Estimating individual subjective values of emotion regulation strategies

- Christoph Scheffel^{†,1}, Josephine Zerna^{†,1}, Anne Gärtner¹, Denise Dörfel¹, & Alexander

 Strobel¹
- ¹ Faculty of Psychology, Technische Universität Dresden, 01069 Dresden, Germany

Author Note

1

5

- The authors made the following contributions. Christoph Scheffel: Conceptualization,
- ⁷ Methodology, Funding acquisition, Formal analysis, Investigation, Project administration,
- 8 Software, Visualization, Writing original draft preparation, Writing review & editing;
- 9 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.
- 14 Correspondence concerning this article should be addressed to Christoph Scheffel,
- ¹⁵ Zellescher Weg 17, 01069 Dresden, Germany. E-mail: christoph_scheffel@tu-dresden.de

16 Abstract

17 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

²⁰ are motivated to reduce their emotional arousal effectively and to avoid cognitive effort.

21 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

²³ 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)

27 and subjective effort. Relations of SVs to personality traits will be explored

28 Keywords: emotion regulation, regulatory effort, effort discounting, registered report,

29 specification curve analysis

Word count: X

Estimating individual subjective values of emotion regulation strategies

32

31

33

1. Introduction

Every day we are confronted with stressful or emotionally demanding situations. The 34 ability to modify emotional experiences, expressions, and physiological reactions¹ to regulate emotions is an important cognitive skill. It is therefore not surprising that 36 emotion regulation (ER) has substantial implications for well-being and adaptive 37 functioning.² Different strategies can be used to regulate emotions, namely situation 38 selection, situation modification, attentional deployment, cognitive change, and response 39 modification, and, following the taxonomy of Powers and LaBar, individuals can implement ER strategies by means of different tactics. So called antecedent-focused 41 strategies, e.g., attentional deployment and cognitive change, take effect early in the emotion generation process. In contrast, response modification takes place late in the process and is therefore conceptualized as a response-focused strategy. 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*. 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 55 (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 64 with the concept of ER flexibility. Similar to other psychological domains, e.g., attention 65 and goal pursuit, maladaptive now refers to inflexible strategy use or use of strategies that 66 are hindering the achievement of goals. 9 Adaptive flexible ER requires having a large 67 repertoire of ER strategies. 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 71 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 75 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 89 perspective with the concept of ER flexibility. Similar to other psychological domains such 90 as attention and goal pursuit, maladaptive now refers to inflexible strategy use or the use of 91 strategies that are hindering the achievement of goals. In contrast, adaptive ER requires 92 having a large repertoire of ER strategies to choose from⁹ 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 101 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 103 which strategy they wanted to use again. Participants tended to re-apply the strategy that 104 was less effortful, even though it was not the most effective one for them - in this case: 105 Suppression. Interestingly, the choice was independent of self reported habitual use of 106 suppression and reappraisal. Research on ER choice has shown that individuals choose a 107 strategy that they prefer for different, relatively well-known reasons. However, nothing has 108 been investigated regarding the strategy that is *not* chosen. 109

We assume that people choose the strategy that has the highest value for them at

110

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. One 112 certainly central goal is the regulation of negative affect. A second, intrinsic and rather less 113 obvious goal is the avoidance of effort. When given the choice, most individuals prefer 114 tasks that are less effortful. 16 We assume that both aspects are traded off against each 115 other by individuals to determine individual subjective values (SVs) of ER strategies: A 116 strategy is more valuable if it can reduce emotional arousal and is less effortful. SVs of ER 117 Strategies could be helpful to describe the ER repertoire⁹ more comprehensively. 118 Depending on the flexibility of a person, different patterns of SVs could be conceivable: A 119 person with high flexibility would show relatively high SVs for a number of strategies. This 120 would mean that all strategies are a good option for goal achievement. A second person 121 with less flexibilty, however, would show high SVs only for one strategy, or for no strategy 122 at all. This in turn would mean that no strategy is a good choice to achieve ER goals. 123

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

128

129

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 138 perceived subjective effort of ER strategies. ¹⁴ Moreover, there is nothing like an objectively 139 easiest ER strategy. Therefore, we have to add an additional step, which preceds the other 140 steps and where the option with the higher subjective value is determined. In this step, the 141 same monetary value (i.e., $1 \in$) is assigned to both options. The assumption is that 142 participants now choose the option that has the higher SV for them. In the next step we 143 return to the original paradigm. The higher monetary value (i.e., $2 \in$) is assigned to the 144 option that was not chosen in the first step and therefore is assumed to have the lower SV. 145 In the following steps, the lower value is changed in every iteration according to Westbrook 146 et al.¹⁷ until the indifference point is reached. This procedure will be repeated until all 147 strategies have been compared. The SV of each strategy is calculated as the mean of this 148 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 152 SVs (see figure 1). Additionally, we will test our adapted paradigm in a n-back task and 153 explore whether this paradigm can describe individuals that do not prefer the easiest n-back option (see Zerna et al., ...). 155

1.2 The present study

161

162

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.

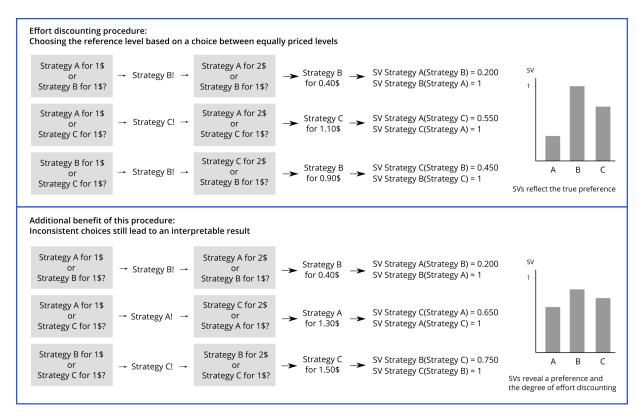


Figure 1. Exemplary visualisation of two response patterns. In the top half, the person has a clear preference for one of the three strategies. In the lower half, they have no clear preference and therefore show an inconsistent response pattern. This pattern can also be represented by our paradigm.

• H1b) Physiological arousal (EMG corrugator activity) is lower after using an ER strategy (distraction, distancing, suppression) compared to active viewing.

163

164

165

166

167

168

169

- 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.

175

181

- 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. The paradigm was written and presented using Psychopy. We used R with R Studio^{20,21} with the main packages afex²² and BayesFactor²³ 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.

199 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 (... and physiological ?!) arousal for active viewing of negative pictures, compared to active viewing of neutral pictures. However, ER strategies could not reduce subjective arousal compared to active viewing of negative pictures. Yet we found accordance with our previous studies that the use of ER strategies compared to active viewing was associated with increased subjective effort.

207 **2.3 Design**

Young, healthy participants (aged 18 to 30 years) will be recruited using the software 208 ORSEE²⁴ at the Technische Universität Dresden. Participants will be excluded from 200 participation if they do not fluently speak German, have current or a history of 210 psychological disorders or neurological trauma, or report to take medication. Participants 211 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 213 lab sessions, which take place in a shielded cabin with constant lighting. Before each session, participants receive information about the respective experimental procedure and 215 provide informed consent. At the beginning of the first session participants fill out a 216 demographic questionnaire and complete an n-Back task with the levels one to four. Then, 217

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 219 exactly one week after session one. Participants provide informed consent and receive 220 written instructions on the ER paradigm and ER strategies that they should apply. A brief 221 training ensures that all participants are able to implement the ER strategies. Next, 222 electrodes to measure EMG being attached and the ER paradigm is conducted. 223 Participants receive 30.00€ or course credit as compensation. Study data are being 224 collected and managed using REDCap electronic data capture tools hosted at Technische 225 Universität Dresden.^{25,26} 226

2.3.1 Psychometric measures. The online survey contains a number of 227 questionnaires: General psychological well-being is assessed using the German version of 228 the WHO-5 scale.^{27,28} To capture the construct of resilience, the German version 220 10-item-form of the Connor-Davidson resilience Scale (CD-RISC)³¹ is used. Dispositional 230 use of ER is assessed using the German version of the Emotion Regulation Questionnaire 231 (ERQ).^{6,32} For the assessment of ER ability we use the Flexible Emotion Regulation Scale 232 (FlexER).³³ Implicit theories of willpower in emotion control are assessed using the implicit 233 theories questionnaire from.³⁴ To assess Need for Cognition, the German version short form 234 of the Need for Cognition Scale^{35,36} is used. To assess self-control, sum scores of the 235 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 246 pictures and permit all emotions that may arise." In the second block, participants actively 247 viewe negative pictures. During the third, fourth, and fifth block, participants see negative 248 pictures and are asked to regulate their emotions using distraction, distancing, and 240 suppression. In order to achieve distraction, participants are asked to think of a geometric 250 object or an everyday activity, like brushing their teeth. During distancing, participants 251 are asked to "take the position of a non-involved observer, thinking about the picture in a 252 neutral way." Participants are told not to re-interpret the situation or attaching a different 253 meaning to the situation. During suppression, participants are told to "suppress their 254 emotional facial expression." They should imagine being observed by a third person that 255 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 257 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 259 asked about their applied ER strategies to avoid misapplication. The order of the three 260 regulation blocks (distraction, distancing, and suppression) is randomized between 261 participants. 262

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. 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 282 Picture Set (EmoPicS)⁴⁴ and the International Affective Picture System (IAPS).⁴⁵ The 20 283 neutral pictures (Valence (V): $M \pm SD = 4.81 \pm 0.51$; Arousal (A): $M \pm SD = 3 \pm 0.65$) 284 depicted content related to the categories persons, objects, and scenes. Further, 100 285 negative pictures, featuring categories animals, body, disaster, disgust, injury, suffering, 286 violence, and weapons, were used. An evolutionary algorithm⁴⁶ was used to cluster these pictures into five sets with comparable valence and arousal values (set one: V: $M \pm SD =$ 2.84 ± 0.57 , A: $M \pm SD = 5.62 \pm 0.34$; set two: V: $M \pm SD = 2.64 \pm 0.46$, A: $M \pm SD = 2.64 \pm 0.46$ 5.58 ± 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 292 can be found in supplementary material X. 293
- 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 299 electrolyte gel. Raw signals will be amplified by a BrainAmp amplifier (Brain Products 300 Inc., Gilching, Germany). Impedance level will be kept below 10 $k\Omega$. Data will be sampled 301 at 1000 Hz, filtered, rectified and integrated. A 20 Hz high pass (order 8), a 300 Hz low 302 pass (order 8), and a 50 Hz notch filter will be applied to both signals. Corrugator and 303 Levator EMG will be analyzed during the 6 s of picture presentation. EMG data will be 304 baseline-corrected using a time window of 2 s prior to stimulus onset. cf, 47 Last, the 305 sampling rate will be changed to 100 Hz, and EMG data will be averaged for each condition and each participant. 307

$_{308}$ 2.4 Sampling plan

Sample size calculation is done using G^*Power . In a meta-analysis of Zaehringer 309 and colleagues,⁵ effect sizes of ER on peripheral-physiological measures were reported. To 310 find an effect of d = -0.32 of ER on corrugator muscle activity with $\alpha = .05$ and $\beta = .95$, 311 data of least N=85 have to be analyzed. Power analyses of all other hypotheses yielded 312 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 315 fatigue), respective data will also be excluded from further analyses. Therefore, we aim to 316 collect data of 90 participants. 317

318 2.5 Analysis plan

All statistical analyses will be performed using RStudio (version 1.4.1717)²¹ and R (version 4.1.0)²⁰ for Windows. 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, 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, 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.

Subjective values of emotion regulation strategies For each ER strategy, SVs will be 339 calculated as follows: first, the value 0.015625 will be added to or subtracted from the last 340 monetary value of the flexible strategy, depending on the participant's last choice. Second, 341 the resulting (monetary) value will be divided by 2.00 €. The final SV for each participant 342 will be computed by averaging all final SVs of each strategy. The resulting values will be 343 entered in a rmANOVA to compare the SVs of the three strategies (distraction, distancing, 344 and suppression) to explore for group effects. Again, estimated, marginal means will be 345 computed as post-hoc contrasts. 346

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. 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.

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

Level-1-predictors will be centered within cluster. Essiduals 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 threir
FlexER score. Therefore, SVs will be ordered by magnitude firstly. 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/.

375 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.

References

379 1.

Gross, J. J. Antecedent- and response-focused emotion regulation: Divergent consequences for experience, expression, and physiology. *Journal of Personality and Social Psychology* **74**, 224–37 (1998).

382 2.

Gross, J. J. The emerging field of emotion regulation: An integrative review. Review of General Psychology 2, 271–299 (1998).

385 3.

Powers, J. P. & LaBar, K. S. Regulating emotion through distancing: A taxonomy, neurocognitive model, and supporting meta-analysis. *Neuroscience and Biobehavioral Reviews* **96**, 155–173 (2019).

₃₈₈ 4.

Webb, T. L., Miles, E. & Sheeran, P. Dealing with feeling: A meta-analysis of the effectiveness of strategies derived from the process model of emotion regulation. *Psy-chological Bulletin* **138**, 775–808 (2012).

391 5.

Zaehringer, J., Jennen-Steinmetz, C., Schmahl, C., Ende, G. & Paret, C. Psychophysiological effects of downregulating negative emotions: Insights from a meta-analysis of healthy adults. Front Psychol 11, 470 (2020).

394 6.

Gross, J. J. & John, O. P. Individual differences in two emotion regulation processes:

Implications for affect, relationships, and well-being. *Journal of Personality and Social Psychology* **85**, 348–62 (2003).

397 7.

Yoon, K. L., Maltby, J. & Joormann, J. A pathway from neuroticism to depression: Examining the role of emotion regulation. *Anxiety, Stress, & Coping* **26**, 558–72 (2013).

400 8.

Aldao, A., Nolen-Hoeksema, S. & Schweizer, S. Emotion-regulation strategies across psychopathology: A meta-analytic review. *Clinical Psychology Review* **30**, 217–237 (2010).

403 9.

Aldao, A., Sheppes, G. & Gross, J. J. Emotion regulation flexibility. *Cognitive Therapy and Research* **39**, 263–278 (2015).

406 10.

Blanke, E. S. et al. Mix it to fix it: Emotion regulation variability in daily life.

Emotion 20, 473–485 (2020).

409 11.

Kable, J. W. & Glimcher, P. W. The neural correlates of subjective value during intertemporal choice. *Nat Neurosci* **10**, 1625–33 (2007).

- 412 12.
- Sheppes, G., Scheibe, S., Suri, G. & Gross, J. J. Emotion-regulation choice. *Psycho-*
- logical Science **22**, 1391–6 (2011).
- 415 13.
- Sheppes, G. et al. Emotion regulation choice: A conceptual framework and supporting
- evidence. Journal of Experimental Psychology: General 143, 163–81 (2014).
- 418 14.
- Scheffel, C. et al. Effort beats effectiveness in emotion regulation choice: Differences
- between suppression and distancing in subjective and physiological measures. Psy-
- chophysiology 00, e13908 (2021).
- 421 15.
- Inzlicht, M., Shenhav, A. & Olivola, C. Y. The effort paradox: Effort is both costly
- and valued. Trends Cogn Sci 22, 337–349 (2018).
- 424 16.
- Hull, C. L. Principles of behavior: An introduction to behavior theory. (Appleton-
- Century-Crofts, 1943).
- 427 17.
- Westbrook, A., Kester, D. & Braver, T. S. What is the subjective cost of cognitive
- effort? Load, trait, and aging effects revealed by economic preference. PLOS ONE
- **8**, e68210 (2013).
- 430 18.
- Simmons, J. P., Nelson, L. D. & Simonsohn, U. A. A 21 word solution. SSRN
- Electronic Journal (2012) doi:10.2139/ssrn.2160588.
- 433 19.

- Peirce, J. et al. PsychoPy2: Experiments in behavior made easy. Behavior Research

 Methods 51, 195–203 (2019).
- 436 20.
- R Core Team. R: A language and environment for statistical computing. (R Foundation for Statistical Computing, 2021).
- 439 21.
- RStudio Team. RStudio: Integrated development for R. (2020).
- 442 22.

441

- Singmann, H., Bolker, B., Westfall, J., Aust, F. & Ben-Shachar, M. S. Afex: Analysis of factorial experiments. (2021).
- 445 23.
- Morey, R. D. & Rouder, J. N. BayesFactor: Computation of Bayes factors for common designs. (2021).
- 448 24.
- Greiner, B. Subject pool recruitment procedures: Organizing experiments with ORSEE. Journal of the Economic Science Association 1, 114–125 (2015).
- 451 25.
- Harris, P. A. et al. Research electronic data capture (REDCap)—A metadata-driven methodology and workflow process for providing translational research informatics support. Journal of Biomedical Informatics 42, 377–381 (2009).
- 454 26.
- Harris, P. A. et al. The REDCap consortium: Building an international community of software platform partners. Journal of Biomedical Informatics 95, 103208 (2019).

- 27. 457
- Bech, P. Measuring the dimensions of psychological general well-being by the WHO-5. 458
- Quality of life newsletter **32**, 15–16 (2004). 459
- 28. 460
- Brähler, E., Mühlan, H., Albani, C. & Schmidt, S. Teststatistische prüfung und 461 normierung der deutschen versionen des EUROHIS-QOL lebensqualität-index und
- des WHO-5 wohlbefindens-index. Diagnostica 53, 83–96 (2007). 462
- 29. 463
- Connor, K. M. & Davidson, J. R. Development of a new resilience scale: The connor-464 davidson resilience scale (CD-RISC). Depression and Anxiety 18, 76–82 (2003).
- 30. 466

465

- Sarubin, N. et al. First analysis of the 10-and 25-item german version of the connor-467 davidson resilience scale (CD-RISC) regarding psychometric properties and components. Zeitschrift Fur Gesundheitspsychologie 23, 112–122 (2015). 468
- 31. 469
- Campbell-Sills, L. & Stein, M. B. Psychometric analysis and refinement of the connor-470 davidson resilience scale (CD-RISC): Validation of a 10-item measure of resilience. Journal of Traumatic Stress 20, 1019–28 (2007).
- 32. 472

471

- Abler, B. & Kessler, H. Emotion regulation questionnaire a german version of the 473 ERQ by gross and john. *Diagnostica* **55**, 144–152 (2009). 474
- 33. 475
- Dörfel, D., Gärtner, A. & Strobel, A. A new self-report instrument for measuring 476 emotion regulation flexibility. Society for Affective Science (SAS) Annual Conference (2019).477

478 34.

Bernecker, K. & Job, V. Implicit theories about willpower in resisting temptations and emotion control. Zeitschrift Fur Psychologie-Journal of Psychology 225, 157–166 (2017).

481 35.

Cacioppo, J. T. & Petty, R. E. The need for cognition. *Journal of Personality and Social Psychology* **42**, 116–131 (1982).

484 36.

Bless, H., Wanke, M., Bohner, G., Fellhauer, R. F. & Schwarz, N. Need for cognition
- a scale measuring engagement and happiness in cognitive tasks. Zeitschrift Für

Sozialpsychologie 25, 147–154 (1994).

487 37.

Paschke, L. M. et al. Individual differences in self-reported self-control predict successful emotion regulation. Social Cognitive and Affective Neuroscience 11, 1193–204 (2016).

490 38.

Schwarzer, R., Diehl, M. & Schmitz, G. S. Self-regulation scale. (1999).

493 39.

492

Tangney, J. P., Baumeister, R. F. & Boone, A. L. High self-control predicts good adjustment, less pathology, better grades, and interpersonal success. *Journal of Personality* **72**, 271–324 (2004).

496 40.

Sproesser, G., Strohbach, S., Schupp, H. & Renner, B. Candy or apple? How self-497 control resources and motives impact dietary healthiness in women. Appetite 56, 784–787 (2011). 498

41. 499

Patton, J. H., Stanford, M. S. & Barratt, E. S. Factor structure of the barratt impul-500 siveness scale. Journal of Clinical Psychology 51, 768–774 (1995). 501

42. 502

Hartmann, A. S., Rief, W. & Hilbert, A. Psychometric properties of the german version of the barratt impulsiveness scale, version 11 (BIS-11) for adolescents. Perceptual and Motor Skills 112, 353–368 (2011). 504

43. 505

Derryberry, D. & Reed, M. A. Anxiety-related attentional biases and their regulation 506 by attentional control. Journal of abnormal psychology 111, 225–236 (2002). 507

44. 508

509

Wessa, M. et al. EmoPicS: Subjective und psychophysiologische evalueation neuen bildmaterials für die klinisch-biopsychologische forschung. Zeitschrift für Klinische Psychologie und Psychotherapie 39, 77 (2010). 510

45. 511

Lang, P. J., Bradley, M. M. & Cuthbert, B. N. International affective picture system 512 (IAPS): Affective ratings of pictures and instruction manual. (University of Florida, 2008). 513

46. 514

Yu, X. & Gen, M. Introduction to evolutionary algorithms. (Springer Science & Busi-515 ness Media, 2010). 516

- 517 47.
- Bradley, M. M. & Lang, P. J. Measuring emotion: Behavior, feeling, and physiology. in *Cognitive neuroscience of emotion* (eds. Lane, R. D. & Nadel, L.) 242–276 (Oxford University Press, 2000).
- 520 48.
- Fridlund, A. J. & Cacioppo, J. T. Guidelines for human electromyographic research.

 Psychophysiology 23, 567–89 (1986).
- 523 49.
- Faul, F., Erdfelder, E., Lang, A.-G. & Buchner, A. G*Power 3: A flexible statistical power analysis program for the social, behavioral, and biomedical sciences. *Behavior Research Methods* **39**, 175–191 (2007).
- 526 50.
- Faul, F., Erdfelder, E., Buchner, A. & Lang, A.-G. Statistical power analyses using G*Power 3.1: Tests for correlation and regression analyses. *Behavior Research Methods* 41, 1149–1160 (2009).
- 529 51.
- Kuznetsova, A., Brockhoff, P. B. & Christensen, R. H. B. lmerTest package: Tests in linear mixed effects models. *Journal of Statistical Software* 82, 1–26 (2017).
- 532 52.
- Enders, C. K. & Tofighi, D. Centering predictor variables in cross-sectional multilevel models: A new look at an old issue. *Psychological Methods* **12**, 121–138 (2007).

535

543

548

549

Acknowledgements

This research is partly funded by the German Research Foundation (DFG) as part of
the Collaborative Research Center (CRC) 940. Additionally, we have applied for funding of
the participants' compensation from centralized funds of the Faculty of Psychology at
Technische Universität Dresden. Applications for the centralized funds will be reviewed in
May. Regardless of whether or not this additional funding will be granted, the study can
commence immediately. The funders have/had no role in study design, data collection and
analysis, decision to publish or preparation of the manuscript.

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.

550

Figures and figure captions

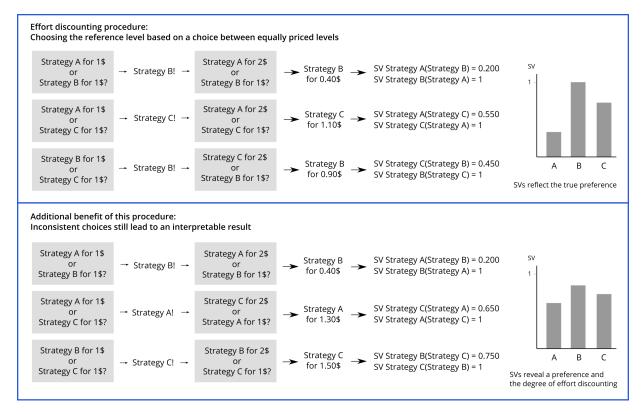


Figure 1

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 $(1-\beta$ err prob) = 0.95 Number of groups = 1 Number of measurements = 4 Corr among rep measures = 0.5 Nonsphericity correction $\epsilon = 1$ Output: Noncentrality parameter $\lambda = 20.0$ 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 $(1-\beta \text{ err prob}) = 0.95$ Number of groups = 1 Number of measurements = 4	Repeated measures ANOVA with four linear contrasts, comparing the <i>corrugator</i> muscle activity of four blocks (active viewing, distraction, distancing, suppression). ANOVA is calculated using aov_ez() function of the afexpackage, 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

1c) l		Corr among rep measures = 0.5 Nonsphericity correction $\epsilon = 1$ Output: Noncentrality parameter $\lambda = 17.5169700$ Critical $F = 2.6404222$ Numerator $df = 3.0$ Denominator $df = 252$ Total sample size = 85 Actual power = 0.9509128 F tests - ANOVA: Repeated	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 <i>corrugator</i> 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. ANOVA yields $p < .05$ is interpreted as
arou musi lowe emo strat dista supp	usal (levator scle activity) is er after using an otion regulation stegy (distraction, ancing,	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 $(1-\beta$ err prob) = 0.95 Number of groups = 1 Number of measurements = 4 Corr among rep measures = 0.5 Nonsphericity correction $\epsilon = 1$	with four linear contrasts, comparing the <i>levator</i> 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	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
		Output: Noncentrality parameter $\lambda = 17.5169700$ Critical F = 2.6404222 Numerator df = 3.0 Denominator df = 252 Total sample size = 85 Actual power = 0.9509128	contrasts are calculated using pairs(). Bayes factors are computed for the ANOVA and each contrast using the BayesFactor-package.	> .05. The Bayes factor <i>BF10</i> is reported alongside every <i>p</i> -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 Input: Effect size $f = 0.2041241$ ($\eta_p^2 = 0.04$) (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 $\frac{Output}{E}$: 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 afexpackage, 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 Input:	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 $p > .05$.

3b) Subjective arousal ratings negatively predict subjective values of ER strategies.	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 Output: Noncentrality parameter $\delta = 3.4$ Critical $t = 1.6991270$ Df = 29	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. 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
3c) Corrugator muscle activity negatively predict subjective values of ER strategies.	Total sample size = 34 Actual power = 0.9529571		strength of evidence. Fixed effects yield $p < .05$ are interpreted as subjective values are related to <i>corrugator</i> activity. Subjective values are interpreted as not being related to <i>corrugator</i> activity if $p > .05$. The Bayes factor $BF10$ is reported alongside every p -value to assess the strength of evidence.
3d) <i>Levator</i> muscle activity negatively predict subjective values of ER strategies.			Fixed effects yield $p < .05$ are interpreted as subjective values are related to <i>levator</i> activity. Subjective values are interpreted as not being related to <i>levator</i> 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, <i>corrugator</i> and <i>levator</i> muscle activity.	t tests - Linear multiple regression: Fixed model, single regression coefficient Analysis: A priori: Compute required sample size Input: 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 Output: Noncentrality parameter δ = 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 Input: 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 Output: 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".	 χ² 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. 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 Input: 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 α = 1 Output: 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 Input: 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 $(1-\beta$ err prob) = 0.95 Number of predictors = 2 Output: 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 <i>BF10</i> is reported alongside every <i>p</i> -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"

567 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.

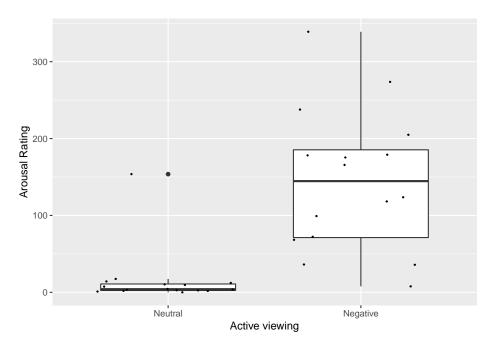


Figure 2. Subjective arousal ratings for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.

 $_{\mbox{\scriptsize 572}}$ Pilot study: Subjective arousal in the conditions "Active viewing - negative",

"Distraction", "Distancing", and "Suppression"

ANOVA:

F.7.F	Effect	df	MSE	F	ges	p.value
575	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:

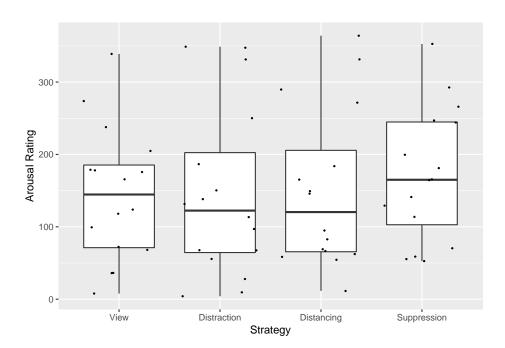


Figure 3. Subjective arousal ratings for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.

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

582

584

585

586

588

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:

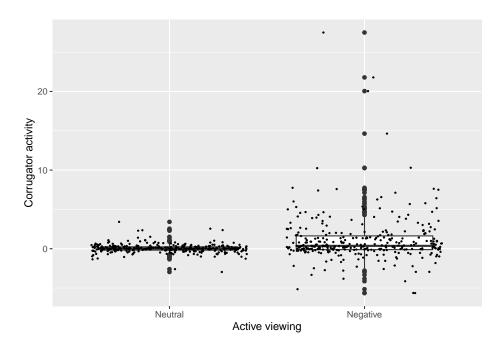


Figure 4. Corrugator activity for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the corrugator activity of a single trial. Bold dots represent outliers.

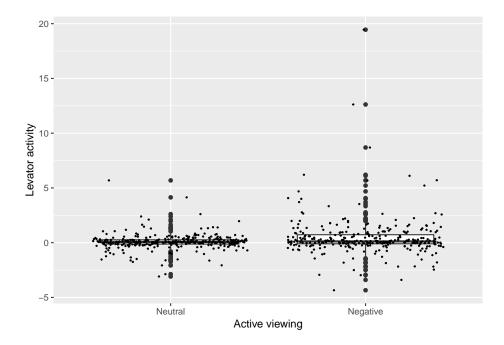


Figure 5. Levator activity for the conditions "Active viewing - neutral" and "Active viewing - negative" visualized as boxplots. Each dot represents the levator activity of a single trial. Bold dots represent outliers.

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

593 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

594

596

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.

597 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:

601

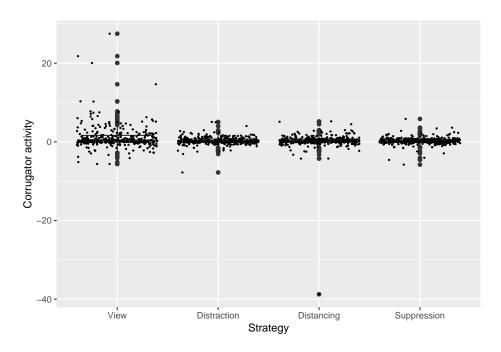


Figure 6. Corrugator activity for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the corrugator activity of a single trial. Bold dots represent outliers.

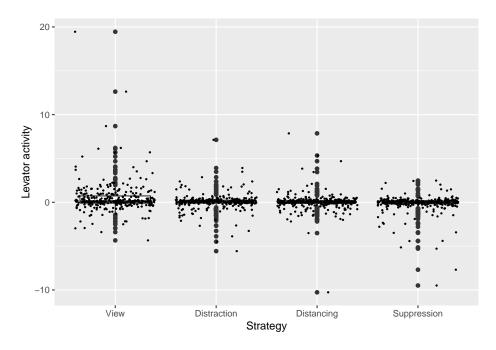


Figure 7. Levator activity for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the levator activity of a single trial. Bold dots represent outliers.

 $_{602}$ Pilot study: Subjective effort in the conditions "Active viewing - negative",

"Distraction", "Distancing", and "Suppression"

ANOVA:

604

607

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

$$BF10 = 7.40$$

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:

608

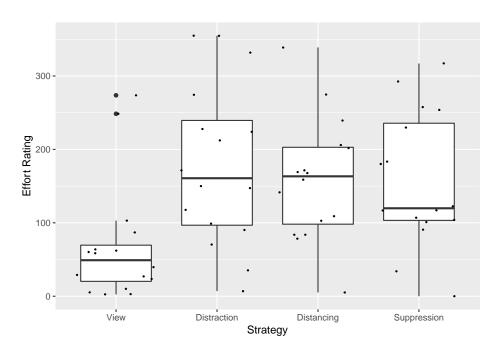


Figure 8. Subjective effort ratings for the conditions "Active viewing - negative", "Distraction", "Distancing", and "Suppression" visualized as boxplots. Each dot represents the effort rating of a single subject. Bold dots represent outliers.