

**Too Beautiful to be Fake: Attractive Faces are Less Likely to be Judged as  
Artificially Generated**

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## Abstract

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24 Abstract abstract abstract.

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## Too Beautiful to be Fake: Attractive Faces are Less Likely to be Judged as Artificially Generated

For the first time in human history, technology has enabled the creation of near-perfect simulations indistinguishable from reality. These artificial, yet realistic constructs permeate all areas of life through immersive works of fiction, deep fakes (real-like images and videos generated by deep learning algorithms), virtual and augmented reality (VR and AR), artificial beings (artificial intelligence “bots” with or without a physical form), fake news and skewed narratives, of which ground truth is often hard to access (Nightingale & Farid, 2022). Such developments not only carry important consequences for the technological and entertainment sectors, but also for security and politics - for instance if used for propaganda and disinformation, recruitment into malevolent organizations, or religious indoctrination (Pantserev, 2020). This issue is central to what has been coined the “post-truth era” (Lewandowsky et al., 2017), in which the distinction (and lack thereof) between authentic and simulated objects will play a critical role.

While not all simulations have achieved perfect realism (e.g., Computer Generated Images - CGI in movies often lack certain key details that makes them visually distinct from real images, McDonnell & Breidt, 2010), it is fair to assume that these technical limitations will become negligible in the near future, particularly in the field of face generation (Moshel et al., 2022; Nightingale & Farid, 2022; Tucciarelli et al., 2020). Such performance, however, leads to a new question: if real and fake stimuli cannot be distinguished based on their objective characteristics, how can we make judgments regarding their nature?

Literature shows that the context surrounding a stimulus often plays an important role in the assessment of its reality (Makowski, 2018; a process henceforth referred to as *simulation monitoring*, Makowski, Sperduti, et al., 2019). With the extensive search and processing of cues within ambiguous stimuli being an increasingly complex and cognitively

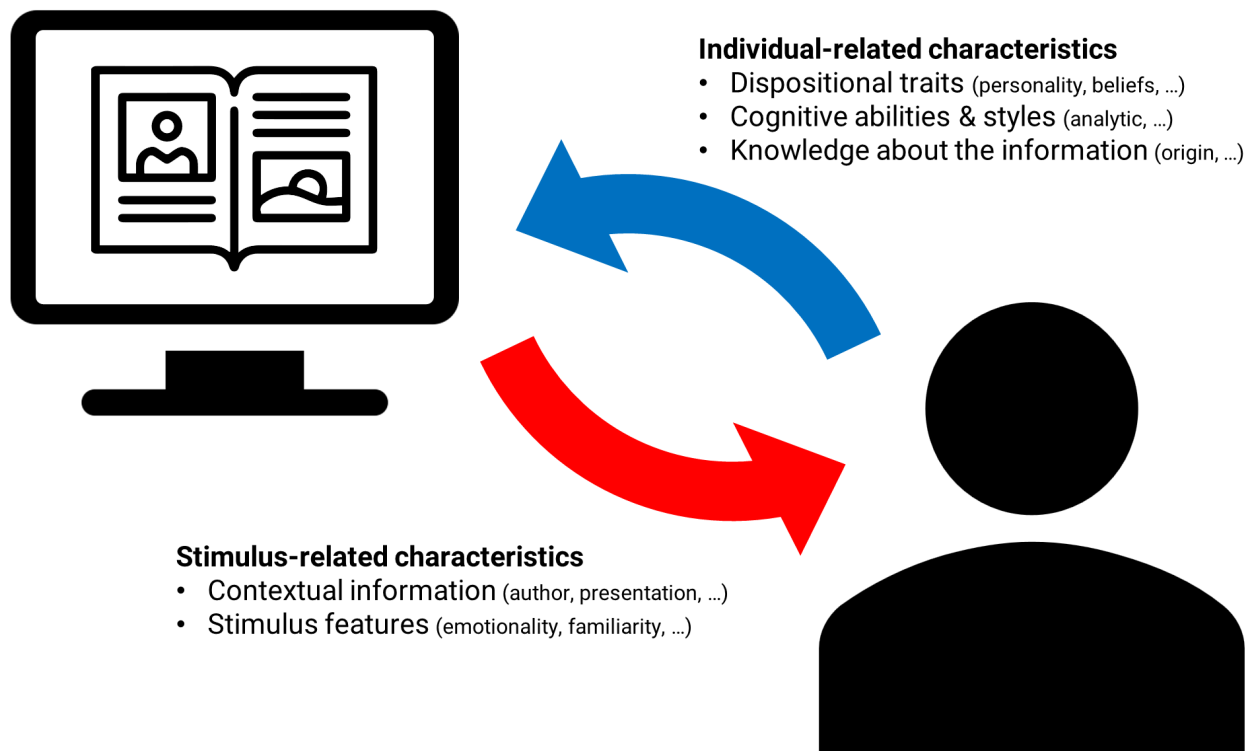
53 effortful strategy (Michael & Sanson, 2021; Susmann et al., 2021), people tend to draw on  
 54 peripheral contextual cues (**Figure 1**), such as the source of the stimulus (e.g., in what  
 55 journal has been an information published), and its credibility, authority and expertise, to  
 56 help facilitate their evaluation (Michael & Sanson, 2021; Petty & Cacioppo, 1986; Susmann  
 57 et al., 2021). However, the atomization and decontextualization of information allowed by  
 58 online social media (where text snippets or video excerpts are mass-shared with little  
 59 context) makes this task increasingly difficult (Berghel, 2018; Y. Chen et al., 2015). In the  
 60 absence of clear contextual information, what drives our beliefs of reality?

## Determinants of Simulation Monitoring

« Is this information *real* or *fake*? »

« *Real* » = genuine, authentic

« *Fake* » = artificial, simulated, deceptive



*Figure 1.* The decision to believe that an ambiguous stimulus (of any form, e.g., images, text, videos, environments, ...) is real or fake depends of individual characteristics (e.g., personality and cognitive styles), stimulus-related features (context, emotionality), and their interaction, which can manifest for instance in our bodily reaction.

Evidence suggests that inter-individual characteristics play a crucial role in simulation monitoring, with factors such as cognitive style, prior beliefs, and personality traits (Bryanov & Vziatysheva, 2021; Ecker et al., 2022; Sindermann et al., 2020). For instance, individuals with stronger analytical reasoning have been found to better discriminate real from fake stimuli (Pehlivanoglu et al., 2021; Pennycook & Rand, 2019), and prior knowledge or beliefs about the stimulus influences one’s perception of it by biasing the attention deployment towards information that is in line with one’s expectations (Britt et al., 2019). Furthermore, dispositional traits, such as high levels of narcissism and low levels of openness and conscientiousness, have been associated with greater susceptibility to fake news (Piksa et al., 2022; Sindermann et al., 2020).

Beyond stimulus- and individual-related characteristics, evidence suggests that the interaction between the two (i.e., the subjective reaction associated with the experience of a given stimulus), contributes to simulation monitoring decisions. For instance, the intensity of experienced emotions have been shown to increase one’s sense of presence - the extent to which one feels like “being there”, as if the object of experience was real - when engaged in a fictional movie or a VR environment (Makowski et al., 2017; Sanchez-Vives & Slater, 2005). Conversely, beliefs that emotional stimuli were fake (e.g., that emotional scenes were not authentic but instead involved actors and movie makeup) were found to result in emotion down-regulation (Makowski, Sperduti, et al., 2019; Sperduti et al., 2017). In line with these findings, studies on susceptibility to fake news have also found heightened stimulus emotionality to be associated with greater belief (Bago et al., 2022; Martel et al., 2020). Additionally, other factors, such as the stimuli’s perceived self-relevance (Goldstein, 2009; Sperduti et al., 2016), as well as familiarity (Begg et al., 1992), could also play a role in guiding our appraisal of a stimulus.

AI-generated images of faces, due to their popularity as a target of CGI technology and to the possibility of experimentally manipulating facial features, are increasingly used

to study face processing in relationship with saliency or emotions, as well as to other important components of face evaluation, such as trustworthiness or attractiveness (Balas & Pacella, 2017; Calbi et al., 2017; Sobieraj & Krämer, 2014; Tsikandilakis et al., 2019). Interestingly, artificially created faces rated as more attractive (by an independent group of raters) were perceived as less real (Tucciarelli et al., 2020). Conversely, Liefoghe et al. (2022) reports that attractiveness ratings were significantly lower when participants who were told that the faces were AI-generated were compared to those who had no prior knowledge. Whereas this line of evidence suggests that reality beliefs have an effect on face attractiveness ratings, the opposite question - whether attractiveness could drive simulation monitoring - has received little attention to date.

This study primarily aims at exploring the effect of facial attractiveness on simulation monitoring, i.e., on the beliefs that an image is real or artificially generated. Based on the embodied reality theory (outlined in Makowski, 2018; Makowski, Sperduti, et al., 2019), which suggests that salient and emotional stimuli are perceived to be more real, we hypothesize a quadratic relationship between perceived realness and attractiveness: faces rated as highly attractive or unattractive will more likely be believed to be real. We expect a similar relationship with trustworthiness ratings given its well-established link with attractiveness (Bartosik et al., 2021; Garrido & Prada, 2017; Liefoghe et al., 2022; Little et al., 2011), and a positive relationship with familiarity (as more familiar faces would appear as more salient, self-relevant and anchored in reality). Additionally, we will further explore the link shared by dispositional traits, such as personality and attitude towards AI, with simulation monitoring tendencies. This study aims beyond the investigation of the discriminative accuracy between “true” photos and “true” artificially-generated images, focusing on the beliefs that a stimulus is real or fake, independently of its true nature.

## Methods

In line with open-science standards, all the material (stimuli generation code, experiment code, raw data, analysis script with complementary figures and analyses, preregistration, etc.) is available at <https://github.com/RealityBending/FakeFace>.

**Procedure.** In the first part of the study, participants answered a series of personality questionnaires, including the *Mini-IPIP6* (24 items, Sibley et al., 2011) measuring 6 personality traits, the *SIAS-6* and the *SPS-6* (6 items each, Peters et al., 2012) assessing social anxiety levels, the *FFNI-BF* (30 items, Jauk et al., 2022) measuring 9 facets of narcissism; the *R-GPTS* (18 items, Freeman et al., 2021) measuring 2 dimensions related to paranoid thinking; and the *IUS-12* (12 items, Carleton et al., 2007) measuring intolerance to uncertainty. Self-rated attractiveness was also assessed using 2 items - one measuring general attractiveness (“How attractive would you say you are?”, Marcinkowska et al. (2021)) and the other measuring physical attractiveness (“How would you rate your own physical attractiveness relative to the average”, Spielmann et al. (2020)). Finally, we devised 5 items pertaining to expectations about AI-generated image technology (“I think current Artificial Intelligence algorithms can generate very realistic images”). To lower their saliency and the possibility of it priming the subjects about the task, we mixed these items with 5 items from the general attitudes towards AI scale (*GAAIS*, Schepman & Rodway, 2020). This scale was presented after the social anxiety questionnaires. 3 attention check questions were also embedded in the surveys.

In the second part of this study, 109 images of neutral-expression faces from the validated American Multiracial Face Database (AMFD, (J. M. Chen et al., 2021)) were presented to the participants for 500ms each, in a randomized order, following a fixation cross display (750 ms). After each stimulus presentation, ratings of *Trustworthiness* (“I find this person trustworthy”) and *Familiarity* (“This person reminds me of someone I know”) were collected using visual analog scales. Notably, as facial attractiveness is a



multidimensional construct, encompassing evolutionary, sociocultural, biological as well as cognitive aspects (Han et al., 2018; Rhodes et al., 2006), we assessed attractiveness using 2 visual analog scales, measuring general *Attractiveness* (“I find this person attractive”) and physical *Beauty* (“This face is good-looking”).

In the last part of the study, participants were informed that about half of the images previously seen were AI-generated (the instructions used a cover story which mentioned that the research was aimed at validating a new face generation algorithm). The same set of stimuli was displayed again for 500 ms in a new randomized order. This time, after each display, participants were asked to express their belief regarding the nature of the stimulus using a visual analog scale (with *Fake* and *Real* as the two extremes). The study was implemented using *jsPsych* (De Leeuw, 2015), and the exact instructions are available in the experiment code.

**Participants.** One hundred and fifty participants were recruited via *Prolific*, a crowd-sourcing platform recognized for providing high quality data (Peer et al., 2022). The only inclusion criterion was a fluent proficiency in English to ensure that the experiment instructions would be well-understood. Participants were incentivised with a reward of about £7.5 for completing the study, which took about 45 minutes to finish. Demographic variables (age, gender, sexual orientation, education and ethnicity) were self-reported on a voluntary basis.

We excluded 5 participants that either failed 2 ( $\geq 66\%$ ) or more attention check questions, took an implausibly short time to finish the questionnaires or had incomplete responses. The final sample included 145 participants (Mean age = 28.3, SD = 9.0, range: [19, 66]; Sex: 48.3% females, 51.0% males, 0.7% others).

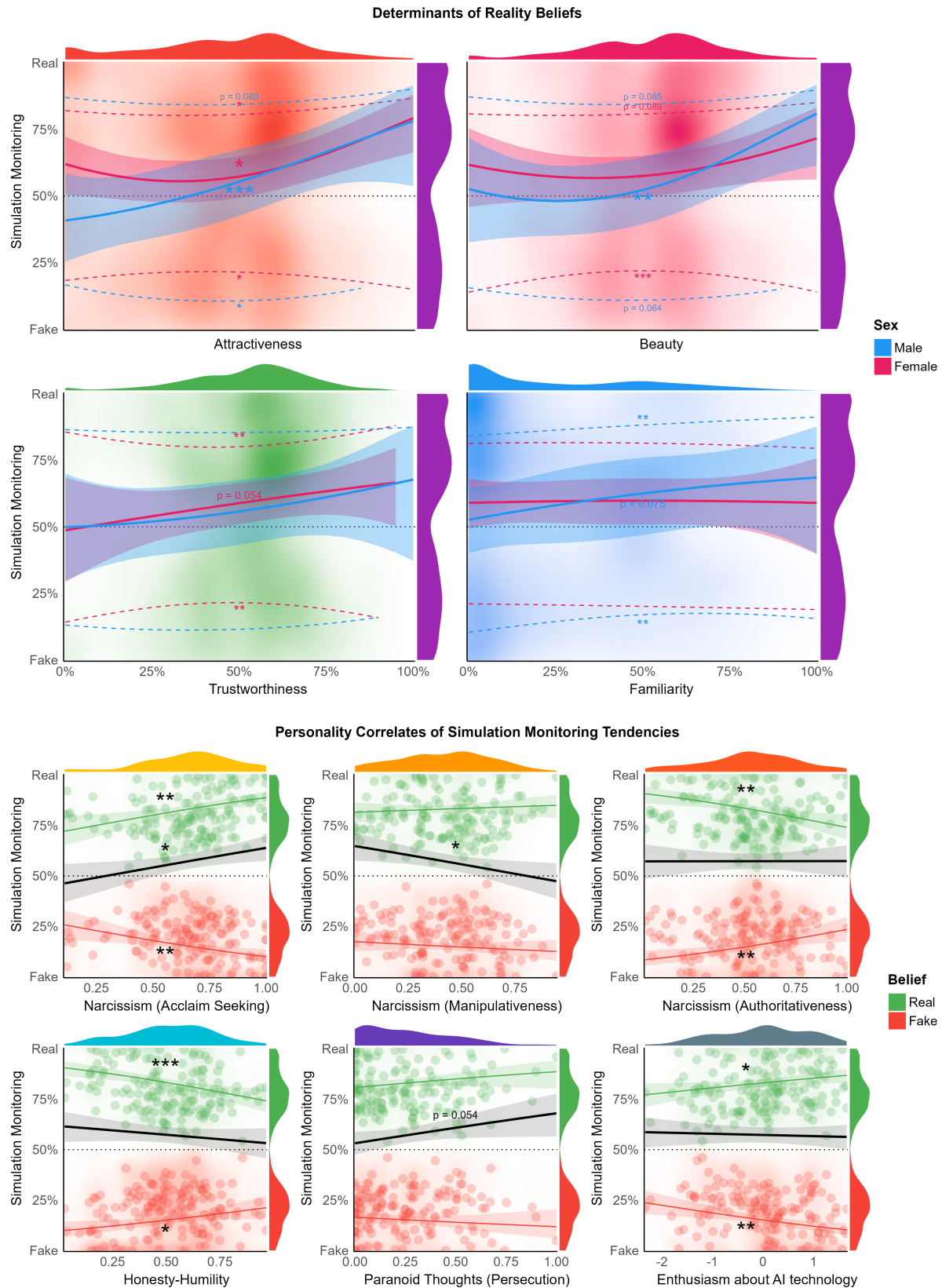
**Data Analysis.** The real-fake ratings (measured originally on a [-1, 1] analog scale) were converted into two scores, corresponding to two conceptually distinct mechanisms: the dichotomous *belief* (real or fake, derived based on the sign of the rating) and the

163 *confidence* (the rating’s absolute value) associated with that belief. The former was  
 164 analyzed using logistic mixed models (with the participants and images entered as random  
 165 factors), which modelled the probability of assigning a face to the real ( $\geq 0$ ) as opposed  
 166 to fake ( $< 0$ ). The latter, as well as the other face ratings (attractiveness, beauty,  
 167 trustworthiness and familiarity), was modelled using beta regressions (suited for outcome  
 168 variables expressed in percentages).

169 We started by investigating the effect of the procedure and instructions to check  
 170 whether the stimuli (which were all images of real faces) were judged as fake in sufficient  
 171 proportion to warrant their analysis. Additionally, we assessed the effect of the re-exposure  
 172 delay, i.e., the time between the first presentation of the image (corresponding to the face  
 173 ratings) and the second presentation (for the real-fake rating).

174 The determinants of reality beliefs were modelled separately for attractiveness,  
 175 beauty, trustworthiness, and familiarity, using second order raw polynomials coefficients to  
 176 allow for possible quadratic relationships (**Figure 2**). Aside from attractiveness  
 177 (conceptualized as a general construct), models for beauty, trustworthiness and familiarity  
 178 were adjusted for the the two remaining variables *mutatis mutandis*. We took into account  
 179 the gender of participants and stimuli by retaining the stimuli that were aligned with the  
 180 participants’ sexual preference (e.g., female faces for homosexual females, male faces for  
 181 heterosexual females, and both for bisexual participants), and modeling the interaction  
 182 with the participants’ gender. For the attractiveness and beauty models, we then added  
 183 the interaction with the reported self-attractiveness (the average of the two questions  
 184 pertaining to it) to investigate its potential modulatory effect. Finally, we investigated the  
 185 inter-individual correlates of simulation monitoring with similar models (but this time, for  
 186 all items regardless of the participant’s gender or sexual orientation) for each questionnaire,  
 187 with all of the subscales as orthogonal predictors.

188 The analysis was carried out using *R 4.2* (R Core Team, 2022), the *tidyverse*



*Figure 2.* Top part shows the effect of face ratings on 1) the probability of judging a face as real vs. fake (solid line) and 2) on the confidence associated with that judgment (dashed lines) depending on the sex. Bottom part shows the effect of personality traits on the belief (black line) and the confidence associated with it (colored lines). The points are the average per participant confidence for both types of judgments. Stars indicate significance ( $p < .001^{***}$ ,  $p < .01^{**}$ ,  $p < .05^{*}$ )

(Wickham et al., 2019), and the *easystats* collection of packages (Lüdtke et al., 2021, 2019, 2020; Makowski et al., 2020; Makowski, Ben-Shachar, et al., 2019). As all the details, scripts and complimentary analyses are open-access, we will focus in the manuscript on findings that are highly statistically significant ( $p < .01$ ).

## Results

**Manipulation Check.** Only one image file yielded a strong simulation monitoring bias ( $> 85\%$ ), being classified as fake by 88.7% of participants. This image was removed from further analysis, leaving 108 trials per participant. On average, across participants, 44% of images (95%~CI [0.12, 0.64]) were judged as fake and 56% of images (95%~CI [0.36, 0.84]) as real. An intercept-only model with the participants and images as random factors showed that the Intraclass Correlation Coefficient (ICC), which can be interpreted as the proportion of variance explained by the random factors, was of 9.0% for the participants and 9.6% for the stimuli.

While the delay of stimulus re-exposure stimulus did not have a significant effect on participants' beliefs of reality ( $OR = 1.00$ , 95%  $CI = [0.99, 1.00]$ ), judgement confidence was found to be negatively associated with re-exposure delay when the faces were judged as real ( $\beta = -0.006$ , 95%  $CI = [-0.1, 0.002]$ ,  $p = .004$ )

**Determinants of Simulation Monitoring.** Attractiveness had a significant positive and linear relationship ( $R^2_{marginal} = 2.0\%$ ) with the belief that a stimulus was real ( $\beta_{poly1} = 16.57$ , 95%  $CI = [7.33, 25.82]$ ,  $z = 3.51$ ,  $p < .001$ ) for males, and a quadratic relationship for females ( $\beta_{poly2} = 7.82$ , 95%  $CI = [1.81, 13.84]$ ,  $z = 2.55$ ,  $p = .011$ ), with both non-attractive and attractive faces being judged as more real. Attractiveness was also found to have a significant positive and quadratic relationship with confidence in judging faces both as real ( $\beta_{poly2} = 4.30$ , 95%  $CI = [0.97, 7.64]$ ,  $z = 2.53$ ,  $p = .011$ ) and as fake ( $\beta_{poly2} = 5.23$ , 95%  $CI = [0.86, 9.60]$ ,  $z = 2.35$ ,  $p = .019$ ) for females. For males, however, a significant negative and quadratic relationship was found between attractiveness ratings

and belief confidence only for faces judged as fake ( $\beta_{poly2} = -9.92$ ,  
 $95\% CI = [-18.99, -0.86]$ ,  $z = -2.15$ ,  $p = .032$ ). There was no interaction with reported  
self-attractiveness.

Beauty, adjusted for trustworthiness and familiarity, had a significant positive and  
linear relationship ( $R^2_{marginal} = 2.0\%$ ) with the belief that a stimulus was real  
( $\beta_{poly1} = 11.82$ ,  $95\% CI = [4.28, 20.21]$ ,  $z = 2.76$ ,  $p = .006$ ) for males only. No effect on  
confidence was found, aside from a quadratic relationship for females for faces judged as  
fake, suggesting that non-beautiful and highly beautiful faces were rated as fake with more  
confidence than average faces ( $\beta_{poly2} = 7.84$ ,  $95\% CI = [3.39, 12.29]$ ,  $z = 3.46$ ,  $p < .001$ ).  
There was no interaction with reported self-attractiveness.

Trustworthiness, adjusted for beauty and familiarity, had a predominantly positive  
and linear relationship ( $R^2_{marginal} = 2.0\%$ ) with the belief that a stimulus was real  
( $\beta_{poly1} = 6.44$ ,  $95\% CI = [-0.11, 13.00]$ ,  $z = 1.93$ ,  $p = .0054$ ) for females only. No effect on  
confidence was found for males, whereas a quadratic relationship was found for females for  
both faces judged as real ( $\beta_{poly2} = 6.14$ ,  $95\% CI = [2.13, 10.14]$ ,  $z = 3.00$ ,  $p = .003$ ) as well  
as fake ( $\beta_{poly2} = 6.12$ ,  $95\% CI = [1.49, 10.75]$ ,  $z = 2.59$ ,  $p = .001$ ), suggesting that  
non-trustworthy and highly trustworthy faces were rated with more confidence than  
average faces.

We did not find any significant relationships for familiarity adjusted for beauty and  
trustworthiness ( $R^2_{marginal} = 2.0\%$ ). However, a significant positive and linear relationship  
was found between familiarity and the confidence judgements of rating faces as real  
( $\beta_{poly1} = 9.98$ ,  $95\% CI = [3.83, 16.13]$ ,  $z = 3.18$ ,  $p = .001$ ) whereas a negative linear  
relationship was found with those judged as fake ( $\beta_{poly1} = -12.41$ ,  
 $95\% CI = [-20.27, -4.54]$ ,  $z = -3.09$ ,  $p = .002$ ) for males only. This hence suggests that  
males more confidently judge faces as real with when they are familiar, and as fake when  
they are unfamiliar.

**Inter-Individual Correlates of Simulation Monitoring.**

The models including the personality traits suggested that *Honesty-Humility* had a significant negative relationship with the confidence associated with real as well as fake judgements ( $\beta_{real} = -1.62$ , 95%  $CI = [-2.55, -0.70]$ ,  $z = -3.43$ ,  $p < .001$ ;  $\beta_{fake} = -1.16$ , 95%  $CI = [-2.09, -0.23]$ ,  $z = -2.45$ ,  $p = 0.014$ ).

Significant positive associations were found between the probability of judging faces as real and dimensions of narcissism such as *Acclaim Seeking* ( $\beta = 2.24$ , 95%  $CI = [1.17, 4.27]$ ,  $z = 2.44$ ,  $p = .015$ ), and *Manipulativeness* ( $\beta = 0.47$ , 95%  $CI = [0.25, 0.87]$ ,  $z = -2.4$ ,  $p = 0.017$ ). Confidence judgements also shared significant links with narcissism through various facets, such as a positive relationship between the confidence for both real and fake judgments with *Acclaim Seeking* ( $\beta_{real} = 1.65$ , 95%  $CI = [0.59, 2.70]$ ,  $z = 3.07$ ,  $p = .002$ ;  $\beta_{fake} = 1.62$ , 95%  $CI = [0.56, 2.68]$ ,  $z = 3.00$ ,  $p = .003$ ), and a negative relationship with *Authoritativeness* ( $\beta_{real} = -1.57$ , 95%  $CI = [-2.58, -0.57]$ ,  $z = -3.08$ ,  $p = .002$ ;  $\beta_{fake} = -1.49$ , 95%  $CI = [-2.50, -0.48]$ ,  $z = -2.89$ ,  $p = .004$ ).

A positive trend was found in the relationship between the *Persecutory Ideation* dimension of paranoid thinking and the belief that the faces were real ( $\beta = 1.87$ , 95%  $CI = [0.99, 3.54]$ ,  $z = 1.93$ ,  $p = .054$ ).

The *Prospective Anxiety* aspect of intolerance to uncertainty shared a negative trend in its association with confidence ratings ( $\beta_{real} = 1.43$ , 95%  $CI = [0.10, 2.76]$ ,  $z = 2.10$ ,  $p = .036$ ;  $\beta_{fake} = -0.91$ , 95%  $CI = [-1.93, 0.11]$ ,  $z = -1.75$ ,  $p = .081$ ). No significant effect was found for social anxiety.

Questions pertaining to the attitude towards AI were reduced to 3 dimensions through factor analysis, labelled AI-Enthusiasm (loaded by items expressing interest and excitement in AI development and applications), AI-Realness (loaded by items expressing

positive opinions on the ability of AI to create realistic material), and AI-Danger (loaded by items expressing concerns on the unethical misuse of AI technology). However, only AI-Enthusiasm displayed a significant positive relationship with the confidence in both real and fake judgments ( $\beta_{real} = 0.21$ , 95%  $CI = [0.02, 0.40]$ ,  $z = 2.20$ ,  $p = .028$ ;  $\beta_{fake} = 0.31$ , 95%  $CI = [0.12, 0.50]$ ,  $z = -8.90$ ,  $p < 0.001$ ).

## Discussion

This study aimed at investigating the effect of facial ratings (attractiveness, beauty, trustworthiness and familiarity) on simulation monitoring, i.e., on the belief that a stimulus was artificially generated. The most striking result, in our opinion, is that despite all the stimuli being real faces from the same database, all participants believed (to high degrees of confidence) that a significant proportion of them were fake. This finding can be seen as a testimony not only to the effectiveness of our instructions, nut also to the current expectations regarding CGI technology in the population, as well as to the volatility of our sense of reality. It underlines the strong impact of prior expectations and information on reality beliefs. In fact, stimuli-related and participant-related characteristics accounted together for less than 20% of the beliefs variance, suggesting that a large part of it is associated with other subjective processes.

Although attractiveness did not seem to be the primary drive underlying simulation monitoring of face images, we do nonetheless report significant associations, with a different pattern observed depending on the participant’s gender. The quadratic relationship found for female participants is aligned with our hypothesis that salient faces (i.e., rated as very attractive or very unattractive) are judged to be more real. The fact that this effect did not reach significance for beauty underlines that attractiveness judgment, and its role in simulation monitoring, is a multidimensional construct that cannot be reduced to physical facial attractiveness, in particular for women (Buunk et al., 2002; Qi & Ying, 2022). In fact, when the given stimulus was rated very high or low on beauty, female participants

were more confident in their judgement of fake faces only, suggesting that physical beauty only partially explains the role attractiveness plays in simulation monitoring decisions.

Interestingly, we found a significant positive linear relationship in male participants for both attractiveness and beauty on simulation monitoring that we could interpret under an evolutionary lens. Specifically, males purportedly place more emphasis on facial attractiveness as a sign of reproductive potential, as compared with females, who tend to value characteristics signaling resource acquisition capabilities (Buunk et al., 2002; Fink et al., 2006; Qi & Ying, 2022). It is thus possible that the evolutionary weight associated with attractiveness skewed men’s perceived saliency towards attractive faces, rendering them significantly more salient than unattractive faces and in turn distorted the relationship with simulation monitoring. However, future studies should test this saliency-based hypothesis by measuring constructs closer to salience and its effects, for instance using neuroimaging (Indovina & Macaluso, 2007; Lou et al., 2015) or physiological markers (e.g., heart rate deceleration, Skora et al., 2022).

Our results found a positive linear trend between trustworthiness and simulation monitoring for females only. Given prior evidence that faces presented as computer-generated were rated less trustworthy (Balas & Pacella, 2017; Hoogers, 2021; Liefoghe et al., 2022), we expected such a linear association to be more clearly present for both genders. One of the underlying mechanisms that possibly contributed to this dimorphism could be the increased risk-taking aversion reported in females (explained evolutionarily as a compromise to their reproductive potential, Van Den Akker et al., 2020), to which perceived facial trustworthiness relates (Hou & Liu, 2019). However, if that was the case, faces judged as highly untrustworthy should have appeared as even more salient (representing an evolutionary threat), and hence be judged as more real, leading to a quadratic relationship between trustworthiness and simulation monitoring instead. Further studies are needed to investigate the causes of the increased simulation monitoring



sensitivity to trustworthiness in females.

Contrary to our hypothesis, familiarity was not found to be significantly related to simulation monitoring decisions. Interestingly, there were significant linear relationships between familiarity and confidence judgements for males only, where familiarity increased the confidence of reality beliefs. This could be taken into perspective with previous studies, which report females to be relatively superior at face recall and recognition (Lewin & Herlitz, 2002; Mishra et al., 2019), displaying higher levels of facial memory (Sommer et al., 2013) and exhibiting greater metacognitive awareness (Kaplan, 2012). Hence, it is possible that the confidence in evaluating faces are less malleable to external factors such as face familiarity for females. Indeed, metacognitive awareness has been associated with overconfidence in making judgements (Klein et al., 2018; Mata et al., 2013). However, it should be noted that this has not been corroborated by recent research, which found males were significantly overconfident in face recall, regardless of their performance accuracy (Bailey, 2021; Herbst, 2020). Furthermore, the present study's distribution of familiarity ratings was strongly skewed, and only a low number of pictures was rated as highly familiar. As such, future studies should clarify this point by experimentally manipulating familiarity, for instance by modulating the amount of exposure to items before querying the simulation monitoring judgments.

Regarding the role of inter-individual characteristics in simulation monitoring tendencies, we found that participants with higher scores of honesty-humility, a trait related to an increased risk perception and aversion (Levidi et al., 2022; Weller & Thulin, 2012), displayed a lower confidence in their simulation monitoring judgments. Notably, greater narcissistic tendencies in dimensions such as acclaim seeking and manipulateness were found to be associated with a higher number of faces judged as real. Indeed, this finding is in line with recent research suggesting that narcissists are more susceptible to fake news (Piksa et al., 2022; Sindermann et al., 2020), possibly because they were less

likely to engage in cognitive strategies such as reflective thinking (Ahadzadeh et al., 2021; Littrell et al., 2020) and are more vigilant when attending to external stimuli (Carolan, 2017; Eddy, 2021; Grapsas et al., 2020). As such, individuals with greater narcissistic tendencies could have perceived the face images as more salient and engaged in less analytical reasoning, leading to their increased perceptions of the faces being real.

Moreover, we put the significant positive links between narcissistic acclaim seeking and confidence judgements in perspective with the negative correlation between honesty-humility and narcissism (Hodson et al., 2018), thus confirming the evidence regarding the relationship between narcissistic grandiosity and over-confidence in decision-making (Brunell & Buelow, 2017; Campbell et al., 2004; Chatterjee & Pollock, 2017; O'Reilly & Hall, 2021). Although an inverse effect was found for the authoritativeness facet of narcissism, we interpret this correlation as related to a higher response assertiveness. Taken together, these results suggest that participants with low humility and high recognition desires are more confident in their judgment regarding the real or fake nature of ambiguous stimuli. Participants with opposite traits might perceive a higher risk in the decision-making process and its potential consequences (e.g., being seen as bad at the task at hand), resulting in more conservative confidence ratings.

Despite the ubiquity of AI, the literature pertaining to the influence of people's AI attitudes on simulation monitoring is scarce. Contrary to our expectations, we did not find evidence for the role of participants' expectations regarding the capabilities of AI technology (in terms of the realism of its productions). Instead, we found only one's enthusiasm about AI technology to be related to an increased confidence in simulation monitoring ratings. This could potentially be because participants with a highly positive attitude towards AI perceive themselves as having greater knowledge about AI and its capabilities (Said et al., 2022), hence permitting themselves to be more confident in their simulation monitoring decisions. This result is in line with reports that AI attitudes

interacts with people’s perceived self-knowledge to influence their perception of the opportunities and risks accorded by AI applications (@ Said et al., 2022).

Our findings suggest - though with weak significance - a positive link between paranoid ideation and the tendency to believe that the stimuli were real. Given previous reports that people with higher levels of paranoia are more sensitive to cues of social threat (Fornells-Ambrojo et al., 2015; Freeman et al., 2003), it could thus be possible that stronger paranoid traits was associated with a greater perception of faces as salient and emotional, hence increasing their beliefs of its realness. This hypothesis, if confirmed by future studies, would be in line with previous findings that persecutory delusions are predicted by a greater sense of presence in a VR environment populated with virtual characters (Freeman et al., 2005).

In contrast, intolerance to uncertainty, a trait related to enhanced threat appraisal, was found to be negatively associated with the confidence in real judgements (J. T.-H. Chen & Lovibond, 2016; Jensen et al., 2014). This is consistent with previous reports that uncertainty intolerance correlates with under-confidence in decision-making in ambiguous scenarios (Jensen et al., 2014; Wei, 2021). **WHY IS THIS CONTRADICTORY? the next part seems to be about emotion regulation. While interesting, it’s a bit of a non-sequitur no? And the previous explanation could almost be enough.**

While the findings seem contradictory, they could possibly be interpreted through the lens of fictional reappraisal, an emotion regulation strategy in which the belief that an emotional stimulus is not real is used as a means to down-regulate the associated emotional reactions (Singer-Landau & Meiran, 2021; Sperduti et al., 2016, 2017). In fact, previous studies report that individuals with greater paranoia and anxiety engaged in more affective processing of external ambiguous stimuli (Fornells-Ambrojo et al., 2015). This is also in line with prior findings which suggest that the likelihood of reality beliefs decreased when pictures were emotional, as compared to when they were neutral (Makowski, Sperduti, et

al., 2019). As such, it is possible that individuals high in such anxiety-related traits are more sensitive to social cues and perceived the ambiguous face images with greater emotionality, therefore having to engage in emotional regulation strategies that attenuated their confidence in real judgements. Future studies should clarify the role of emotions, particularly those related to anxiety, and attempt at delineating processes related to emotion perception and reaction from emotion regulation.

On a methodological level, although the order of presentation of the facial images was randomized to reduce effects of adaptation, participants were more confident in their judgements for faces perceived as real following a shorter re-exposure delay. This may be because shorter re-exposure delays led to the faces being better remembered and appearing more familiar than faces that were displayed after a longer delay, thereby triggering self-referential and autobiographical memory processing during the repeated display (Abraham & Von Cramon, 2009; Gobbini et al., 2013; Taylor et al., 2009). Indeed, this finding is in line with studies that show an up-regulation of emotions when fictional stimuli were associated with familiarity (Makowski et al., 2017; Sperduti et al., 2016), thus biasing its salience and perceived realness. If that was the case, we would expect shorter re-exposure delays to impact the decision bias as well towards reality, rather than simply the confidence. Future studies should further investigate the modulatory effects of types and degrees of familiarity on perceived realness judgements.

In conclusion, understanding the factors driving our beliefs that ambiguous stimuli are real or not is becoming increasingly pertinent, given the current expansion on technologies related to virtual and augmented realities. The aim of the present study was to examine whether facial attractiveness significantly influences our simulation monitoring decisions. We found faces rated as attractive to be perceived as more real, with a possible sexual dimorphism affecting the shape of the relationship. We also found that inter-individual traits, such as narcissistic acclaim-seeking and manipulativeness, as well as

422 persecutory ideation, were related to a systematic bias towards beliefs that the stimuli were  
423 real or fake. Limitations include [TO ADD].

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