

**Ruud Hortensius** Social Robots Journal Club 24.04.20





Rosenthal-von der Pütten, A. M., Krämer, N. C., Maderwald, S., Brand, M., & Grabenhorst, F. (2019). Neural Mechanisms for Accepting and Rejecting Artificial Social Partners in the Uncanny Valley. The Journal of Neuroscience, 39(33), 6555–6570.



## The uncanny valley







## Polar Express move over

#### It's a DNN gone wrong





# The uncanny valley

• Theorised by Masahiro Mori (1970, Energy)



Picture by Bernd Schifferdecker

- reactions
- Also mentioned by <u>Freud (1919)</u> and <u>Jentsch (1906)</u>



• Non-linear relationship between humanlikeness and likability, resulting in eeriness, fear, unease, negative



## Where we left off

- S02E08: <u>article</u> | <u>slides</u>
- Some behavioural evidence for UV
- •But what about neural mechanisms?

1.  $3^{rd}$  degree model; converging with Mori's UV 2. While emotion  $\rightarrow$  likeability, no effect of emotion of fit

### Experiment 1B - Results





# Aims of the study

 Does a linear-to-nonlinear transformation underlie the uncanny valley? Linear: humanlikeness Nonlinear: likeability

•What is the role of 'social' brain regions (e.g. TPJ, DMPFC, VMPFC)

Three questions:



#### Behavioral/Cognitive

#### Neural Mechanisms for Accepting and Rejecting Artificial Social Partners in the Uncanny Valley

<sup>©</sup>Astrid M. Rosenthal-von der Pütten,<sup>1,2,3</sup> Nicole C. Krämer,<sup>1</sup> Stefan Maderwald,<sup>3</sup> Matthias Brand,<sup>3,4</sup> and <sup>©</sup>Fabian Grabenhorst<sup>5</sup>

<sup>1</sup>Social Psychology: Media and Communication, University Duisburg-Essen, 47048 Duisburg, Germany, <sup>2</sup>Individual and Technology, RWTH Aachen University, 52062 Aachen, Germany, <sup>3</sup>Erwin L. Hahn Institute for Magnetic Resonance Imaging, 45141 Essen, Germany, <sup>4</sup>General Psychology: Cognition and Center for Behavioral Addiction Research (CeBAR), University Duisburg-Essen, 47048 Duisburg, Germany, and 5Department of Physiology, Development and Neuroscience, University of Cambridge, CB2 3DY Cambridge, United Kingdom



6

- Final sample: n = 21
- (108 trials)
- Tasks:



rating task

#### • Design: 6 runs (they call it sessions) with random order for task-rating (72 trials) and task-choice

•Cover story choice task:

gifted to the participants at the end of the study

previously chosen present"

Human no physical vs. other agents Android vs. other agents

"nobody explicitly stated that they did not believe the cover story"

- -participants told that all agents chose one item ("movie theater voucher, a package of dishwasher tabs, a bottle of sparkling wine, and a package of quality toilet bowl deodorizer blocks") that will be
- -participants' task was to decide between a person or a robot "from whom you prefer to receive the







•Stimuli:

**MechR**: mechanoid robots; HumR: humanoid robots; **AndR**: android robots; **ArtificH**: artificial humans; HumPhys: humans with physical impairments; HumNPhys: humans without physical impairments

Mechanoid and humanoid robots from Rosenthal-vor der Pütten & Kramer (2014)



•Stimuli:

**ArtificH**: artificial humans;

(dramatic lighting, and reduced colouring).

evaluated"

Focused on comparison between these and other agents (not androids) – Pollick, 2010



- Based on Toledano's 2011 "A new kind beauty". Recreated using Toledano's pictures as reference
- "[...] resulting in full body images of humans who share some irritating features: reduced coloring which resulted in lightgray complexion, mismatches in the proportion of head and body, exaggerated facial features (due to plastic surgery). In total, nine synthetic 1 humans (four female, five male) were
- Piloted on likable, unpleasant, familiar, **uncanny**, intelligent, disgusting, humanlike, and attractive.



•fMRI:

= 10

Standard preprocessing, 6 mm smoothing, leave-one-subject-out method for ROI analysis

and confidence, and choice > rating.

GLM2: focused on regions related to choice-task activity considering humanlikeness

GLM3: focused on regions related to agent detection (human > artificial agent)

Show me some nice graphs!



- 7T, with EPI sequence (TR = 2000ms, TE = 22 ms, 1.51 mm slice thickness, 144 coronal slices) 12 min runs, 331 – 343 volumes per run (2022 volumes in total), FWE or SVC < .05, initial p < .005, k
- GLM1: focused on regions related to likability, familiarity, and humanlikeness; main decision variable







- There is a humanlikeness continuum: r = .98, p = .0006, linear regression Difference in likability (eta<sup>2</sup> = .70), familiarity (eta<sup>2</sup> = .73), and humanlikeness (eta<sup>2</sup> = .86)
- 2.

Can humans also fall in UV? ← Frank Pollick's hypothesis



1. Likability: Cubic polynomial fit ( $R^2 = .57$ ), also for individual data ( $R^2 = .36 \pm .03$ ) 2.17 of 21 show deviation from linear fit of likability ratings for artificial humans (familiarity weaker)





#### **Activity in VMPFC:**

- a. associated with trial-by-trial likability ratings (onset regressor modulated trial-specific likability rating whole brain)
- (different activity-component) associated with humanlikeness (ROI regression) reflects explicit UV reactions; activity for artificial human deviates from linear fit (p = .008)
- b. С. d. VMPFC UV depth is associated with behavioural UV depth (r = .57, p = .006)



### rTPJ

- reflect humanlikeness on a trial-by-trial basis (whole-brain) a.
- TPJ activity associated with humanlikeness (ROI regression) b.
- Activity in TPJ reflects humanlikeness in a linear fashion С.
- d. Stronger association of TPJ with humanlikeness is correlated with behavioural UV depth





#### **DMPFC:**

- е.
- ROI regression: human detection best explains DMPFC activity (not humanlikeness) f.
- More activity in DMPFC for humans compared to artificial agents g.
- h. Better fit of DMPFC activity on human detection  $\rightarrow$  stronger behavioural UV effect

More activity in DMPFC for humans vs. nonhumans (MechR, HumanR, AndR and ArtificH - WB)





#### FFG:

- Negative relationship between FFG activity and humanlikeness (WB) Ι. Activity reflects both humanlikeness and human detection (ROI regression) More activity for mechanical and humanoid robots compared to other categories k. FFG differential humanlikeness (sensitivity) (?) correlated with behavioural UV dept





**Does a linear-to-nonlinear transformation reflect UV?:** a. TPJ (linear) \* FFG (inverse linear, but selective for nonhumans) = VMPFC (BOLD signal) b. and c. VMPFC UV depths are correlated with TPJ and FFG regression fits DMPFC?

### **Results - Decision**



b.

Participants prefer humans over artificial agents, more variability for artificial agent comparison

# **Results - Decision**



- c. Relative likability, familiarity and humanlikeness of agent influences the decision; relative ratings classified choices correct on 85.7% of the trials.
- Weighted sum of relative rating predicts choices consistently d.
- e. Highest difference between mechanical robots and humans, however UV effect also visible for artificial humans (deviation from linear line for artificial humans)
- f. Consistent across participants





- a. VMPFC associated with decision variable of participant (WB)
- b. ROI regression: VMPFC activity fits decision variable and confidence during decision
- c. Activity also reflects UV effect; UV depth visible for artificial humans
- d. VMPFC UV depth for choice task is correlated with choice UV depth

### **Results - Decision**



- e.
- f.
- TPJ reflects humanlikeness (WB) g.
- ROI regression showing fit of ROI activity on humanlikeness h.
- TPJ humanlikeness fit is correated with choice UV depth (not shown)

More activity in DMPFC and VMPFC for choices with humans vs. choices without humans DMPFC codes decisions for nonhumans, while VMPFC codes for decisions involving humans

FFG: similar to rating, negative humanlikeness association and human-detection (not shown)

## Discussion

Three questions:



is there a neural 'representation' of a subjective UV reaction? Yes, activity in the VMPFC corresponds to a nonlinear shape overlapping with the UV

- for robots (linear up till human agents), VMPFC: nonlinear

does this map onto perception and decision making? Yes, UV reactions are generalizable across tasks; data for rating and choice showed UV effects

### is there a differentiation between linear and nonlinear regions? Humanness vs. likability

Yes, TPJ: linear humanlikeness, DMPFC: human/nonhuman detection, nonlinear, FFG: selective





#### Why select artificial humans as the critical condition?



#### Modeling the UV

To examine neural correlates of UV reactions, it was critical to first model and quantify the UV psychometrically within individual subjects. As we observed the most pronounced UV effect for artificial humans, we focused on this stimulus category (see Materials and Methods; including android stimuli yielded similar results).



#### What is the shape of the UV?



Mori (1970)

Rosenthal-von der Pütten et al. (2019)

Mathur & Rechling (2016)



#### The uncanniness of the uncanny valley What is it?

Affinity (Shinwakari)

- Likability
- Threat
- Uncanny
- Eeriness

#### **Measurement practices (Flake and Fried, preprint)**

External, internal and construct validity **Prediction and relevance for behaviour** 



The future has agency and experience: Henschel et al. 2020; Hortensius et al. in prep



#### **Emerging evidence on FFG reactivity**



#### Activation in object-specific brain regions during engagement with robots





Henschel\*, Hortensius\* & Cross (2020) TINS

#### **Role of TPJ and other regions?**



- O Beliefs and expectations
- Trust toward artificial agents
- O Dehumanizing
- Anthropomorphism
- Self/other identification with avatars
- Theory-of-Mind network (meta-analysis)

Hortensius & Cross (2019) NYAS





#### Analyses

- How robustness are the analyses? How straightforward are their choices? Discovery sample  $\rightarrow$  Replication sample; more formal confirmatory analyses Sample size (but 7T)
- Data, code and material missing (paper was published in 2019)





#### Uncanny but not confusing: Multisite study of perceptual category confusion in the Uncanny Valley

Mathur Maya B.<sup>a,b,\*</sup>, David B. Reichling<sup>c</sup>, Francesca Lunardini<sup>d,1</sup>, Alice Geminiani<sup>d,1</sup>, Alberto Antonietti<sup>d,1</sup>, Peter A.M. Ruijten<sup>e,1</sup>, Carmel A. Levitan<sup>f,1</sup>, Gideon Nave<sup>g,1</sup>, Dylan Manfredi<sup>g,1</sup>, Brandy Bessette-Symons<sup>h,1</sup>, Attila Szuts<sup>i,1</sup>, Balazs Aczel<sup>i,1</sup>



#### link



#### But I have to ask...

How relevant is the UV? For long-term embodied interactions? Is it temporary? Predictive coding account? Updating priors? Or as I have argued and as we also state in our recent TINS paper:

Uncanny valley hypothesis: humans prefer anthropomorphic agents, but reject them if they appear too humanlike. To what extent the uncanny valley is an artefact of contemporary experimental procedures remains unknown.

## Thank you!

