Learning about threat from friends and strangers is equally effective:

an fMRI study on observational fear conditioning

Anna M. Kaźmierowska[1, 2, †], Michał Szczepanik[1, 2, †],

Marek Wypych[1], Dawid Droździel[1], Artur Marchewka[1], Jarosław M. Michałowski[3],

Andreas Olsson[4], Ewelina Knapska[2]

[1] Laboratory of Brain Imaging, Nencki Institute of Experimental Biology of Polish Academy of

Sciences, 3 Pasteur Str., 02-093 Warsaw, Poland

[2] Laboratory of Emotions Neurobiology, BRAINCITY - Centre of Excellence for Neural Plasticity

and Brain Disorders, Nencki Institute of Experimental Biology of Polish Academy of Sciences,

3 Pasteur Str., 02-093 Warsaw, Poland

[3] Laboratory of Affective Neuroscience, SWPS University of Social Sciences and Humanities,

10 Kutrzeby Str., 61-719 Poznań, Poland

[4] Department of Clinical Neuroscience, Division of Psychology, Karolinska Institutet,

Stockholm, Sweden

† These authors contributed equally to this work

Abstract

Humans often benefit from social cues when learning about the world. For instance,

learning about threats from others can save the individual from dangerous first-hand

experiences. Familiarity is believed to increase the effectiveness of social learning, but it is not

clear whether it plays a role in learning about threats. Using functional magnetic resonance

imaging, we undertook a naturalistic approach and investigated whether there was a difference

between observational fear learning from friends and strangers. Participants (observers)

witnessed either their friends or strangers (demonstrators) receiving aversive (shock) stimuli

paired with colored squares (observational learning stage). Subsequently, participants

watched the same squares, but without receiving any shocks (direct-expression stage). We

observed a similar pattern of brain activity in both groups of observers. Regions related to

threat responses (amygdala, anterior insula, anterior cingulate cortex) and social perception

(fusiform gyrus, posterior superior temporal sulcus) were activated during the observational

phase, reflecting the fear contagion process. The anterior insula and anterior cingulate cortex

were also activated during the subsequent stage, indicating the expression of learned threat.

Because there were no differences between participants observing friends and strangers, we

argue that social threat learning is independent of the level of familiarity with the demonstrator.

Introduction

Detecting potential threats in the environment is vital for survival. To learn about danger, social species, including humans, use direct and vicarious cues (Olsson et al., 2020). Using vicarious cues is beneficial from the evolutionary perspective as it might eliminate the risk of first-hand encounters with a threat. One of the mechanisms underlying vicarious learning about dangers is emotional contagion, an automatic, unconscious tuning into the affective state of others. Due to such emotional mimicry, the observers can 'feel' the threat while enjoying relative safety themselves. Through exposure to the observed danger, they can learn about the potential risks within the environment.

The strength of the relationship between the 'demonstrator' and the 'observer' is believed to modulate emotional contagion (Preston & de Waal, 2002). In general, familiarity improves the social transfer of emotions (Bruder et al., 2012; de Vignemont & Singer, 2006; Gonzalez-Liencres et al., 2014; Hatfield et al., 2014). Importantly, however, previous research suggests that a close social relationship does not facilitate the emotional convergence of fear (Bruder et al., 2012). The only previous report comparing the neural correlates of a threat (an electric shock) to self, a friend, and a stranger gave no conclusive results (Beckes et al., 2013). Thus, it is unclear whether there is a difference between observational fear learning from friends and strangers.

To address this open question, we used a modified observational fear learning protocol (Haaker, Golkar, et al., 2017) improved in terms of ecological validity (Szczepanik, Kaźmierowska, et al., 2020). We directly tested how the demonstrator's familiarity influenced both psychophysiological and brain correlates of observational fear learning in the observer. The experimental design (Figure 1) involved two main stages. During the observational learning stage, two groups of participants (observers) witnessed their friends or strangers (demonstrators) undergoing a differential fear conditioning procedure. In this task, uncomfortable but not painful electric shocks were applied to the demonstrator's forearm during the presentation of emotionally neutral visual stimuli. Subsequently, during the direct-

expression stage, we tested the observers' brain responses to previously watched stimuli. No

electrical stimulation was applied to the observers throughout the whole experiment. Thus, the

observers derived the knowledge about the aversive value of stimuli only from social learning.

As we wanted our experimental situation to be as naturalistic as possible, no apparent ingroup-

outgroup division was introduced (we did not emphasize the demonstrator's affiliation). Also,

unlike most previous studies that have used an actor serving as the demonstrator, real-time

videos were transmitted to the observers in the friend-observation condition. In the stranger-

observation condition, we used recordings from the friend-observation condition. Developing

such an ecologically valid paradigm enabled investigating social learning of fear in a setting

relevant for naturally occurring behaviors.

Figure 1

Experimental Design

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Note. (A) Depending on the group, observers arrived at the laboratory with their friend (upper

panel, the 'friend' group, orange) or alone (lower panel, the 'stranger' group, blue). (B) During

the observational learning stage, observers were lying in the fMRI scanner and watching the

demonstrator (a friend or a stranger) performing the differential fear conditioning task. In the

task, one stimulus (conditioned stimulus +; CS+) was sometimes paired with an uncomfortable

electric shock (unconditioned stimulus, US), while the other one (conditioned stimulus -; CS-)

was always safe. (C) During the direct-expression stage, observers remained in the scanner

and were told that they would perform the same task as previously watched. They watched an

identical set of visual stimuli, but no electrical stimulation was applied.

105   **Results**

Behavioral Results

Stimulus Contingency Awareness

After completing both stages of the experiment, the observers answered a set of

questions about the relationship between the electrical stimulation and the visual cues in the

observational learning stage (see Assessment of Stimulus Contingency Knowledge). 35 out of

48 participants in the friend group and 35 out of 44 participants in the stranger group correctly

identified the CS+/US contingency. The proportion of contingency-aware and non-aware

participants was not significantly different between groups (chi-squared = 0.554, df = 1, p =

0.46). We included only the contingency-aware participants in further analyses (Szczepanik,

Kaźmierowska, et al., 2020).

Evaluation of the Demonstrator's Expression (the Observational US)

Next, the observers evaluated the demonstrator's expression (the observational US)

by answering a set of questions scored on 0 - 9 scales. The questions concerned the

demonstrator's reactions to electrical stimulation (how much discomfort they experienced, how

expressive they were, and how natural their reactions were) and their attitude (how much

empathy they felt for the demonstrator). Additionally, in the pairs of friends, the observers

assessed how well they could identify with the demonstrator. Median ratings in both the friend

and stranger groups were between 5 and 7 (see Figure 2). Additionally, we asked the

observers to rate the degree of unpleasantness attributed to the demonstrators on a 1 - 5 scale

(from very unpleasant to very pleasant). The median ratings were between 1 (very unpleasant)

and 2 (rather unpleasant). The Wilcoxon-Mann-Whitney test showed no significant differences

between the groups in any category. See Table S1 in the Appendix for detailed statistics.

Figure 2

Evaluation of the Demonstrator's Expression (the Observational US)

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134 *Note.* Error bars extend to data points placed no further than 1.5*IQR (interquartile range)

135 beyond the 1st quartile and above the 3rd quartile.

137 ***Empathy and Anxiety - Questionnaire Results***

138   The state anxiety was tested twice, before and after the experiment. The results were

analyzed for the contingency-aware observers, who were included into the fMRI analysis.

Repeated-measures ANOVA and analogous Bayesian repeated measures ANOVA with the

group (friend, stranger) as a between-subject factor and measurement (before, after) as a

within-subject factor showed no significant differences: main effect of the group: $F(1, 67) =$

0.76, $\eta_g^2 = .009$, $p = .39$, $BF_{incl} = 0.35$; main effect of the measurement: $F(1,67) = 1.92$, $\eta_g^2 =$

.007, p = .17, BF_{incl} = 0.36; group × measurement interaction $F(1,67)$ = 1.29, η_g^2 = .004, p =

.259, BF_{incl} = 0.14.

The trait anxiety index, and three subscales of the empathic sensitiveness scale were

treated independently, and compared between friend- and stranger-observers groups using

two sample t-tests. Again, none of the comparisons was statistically significant and indicated

either absence of evidence or moderate evidence against the difference: trait anxiety: t = 1.07,

df = 63.5, p = 0.29, BF = 0.40; empathic concern: t = 0.20, df = 59.0, p = 0.85, BF = 0.25;

personal distress t = 1.11, df = 64.1, p = 0.27, BF = 0.42; perspective taking: t = 0.12, df = 63.2,

p = 0.91, BF = 0.25.

Skin Conductance Responses

Amplitudes of the observers' skin conductance responses (SCR) were compared

(separately for each stage of the experiment) using classical and Bayesian repeated measures

ANOVA, with group (friend, stranger) as a between subjects factor. In the observational

learning stage, ANOVA revealed a significant main effect of the stimulus (US / no US), $F(1,$

$52)$ = 29.19, η_g^2 = .089, p < .001, BF_{incl} = 6976 where participants showed stronger reactions

to the US compared to no US, $t(52)$ = 5.40, p < .001, BF = 11341 (Figure 3 A). The main effect

of the group, $F(1,52)$ = 2.21, η_g^2 = .034, p = .14, BF_{incl} = 0.75, and the stimulus × group

interaction, $F(1,52)$ = 0.55, η_g^2 = .002, p = .46, BF_{incl} = 0.63, were not significant. Similarly, in

the direct-expression stage, we found a significant effect of the stimulus (CS+/CS-), $F(1, 52)$ =

5.73, η_g^2 = .018, p = .02, BF_{incl} = 1.79, with stronger reactions to CS+ than CS-, $t(52)$ = 2.39, p

= .02 , BF = 2.02 (Figure 3 B), while the main effect of the group, $F(1, 52)$ = 1.89, η_g^2 = .029, p

= .18, BF_{incl} = 0.62, and the stimulus × group interaction, $F(1,52)$ = 0.83, η_g^2 = .003, p = .37,

BF_{incl} = 0.464, were not significant.

Figure 3

Skin Conductance Responses

Note. (A) observational learning stage, (B) direct-expression stage. Error bars extend to data

points placed no further than 1.5*IQR beyond the 1st quartile and above the 3rd quartile.

Imaging Results

Brain Activation Analysis - Observational Learning Stage

In the observational learning stage, we analyzed the whole-brain activity of the observers witnessing the demonstrator's reaction to the aversive stimulation. We compared it to the corresponding periods when the CS appeared without aversive stimulus (US > no US). To test the main effect of the task, we first evaluated the US > no US contrast for both groups. It revealed a robust and extensive activation ($p < 0.05$, FWE peak-level correction) of multiple brain regions, including the bilateral amygdala, bilateral anterior insula (AI), and anterior mid-cingulate cortex (aMCC), the bilateral posterior superior temporal sulcus (pSTS) and bilateral fusiform gyrus, including the fusiform face area (FFA).

Next, we compared the friend and stranger groups directly, i.e., with the (friend US > no US) vs. (stranger US > no US) using t contrasts (see Figure 4 A). The whole-brain analysis yielded no significant between-group differences (with either peak- or cluster-level FWE

corrections). To further compare the groups, we performed region-of-interest analysis

(averaging parameter estimates within a region) for the six pre-selected areas, defined

independently of the functional data (AI, aMCC, amygdala, FFA, rpSTS, rTPJ). We found no

significant differences between the friend and stranger groups; Bayes Factor analysis indicated

moderate evidence for the absence of effects ($BF_{01} > 3$) in all regions except the rTPJ, where

the evidence was inconclusive ($BF_{01} = 2.07$), see Table 1 and Figure 4 B.

An additional between-group comparison, based on the observational US reactions,

i.e., friend US > stranger US t contrast, also did not yield significant differences on a whole-

brain level. Region-of-interest analysis for this contrast (using the same set of regions as

above) confirmed the lack of significant between-group differences. At the same time, Bayes

factors indicated evidence that was either inconclusive or moderately favored the null

hypothesis (all BF_{01} between 1.36 and 4.04), see Table 1 and Figure 4 - figure supplement 1.

Figure 4

Activation in the Observational Learning Stage, US > no US Contrast

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Note. We found no between-group differences. (A) Statistical maps show whole-brain

activations for both groups analyzed together, p < 0.05, FWE peak-level correction. (B) Region-

of-interest statistics, showing group averages and 95% confidence intervals. Abbreviations: AI

- anterior insula, aMCC - anterior mid-cingulate cortex, Amy - amygdala, FFA - fusiform face

area, rpSTS - right posterior superior temporal sulcus, rTPJ - right temporo-parietal junction.

Figure 4 - figure supplement 1

Region-of-interest Statistics for the Observational Learning Stage

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Note. Parameter estimates (group average and 95% confidence intervals, with subjects

represented by dots) are shown for the US and no US separately. AI - anterior insula, aMCC -

anterior mid-cingulate cortex, FFA - fusiform face area, rpSTS - right posterior superior

temporal sulcus, rTPJ - right temporo-parietal junction.

Brain Activation Analysis - Direct-expression Stage

To measure observational fear learning, i.e., the association formed between the CS

and the observational US, in the direct-expression stage, we analyzed the brain activity of the

observers in response to the conditioned stimuli (CS+ > CS-). To evaluate the main effect of

the task, the CS+ > CS- contrast was first analyzed for both groups together and subsequently

compared between the groups.

Analyzing both groups together, we observed significant activations in the bilateral AI

and the aMCC (Figure 5). Next, we performed a direct between-group comparison, i.e., friend

(CS+ > CS-) vs. stranger (CS+ > CS-), which yielded no significant differences on the whole-

brain level. We performed a region-of-interest analysis with the same set of regions as in the

observational learning stage to further compare the groups. We found no significant

differences, and Bayes Factor analysis indicated moderate evidence for the absence of effects

($BF_{01} > 3$) in all regions except the rpSTS where the evidence was inconclusive ($BF_{01} = 2.32$);

see Table 1 and Figure 7.

Figure 5

Activation in the Direct-expression Stage, CS+ > CS- Contrast

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Note. (A) Statistical map showing whole-brain activations for the friend and stranger groups

analyzed together, FWE corrected, p < 0.05. (B) Region-of-interest statistics, showing group

averages and 95% confidence intervals. Abbreviations: AI - anterior insula, aMCC - anterior

mid-cingulate cortex, Amy - amygdala, FFA - fusiform face area, rpSTS - right posterior

superior temporal sulcus, rTPJ - right temporo-parietal junction.

To capture the potential effects of extinction during the direct-expression stage, we

included 1st order temporal modulation in the statistical model. Analyzing both groups

together, we observed a significant decrease in the brain responses to CS+ across several

regions, most notably in the aMCC. To test whether the groups differed concerning the

temporal dynamics of brain responses, we evaluated the (friend CS+ × t) vs. (stranger CS+ x

t) contrast. We observed the differences in some brain regions, including the left inferior frontal

gyrus, left middle temporal gyrus, right lingual gyrus, and left putamen. Strangers showed a

stronger linear decrease in activity than friends (see Figure 6 and Table S5). Importantly,

however, we found no differences in the structures activated in the observational and direct-

expression stages.

Figure 6

Temporal Modulation of the CS+ Response in the Direct-expression Stage

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263 *Note.* (A) Statistical map showing a significant effect for the friend and stranger groups

264 analyzed together. (B) Statistical map showing the (stranger CS+ × time decrease) > (friend

265 CS+ × time decrease) contrast.

267 **Table 1**

268 *Region-of-interest Analysis of Between-group Differences*

ROI	Observational Learning Stage (US > noUS)			Observational Learning Stage (US)			Direct-expression Stage (CS+ > CS-)		
	t	p	BF_{01}	t	p	BF_{01}	t	p	BF_{01}
AI	-0.64	.52	3.39	1.06	.29	2.50	0.03	.97	4.04
aMCC	-0.77	.45	3.13	1.60	.11	1.36	-0.08	.93	4.03

Amy	0.01	.99	4.04	1.38	.17	1.79	0.40	.68	3.76
FFA	-0.80	.43	3.08	0.03	.98	4.04	-0.28	.78	3.90
rpSTS	0.24	.81	3.94	1.08	.29	2.47	1.14	.26	2.32
rTPJ	-1.25	.21	2.07	-0.49	.69	3.78	-0.13	.90	4.01

Note. We performed two comparisons for the observational learning stage based on within-
subject parameter estimates for either the (US > no US) or the US alone. We compared the
(CS+ > CS-) contrast estimate between the groups for the direct-expression stage. The
statistical results: positive t values represent friend > stranger; t-tests with Welch correction to
the degrees of freedom; p values not corrected for the number of comparisons; Bayes factors
reported in favor of the null hypothesis. AI - anterior insula, aMCC - anterior mid-cingulate
cortex, Amy - amygdala, FFA - fusiform face area, rpSTS - right posterior superior temporal
sulcus, rTPJ - right temporo-parietal junction.

Psychophysiological Interactions

We performed a psychophysiological interaction (PPI) analysis for several selected
ROIs to investigate the coupling between the activated structures. First, we analyzed the
observational learning stage, focusing on the US > no US contrast. As the seed structures, we
used the AI, rpSTS, aMCC, amygdala, FFA, and rTPJ. Analyzing both groups together, we
observed increased coupling of the AI with several regions, including the rpSTS (Figure 7 A),
and stronger coupling of the rpSTS with the AI, right fusiform gyrus (Figure 7 B), and amygdala
(Figure 7 C). However, there were no significant differences between the groups. Similarly, we
did not find the group differences using the aMCC, amygdala, FFA, and rTPJ as the seed
structures. Further, we applied the PPI analysis to the activations observed during the direct-
expression stage, using the AI, aMCC, and amygdala as the seeds. We did not find significant
differences for the groups analyzed together or between the friend and stranger groups.

Figure 7

Psychophysiological Interaction Maps, US > no US Contrast

y=-40   x=52   3.2   z=0   z=-16   3.2   y=-4   3.4
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296   *Note.* AI and pSTS seeds were analyzed for both groups together during the observational

297   learning stage. (A) the AI exhibits interaction with several regions, including the rpSTS (FWEc),

298   (B) the rpSTS exhibits interaction with several areas, including the right fusiform gyrus and

299   bilateral AI (FWEc), (C) psychophysiological interaction of the rpSTS with the amygdala (SVC).

301   **Discussion**

Here, we investigated the relevance of the demonstrator's familiarity in observational

fear learning using physiological and neural measures. We focused on the US > no US contrast

in the observational learning and the CS+ > CS- contrast in the direct-expression stage. We

found no significant differences between participants observing friends ('friend' group) and

subjects watching strangers ('stranger' group), neither in fear contagion nor in observational

fear learning. Our study is the first to directly examine the difference between neural correlates

of observational fear learning from friends and strangers. Using the ecologically valid paradigm

that has been developed and tailored for this study, we replicated the previously reported brain

activations underlying fear conditioning in humans. However, brain activity did not differ

between participants observing friends and strangers. This clarifies the role of familiarity in fear

contagion in humans and is in line with the previous rodent reports (Hernandez-Lallement et

al., 2020). Overall, our results show that humans learn fear equally from friends and strangers,

which might have been evolutionarily beneficial.

Main Physiological and fMRI Effects - Validation of the Protocol

We improved the original protocol of Haaker, Golkar et al. (2017) to study a real-time

demonstrator-observer interaction. To validate the modified protocol, we firstly examined the

data regardless of the level of familiarity. We analyzed both physiological and neural correlates

of observational fear learning. The majority of observers correctly identified the CS+/US

contingency and the contingency-aware to non-aware participants ratio, around three-

quarters, was similar across groups. As our previous study showed that the

psychophysiological effects of observational fear learning were present only in the

contingency-aware participants (Szczepanik, Kaźmierowska, et al., 2020), we included only

those in the analyses.

We observed enhanced observers' skin conductance response (SCR) to the shock-

associated stimuli during both observational learning and direct-expression stages, which is in

line with earlier reports (Sevenster et al., 2014; Szczepanik, Kaźmierowska, et al., 2020).

Neuroimaging analysis of data collected during the observational fear learning stage (US > no

US contrast) revealed brain activity in multiple regions, including those crucial for fear-related

processing (amygdala, AI, aMCC), which is consistent with previous studies (Lindström et al.,

2018). Additionally, we observed activation of brain areas involved in dynamic social

perception (fusiform gyrus, pSTS, and supplementary motor area), which may indicate

processing of visual body cues (Allison et al., 2000; Yang et al., 2015). PPI analysis revealed

coupling of the AI with the rpSTS and of the rpSTS with the AI, right fusiform gyrus, and

amygdala, suggesting that these brain regions are jointly involved in the social transfer of fear.

Brain activity during the direct-expression stage (CS+ > CS- contrast) involved the bilateral

insular cortex and aMCC, which is in line with previous findings (Lindström et al., 2018; Olsson

et al., 2007). We did not observe significant amygdala activations. In the direct fear

conditioning, the amygdala is considered a region where the CS-US association is stored

(Olsson & Phelps, 2007; Phelps & LeDoux, 2005). However, none of the previous studies has

provided consistent evidence for the amygdala involvement during the direct-expression stage.

Additionally, a meta-analysis on the neural signatures of the human fear conditioning has

shown that the amygdala is not reliably activated when fMRI fear conditioning tasks are

employed (Fullana et al., 2016). Taken together, using the ecologically valid paradigm, we

replicated most of the previously reported brain activations underlying fear conditioning in

humans.

Familiarity Effects

We employed the experimental design in which familiarity with the demonstrator was

the only between-group difference. We also made sure that the observers learned about the

threat through social means only - they did not directly experience any electrical stimulation at

any point of the experiment. The questionnaire results indicated that the observers from both

groups did not differ in their general level of anxiety and empathy-related traits. Also, their initial

state of anxiety (measured at the beginning of the experiment) was similar. Notably, the

expressions of the demonstrators (the observational US) were rated as natural and did not

differ between the groups. This lack of differences indicates that the experimental conditions

were similar across groups, which is essential considering the naturalistic paradigm employed

and potential confounds related to individual differences in the demonstrators' fear expression.

We found no significant differences in the brain activations of observers learning about

the threat from their friends or strangers. The only previous report directly comparing the neural

correlates of threat-to-self, threat-to-friend, and threat-to-stranger experience (Beckes et al.,

2013) used a within-subject design. The within-subject protocol makes it difficult to disentangle

the impact of direct vs. social sources of information. Our results show that familiarity between

the participants does not modulate social fear learning when a demonstrator is a sole source

of information.

In contrast to our results, the social learning of pain activated the pain network

components (ACC and AI) depending on whether the participants imagined a loved one or a

stranger (Cheng et al., 2010). This result seems to be at odds with our findings. However, the

fear and pain paradigms differ. Although both paradigms use aversive stimulation (often

involving electric shocks), the former emphasizes that it should be uncomfortable but not

painful. In our study, we adjusted the electric shock's intensity by asking every demonstrator

to choose the level of electric stimulation and instructing him to react 'in a natural, yet

noticeable manner'. Based on the observers' post-experimental ratings (on average describing

the degree of unpleasantness attributed to the demonstrators as 'rather unpleasant'), we may

suppose they did not interpret the demonstrators' condition in terms of pain.

Animal studies provide somewhat disparate results on the role of familiarity in threat

contagion (Gonzalez-Liencres et al., 2014; Jeon et al., 2010; Knapska et al., 2010; Sanders et

al., 2013). The studies carried out in a well-known environment, in which the presence of a

partner attenuates threat response (the effect known as social buffering (Kiyokawa et al.,

2014)) more consistently report familiarity effect. Thus, familiarity may play a role in social

attenuation rather than the social enhancement of fear. In line with this reasoning, a recent

meta-analysis of rodent studies indicated no familiarity effect on emotional contagion of threat

(Hernandez-Lallement et al., 2020). The lack of familiarity effect in threat contagion may stem

from the fact that social learning about danger, especially in a novel environment, is equally

effective and might have been evolutionary beneficial regardless of the demonstrator's level of

familiarity.

Emotional contagion is considered a primary form of empathy, which results from the

basic mechanism of the Perception-Action Model (De Waal, 2012; de Waal & Preston, 2017).

The PAM involves automatic matching between the target's and the observer's neural

responses and constitutes a foundation for more complex phenomena such as sympathetic

concern and perspective-taking. The model predicts that emotional contagion is more robust

among individuals in close social relationships as their perception-action coupling is stronger

(Preston & de Waal, 2002). However, our data show that fear contagion does not depend on

the level of familiarity, which may stem from its crucial informative role. In the face of threat,

observing the emotional responses of others can facilitate life-saving responses (Olsson et al.,

2020). It is noteworthy that the actors serving as demonstrators in previous observational fear

learning studies were all unfamiliar to the observers and still remained an effective source of

learning (Golkar et al., 2015; Golkar & Olsson, 2017). Thus, perceiving the emotions of others

not only provides information about their source, i.e., the emotional state of the 'demonstrator'

but also about the environment. Friends and strangers are equally good as a source of

information about threats.

Limitations

Since the participants came together to the laboratory in the 'friend' group, the situation

was more natural and socially engaging than in the 'stranger' group, where the observers had

no interaction with the demonstrators before the experiment. However, if this was an essential

factor, we could expect stronger activations of the social brain network in the 'friend' group,

which was not the case. Moreover, we did not directly measure observers' empathy toward the

demonstrators. However, we assessed observers' general level of empathy, identification with

a demonstrator, and degree of unpleasantness attributed to the demonstrator and did not find

differences between the groups. Finally, we studied only males; thus, it is unclear whether the

results extrapolate to the female population. Considering the sex differences in emotional

processing and empathy (Proverbio, 2021) and a pioneering character of the study, limiting

the probe to one sex only enabled us to watch relatively strong and statistically powered

effects. However, this issue certainly requires future investigations.

Conclusion

We describe here the neural correlates of observational fear learning in humans using

a naturalistic approach. Using this paradigm, we replicated most of the previously reported

brain activation underlying fear conditioning in humans. Importantly, our findings constitute the

first neuroimaging evidence for the lack of relationship between familiarity and fear contagion

and observational fear learning in humans. We argue that fear contagion is an automatic,

unconscious process independent of the level of the demonstrator's familiarity. We claim that

fear is learned equally from friends and strangers in humans, which is evolutionarily beneficial.

These results resonate with ongoing debates on the questions such as the components of

empathy, the boundaries between unconscious threat processing and conscious fear

experience, and the role of the amygdala in the fear conditioning process. A question of the

demonstrator's characteristics modulating the observational fear learning process remains
open, creating a space for further investigations.

Methods

Participants

48 pairs of friends ('friend' group) and 47 individuals ('stranger' group) participated in
the experiment. Pairs of friends took part in the experiment together, with one participant being
a demonstrator and another an observer. The demonstrators undergoing fear conditioning
were live video streamed in the 'friend' group, and their video recordings were used in the
'stranger' group. Only the observers underwent fMRI. To estimate groups' sizes that provide
sufficient statistical power, we considered a similar fMRI experiment with the between-group
design (n=21 and n=22 (Haaker et al., 2017)). We increased the number of recruited
individuals as the rate of contingency-aware participants is lower in a real-time demonstrator-
observer interaction (Szczepanik, Kaźmierowska, et al., 2020). The eligibility criteria included
being heterosexual male, aged between 18 and 30 years, right-handed, and native or fluent
Polish speaker. We excluded students and graduates of either psychology or cognitive
science, participants with neurological disorders or other medical conditions precluding MR
scanning or electrical stimulation, and participants taking psychoactive drugs. Additionally, in
the 'friend' group, the participants had to have known each other for at least three years and
score at least 30 out of 60 points in the McGill Friendship Questionnaire - Respondent's
Affection, see Behavioral Measures.

We excluded four subjects from the 'stranger' group: three subjects who experienced
technical problems with the video playback, and one subject who showed excessive head
motion during one of the tasks (more than 25% volumes classified as motion outliers, see fMRI
Data Preprocessing section). We included only the contingency-aware participants in the
analyses; please refer to Assessment of Stimulus Contingency Awareness (Szczepanik,
Kaźmierowska, et al., 2020). The final group sizes were n = 35 ('friend' group) and n = 34

('stranger' group). In the analyzed sample, the mean age of all observers was 22.9 years (SD

= 2.87), the mean length of the observer-demonstrator friendship was 7.7 years (SD = 4.22),

and the mean observers' score in the McGill Friendship Questionnaire was 52 (SD = 7.74),

see Table S2 in Appendix for the detailed statistics. All participants received financial

remuneration of 100 PLN (~23 EUR) for their participation.

Tasks and Stimuli

Experimental Setup

In the MRI scanner, the observer watched a video (either a streaming or a replay, both

without sound) or picture stimuli on an MR-compatible monitor through a mirror box placed on

the head coil. The demonstrator sat in a small room adjacent to the MR room (in the 'friend'

group). A GoPro Hero7 camera provided video transmission and recording. To ensure reliable

reproduction of stimuli' colors, we lighted the room with an LCD panel with adjustable color

temperature, set the computer screen brightness to low, and adjusted the camera's white

balance. The room walls were covered with gray acoustic foam to minimize distractors.

Stimuli

The conditioned stimuli were two squares, blue and yellow, displayed on a gray

background. The squares were presented centrally and covered either a half (during the

observation stage) or a quarter (during the direct-expression stage) of screen height (Figure

8). The assignment of color to CS+ and CS- was counterbalanced across participants.

Cutaneous electrical stimulation, applied to the ventral part of a forearm of the

demonstrator, was used as the unconditioned stimulus. Stimulating electrodes were placed

above the *flexor carpi radialis* muscle so that even low-intensity stimulation caused involuntary

muscle flexion, visible to the observer. The stimulation consisted of five unipolar pulses of 1

ms duration applied in 200 ms intervals. The demonstrators individually adjusted shock

intensity to be unpleasant but not painful (see Demonstrator's Preparation and Instructions).

 Observational Learning Stage

486   The observer watched the demonstrator performing a differential conditioning task. The

487 demonstrator watched 24 CS+, and 24 CS- displayed on the screen in pseudo-random order,

488 with a given CS repeated at most twice in a row. Each CS lasted 9 seconds, half of the CS+

489 reinforced with the US. The US reinforced the first and the last presentation of CS+. The US

490 started 7.5 seconds after the CS onset to allow the demonstrator's reaction to co-terminate

491 with the CS. The CS- was never reinforced. The intertrial intervals (ITIs) lasted randomly

492 between 10 and 15 seconds, with a fixation symbol (+) displayed centrally on the screen

493 (Figure 8).

495 ***Direct-expression Stage***

496   The observer watched 12 CS+ and 12 CS-. The stimuli and the stimuli' order, timing,

497 and ITI were the same as during the observational learning stage. None of the CS was

498 reinforced.

500 **Figure 8**

501 *Tasks and Stimuli Overview*

503
504 *Note.* Schematic representation of the task design. (A) The observational learning stage:

505 participants watched another person (demonstrator) who performed a differential conditioning

506 task. The US (electrical stimulation) reinforced half of the visual CS+, CS- were never

507 reinforced. (B) The direct-expression stage: participants, informed that they would undergo an

identical task as the demonstrators, watched the same visual stimuli presentation. However,

the electric stimulation was not applied.

Procedure

To investigate observational fear learning, we adapted the protocol of Haaker, Golkar,

et al. (2017) for live observation of demonstrator-observer interaction (Szczepanik,

Kaźmierowska, et al., 2020). The protocol was approved by the Ethics Committee of the

Faculty of Psychology at the University of Warsaw (decision from 28 November 2017). The

procedure complied with the Ethical Principles of Psychologists and Code of Conduct

published by the Polish and the American Psychological Associations.

Before the Experiment

The participants received brief information about the experimental procedure, including

the possibility of receiving aversive electrical stimulation. Next, the participants gave informed

consent (including optional permission for recording and recording's reuse) and filled in safety

screening forms. The roles of a demonstrator and an observer were then randomly assigned

to the participants by giving two color-coded envelopes. Then the participants were isolated;

the demonstrators went to a room adjacent to the MR room.

Observer's Preparation and Instructions

First, the observers filled in the questionnaires (see Behavioral Measures for details).

Next, they opened the envelope with the role assignment and received instructions. The

instruction included information on the two-stages procedure in the fMRI scanner. The

instruction said that, firstly, the participants would watch their friends perform a task involving

the presentation of colored symbols and the administration of unpleasant electrical stimulation.

Then, they would perform an identical task themselves. Importantly, no information about the

stimuli contingency was provided. After receiving the instruction, the observers had stimulation

and skin conductance electrodes attached and went to the MR room. In the scanner, the

subjects had sham leads connected to the stimulation electrodes attached to make receiving

electrical stimulation believable.

Demonstrator's Preparation and Instructions

First, the demonstrators filled in the questionnaires (see Behavioral Measures for

details). Next, they opened the envelope with the role assignment and received instructions.

The instruction said they would perform a task involving the presentation of colored symbols

and administering unpleasant, but not painful, electrical stimulation. Unlike the observers, the

demonstrators received information on the stimuli contingency and reinforcement rules. We

asked the demonstrators to react to the stimulation in a natural yet noticeable manner. The

demonstrators watched a recording with a model reaction. After receiving the instruction, the

demonstrators had stimulation and sham skin conductance electrodes attached. Afterward,

the demonstrators adjusted stimulation intensity. The experimenter increased it stepwise, and

the participants rated the intensity using a scale from 1 (*imperceptible*) to 8 (*painful*). The target

level was 6 (*very unpleasant but not painful*).

MRI Procedure

The MRI scanning started with the acquisition of an anatomical image. Then, the

observational learning stage followed (see Tasks and Stimuli). The experimenter adjusted the

camera's position to ensure that the observer could see the demonstrator's face, hand, and

computer screen and turned on video transmission and recording. The observer received a

brief reminder: 'in this part of the study, you will observe your friend performing a certain task'.

After completion of the observational learning stage, the observer received a brief reminder:

'in this part of the study, you will perform the same task that you just watched', and the direct-

expression stage ensued (see Tasks and Stimuli).

Conclusion of the Experiment

While the observer took part in the direct-expression stage, the demonstrator

completed the post-experimental questionnaires (see Behavioral Measures for details). After

the MRI session ended, the observer also completed their set of post-experimental

questionnaires, and both participants were debriefed about the study.

Stranger Group

In the 'stranger' group, only one subject (the observer) participated in the experimental

session. Instead of a live stream, we used 45 recordings made in the 'friend' group. Otherwise,

the procedure in the stranger group was very similar to that involving a pair of friends and

differed only in the necessary aspects. There was no role assignment, and instructions given

to the observers referred to another person rather than a friend.

Behavioral Measures

McGill Friendship Questionnaire - Respondent's Affection

To recruit pairs of friends, we used the McGill Friendship Questionnaire - Respondent's

Affection (Mendelson & Aboud, 1999); translated by A. Kaźmierowska, P. Pączek, and A.

Schudy for online screening in this study. It consists of 16 positive statements describing

feelings for a friend and satisfaction with the friendship, rated along a 9-point scale ranging

from -4 (*very much disagree*) to 4 (*very much agree*). One item (no. 9) was omitted from the

questionnaire due to human error. A score of 30 points was a threshold for inclusion in the

study.

Empathic Sensitiveness Scale

To measure the participants' empathy, we used the Empathic Sensitiveness Scale

(original title: Skala Wrażliwości Empatycznej, SWE (Kaźmierczak et al., 2007)). It is a self-

report questionnaire based on the Interpersonal Reactivity Index (Davis, 1983), but with far-

reaching modifications developed in the Polish language. It contains 28 statements, which

reflect three components of empathy: two emotional (empathic concern, personal distress) and

one cognitive (perspective taking). The answers are given on a 5-point scale. The

demonstrators and observers filled in the questionnaire at the end of the experimental

procedure.

State and Trait Anxiety Inventory

To measure participants' anxiety, we used the State and Trait Anxiety Inventory

(Spielberger et al., 1983, polish adaptation 2012). It is a self-report questionnaire consisting of

two 20-item subscales, STAI-State and STAI-Trait. Participants rate statements related to how

they feel either at a given moment (state) or usually (trait) using 4-point Likert scales. Each

participant completed the STAI-state twice (at the beginning and the end of the experiment)

and STAI-trait once (at the end of the experiment).

Assessment of Stimulus Contingency Awareness

To assess the observers' declarative knowledge of the CS/US contingency, we used

an online questionnaire adapted from (Weidemann et al., 2016). The questionnaire referred to

the observational learning stage and was divided into several parts to eliminate guessing.

Firstly, it asked whether the participant could predict shocks to the demonstrator. Then, if

answered positively, an open-ended question 'how' followed. Next, participants gave

percentage ratings of shock occurrence for each stimulus. Finally, they made a forced choice

of one stimulus paired with the shock. CS+, CS- and a fixation symbol were all presented as

stimuli in the latter two questions. Observers completed this questionnaire after the direct-

expression stage to avoid interfering with the learning process (Haaker, Golkar, et al., 2017).

Evaluation of the Demonstrator's Expression (the Observational US)

To control how the observers perceived the demonstrators' behavior, we used a set of

questions suggested by Haaker, Golkar, et al. (2017). At the end of the experiment, we asked

the observers to rate: how much discomfort the demonstrator experienced when receiving the

electrical stimulation, how expressive the demonstrator was, how natural the demonstrator's

reactions were, and how much empathy they felt for the demonstrator. Additionally, we asked

the observers to rate the degree of unpleasantness attributed to the demonstrators. Finally, in

the 'friend' group, the observer scored the degree to which they identified with their friend. All

ratings used a 10-point Likert scale, ranging from 0 (*not at all*) to 9 (*very much*), except for the

unpleasantness rating, which used a 5-point Likert scale ranging from 1 (*very unpleasant*) to

5 (*very pleasant*).

Skin Conductance Response Recordings

During fMRI sessions, we registered skin conductance responses as a secondary

measure. It was recorded using BrainVision BrainAmp ExG MR amplifier with GSR MR sensor,

sampled at 250 Hz.

fMRI Data Acquisition

The MRI data were acquired on a 3T Siemens Magnetom Trio scanner equipped with

a 12 channel head coil. At the beginning of a session, a T1-weighted anatomical image was

acquired using an MPRAGE sequence with 1 × 1 × 1 mm resolution and the following

parameters: inversion time TI = 1100 ms, GRAPPA parallel imaging with acceleration factor

PE = 2, acquisition time TA = 6 minutes and 3 seconds. After acquiring the anatomical scans,

two functional imaging runs followed (observational learning and direct-expression tasks). The

first run contained 362 volumes in the 'friend' group and 380 volumes in the 'stranger' group.

The scanning started earlier in the 'stranger' group, but the leading volumes were removed

from the analysis. The second run of scanning contained 184 volumes. Each functional volume

comprised 47 axial slices (2.3 mm thick, with 2.3 × 2.3 mm in-plane resolution and 30%

distance factor) that were acquired using a T2*-sensitive gradient echo-planar imaging (EPI)

sequence with the following parameters: repetition time TR = 2870 ms, echo time TE = 30 ms,

flip angle FA = 90 degrees, field of view FoV = 212 mm, matrix size: 92 × 92, interleaved

acquisition order, GRAPPA acceleration factor PA = 2.

fMRI Data Preprocessing

The fMRI data were preprocessed using fMRIPrep 1.4.0 (Esteban, Blair, et al., 2019;

Esteban, Markiewicz, et al., 2019), based on Nipype 1.2.0 (Gorgolewski et al., 2011;

Gorgolewski et al., 2019) and Nilearn 0.5.2 (Abraham et al., 2014). We followed the fMRIprep

preprocessing with smoothing in SPM (SPM 12 v7487, Wellcome Centre for Human

Neuroimaging, London, UK). At the beginning of the fMRIprep pipeline, the anatomical images

were corrected for intensity non-uniformity with N4BiasFieldCorrection (Tustison et al., 2010),

distributed with ANTs 2.2.0 (Avants et al., 2008), and used as an anatomical reference. The

anatomical reference was then skull-stripped with antsBrainExtraction (from ANTs) and

segmented using fast from FSL 5.0.9 (Zhang et al., 2001). Finally, the anatomical images were

normalized to the MNI space through nonlinear registration with antsRegistration (ANTs 2.2.0).

The ICBM 152 Nonlinear Asymmetrical template version 2009c was used (Fonov et al., 2009).

The functional images from the observational learning and direct-expression stages

were preprocessed in the following manner. First, a reference volume was generated using a

custom methodology of fMRIPrep. This reference was co-registered to the anatomical

reference using flirt from FSL 5.0.9 (Jenkinson & Smith, 2001) with the boundary-based

registration cost-function (Greve & Fischl, 2009). Head-motion parameters with respect to the

functional reference volume (transformation matrices and six corresponding rotation and

translation parameters) were estimated before any spatiotemporal filtering using mcflirt from

FSL 5.0.9 (Jenkinson et al., 2002). The functional scanning runs were slice-time corrected

using 3dTshift from AFNI 20160207 (Cox & Hyde, 1997). Next, the functional images were

resampled into the MNI space. All resamplings were performed in a single interpolation step

(composing head-motion transform matrices and co-registrations to anatomical and output

spaces) using antsApplyTransforms (ANTs). Finally, we smoothed the functional images with

a 6 mm FWHM 3D Gaussian kernel using spm_smooth (SPM 12). Framewise displacement

(FD) and the derivative of root mean square variance over voxels (DVARS) were calculated

by fMRIPrep for each functional scan, both using their implementations in Nipype and following

the definitions by (Power et al., 2014). Frames that exceeded a threshold of 0.5 mm FD or 1.5

standardized DVARS were annotated as motion outliers. For more details on the fMRIprep

pipeline, see fMRIPrep's documentation at https://fmriprep.org/en/latest/workflows.html.

Physiological Data Analysis

To analyze skin conductance data we used PsPM 4.3.0 software

(https://bachlab.github.io/PsPM/) running under MATLAB 2018b (MathWorks, Natick, MA,

USA). We used a non-linear model for event-related SCR. In this analysis, three parameters

of sudomotor nerve response function (its amplitude, latency and dispersion) are estimated to

best match the observed skin conductance data. This model is suitable for anticipatory

responses in fear conditioning and assumes that the response onset following CS presentation

is not precisely known (Bach et al., 2010; Staib et al., 2015). Before the analysis, we visually

inspected the signals for artifacts and manually marked missing epochs. We excluded subjects

with missing data or excessive presence of artifacts precluding further analysis. As a result,

we analyzed SCR data for 27 subjects per group. We used the default settings for

preprocessing: signals were filtered using bi-directional 1st order Butterworth filters, with 5 Hz

low-pass and 0.0159 Hz high-pass cut-off frequencies, and resampled to 10 Hz. We performed

no response normalization. We accelerated the processing of multiple subjects using GNU

Parallel v.20161222 (Tange, 2011).

fMRI Data Analysis

Activation Analysis

To analyze data we used a mass univariate approach based on a general linear model.

We used SPM 12 software (SPM 12 v7487, Wellcome Centre for Human Neuroimaging,

London, UK) running under MATLAB 2020a (MathWorks, Natick, MA, USA). First-level models

contained four types of events in the observational learning stage: CS+, CS-, US, and no US

(i.e., lack of US during 50% of CS+). The observational CS were modeled as instantaneous

events (i.e., CS onset), while the observational US/no US events were 1.5 seconds (i.e., from

US onset to CS offset). There were only two types of events in the direct-expression stage,

CS+ and CS-, modeled as 9 seconds (entire presentation of CS). Additionally, we included

temporal modulation of the 1st order to capture effects of extinction in the direct-expression

stage. In addition to the event regressors described above, we had six motion parameters

(translation and rotation) as regressors of no interest. We added one delta regressor for each

volume annotated by fMRIPrep as a motion outlier (up to 60 such volumes per subject in the

observational learning stage, up to 38 in the direct-expression stage, median 8 and 2,

respectively).

The US > no US was the primary contrast of interest in the observational learning stage

and the CS+ > CS- in the direct-expression stage. We calculated the contrasts within-subject

and used the parameter estimates in second-level analyses. We performed two types of

second-level analyses in SPM: one-sample t-test designs for data pooled across groups (the

'friend' and 'stranger' groups) and two-sample t-tests for between-groups comparisons (the

'friend' vs. the 'stranger' group). Additionally, we used a temporal modulation regressor (CS+

× t) for both groups analyzed together and for between-group comparisons. We thresholded

the second-level statistical maps using family-wise error (FWE) correction, with a p = 0.05

threshold. We used the voxel-level correction for the primary contrasts and cluster size

correction with p = .001 cluster defining threshold for the results of temporal modulation and

psychophysiological interactions.

Region-of-interest - Definitions and Analysis

To perform the ROI analysis (parameter estimate extraction), the psychophysiological

interaction analysis (time course extraction), and, in the case of the amygdala, the small

volume correction (SVC) analysis, we defined anatomical regions of interest (ROI). We used

the following brain structures as ROIs: the anterior insula (AI), anterior mid-cingulate cortex

(aMCC), amygdala, fusiform face area (FFA), right posterior superior temporal sulcus (rpSTS),

and right temporo-parietal junction (rTPJ). We chose regional definitions for these structures

based on a meta-analytic connectivity mapping study focusing on social processes

(Gorgolewski et al., 2011) and the Neurovault collection

(https://identifiers.org/neurovault.collection:2462).

We combined anatomical masks from the left and right hemispheres and treated them

as single ROIs in the AI, amygdala, and FFA analysis. For the TPJ and pSTS, the regions

considered functionally lateralized (Alcalá-López et al., 2018; Boccadoro et al., 2019; Sliwinska

& Pitcher, 2018), we used only the right-hemispheric masks. For the aMCC, which is a

centrally-located structure, we used no hemispheric division. We used the same definitions

across the ROI and PPI analyses. For small volume correction, which was applied only for the

amygdala, we used an alternative bilateral definition based on the Harvard - Oxford atlas

(Desikan et al., 2006) thresholded at 0.7. This definition was consistent with a previous study

on observational fear conditioning (Lindström et al., 2018) and provided better structure

coverage than the more restrictive meta-analytic definition. For the ROI analysis, the first-level

parameter estimates were extracted and averaged within a given region for each participant

using the spm_summarise function from SPM and compared using the BayesFactor package

in R (Morey & Rouder, 2018).

Psychophysiological Interactions

To perform the psychophysiological interaction analysis (PPI), we extracted the

functional time series (1st eigenvariate) from the ROIs defined above and multiplied with the

psychological variable (observation stage: US > no US; direct-expression stage: CS+ > CS-)

using SPM VOI and PPI modules. We entered the PPI time series into a first-level model as

the regressor of interest. We also included the functional time series and the psychological

variable in the model as regressors and the motion parameters and motion outliers (the same

as in the models used for activation analysis). Then we subjected parameter estimates

obtained for the PPI regressor to second level analyses, using one-sample (both groups

analyzed together) and two-sample (between-group) designs. We thresholded the statistical

maps using cluster-based FWE correction, with cluster-defining threshold of $p = 0.001$ and

corrected p-value of 0.05. Additionally, we applied a small volume correction to the amygdala.

Statistical Analysis

To perform most statistical analyses we used R with *afex*, *emmeans*, and *BayesFactor* libraries, with plots made using *the ggpubr* library. We calculated Bayesian ANOVA using JASP. We used repeated-measures ANOVA (with type III errors) and Bayesian repeated-measures ANOVA to analyze STAI-state scores (with the group as a between-subject factor and the questionnaire filling time, before and after the experiment, as a within-subject factor) and skin conductance response amplitudes (with the group and stimulus factors). We used the Wilcoxon-Mann-Whitney test to compare the discrete ratings describing the demonstrator's expression (observational US) between the groups and a chi-squared test to compare the distribution of contingency-aware participants. We used t-test (with Welch's correction for unequal variance) and Bayesian t-test (using default effect size priors, i.e., Cauchy scale 0.707) to compare the results of STAI-trait and subscales of SWE, and region-of-interest parameter estimates from fMRI.

For the classical tests, we applied a significance level of $p = 0.05$. For Bayesian t-tests, we used BF_{10} to denote evidence for the alternative hypothesis and BF_{01} for the null hypothesis. In Bayesian ANOVA, we reported the effects as the Bayes factor for the inclusion of a given effect (BF_{incl}), calculated as a ratio between the models' average likelihood, including the factor (given the data) vs. that of the models without the factor. When interpreting these results, we followed the convention suggested by Keysers et al. (2020) and considered $\frac{1}{3} < BF < 3$ as absence of evidence, $1/10 < BF < \frac{1}{3}$ or $3 < BF < 10$ as moderate evidence and $BF < 1/10$ or $BF > 10$ as strong evidence.

Data Availability

Relevant data are stored in an OSF repository and are available at https://osf.io/g3wkq/. Unthresholded statistical maps from the reported comparisons are also available at Neurovault, https://neurovault.org/collections/RSLLSFTQ/. Code replicating analyses reported here is available at https://github.com/nencki-lobi/emocon-mri.

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Competing Interests

The authors declare no competing interests.

Author Contributions

Anna M. Kaźmierowska - Conceptualization, Methodology, Data curation, Formal Analysis, Investigation, Software, Visualization, Writing - original draft, Writing - review & editing

Michał Szczepanik - Conceptualization, Methodology, Data curation, Formal Analysis, Investigation, Software, Visualization, Writing - original draft, Writing - review & editing

Marek Wypych - Conceptualization, Methodology, Supervision, Writing - review & editing

Dawid Droździel - Investigation, Resources

Artur Marchewka - Conceptualization, Methodology, Resources, Supervision, Writing - review & editing

Jarosław M. Michałowski - Conceptualization, Methodology, Supervision, Writing - review & editing

Andreas Olsson - Conceptualization, Supervision, Writing - review & editing

Ewelina Knapska - Conceptualization, Methodology, Funding acquisition, Project administration, Supervision, Writing - review & editing

Appendix

Table S1

Evaluation of the Demonstrator's Expression (the Observational US)

	Discomfort of the Demonstrator		Expressiveness of the Demonstrator		Naturalness of the Demonstrator		Empathy Toward the Demonstrator		Identifying with the Demonstrator		Unpleasantness Attributed to the Demonstrator	
Group	F	S	F	S	F	S	F	S	F	S	F	S
Median (IQR)	6 (2.0)	6 (2.0)	6 (2.0)	6.5 (2.0)	7 (4.0)	7 (3.75)	5 (3.5)	5 (3.75)	7 (2.0)	n/a *	1 (1.0)	2 (1.0)
Range	2 - 8	0 - 9	2 - 8	4 - 9	0 - 9	2 - 9	0 - 9	0 - 8	1 - 9	n/a	1 - 4	1 - 5
Stat. Test	W = 581.5, p = 0.87		W = 501, p = 0.25		W = 630.5, p = 0.67		W = 671.5, p = 0.36		n/a		W = 586, p = 0.91	

Note. To evaluate the demonstrator's expression (the observational US), we asked the observers to rate (on a 0 - 9 scale) the demonstrator's reactions. The questions concerned the demonstrator's responses to electrical stimulation (how much discomfort they experienced, how expressive they were, and how natural their reactions were) and the observers' attitude (how much empathy they felt for the demonstrator). Additionally, we asked the observers how well they could identify with the demonstrator. We asked the question in the pairs of friends only, as the pilot studies revealed problems with the question's understanding in the 'stranger' group (*n/a: no answer). Additionally, we asked the observers to rate the degree of unpleasantness attributed to the demonstrators (on a 1 - 5 scale). The last row shows comparisons between the 'friend' and 'stranger' groups, the Wilcoxon-Mann-Whitney test, p-values uncorrected. F - 'friend' group; S - 'stranger' group; IQR: interquartile range.

Table S2

Descriptive Statistics for Age, Demonstrator-observer Friendship Length and the McGill Friendship Questionnaire (MFQ) Result for the Subgroups of Participants

Variable	Subgroup	Mean (SD)	Median (IQR)	Range
Age	OF	22.5 (2.77)	22 (4.5)	18 - 29
	OS	23.3 (2.95)	22.5 (4)	18 - 30

	D	22.1 (2.53)	22 (4)	18 - 28
Friendship Length	OF	7.7 (4.22)	6 (7)	3 - 19
MFQ Result	OF	52 (7.7)	55 (9.5)	33 - 60
	DF	49.8 (9.21)	52.5 (15)	30 - 60

Note. Data regarding friendship length and the MFQ result were collected in the 'friend' group

only. OF: observers from the 'friend' group who learned the contingency (n = 35); OS:

observers from the 'stranger' group who learned the contingency (n = 34); D: demonstrators

for whom either a friend or a stranger observer learned the contingency (n = 44); DF:

demonstrators for whom a friend observer learned the contingency (n = 34); SD: standard

deviation; IQR: interquartile range.

Table S3

Activation Peaks in the Observational Learning Stage, US > no US Contrast

Cluster	Peak x	Peak y	Peak z	Peak value	Volume mm	Vox	Label
1	50	-66	6	14.98	89912	11239	Right Lateral Occipital Cortex inferior division
	44	-76	-8	12.92			Right Lateral Occipital Cortex inferior division
	42	-44	-16	12.21			Right Temporal Occipital Fusiform Cortex
	64	-40	26	11.94			Right Supramarginal Gyrus posterior division
	52	-28	-6	11.74			Right Middle Temporal Gyrus posterior division
	50	-40	10	11.51			Right Supramarginal Gyrus posterior division
	34	22	-18	10.71			Right Frontal Orbital Cortex
	52	8	-22	10.34			Right Temporal Pole
	22	-88	-8	10.07			Right Occipital Fusiform Gyrus
	46	26	-4	9.82			Right Frontal Orbital Cortex
	42	-2	-14	9.71			Right Planum Polare
	52	24	22	9.14			Right Inferior Frontal Gyrus pars opercularis
	46	2	44	8.88			Right Precentral Gyrus

	x	y	z				Region
	12	-68	38	8.81			Right Precuneous Cortex
	36	-62	-20	8.55			Right Occipital Fusiform Gyrus
	22	-2	-18	8.42			Right Amygdala
	14	-98	6	8.17			Right Occipital Pole
	28	-92	12	8.16			Right Occipital Pole
	38	6	4	7.50			Right Insular Cortex
	54	40	-2	7.40			Right Frontal Pole
	64	-32	40	7.35			Right Supramarginal Gyrus anterior division
	26	-80	34	6.78			Right Lateral Occipital Cortex superior division
2	-48	-78	6	13.87	32568	4071	Left Lateral Occipital Cortex inferior division
	-52	-58	10	11.73			Left Middle Temporal Gyrus temporooccipital part
	-62	-46	22	11.32			Left Supramarginal Gyrus posterior division
	-44	-50	-18	10.52			Left Inferior Temporal Gyrus temporooccipital part
	-30	-92	-10	9.49			Left Occipital Pole
	-24	-66	-28	6.64			no label
	-28	-94	14	6.22			Left Occipital Pole
3	-32	20	-14	10.65	15520	1940	Left Frontal Orbital Cortex
	-30	28	2	8.87			Left Insular Cortex
	-42	0	-16	8.24			Left Planum Polare
	-38	-12	-6	8.12			Left Insular Cortex
	-20	-4	-14	7.58			Left Amygdala
	-54	10	8	6.42			Left Inferior Frontal Gyrus pars opercularis
4	6	20	38	9.50	10368	1296	Right Paracingulate Gyrus
	4	20	56	9.07			Right Superior Frontal Gyrus
	6	40	10	8.55			Right Cingulate Gyrus anterior division
	10	4	72	6.61			Right Superior Frontal Gyrus
5	-8	-24	-12	9.77	6816	852	Brain-Stem
	12	-22	-10	8.72			no label
	10	-8	6	8.38			Right Thalamus

Cluster	Peak x	Peak y	Peak z	Peak value	Volume mm	Vox	Label
	10	8	4	7.85			Right Caudate
	6	-4	-10	7.13			no label
6	-10	-66	36	8.61	2336	292	Left Precuneous Cortex
7	-2	-24	28	8.69	2096	262	Left Cingulate Gyrus posterior division
	0	-8	32	6.43			Left Cingulate Gyrus anterior division
8	-16	-94	26	7.12	1896	237	Left Occipital Pole
	-26	-78	24	5.95			Left Lateral Occipital Cortex superior division
9	34	-52	50	6.78	1072	134	Right Superior Parietal Lobule
10	-30	-52	52	6.86	680	85	Left Superior Parietal Lobule
11	4	52	34	6.47	632	79	Right Superior Frontal Gyrus
12	-10	-76	-44	6.58	440	55	no label
13	-10	-84	8	6.03	392	49	Left Intracalcarine Cortex
14	-40	-4	48	6.68	376	47	Left Precentral Gyrus
15	-50	-26	-6	7.12	368	46	Left Middle Temporal Gyrus posterior division

837 *Note.* Table shows local maxima more than 16 mm apart. For brevity, clusters greater than 30

838 voxels were included. Last column gives the most probable label from the Harvard - Oxford

839 atlas.

841 **Table S4**

842 *Activation Peaks in the Direct-expression Stage, CS+ > CS- Contrast*

Cluster	Peak x	Peak y	Peak z	Peak value	Volume mm	Vox	Label
1	30	30	0	8.30	5024	628	Right Frontal Orbital Cortex
	50	22	4	6.70			Right Inferior Frontal Gyrus pars triangularis
2	-36	24	-4	7.18	1400	175	Left Frontal Orbital Cortex
3	-14	-76	-30	6.37	112	14	no label
4	10	-12	-12	6.25	64	8	no label
5	6	24	36	5.77	40	5	Right Paracingulate Gyrus
6	34	46	20	5.63	8	1	Right Frontal Pole
7	8	4	2	5.45	8	1	Right Caudate

843   *Note.* Table shows local maxima more than 16 mm apart. Last column gives the most probable

844   label from the Harvard - Oxford atlas.

846   **Table S5**

847   *Activation Peaks for the Temporal Modulation of the CS+ Response in the Direct-Expression*

848   *Stage*

Cluster	Peak x	Peak y	Peak z	Peak value	Volume mm	Vox	Label
				Both Groups			
1	-18	-88	-6	6.38	10480	1310	Left Occipital Fusiform Gyrus
	-8	-100	0	5.43			Left Occipital Pole
	-12	-80	-32	4.49			no label
	10	-86	-10	4.47			Right Lingual Gyrus
	0	-86	4	4.01			Left Intracalcarine Cortex
	24	-76	-14	3.65			Right Occipital Fusiform Gyrus
	-26	-98	20	3.58			Left Occipital Pole
2	-4	-54	62	5.42	5344	668	Left Precuneous Cortex
	-12	-76	44	4.84			Left Lateral Occipital Cortex superior division
	4	-36	46	3.75			Right Cingulate Gyrus posterior division
	-8	-68	28	3.71			Left Precuneous Cortex
3	2	8	34	4.58	1704	213	Right Cingulate Gyrus anterior division
	-2	26	24	3.66			Left Cingulate Gyrus anterior division
4	44	28	8	4.71	1184	148	Right Inferior Frontal Gyrus pars triangularis
5	6	46	32	5.67	848	106	Right Paracingulate Gyrus
				Friend > Stranger			
1	-58	14	10	5.83	2280	285	Left Inferior Frontal Gyrus pars opercularis
2	-62	-44	8	4.65	1328	166	Left Supramarginal Gyrus posterior division
3	16	-58	-2	4.41	1096	137	Right Lingual Gyrus
4	-52	36	8	4.83	1072	134	Left Inferior Frontal Gyrus pars triangularis
5	-24	-6	12	4.55	768	96	Left Putamen

849 *Note.* Table shows local maxima more than 16 mm apart. Last column gives the most probable

850 label from the Harvard - Oxford atlas.

852 **Table S6**

853 *Activation peaks for the psychophysiological interaction analysis based on the US > no US*

854 *contrast*

Cluster	Peak x	Peak y	Peak z	Peak value	Volume mm	Vox	Label
					Anterior Insula		
1	64	-38	20	6.16	8520	1065	Right Supramarginal Gyrus posterior division
	54	-48	4	5.00			Right Middle Temporal Gyrus temporooccipital part
	58	-66	6	4.63			Right Lateral Occipital Cortex inferior division
	48	-28	2	4.62			Right Superior Temporal Gyrus posterior division
	48	-80	0	3.60			Right Lateral Occipital Cortex inferior division
2	-52	-56	12	5.26	6792	849	Left Middle Temporal Gyrus temporooccipital part
	-42	-70	12	4.80			Left Lateral Occipital Cortex inferior division
	-68	-46	8	4.01			Left Middle Temporal Gyrus temporooccipital part
	-50	-80	-4	3.76			Left Lateral Occipital Cortex inferior division
3	-12	-74	-14	4.88	3536	442	Left Lingual Gyrus
	-30	-74	-22	4.64			Left Occipital Fusiform Gyrus
	-40	-64	-14	3.66			Left Occipital Fusiform Gyrus
4	52	14	26	4.46	808	101	Right Inferior Frontal Gyrus pars opercularis
5	32	-68	-28	4.16	744	93	no label
					posterior STS		
1	46	-64	2	9.29	20656	2582	Right Lateral Occipital Cortex inferior division
	66	-40	20	6.51			Right Supramarginal Gyrus posterior division
	38	-80	-14	5.01			Right Lateral Occipital Cortex inferior division
	46	-40	10	4.64			Right Supramarginal Gyrus posterior division
	62	-38	44	4.21			Right Supramarginal Gyrus posterior division
	32	-94	-4	3.51			Right Occipital Pole
2	-46	-70	8	8.07	12424	1553	Left Lateral Occipital Cortex inferior division

	-58	-48	12	5.20			Left Supramarginal Gyrus posterior division
3	38	32	-2	5.68	4840	605	Right Frontal Orbital Cortex
4	-58	-40	36	4.70	3072	384	Left Supramarginal Gyrus posterior division
	-66	-32	24	4.63			Left Supramarginal Gyrus anterior division
5	-38	22	-8	5.01	2608	326	Left Frontal Orbital Cortex
	-34	6	-6	3.57			Left Insular Cortex
6	48	14	30	4.91	2240	280	Right Inferior Frontal Gyrus pars opercularis
	54	30	20	3.71			Right Inferior Frontal Gyrus pars triangularis
7	48	-42	-16	6.12	1992	249	Right Inferior Temporal Gyrus temporooccipital part
8	38	-12	-6	4.76	1768	221	Right Insular Cortex
	50	-24	-4	4.49			Right Middle Temporal Gyrus posterior division
9	12	-2	12	4.93	1664	208	Right Thalamus
	-2	-4	4	3.94			Left Thalamus
10	54	20	8	4.96	1520	190	Right Inferior Frontal Gyrus pars opercularis
11	6	6	62	4.63	1344	168	Right Juxtapositional Lobule Cortex (formerly Supplementary Motor Cortex)
12	20	10	-6	4.96	1120	140	Right Putamen
13	-4	-14	-10	4.75	904	113	no label
14	6	-24	0	4.49	816	102	Right Thalamus
15	-40	-62	-28	4.65	744	93	no label

posterior STS (small volume corrected within the amygdala)							
1	24	-4	-16	3.82	104	13	Right Amygdala
2	-22	-4	-14	3.56	24	3	Left Amygdala
3	-20	-6	-18	3.42	8	1	Left Amygdala

Note. Table shows local maxima more than 16 mm apart. Last column gives the most probable label from the Harvard - Oxford atlas.

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