jennifer M Tetzlaff, Elie A Akl, Sue E Brennan, et al. The PRISMA 2020 statement: An updated guideline for reporting systematic reviews. International Journal of Surgery. 88:105906, 021.



arousal)

No psychological basis mentioned for behavior and

attitude ARS