

AN INTERFACE WITH COMPUTER-GENERATED FACIAL EXPRESSIONS AS AN ALTERNATIVE FOR MOOD SELF-REPORTS IN AN EMA CONTEXT

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Abstract

Traditionally numerical scales, colors, abstract representations or animated characters have been for reporting one's own mood. While working in practice, they do not provide a universal stable representation between participants. Abstract representations or colors rely on subjective interpretation, while numerical or animated characters require mapping the subjective feeling of mood onto a scale, which is also differentially interpretable. The described approach relies on computer-generated facial expressions to span a scale, which can be used for mood self-reports.

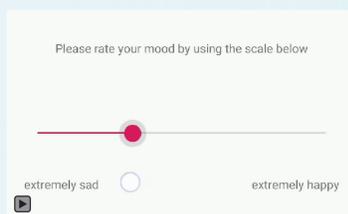
Participants

37 healthy participants in the Netherlands (20 male and 17 female) with an mean age of 30 years old were recruited. All participants provided an informed consent form prior to their recruitment.

Method

Participants were asked to use an android application on their own mobile phone for two weeks recording their mood. They were prompted 5 times per day at semi-fixed intervals to record their mood with both assessment interfaces presented in a randomized order. For each assessment we recorded the traces of input assessments with their respective timestamps.

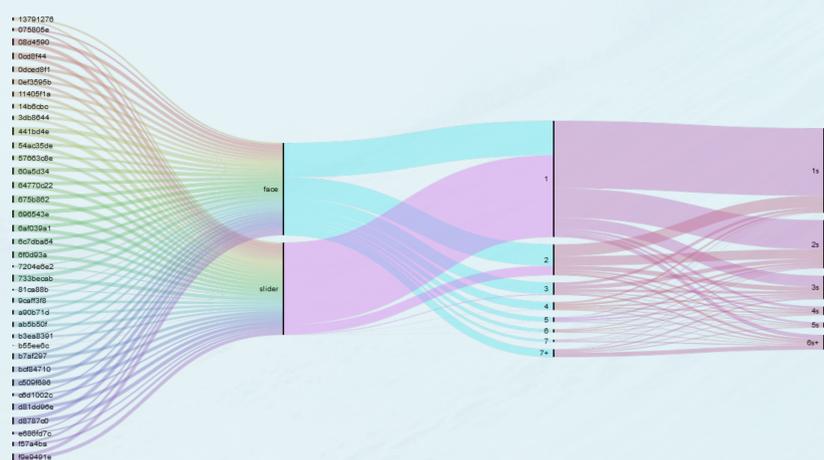
Assessment interfaces



On the left is the slider interface used to provide mood assessments. On the right side is the facial expression interface. Each of the interfaces features 100 unique points for the happy and 100 unique points for the sad dimensions.

If you're viewing this online, you can click on each element to see how the assessment works

Relationship between type of assessment, oscillations and assessment duration

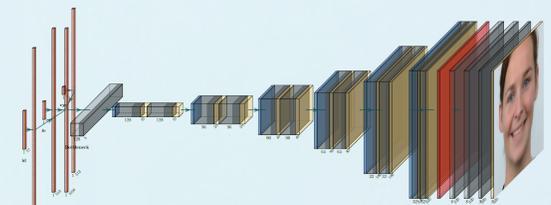


On the left hand side you see all participants (N=37) that took part in the study. The amount of assessments is represented by the thickness of the lines. Each provided assessment with the facial expression interface and the slider is depicted in the first band. For each assessment we captured the 'oscillations' or turning points in the pattern of their selection, represented in the second band. The final band depicts the time it took to provide an assessment, where the label describes the higher bound of the duration except for the highest value.

Graph has been generated with raw-graphs ("How to make an alluvial diagram", by RAWGraphs Team. Licensed under CC BY-NC-SA 4.0. Accessed: May 12, 2020, from <https://rawgraphs.io/learning/how-to-make-an-alluvial-diagram/>)

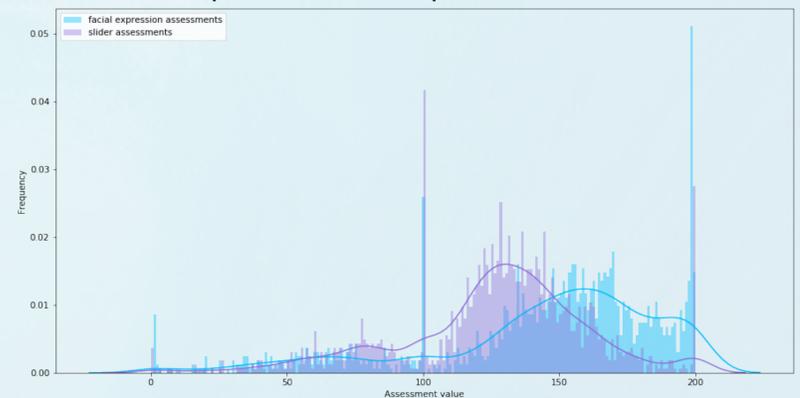
Generating facial expressions

- Artificially generated facial expressions through a deconvolutional neural network
- Topology of the network can be seen to the right.
- The network is trained on the RaFD dataset^[1] with 57 identities and 8 distinct facial expressions for happiness, sadness, anger, fear, surprise, disgust, contempt and the neutral expression.



[1] O. Langner, R. Dotsch, G. Bijlstra, D. H. J. Wigboldus, S. T. Hawk, and A. van Knippenberg, "Presentation and validation of the radboud faces database," *Cogn. Emot.*, vol. 24, no. 8, pp. 1377–1388, 2010.

Distribution plot of all assessments provided with both interfaces



Distribution of the assessments with each interface. Values between 0 and 99 represent sadness, where 0 is the saddest expression or corresponding to the utmost left position on the slider. Values between 101 and 200 represent happiness, where 200 is the happiest expression or the utmost right position on the slider. The value 100 or the center position on the slider is the neutral expression.

Graph has been created with the seaborn library for python

Results

	Facial expression	Slider
Pearson's r:	0.876	0.876
Variance	1978	1229
Average duration	2.7 seconds	1.4 seconds

Discussion

- We have recruited healthy participants, which is why as expected the distribution is heavily represented in the happiness dimension. This may have unforeseen implications for sadness.
- The high correlation between both assessments gives us the insight that both assessment methods can be used interchangeably with high certainty that they will yield comparative results.
- Albeit not conclusive, the significantly higher variance captured by the facial expression interface could mean that it describes better the modality of mood. Furthermore, two-thirds of the assessments provided with the facial expression interface featured more than one turning points in their trace. This is indicative of the desire of the participants to be more particular in their input.
- The quality of the generated facial expressions was good as they were easily understood, even though the only real facial expressions used to train the model were the expressions for extreme happiness (i.e. 200), extreme sadness (i.e. 0) and the neutral expression.

Considerations

- The happiest or saddest facial expressions might not necessarily be representative of their respective extremes, which could explain the offset between the distributions for the facial expression and slider assessments.
- Timestamps were only recorded during the interaction with the assessment elements. This might have reduced the time required to provide an assessment with the slider as a single tap would have a 0 seconds duration.

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