Estimating individual subjective values of emotion regulation strategies

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The authors made the following contributions. Christoph Scheffel: Conceptualization, Methodology, Funding acquisition, Formal analysis, Investigation, Project administration, Software, Visualization, Writing - original draft preparation, Writing - review & editing; Josephine Zerna: Conceptualization, Methodology, Funding acquisition, Investigation, Project administration, Software, Writing - review & editing; Anne Gärtner: Formal analysis, Writing - review & editing; Denise Dörfel: Conceptuatlization, Writing - review & editing; Alexander Strobel: Conceptualization, Writing - review & editing. [†] Christoph Scheffel and Josephine Zerna contributed equally to this work.

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Abstract

Individuals have a repertoire of emotion regulation (ER) strategies at their disposal, which they can use more or less flexibly. In ER flexibility research, strategies that facilitate goal achievement are considered adaptive and therefore are subjectively valuable. Individuals are motivated to reduce their emotional arousal effectively and to avoid cognitive effort. Perceived costs of ER strategies in the form of effort, however, are highly subjective. Subjective values (SVs) should therefore represent a trade-off between effectiveness and subjectively required cognitive effort. However, SVs of ER strategies have not been determined so far. We present a paradigm that is suitable for determining individual SVs of ER strategies. Using a multilevel modelling approach, it will be investigated whether individual SVs can be explained by effectiveness (subjective arousal, facial muscle activity) and subjective effort. Relations of SVs to personality traits will be explored

Keywords: emotion regulation, regulatory effort, effort discounting, registered report, specification curve analysis

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Estimating individual subjective values of emotion regulation strategies

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1. Introduction

Every day we are confronted with stressful or emotionally demanding situations. The ability to modify emotional experiences, expressions, and physiological reactions[1] to regulate emotions is an important cognitive skill. It is therefore not surprising that emotion regulation (ER) has substantial implications for well-being and adaptive functioning.[2] Different strategies can be used to regulate emotions, namely situation selection, situation modification, attentional deployment, cognitive change, and response modification,[1] and, following the taxonomy of Powers and LaBar,[3] individuals can implement ER strategies by means of different tactics. So called antecedent-focused strategies, e.g., attentional deployment and cognitive change, take effect early in the emotion generation process.[1] In contrast, response modification takes place late in the process and is therefore conceptualized as a response-focused strategy.[1] This postulated temporal sequence of ER strategies influences their effectiveness. It is meta-analytically proven that all mentioned strategies reduce subjective emotional arousal.[4] Distraction as a tactic of attentional deployment and (expressive) suppression as a tactic of response modulation showed small to medium effect sizes on measures of emotional experience (distraction: $d_+ = 0.27$; suppression: $d_+ = 0.27$). Distancing as tactic of cognitive change showed the highest effectiveness with an effect size of $d_+ = 0.45$.[4] Moreover it is known that ER strategies reduce physiological arousal, measured via electromyography (EMG) in the region of the *corrugator supercilii*.[5] So these tactics from three different strategies proved to be effective in the short term. However, in order to be able to make a statement on their general benefit, longer-term consequences must also be considered. Especially strategies that do not put the emotional content of the situation into a neutral perspective (i.e., distraction and suppression) are presumed to be disadvantageous in the longer term.

Long-term consequences of subjectively reported habitual use of emotion regulation strategies for affect and well-being have been discussed.[6] Especially suppression is generally associated with poorer outcomes (i.e., more negative affect, lower general well-being), which led to the assumption of *adaptive* and *maladaptive* strategies. For example, it could be shown that maladaptive ER strategies mediate the effect between neuroticism and depressive symptoms.[7] Also, a number of ER strategies is linked to psychological disorders (for meta-analytic review, see Aldao et al.).[8]

The postulation of *adaptive* and *maladaptive* strategies was put in a new perspective with the concept of ER flexibility. Similar to other psychological domains, e.g., attention and goal pursuit, *maladaptive* now refers to inflexible strategy use or use of strategies that are hindering the achievement of goals.[9] Adaptive flexible ER requires having a large repertoire of ER strategies.[9] For example, greater ER variability is related to reduced negative affect and therefore beneficial in daily life.[10] Strategies have to be chosen from the repertoire that are useful for goal achieving. Evidence from other contexts (e.g., intertemporal choice,)[11] shows that subjective values (SVs) are attributed to the choice options on the basis of which the decision is made. Research on ER choice has identified numerous factors that influence the choice of ER strategies, which can be seen as indirect evidence for factors influencing SVs. For example, it was shown that the intensity of a stimulus or situation plays a role in the choice.[12] Higher intensity of the stimulus leads to a choice of rather disengaging strategies, like distraction.[12,13] Further influencing factors are for example extrinsic motivation (e.g., monetary incentives), motivational determinants (i.e., hedonic regulatory goals), and effort.[13,14] Especially for effort, in our previous work we could show that the choice for an ER strategy is mainly influenced by the effort required to implement a given strategy.[14] In this study, participants used the strategies distancing and suppression while inspecting emotional pictures. Afterwards they could choose, which strategy they want to use again. Participants tended to re-apply the strategy that was subjectively less effortful, even though it was subjectively not the most effective one - in this case: suppression. Interestingly, the choice was independent of self reported habitual use of suppression and reappraisal. What has been missing in research on ER choice so far is information regarding the strategy *not* chosen. People choose a strategy that they prefer for different, relatively well-known reasons. However, nothing is revealed about the strategy that is *not chosen*.

The postulation of *adaptive* and *maladaptive* strategies was put into a new perspective with the concept of ER flexibility. Similar to other psychological domains such as attention and goal pursuit, *maladaptive* now refers to inflexible strategy use or the use of strategies that are hindering the achievement of goals.[9] In contrast, *adaptive* ER requires having a large repertoire of ER strategies to choose from[9] and choosing the strategy that is useful for goal achievement in a particular situation. This greater ER variability has been shown to be related to reduced negative affect.[10] The process of choosing an ER strategy from one's own repertoire likely shares similarities with other contexts such as intertemporal choice, in which individuals assign subjective values to every choice option.[11] Additionally, there are ER-specific influences on choice behaviour such as the intensity of the stimulus or situation, which leads to a choice of rather disengaging strategies like distraction, extrinsic motivation (e.g., monetary incentives), motivational determinants (i.e., hedonic regulatory goals), and effort.[13,14] Effort in particular appeared to be the main factor in ER choice behaviour in our previous work.[14] In that study, participants used the strategies distancing and suppression while inspecting emotional pictures, and could choose which strategy they wanted to use again. Participants tended to re-apply the strategy that was less effortful, even though it was not the most effective one for them - in this case: Suppression. Interestingly, the choice was independent of self reported habitual use of suppression and reappraisal. Research on ER choice has shown that individuals choose a strategy that they prefer for different, relatively well-known reasons. However, nothing has been investigated regarding the strategy that is *not* chosen.

We assume that people choose the strategy that has the highest value for them at that moment. The value is determined against the background of goal achievement in the specific situation: A strategy is highly valued if it facilitates goal achievement.[9] One certainly central goal is the regulation of negative affect. A second, intrinsic and rather less obvious goal is the avoidance of effort.[15] When given the choice, most individuals prefer tasks that are less effortful.[16] We assume that both aspects are traded off against each other by individuals to determine individual subjective values (SVs) of ER strategies: A strategy is more valuable if it can reduce emotional arousal *and* is less effortful. SVs of ER Strategies could be helpful to describe the ER repertoire[9] more comprehensively. Depending on the flexibility of a person, different patterns of SVs could be conceivable: A person with high flexibility would show relatively high SVs for a number of strategies. This would mean that all strategies are a good option for goal achievement. A second person with less flexibilty, however, would show high SVs only for one strategy, or for no strategy at all. This in turn would mean that no strategy is a good choice to achieve ER goals.

However, so far we have not seen any attempt in ER choice research to determine individual SVs of ER strategies. To investigate this question, the individual SVs of each strategy available for selection would have to be determined. Promising approaches can be found in studies on difficulty levels of effortful cognitive tasks.

Individual SVs of effortful cognitive tasks have been quantified using the Cognitive Effort Discounting Paradigm (COG-ED).[17]

In the original study by Westbrook et al.,[17] cognitive load was varied using the n-back task, a working memory task that requires fast and accurate responses to sequentially presented stimuli. Participants had to decide in an iterative procedure whether they wanted to repeat a higher n-back level for a larger, fixed monetary reward, or a lower level for a smaller, varying reward, with the implicit assumption taht teh objectively easiest n-back level has the highest SV. In the current study, we want to use this paradigm to determine SVs of ER strategies. In doing so, we need to make an important change: We have to adapt the assumption that the easiest n-back level has the highest SV. As we have shown in previous studies, there are large inter-individual differences in the preference and perceived subjective effort of ER strategies.[14] Moreover, there is nothing like an objectively easiest ER strategy. Therefore, we have to add an additional step, which preceeds the other steps and where the option with the higher subjective value is determined. In this step, the same monetary value (i.e., 1 €) is assigned to both options. The assumption is that participants now choose the option that has the higher SV for them. In the next step we return to the original paradigm. The higher monetary value (i.e., 2 €) is assigned to the option that was not chosen in the first step and therefore is assumed to have the lower SV. In the following steps, the lower value is changed in every iteration according to Westbrook et al.[17] until the indifference point is reached. This procedure will be repeated until all strategies have been compared. The SV of each strategy is calculated as the mean of this strategy's SV from all comparisons. In case a participant has a clear preference for one strategy, the SV of this strategy will be 1. But our paradigm can also account for the case that a person does not have a clear preference. Then no SV will be 1, but still, the SVs of all strategies can be interpreted as absolute values and in relation to the other strategy's SVs (see figure 1). Additionally, we will test our adapted paradigm in a *n*-back task and explore whether this paradigm can describe individuals that do not prefer the easiest *n*-back option (see Zerna et al., …).

1.2 The present study

The aim of the present study is to evaluate if this paradigm is suitable for determining SVs of ER strategies. As a manipulation check, we want to explore whether ER strategies distraction, distancing, and suppression effectively reduce emotional arousal and require cognitive effort. The following hypotheses are proposed for this purpose:

- H1a) Subjective arousal (arousal ratings) is lower after using an ER strategy (distraction, distancing, suppression) compared to active viewing.

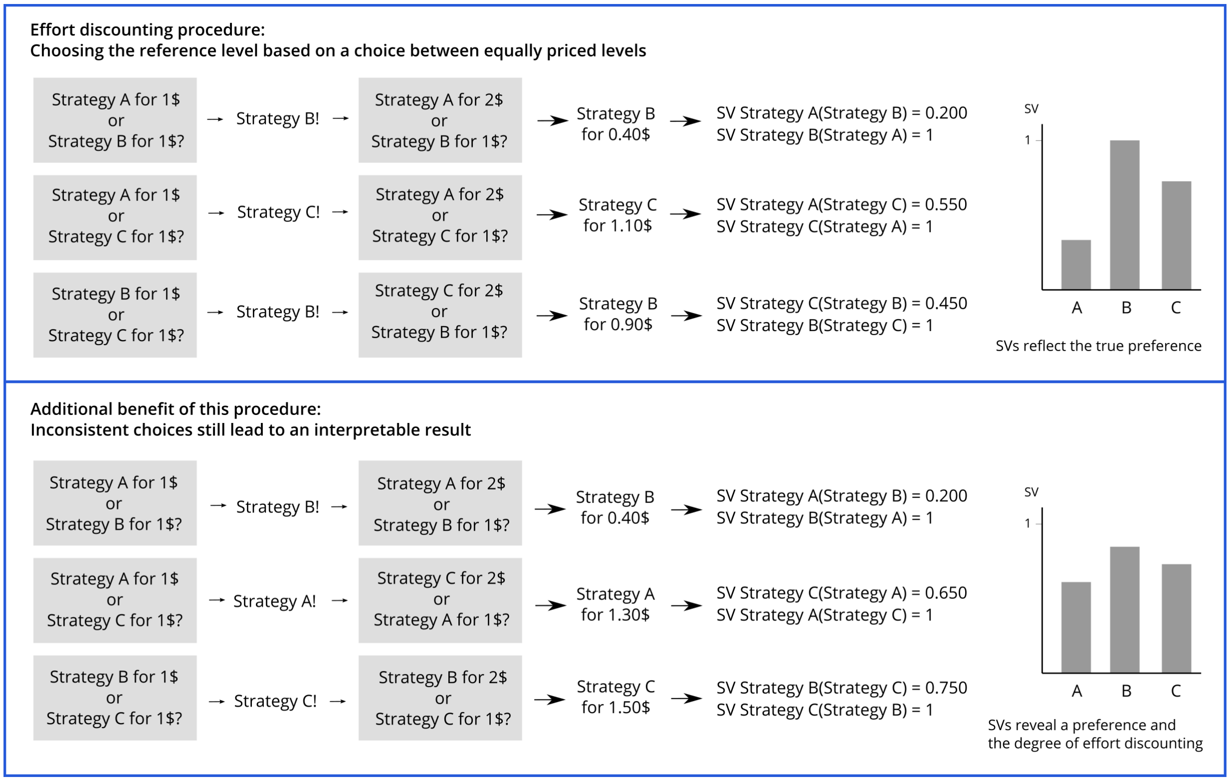

Figure 1. Exemplary visualisation of two response patterns. In the top half, the person has a clear preference for one of the three strategies. In the lower half, they have no clear preference and therefore show an inconsistent response pattern. This pattern can also be represented by our paradigm.

- H1b) Physiological arousal (EMG *corrugator* activity) is lower after using an ER strategy (distraction, distancing, suppression) compared to active viewing.
- H1c) Physiological arousal (EMG *levator* activity) is lower after using an ER strategy (distraction, distancing, suppression) compared to active viewing.
- H2a) Subjective effort (effort ratings) is greater after using an Er strategy (distraction, distancing, suppression) compared to active viewing.
- H2b) The majority of participants ruse the strategy that was least effortful for them.

Further we want to explore which variables predict individual subjective values of ER strategies and whether effort is the best predictor for SVs of ER strategies with the following hypotheses:

- H3a) Subjective effort ratings negatively predict SVs of ER strategies.
- H3b) Subjective arousal ratings negatively predict SVs of ER strategies.
- H3c) EMG *corrugator* activity negatively predict SVs of ER strategies.
- H3d) EMG *levator* activity negatively predict SVs of ER strategies.
- H4a) SVs decline with increasing effort, even after controlling for task performance measured by subjective arousal ratings, *corrugator* and *levator* activity.

We also want to explore whether SVs are related to flexible emotion regulation:

- H5a) The higher the SV, the more likely the respective strategy is chosen.
- H5b) SVs are lower and decline stronger when ER flexibility is lower.

Exploratorily, we want to investigate whether individual SVs are related to personality traits and how individual SVs of ER strategies relate to SVs of other tasks with different demand levels, namely *n*-back.

2. Method

We report how we determined our sample size, all data exclusions (if any), all manipulations, and all measures in the study.[18] The paradigm was written and presented using *Psychopy*.[19] We used *R* with *R Studio*[20,21] with the main packages *afex*[22] and *BayesFactor*[23] for all analyses. The R Markdown file used to analyze the data and write this document, as well as the raw data and the materials are freely available at https://github.com/ChScheffel/CERED. A complete list of all measures assessed in the study can be found at OSF (https://osf.io/vnj8x/) and GitHub (https://github.com/ChScheffel/CERED).

2.1 Ethics information

The study protocol complies with all relevant ethical regulations and was approved by the ethics committee of the Technische Universität Dresden (reference number EK...). Prior to testing, written informed consent will be obtained. Participants will receive 30€ in total or course credit for participation.

2.2 Pilot data

The procedure described above was tested in a pilot study with $N = 16$ participants (9 female; age: $M = 24.1 \pm SD = 3.6$). Results showed significant higher subjective arousal for active viewing of negative pictures, compared to active viewing of neutral pictures. ER strategies did not lead to a reduction of subjective arousal compared to active viewing of negative pictures. However, ER strategies were associated with reduced facial muscle activity of the *corrugator* and *levator* compared to active viewing of negative pictures. Yet we found in accordance with our previous studies[14] that the use of ER strategies compared to active viewing was associated with increased subjective effort.

2.3 Design

Young, healthy participants (aged 18 to 30 years) will be recruited using the software *ORSEE*[24] at the Technische Universität Dresden. Participants will be excluded from participation if they do not fluently speak German, have current or a history of psychological disorders or neurological trauma, or report to take medication. Participants will be invited to compete an online survey containing different questionnaires to assess broad and narrow personality traits and measures of well-being. The study consists of two lab sessions, which take place in a shielded cabin with constant lighting. Before each session, participants receive information about the respective experimental procedure and provide informed consent. At the beginning of the first session participants fill out a demographic questionnaire and complete an n-Back task with the levels one to four. Then, they complete an effort discounting (ED) procedure on screen, followed by a random repetition of one n-back level. The second session, containing the ER paradigm, takes place exactly one week after session one. Participants provide informed consent and receive written instructions on the ER paradigm and ER strategies that they should apply. A brief training ensures that all participants are able to implement the ER strategies. Next, electrodes to measure EMG being attached and the ER paradigm is conducted. Participants receive 30.00€ or course credit as compensation. Study data are being collected and managed using REDCap electronic data capture tools hosted at Technische Universität Dresden.[25,26]

2.3.1 Psychometric measures. The online survey contains a number of questionnaires: General psychological well-being is assessed using the German version of the WHO-5 scale.[27,28] To capture the construct of resilience, the German version 10-item-form of the Connor-Davidson resilience Scale (CD-RISC)[31] is used. Dispositional use of ER is assessed using the German version of the Emotion Regulation Questionnaire (ERQ).[6,32] For the assessment of ER ability we use the Flexible Emotion Regulation Scale (FlexER).[33] Implicit theories of willpower in emotion control are assessed using the implicit theories questionnaire from.[34] To assess Need for Cognition, the German version short form of the Need for Cognition Scale[35,36] is used. To assess self-control, sum scores of the German versions of the following questionnaires are used:[37] the Self-Regulation Scale (SRS),[38] the Brief Self-Control Scale (BSCS);[39]],[40] and the Barratt Impulsiveness Scale (BIS-11).[41,42] Attentional control is assessed using the Attentional Control Scale (ACS).[43]

2.3.2 Emotion regulation paradigm. The ER paradigm roughly consists of three parts that will be described in the following.

Part one: ER task. Part one is a standard ER task in a block design (see Figure X), similar to paradigms previously used by our group.[14] Participants are told to actively view neutral and negative pictures (see 2.3.3) or to regulate all upcoming emotions by means of distraction, distancing, and expressive suppression, respectively. Every participant first has the condition “active viewing-neutral” that serves as a baseline condition. During this block, 20 neutral pictures are presented. Participants are asked to “actively view all pictures and permit all emotions that may arise.” In the second block, participants actively viewe negative pictures. During the third, fourth, and fifth block, participants see negative pictures and are asked to regulate their emotions using distraction, distancing, and suppression. In order to achieve distraction, participants are asked to think of a geometric object or an everyday activity, like brushing their teeth. During distancing, participants are asked to “take the position of a non-involved observer, thinking about the picture in a neutral way.” Participants are told not to re-interpret the situation or attaching a different meaning to the situation. During suppression, participants are told to “suppress their emotional facial expression.” They should imagine being observed by a third person that should not be able to tell just by looking at the facial expression whether the person is looking at an emotional picture. Participants are instructed not to suppress their thoughts or change their facial expression to the opposite.[14] All participants receive written instruction and complete a training session. After the training session, participants are asked about their applied ER strategies to avoid misapplication. The order of the three regulation blocks (distraction, distancing, and suppression) is randomized between participants.

Part two: ER effort discounting. In the second part, ER effort discounting takes place. The procedure of the discounting follows the COG-ED paradigm by Westbrook et al.[17] with major change. We use the following adaption that allows the computation of SVs for different strategies without presuming that all individuals would inherently evaluate the same strategy as the easiest one: For each possible pairing (distraction vs. distancing, distraction vs. suppression, and distancing vs. suppression), two strategies with monetary values are presented. The order of the comparisons is randomized. Because there is no strategy that is objectively more difficult, we added an initial comparison that begins with the option “1 EUR for strategy A or 1 EUR for strategy B”. The strategy that is not chosen is assigned the value of 2 EUR. From this point on, comparisons between strategies follow the original COG-ED paradigm.[17] Participants are instructed to decide as realistically as possible, imagining the displayed money would really be up for election.

Part three: ER choice. After the discounting part, participants choose which of the three ER strategies (distraction, distancing or suppression) they want to re-apply. Importantly, there are no further instruction on what basis they should make their decision. Participants should make their decision freely, according to the criteria they consider important for themselves. However, participants are asked to state the reasons for the decision afterwards. As soon as they have decided, they see the respective instruction and the block with another 20 negative pictures starts.

2.3.3 Stimuli. Pictures used in the paradigm were selected from the Emotional Picture Set (EmoPicS)[44] and the International Affective Picture System (IAPS).[45] The 20 neutral pictures (Valence (V): $M \pm SD = 4.81 \pm 0.51$; Arousal (A): $M \pm SD = 3 \pm 0.65$) depicted content related to the categories persons, objects, and scenes. Further, 100 negative pictures, featuring categories animals, body, disaster, disgust, injury, suffering, violence, and weapons, were used. An evolutionary algorithm[46] was used to cluster these pictures into five sets with comparable valence and arousal values (set one: V: $M \pm SD = 2.84 \pm 0.57$, A: $M \pm SD = 5.62 \pm 0.34$; set two: V: $M \pm SD = 2.64 \pm 0.46$, A: $M \pm SD = 5.58 \pm 0.35$; set three: V: $M \pm SD = 2.82 \pm 0.62$, A: $M \pm SD = 5.60 \pm 0.39$; set four: V: $M \pm SD = 2.65 \pm 0.75$, A: $M \pm SD = 5.61 \pm 0.41$; set five: V: $M \pm SD = 2.74 \pm 0.70$, A: $M \pm SD = 5.63 \pm 0.37$). A complete list of all pictures and their classification into sets can be found in supplementary material 1.

2.3.4 Facial electromyography. Bipolar facial electromyography (EMG) will be measured for *corrugator supercilii* and *levator labii* as indices of affective valence.[47] Two passive surface Ag/AgCl electrodes (8 mm inner diameter, 10 mm distance between electrodes) will be placed over each left muscle according to the guidelines of.[48] The ground electrode will be placed over the left *Mastoid*. Before electrode placement, the skin will be abraded with Every abrasive paste, cleaned with alcohol, and filled with Lectron III electrolyte gel. Raw signals will be amplified by a BrainAmp amplifier (Brain Products Inc., Gilching, Germany). Impedance level will be kept below 10 $k\Omega$. Data will be sampled at 1000 Hz, filtered, rectified and integrated. A 20 Hz high pass (order 8), a 300 Hz low pass (order 8), and a 50 Hz notch filter will be applied to both signals. Corrugator and Levator EMG will be analyzed during the 6 s of picture presentation. EMG data will be baseline-corrected using a time window of 2 s prior to stimulus onset.[47] Last, the sampling rate will be changed to 100 Hz, and EMG data will be averaged for each condition and each participant.

2.4 Sampling plan

Sample size calculation is done using *G*Power*.[49,50] In a meta-analysis of Zaehringer and colleagues,[5] effect sizes of ER on peripheral-physiological measures were reported. To find an effect of $d = -0.32$ of ER on *corrugator* muscle activity with $\alpha = .05$ and $\beta = .95$, data of least $N = 85$ have to be analyzed. Power analyses of all other hypotheses yielded smaller sample sizes. However, if participants withdraw from study participation, technical failures occur, or experimenter considers the participant for not suitable for study participation (e.g., because the participant does not follow instructions or shows great fatigue), respective data will also be excluded from further analyses. Therefore, we aim to collect data of 90 participants. Detailed information on power calculation for each hypothesis can be found in the design table.

2.5 Analysis plan

Data collection and analysis will not be performed blind to the conditions of the experiments. The level of significance will be set to $\alpha = .05$.

Effects of emotion regulation on arousal, facial EMG and effort To examine the impact of valence of emotional pictures on subjective arousal ratings, a repeated measures analysis of variance (rmANOVA) with the factor valence (neutral and negative) for the strategy active viewing will be conducted. To investigate the effects of the three ER strategies on subjective arousal, another rmANOVA with the factor strategy (active viewing - negative, distraction, distancing, and suppression) for subjective arousal ratings will be conducted.

To examine the impact of valence on facial EMG, a rmANOVA with the factor valence (neutral and negative) for the strategy active viewing will be conducted for *corrugator* and *levator* activity. To examine the effects of the three ER strategies on facial EMG, another rmANOVA with the factor strategy (active viewing - negative, distraction, distancing, and suppression) for *corrugator* and *levator* activity will be conducted.

To examine the effect of ER strategies on cognitive effort, a rmANOVA with the factor strategy (active viewing, distraction, distancing, and suppression) for subjective effort ratings will be conducted. Greenhouse-Geisser-corrected p-values and degrees of freedom will be reported when the assumption of sphericity is violated. Proportion of explained variance η_p^2 will be reported as a measure of effect size. If indicated by the data, estimated marginal means will be computed as post-hoc contrasts. rmANOVAS will be run using the *afex* package[22]

Subjective values of emotion regulation strategies For each ER strategy, SVs will be calculated as follows: first, the value 0.015625 will be added to or subtracted from the last monetary value of the flexible strategy, depending on the participant's last choice. Second, the resulting (monetary) value will be divided by 2.00 €. The final SV for each participant will be computed by averaging all final SVs of each strategy. The resulting values will be entered in a rmANOVA to compare the SVs of the three strategies (distraction, distancing, and suppression) to explore for group effects. Again, estimated, marginal means will be computed as post-hoc contrasts.

To investigate, whether individual SVs predict ER choice, a Chi-squared test with predicted choice (highest SV of each participant) and actual choice will be computed. Furthermore, an ordinal logistic regression with dependent variable choice and independent variables SVs of each strategy will be computed.

To explore the association between subjective arousal, physiological arousal, and subjective effort on SVs, a multilevel model (MLM) will be specified using the $lmerTest$ package.[51] First, ER strategies will be recoded and centered for each subject according to their individual SVs: The strategy with the highest SV will be coded as -1, the strategy with the second highest SV 0, and the strategy with the lowest SV will be coded as 1. Restricted maximum likelihood (REML) will be applied to fit the model. A random slopes model of SVs including subjective effort (effort ratings), subjective arousal (arousal ratings), and physiological arousal (*corrugator* activity and *levator* activity) as level-1-predictor will be specified.

$$SV \sim strategy + \text{effort rating} + \text{arousal rating} + corrugator\ \text{activity} + levator\ \text{activity} + (strategy|subject)$$

Level-1-predictors will be centered within cluster.[52] Residuals of the final model will be inspected visually. Intraclass correlation coefficient (ICC), ρ, will be reported for each model (null model, as well as full model).

The influence of personality traits on SVs will be investigated exploratorily. Therefore, the MLM specified above will be extended by the level-2-predictors NFC and self-control.

The association between flexible ER and SVs of ER strategies will be investigated with a regression using the *intercept* and *slope* of each participants' SVs to predict their FlexER score. Firstly, SVs will be ordered by magnitude. Secondly, for each participant a linear model will be built to estimate the individual *intercept* and *slope*.

For each result of the analyses both, p-values and Bayes factor $BF10$, calculated using the *BayesFactor* package,[23] will be reported. Bayses factors are calculated using the default prior widths of the functions *anovaBF*, *lmBF* and *regressionBF*.

Data availability

The data of this study can be downloaded from osf.io/vnj8x/.

Code availability

The paradigm code, as well as the R Markdown file used to analyze the data and write this document is available at our Github repository.

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Author Contributions

CS, AS, and JZ conceptualized the study and its methodology. CS and JZ acquired funding, investigated, administered the project, and wrote the software. CS, JZ, and AG did the formal analysis. CS and JZ visualized the results, and prepared the original draft. All authors reviewed, edited, and approved the final version of the manuscript.

Competing Interests

The authors declare no competing interests.

Figures and figure captions

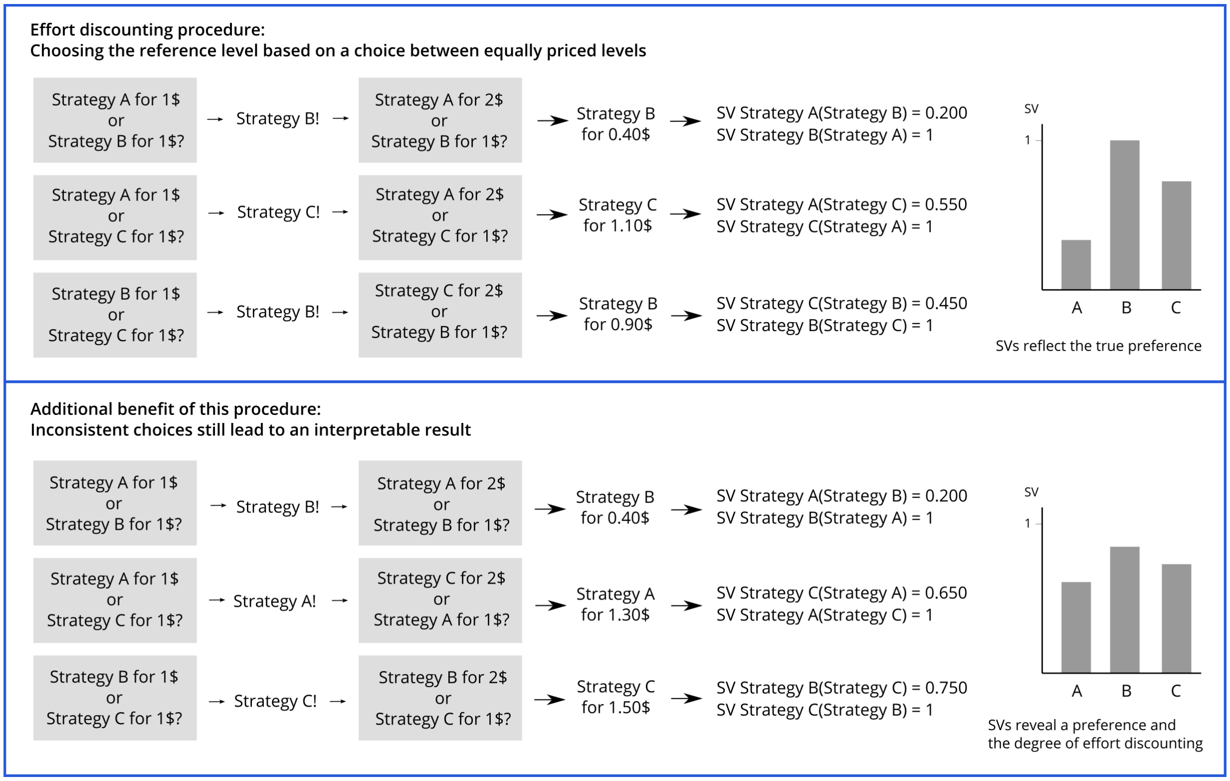

Figure 1

Figure 1. Exemplary visualisation of two response patterns. In the top half, the person has a clear preference for one of the three strategies. In the lower half, they have no clear preference and therefore show an inconsistent response pattern. This pattern can also be represented by our paradigm.

Design Table

Question	Hypothesis	Sampling plan (e.g. power analysis)	Analysis Plan	Interpretation given to different outcomes
1.) Do ER strategies reduce emotional arousal? (Manipulation check)	1a) Subjective arousal (arousal rating) is lower after using an emotion regulation strategy (distraction, distancing, suppression) compared to active viewing.	F tests - ANOVA: Repeated measures, within factors Analysis: A priori: Compute required sample size Input: Effect size f = 0.50 ($\eta_p^2 = 0.20$) (Scheffel et al., 2021) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of groups = 1 Number of measurements = 4 Corr among rep measures = 0.5 Nonsphericity correction ε = 1 Output: Noncentrality parameter λ = 20.0 Critical F = 2.9603513 Numerator df = 3.0 Denominator df = 27.0 Total sample size = 10 Actual power = 0.95210128	Repeated measures ANOVA with four linear contrasts, comparing the subjective arousal ratings of four blocks (active viewing, distraction, distancing, suppression). ANOVA is calculated using aov_ez() function of the afex-package, estimated maginal means are calculated using emmeans() function from the emmeans-package, pairwise contrasts are calculated using pairs(). Bayes factors are computed for the ANOVA and each contrast using the BayesFactor-package.	ANOVA yields $p < .05$ is interpreted as arousal ratings changing significantly with blocks. Values of arousal ratings are interpreted as equal between blocks if $p > .05$. Each contrast yielding $p < .05$ is interpreted as arousal ratings being different between those two blocks, magnitude and direction are inferred from the respective estimate. Values of arousal ratings are interpreted as equal between blocks if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	1b) Physiological arousal (*corrugator* muscle activity) is lower after using an emotion regulation strategy (distraction, distancing, suppression) compared to active viewing.	F tests - ANOVA: Repeated measures, within factors Analysis: A priori: Compute required sample size Input: Effect size f = 0.1605 (Zaehringer et al., 2020) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of groups = 1 Number of measurements = 4	Repeated measures ANOVA with four linear contrasts, comparing the *corrugator* muscle activity of four blocks (active viewing, distraction, distancing, suppression). ANOVA is calculated using aov_ez() function of the afex-package, estimated maginal means are calculated using	ANOVA yields $p < .05$ is interpreted as *corrugator* muscle activity changing significantly with blocks. Values of *corrugator* muscle activity are interpreted as equal between blocks if $p > .05$. Each contrast yielding $p < .05$ is interpreted as *corrugator* muscle activity being different between those two blocks, magnitude and direction are

		Corr among rep measures = 0.5 Nonsphericity correction ε = 1 <u>Output</u>: Noncentrality parameter λ = 17.5169700 Critical F = 2.6404222 Numerator df = 3.0 Denominator df = 252 Total sample size = 85 Actual power = 0.9509128	emmeans() function from the emmeans-package, pairwise contrasts are calculated using pairs(). Bayes factors are computed for the ANOVA and each contrast using the BayesFactor-package.	inferred from the respective estimate. Values of *corrugator* muscle activity are interpreted as equal between blocks if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	1c) Physiological arousal (*levator* muscle activity) is lower after using an emotion regulation strategy (distraction, distancing, suppression) compared to active viewing.	F tests - ANOVA: Repeated measures, within factors Analysis: A priori: Compute required sample size <u>Input:</u> Effect size f = 0.1605 (Zaehringer et al., 2020) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of groups = 1 Number of measurements = 4 Corr among rep measures = 0.5 Nonsphericity correction ε = 1 <u>Output</u>: Noncentrality parameter λ = 17.5169700 Critical F = 2.6404222 Numerator df = 3.0 Denominator df = 252 Total sample size = 85 Actual power = 0.9509128	Repeated measures ANOVA with four linear contrasts, comparing the *levator* muscle activity of four blocks (active viewing, distraction, distancing, suppression). ANOVA is calculated using aov_ez() function of the afex-package, estimated maginal means are calculated using emmeans() function from the emmeans-package, pairwise contrasts are calculated using pairs(). Bayes factors are computed for the ANOVA and each contrast using the BayesFactor-package.	ANOVA yields $p < .05$ is interpreted as *levator* muscle activity changing significantly with blocks. Values of *levator* muscle activity are interpreted as equal between blocks if $p > .05$. Each contrast yielding $p < .05$ is interpreted as *levator* muscle activity being different between those two blocks, magnitude and direction are inferred from the respective estimate. Values of *levator* muscle activity are interpreted as equal between blocks if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.

2.) Do ER strategies require cognitive effort? (Manipulation check)	2a) Subjective effort (effort ratings) is greater after using an emotion regulation strategy (distraction, distancing, suppression) compared to active viewing.	F tests - ANOVA: Repeated measures, within factors Analysis: A priori: Compute required sample size <u>Input:</u> Effect size f = 0.2041241 (η_p^2 = 0.04) (Scheffel et al., 2021) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of groups = 1 Number of measurements = 4 Corr among rep measures = 0.5 Nonsphericity correction ε = 1 <u>Output</u>: Noncentrality parameter λ = 17.6666588 Critical F = 2.6625685 Numerator df = 3.0 Denominator df = 156.0 Total sample size = 53 Actual power = 0.95206921	Repeated measures ANOVA with four linear contrasts, comparing the subjective effort ratings of four blocks (active viewing, distraction, distancing, suppression). ANOVA is calculated using aov_ez() function of the afex-package, estimated maginal means are calculated using emmeans() function from the emmeans-package, pairwise contrasts are calculated using pairs(). Bayes factors are computed for the ANOVA and each contrast using the BayesFactor-package.	ANOVA yields *p* < .05 is interpreted as effort ratings changing significantly with blocks. Values of effort ratings are interpreted as equal between blocks if *p* > .05. Each contrast yielding *p* < .05 is interpreted as effort ratings being different between those two blocks, magnitude and direction are inferred from the respective estimate. Values of effort ratings are interpreted as equal between blocks if *p* > .05. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	2b) Majority of participants reuse the strategy that was least effortful for them.	-	Subjects are asked about the reasons for their choice in the follow-up survey. These answers are classified into categories and counted.	The percentage choice of strategies is described descriptively.
3.) Which variables can predict individual subjective values of ER strategies?	3a) Subjective effort ratings negatively predict subjective values of ER strategies.	t tests - Linear multiple regression: Fixed model, single regression coefficient Analysis: A priori: Compute required sample size <u>Input</u>:	Multilevel model of SVs with level-1-predictors subjective effort, subjective arousal, *corrugator*, and *levator* muscle activity using subject specific	Fixed effects yield *p* < .05 are interpreted as subjective values are related to subjective effort. Subjective values are interpreted as not being related to subjective effort if *p* > .05.

		Tail(s) = One Effect size $f^2 = 0.34$ (Since there are no findings in this respect yet, we have inferred from the effect size in the closest-similar model: Westbrook et al., 2013) α err prob = 0.05 Power ($1-\beta$ err prob) = 0.95 Number of predictors = 4 <u>Output</u>: Noncentrality parameter $\delta = 3.4$ Critical t = 1.6991270 Df = 29 Total sample size = 34 Actual power = 0.9529571	intercepts and allowing random slopes for ER strategies. The null model and the random slopes model are calculated using lmer() of the lmerTest-package. Bayes factors are computed for the MLM using the BayesFactor-package.	The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	3b) Subjective arousal ratings negatively predict subjective values of ER strategies.			Fixed effects yield $p < .05$ are interpreted as subjective values are related to subjective arousal. Subjective values are interpreted as not being related to subjective arousal if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	3c) *Corrugator* muscle activity negatively predict subjective values of ER strategies.			Fixed effects yield $p < .05$ are interpreted as subjective values are related to *corrugator* activity. Subjective values are interpreted as not being related to *corrugator* activity if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	3d) *Levator* muscle activity negatively predict subjective values of ER strategies.			Fixed effects yield $p < .05$ are interpreted as subjective values are related to *levator* activity. Subjective values are interpreted as not being related to *levator* activity if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.

4.) Is the effort required for an ER strategy the best predictor for subjective values of ER strategies?	4a) Subjective values decline with increasing effort, even after controlling for task performance measured by subjective arousal ratings, *corrugator* and *levator* muscle activity.	t tests - Linear multiple regression: Fixed model, single regression coefficient Analysis: A priori: Compute required sample size <u>Input</u>: Tail(s) = One Effect size $f^2 = 0.34$ (Since there are no findings in this respect yet, we have inferred from the effect size in the closest-similar model: Westbrook et al., 2013) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of predictors = 4 <u>Output</u>: Noncentrality parameter $\delta = 3.4$ Critical t = 1.6991270 Df = 29 Total sample size = 34 Actual power = 0.9529571		Fixed effects yield $p < .05$ are interpreted as subjective values changing significantly with ER strategy. Subjective values are interpreted as equal between ER strategies if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
5.) Are subjective values related to flexible emotion regulation?	5a) The higher the subjective value, the more likely the respective strategy is chosen.	1) χ^2 tests – Goodness-of-fit tests_ Contingency tables Analysis: A priori: Compute required sample size <u>Input:</u> Effect size $\omega = 0.5$ (Based on our theoretical considerations, we assume a large effect) α err prob = 0.05 Power (1-β err prob) = 0.95 Df = 1 <u>Output:</u> Noncentrality parameter $\lambda = 19.8$ Critical $\chi^2 = 11.0704977$	1) Chi-squared test with the variables “predicted choice” (= highest SV of each participant) and “choice” (Strategy 1, 2, or 3) 2) Ordinal regression with dependent variable “Choice” (Strategy 1, 2, or 3) and independent variables “SV strategy 1”, “SV strategy 2” and “SV strategy 3”.	1) χ^2 yields $p < .05$ is interpreted as predicted choice (highest SV of each participant) and actual choice show significant consistency. Predicted choice and actual choice are interpreted as independent if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence. 2) Ordinal logistic regression yields $p < .05$ is interpreted as the respective subjective value has a significant

		Total sample size = 52 Actual power = 0.9500756 2) z tests –Logistic regression Analysis: A priori: Compute required sample size <u>Input:</u> Tails: One Pr(Y=1|X=1) H1 = 0.80 (Based on our theoretical considerations, that a higher SVs should lead almost certainly to the choice of the respective strategy) Pr(Y=1|X=1) H0 = 0.333 (Based on theoretical considerations: if all SVs are equal, choice is on chance level) α err prob = 0.05 Power (1-β err prob) = 0.95 R^2 other X = 0 X distribution: normal X param μ = 0 X param σ = 1 <u>Output:</u> Critical z = 1.6448536 Total sample size = 25 Actual power = 0.9528726		influence on the OR of the choice of a strategy. Respective SV is interpreted as not related to choice if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
	5b) Subjective values are lower and decline stronger when ER flexibility is lower.	t tests – Linear multiple regression: Fixed model, single regression coefficient Analysis: A priori: compute required sample size <u>Input:</u> Tail(s) = One	SVs will be ordered by magnitude. Values will be fitted in a GLM to estimate the individual intercept and slope. A linear regression will be computed with intercept and	β yield $p < .05$ are interpreted as significant association between predictor (intercept, slope) and ER flexibility. The direction of effect is interpreted according to sign (negative or positive). *p* – values > .05 are

		Effect size $f^2 = 0.15$ (as there is no evidence in the literature, we assume a medium sized effect) α err prob = 0.05 Power (1-β err prob) = 0.95 Number of predictors = 2 <u>Output:</u> Noncentrality parameter δ = 3.316662 Critical t = 1.69665997 Df = 71 Total sample size = 74 Actual power = 0.95101851	slope as predictors and FlexER score as criterion.	interpreted as no association between predictor and ER flexibility. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.
Exploratory: Are individual subjective values of ER strategies related to personality traits?			Multilevel model of SVs with level-1-predictors subjective effort, subjective arousal, *corrugator*, and *levator* muscle activity and level-2-predictors NFC and self-control using subject specific intercepts and allowing random slopes for ER strategies. The null model and the random slopes model are calculated using lmer() of the lmerTest-package. Bayes factors are computed for the MLM using the BayesFactor-package.	Fixed effects yield $p < .05$ are interpreted as subjective values are related to NFC and self-control. Subjective values are interpreted as not being related to subjective effort if $p > .05$. The Bayes factor *BF10* is reported alongside every *p*-value to assess the strength of evidence.

Supplement

Supplementary Material 1

Table S1

List of IAPS (Lang, Bradley, and Cuthbert, 2008) and EmoPicS (Wessa et al., 2010) used in the ER paradigm.

	Neutral	Negative 1	Negative 2	Negative 3	Negative 4	Negative 5
	083†	225†	210†	208†	227†	223†
	107†	230†	218†	219†	252†	238†
	124†	255†	222†	226†	1051*	245†
	140†	327†	228†	253†	2800*	2981*
	143†	1111*	246†	254†	3061*	3016*
	7000*	3017*	251†	326†	3230*	3101*
	7002*	3022*	2703*	1301*	6561*	3181*
	7004*	3180*	3051*	3350*	6838*	3215*
	7006*	3280*	3160*	6242*	9120*	3220*
	7009*	6190*	3185*	6410*	9181*	3225*
	7021*	6244*	3301*	6555*	9185*	6020*
	7025*	6836*	6562*	6825*	9230*	6571*
	7041*	9180*	9031*	6940*	9254*	6831*
	7100*	9182*	9040*	8230*	9295*	8231*
	7150*	9253*	9042*	9041*	9332*	9373*
	7185*	9300*	9043*	9140*	9411*	9400*
	7211*	9326*	9145*	9340*	9420*	9402*
	7224*	9424*	9160*	9409*	9421*	9403*
	7233*	9425*	9184*	9570*	9599*	9405*
	7235*	9920*	9904*	9800*	9905*	9423*
Valence	4.86 ± 0.49	2.84 ± 0.57	2.64 ± 0.46	2.82 ± 0.62	2.65 ± 0.75	2.74 ± 0.70
Arousal	3.01 ± 0.61	5.62 ± 0.34	5.58 ± 0.38	5.60 ± 0.39	5.61 ± 0.41	5.63 ± 0.37

Note. * Pictures taken from the IAPS (Lang, Bradley, and Cuthbert, 2008); † Pictures taken from the EmoPicS (Wessa et al., 2010).

Pilot study: Subjective arousal in the conditions “Active viewing - neutral” and “Active viewing - negative”

ANOVA:

Effect	df	MSE	F	ges	p.value
block	1, 15	3895.91	34.32 ***	.475	<.001

$BF10$ = 1,244.99

Paired contrasts:

Table 1

Paired contrasts for the rmANOVA comparing subjective arousal of negative and neutral pictures in the condition "active viewing".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\% CI$
$View_{neutral} - View_{negative}$	-129.28	22.07	15.00	-5.86	0.00	794.78	0.70	[0.43, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Figure:

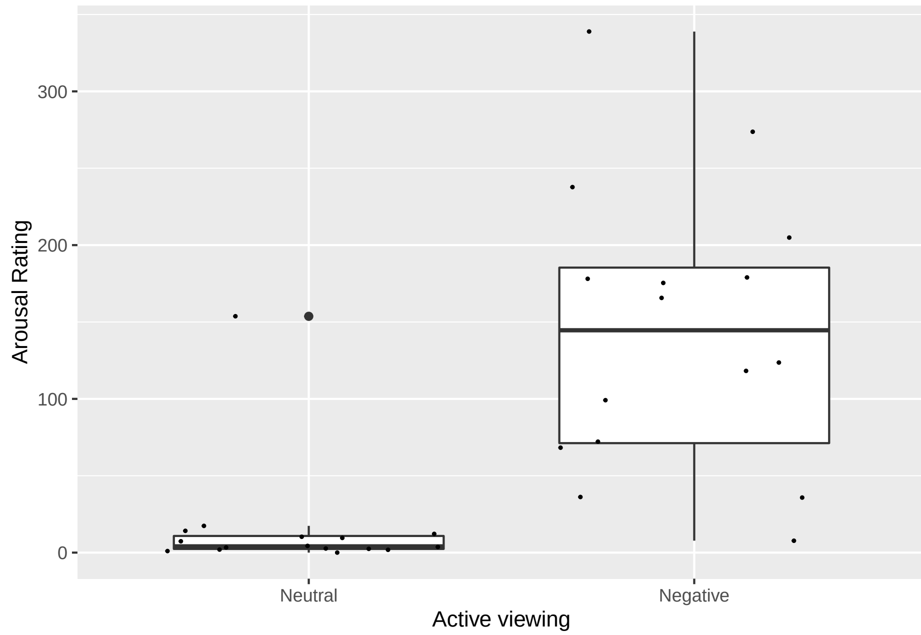

Figure 2. Subjective arousal ratings for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.

Pilot study: Subjective arousal in the conditions “Active viewing - negative”, “Distraction”, “Distancing”, and “Suppression”

ANOVA:

Effect	df	MSE	F	ges	p.value
block	2.79, 41.89	2238.27	1.17	.011	.332

$BF10 = 0.11$

Paired contrasts:

Table 2

Paired contrasts for the rmANOVA comparing subjective arousal of conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\%CI$
$View_{negative} - Distraction$	-0.74	16.14	45.00	-0.05	1.00	0.26	4.68e-05	[0.00, 1.00]
$View_{negative} - Distancing$	-5.35	16.14	45.00	-0.33	1.00	0.27	2.43e-03	[0.00, 1.00]
$View_{negative} - Suppression$	-26.23	16.14	45.00	-1.63	0.67	1.25	0.06	[0.00, 1.00]
$Distraction - Distancing$	-4.61	16.14	45.00	-0.29	1.00	0.26	1.81e-03	[0.00, 1.00]
$Distraction - Suppression$	-25.49	16.14	45.00	-1.58	0.73	0.77	0.05	[0.00, 1.00]
$Distancing - Suppression$	-20.88	16.14	45.00	-1.29	1.00	0.52	0.04	[0.00, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Figure:

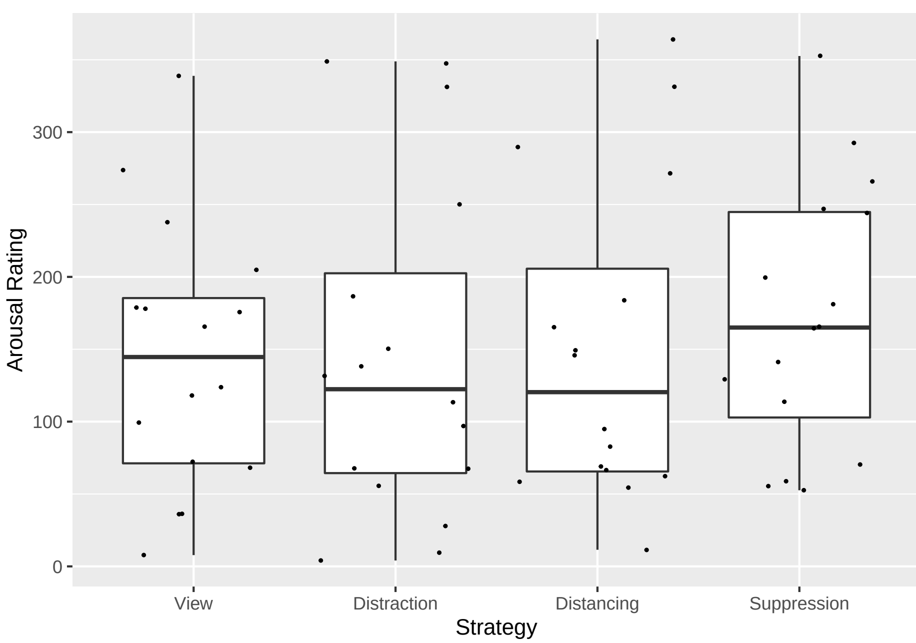

Figure 3. Subjective arousal ratings for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.

Pilot study: Physiological arousal (*Corrugator* and *Levator* activity) in the conditions "Active viewing - neutral" and "Active viewing - negative"

Corrugator: ANOVA:

Effect	df	MSE	F	ges	p.value
block	1, 15	1.01	9.70 **	.237	.007

$BF10 = 6{,}690{,}401.91$

Paired contrasts:

Table 3

Paired contrasts for the rmANOVA comparing physiological arousal (Corrugator activity) of negative and neutral pictures in the condition "active viewing".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\%CI$
$View_{neutral} - View_{negative}$	-1.11	0.36	15.00	-3.11	0.01	5,019,313.20	0.39	[0.09, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Levator: ANOVA:

Effect	df	MSE	F	ges	p.value
block	1, 15	0.17	7.72 *	.162	.014

$BF10 = 48.44$

Paired contrasts:

Table 4

Paired contrasts for the rmANOVA comparing physiological arousal (Levator activity) of negative and neutral pictures in the condition "active viewing".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\%CI$
$View_{neutral} - View_{negative}$	-0.40	0.14	15.00	-2.78	0.01	41.02	0.34	[0.05, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Figures:

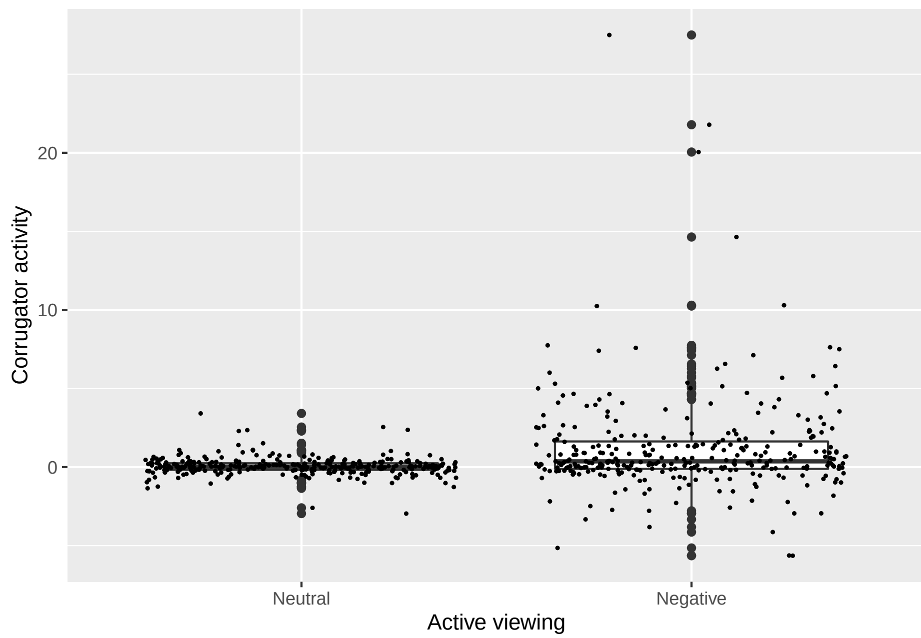

Figure 4. Corrugator activity for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the corrugator activity of a single trial. Bold dots represent outliers.

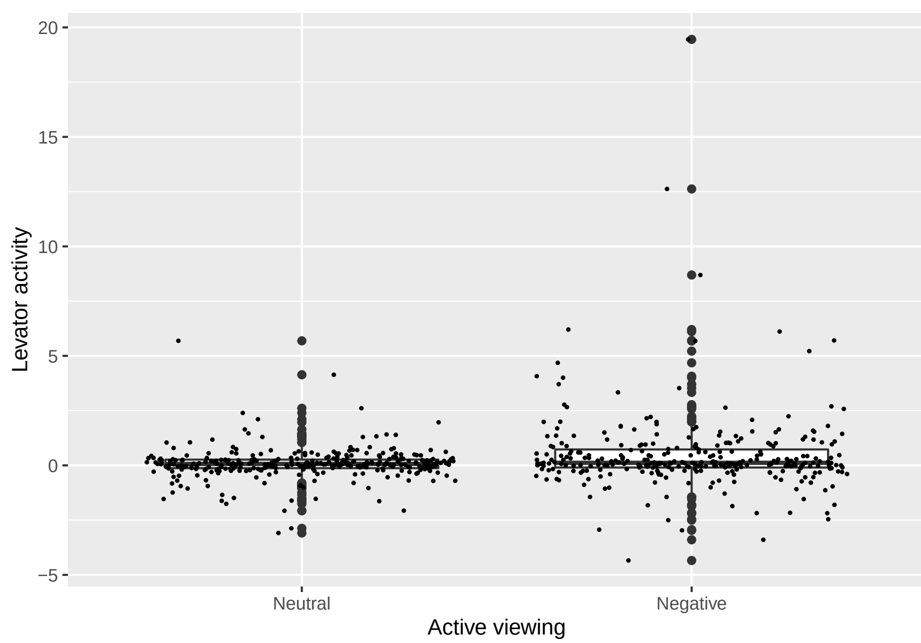

Figure 5. Levator activity for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the levator activity of a single trial. Bold dots represent outliers.

Pilot study: Physiological arousal (*Corrugator* and *Levator* activity) in the conditions“Active viewing - negative”, “Distraction”, “Distancing”, and “Suppression”

Corrugator: ANOVA:

Effect	df	MSE	F	ges	p.value
block	1.53, 22.98	1.16	5.71 *	.189	.015

$BF10 =$ 5,257,689.54

Paired contrasts:

Table 5

Paired contrasts for the rmANOVA comparing physiological arousal (Corrugator activity) of conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\% CI$
$View_{negative} - Distraction$	0.88	0.27	45.00	3.22	0.01	4,962.89	0.19	[0.05, 1.00]
$View_{negative} - Distancing$	0.95	0.27	45.00	3.50	0.01	616.63	0.21	[0.06, 1.00]
$View_{negative} - Suppression$	0.92	0.27	45.00	3.40	0.01	11,678.82	0.20	[0.06, 1.00]
$Distraction - Distancing$	0.08	0.27	45.00	0.28	1.00	0.07	1.78e-03	[0.00, 1.00]
$Distraction - Suppression$	0.05	0.27	45.00	0.18	1.00	0.08	7.22e-04	[0.00, 1.00]
$Distancing - Suppression$	-0.03	0.27	45.00	-0.10	1.00	0.06	2.36e-04	[0.00, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Levator: ANOVA:

Effect	df	MSE	F	ges	p.value
block	2.07, 31.00	0.20	8.27 **	.225	.001

$BF10 =$ 672,341.29

Paired contrasts:

Table 6

Paired contrasts for the rmANOVA comparing physiological arousal (Levator activity) of conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\%CI$
$View_{negative} - Distraction$	0.42	0.13	45.00	3.24	0.01	58.02	0.19	[0.05, 1.00]
$View_{negative} - Distancing$	0.45	0.13	45.00	3.46	0.01	93.49	0.21	[0.06, 1.00]
$View_{negative} - Suppression$	0.62	0.13	45.00	4.79	0.00	6,253.91	0.34	[0.16, 1.00]
$Distraction - Distancing$	0.03	0.13	45.00	0.22	1.00	0.07	1.06e-03	[0.00, 1.00]
$Distraction - Suppression$	0.20	0.13	45.00	1.54	0.78	1.52	0.05	[0.00, 1.00]
$Distancing - Suppression$	0.17	0.13	45.00	1.32	1.00	0.52	0.04	[0.00, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Figures:

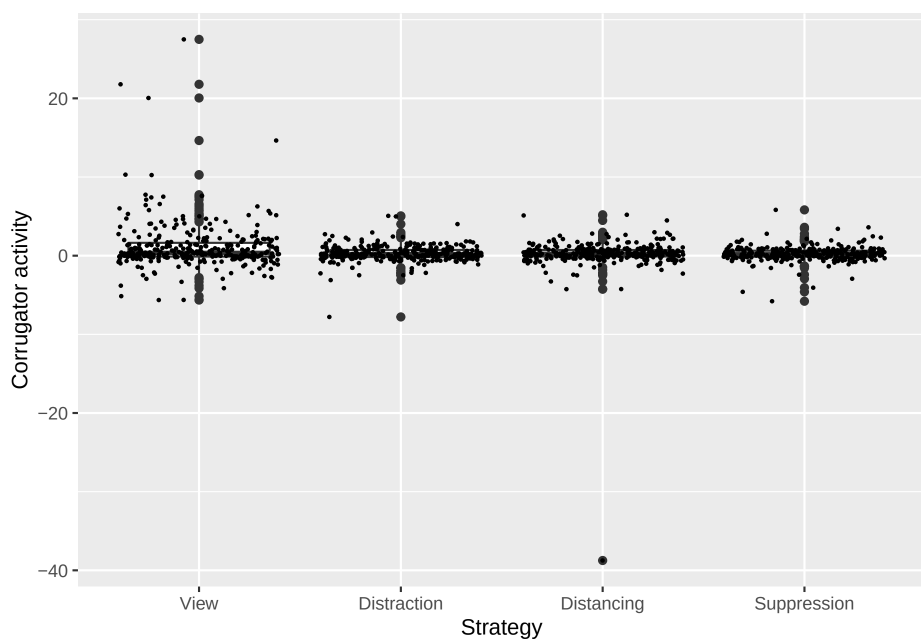

Figure 6. Corrugator activity for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the corrugator activity of a single trial. Bold dots represent outliers.

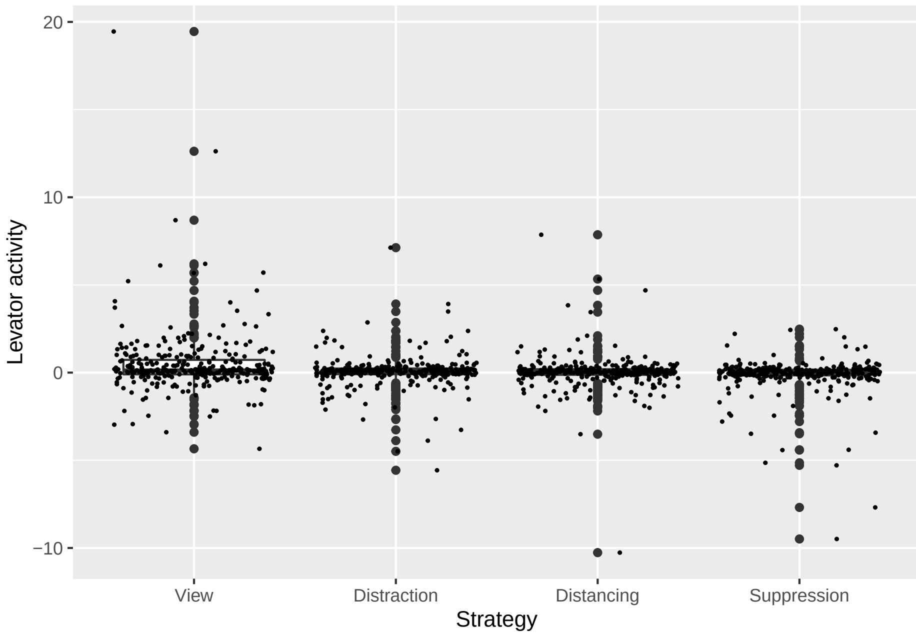

Figure 7. Levator activity for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the levator activity of a single trial. Bold dots represent outliers.

606 **Pilot study: Subjective effort in the conditions “Active viewing - negative”,**
607 **“Distraction”, “Distancing”, and “Suppression”**

608 ANOVA:

609

Effect	df	MSE	F	ges	p.value
block	2.38, 35.66	4388.19	11.13 ***	.185	<.001

610 $BF10 = 7.40$

611 Paired contrasts:

Table 7

Paired contrasts for the rmANOVA comparing subjective effort of conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression".

Contrast	Estimate	SE	df	t	p	$BF10$	η_p^2	$95\%CI$
$View_{negative} - Distancing$	-110.72	20.85	45.00	-5.31	0.00	59.77	0.39	[0.20, 1.00]
$View_{negative} - Distraction$	-89.72	20.85	45.00	-4.30	0.00	20.49	0.29	[0.12, 1.00]
$View_{negative} - Suppression$	-88.15	20.85	45.00	-4.23	0.00	33.13	0.28	[0.11, 1.00]
$Distraction - Distancing$	21.00	20.85	45.00	1.01	1.00	0.50	0.02	[0.00, 1.00]
$Distraction - Suppression$	22.57	20.85	45.00	1.08	1.00	0.57	0.03	[0.00, 1.00]
$Distancing - Suppression$	1.57	20.85	45.00	0.08	1.00	0.26	1.27e-04	[0.00, 1.00]

Note. SE = standard error, df = degrees of freedom, t = t-statistic, p = p-value, CI = confidence interval.

Figure:

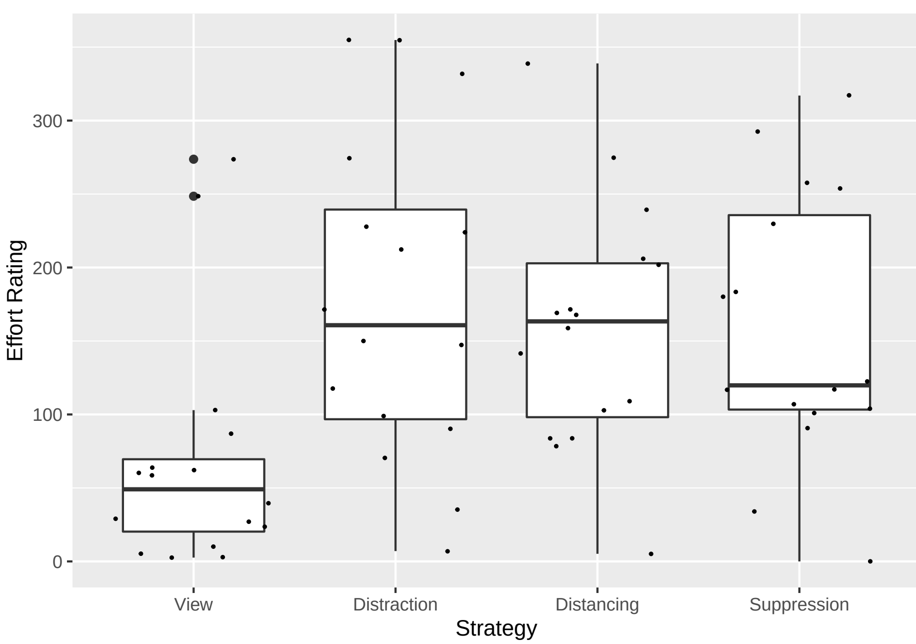

Figure 8. Subjective effort ratings for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.