



## Reanalyzing published science

The power and perils of commentary articles as an early career researcher

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#### A tale of two papers



#### Cognition

Volume 227, October 2022, 105213



# Conceptual alignment in a joint picture-naming task performed with a social robot

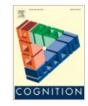
Giusy Cirillo <sup>a b</sup> △ ⊠, Elin Runnqvist <sup>a b</sup>, Kristof Strijkers <sup>a b</sup>, Noël Nguyen <sup>a b</sup>, Cristina Baus <sup>c</sup>

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Cognition

Volume 259, June 2025, 106099



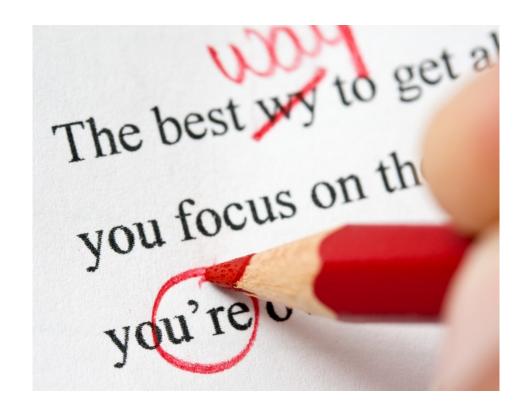
## Linguistic alignment with an artificial agent: A commentary and re-analysis

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## What I am NOT going to do in this talk

- Deep dive into all the specific methodological issues of the target paper and the details of our commentary (only main points)
- Focus on the details of the statistical analyses
- Lecture you on mistakes and how to revolutionize scientific publishing

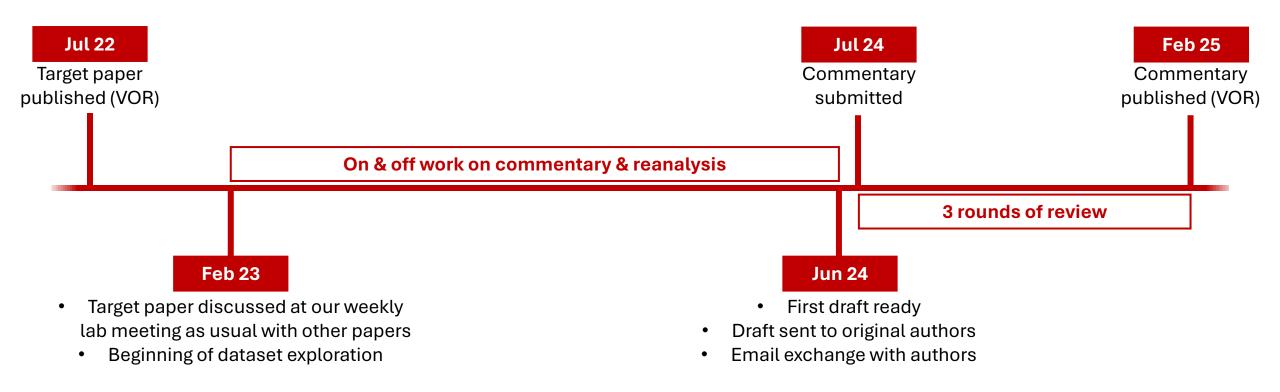


## What I AM going to do in this talk

- Focus on why we worked on a reanalysis & commentary paper
- Focus on *how* we proceeded and how we communicated with the authors
- Propose that *Early Career Researchers* should make commentaries and re-analyses
- Ask you about your opinion on and your experience with commentaries



#### A timeline

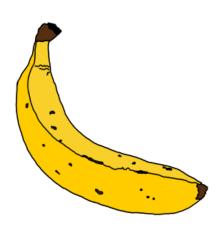


- Joint picture naming task (in French)
- Human social robot partner (Furhat Robotics)
- Do participants (N = 24) align with the naming behavior of the robot?

#### **BASIC CONDITION**

For the category "fruit", the robot uses basic labels. For 10/15 categories (mixed); 3600 trials in total.







- Joint picture naming task (in French)
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#### **CATEGORY CONDITION**

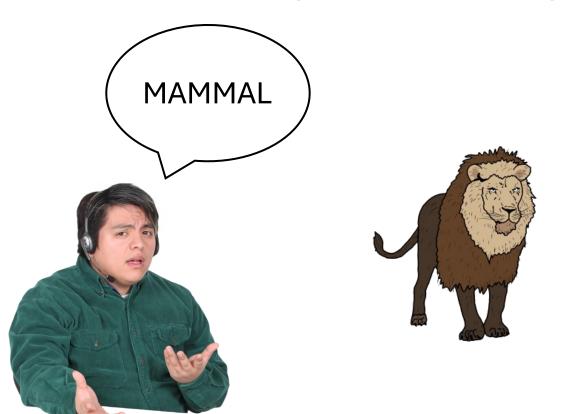
For the category "mammals", the robot uses category labels (superordinate). For 5/15 categories (mixed); 1800 trials in total.







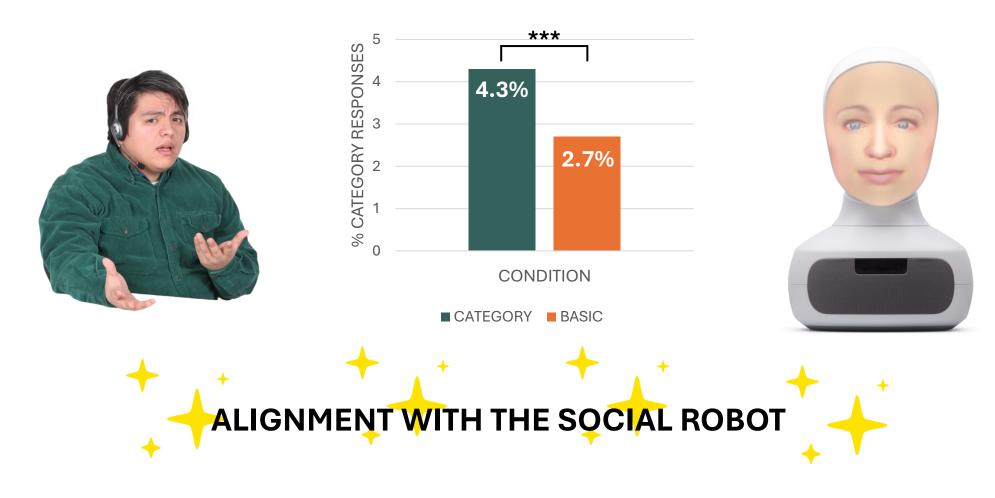
- Joint picture naming task (in French)
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#### % category responses in CATEGORY condition > BASIC condition

(GLM binomial family, random intercepts for participants and items)



#### Complexity remains hidden under a blanket

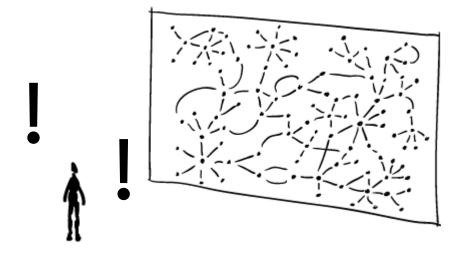
- ❖ 15 categories, 4 exp. blocks, lexical frequency, MultiPic (validated database)/new pictures: statistics is blind to all this complexity (if not/cannot be modeled)
- No visualization of **how responses are distributed**, despite category responses overall being very few to begin with (hence a rare naming behavior in the exp)

To assess how reliable and consistent the effect is, data need to be thoroughly explored

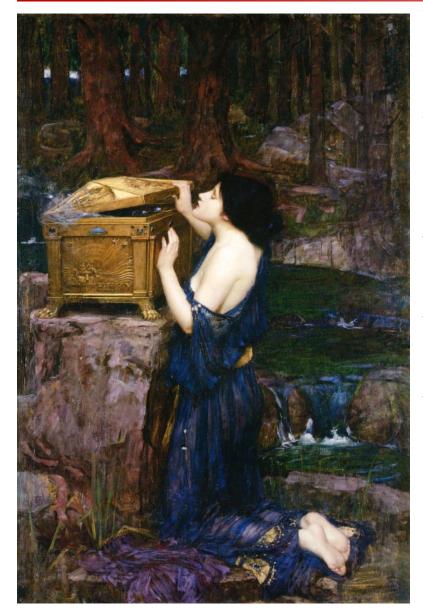
## Praised be open data!



Possibility for the scientific community to explore them and uncover (and appreciate) complexity

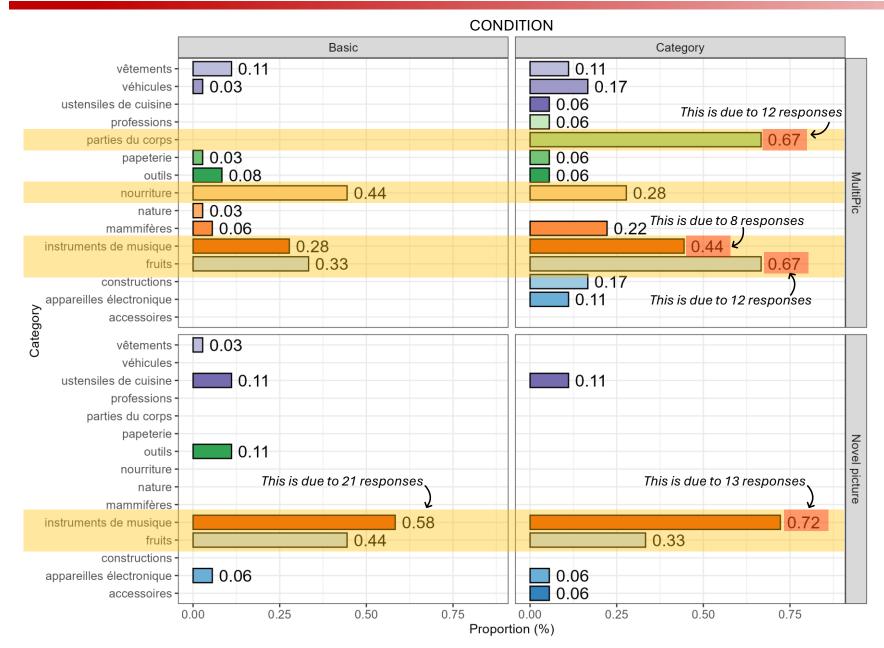


## Opening Pandora's box & uncovering complexity



- Analysis of response times to assess the claim of automaticity (not the focus here; see the papers)
- **Extensive data visualization** across many variables
- Check for lexical frequency biases in categories
- Multiverse approach and robustness check: how category-dependent is the effect?

#### Category responses cluster in some categories and new pictures



- 377 picture stimuli from the MultiPic database
- ❖ 73 picture stimuli were newly designed (novel pictures, not shared on OSF)
- ❖ 5400 trials in the exp.

#### # RESPONSES

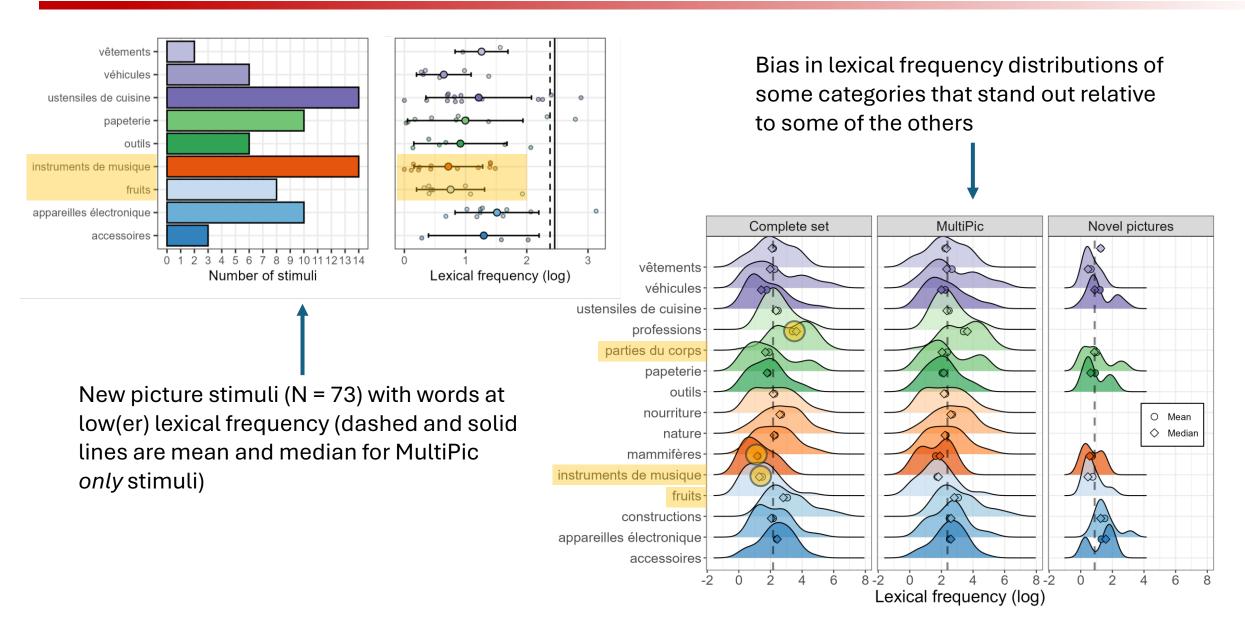
	BASIC	CATEGORY	
ALIGNED	2862	78	
NON-ALIGNED	98	1467	
0/ PECPONICES			

#### **% RESPONSES**

	BASIC	CATEGORY
ALIGNED	79.5%	4.3%
NON-ALIGNED	2.7%	81.5%

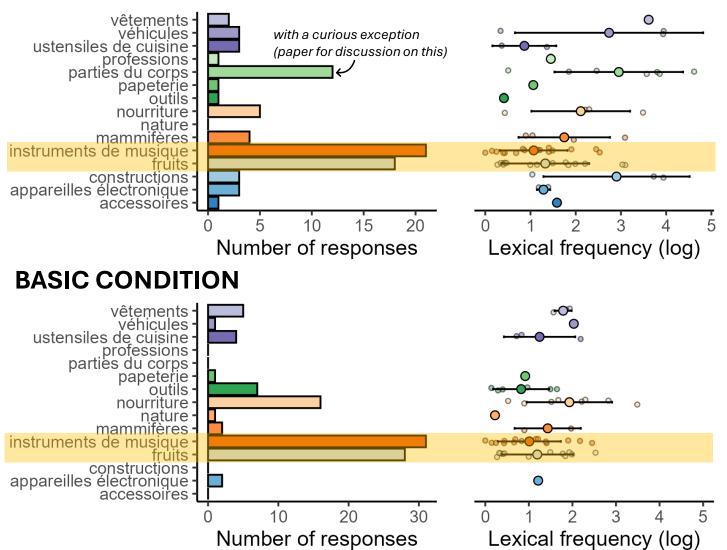
"Pure" alignment or non-alignment, excluding a variety of naming errors (N = 750)

## Newly developed figures introduce biases in lexical frequency



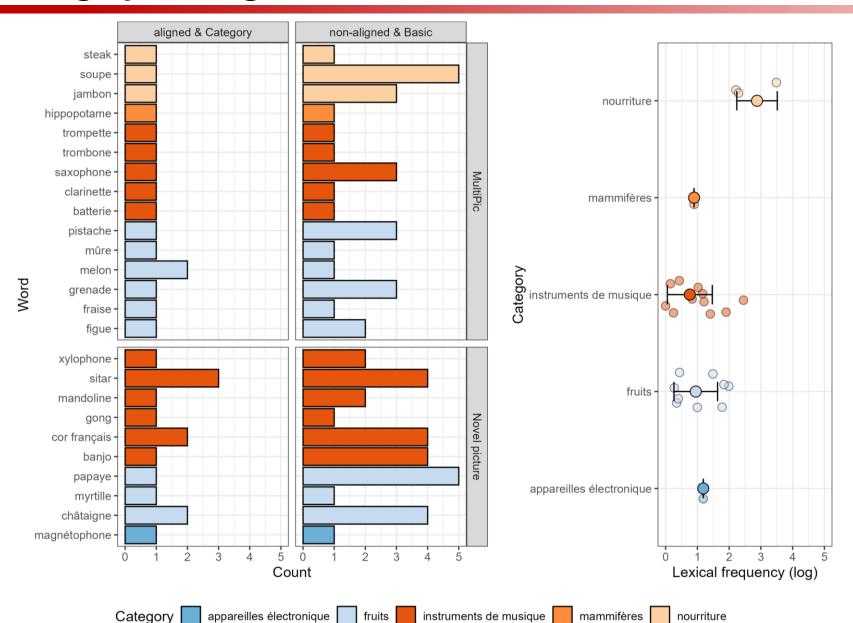
#### Category responses cluster primarily in low lexical frequency

#### **CATEGORY CONDITION**



In a naming task in which I learn that I am allowed to say the category, maybe I say the category more easily when I can't retrieve the basic label (and not because I align with the robot).

#### Category naming bias for a subset of items across conditions

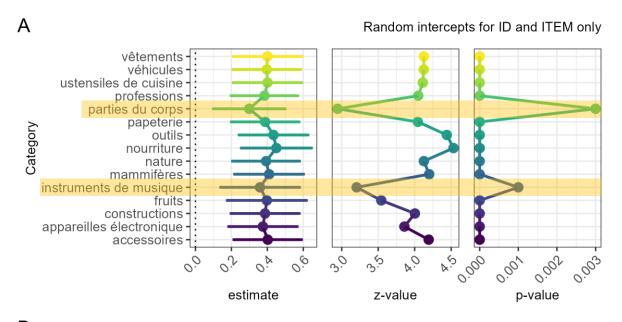


There is a bias for some items (N = 15) belonging to certain categories of being named with the category:

- low lexical frequency
- possibly unclear picture?
- low familiarity? (trompette, trombone, clarinette, sitar, cor français...) – no databases/ratings

In this context, a slight increased number of category responses in the category condition can greatly impact the statistics. But the effect cannot be said to be robust, reliable, and generalizable!

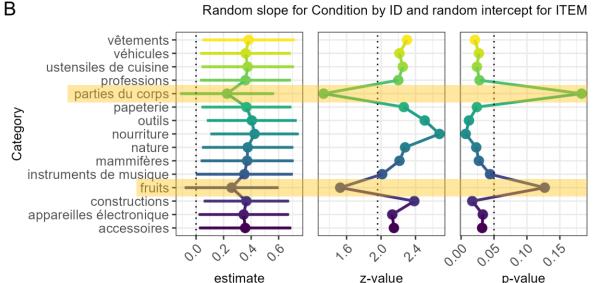
#### The effect is not robust and depends on category and participant



Leave-one-out approach for robustness check

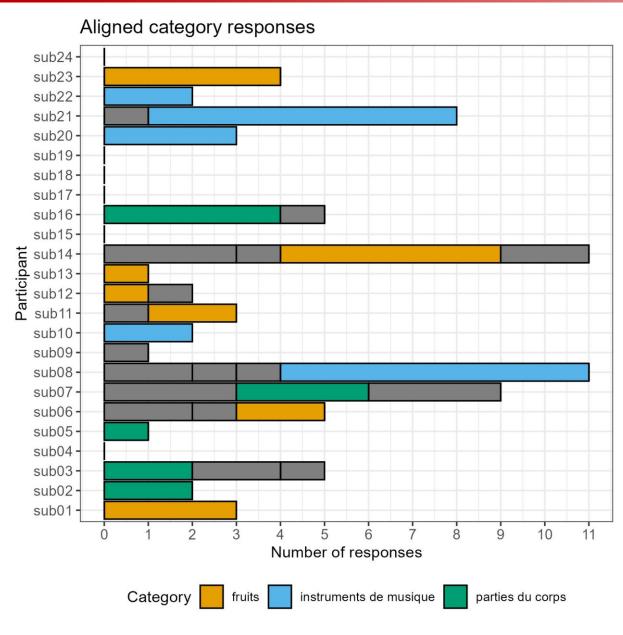
The same influential categories:

- body parts
- fruit
- musical instruments



The effect also varies by participant (remember that participants had different categories in the category/basic condition!)

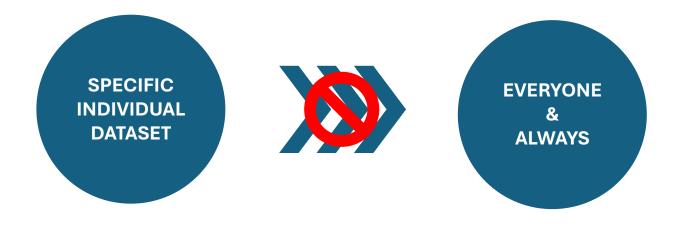
#### Few responses, for few categories, for few participants



- 50% of aligned category responses
   (N = 39) are given by 4 participants
- 65% of aligned category responses
   (N = 51) are given for 3 categories
   (body parts, fruit, musical instrument)

#### The curse of Pandora's box: limited generalizability





The effect is due to **idiosyncrasies** in the experiment. Stimuli are crucial! Data needs to be thoroughly understood!

Influential variables influencing generalizability:

- Category
- Picture origin
- Lexical frequency
- Individual pictures' familiarity / recognizability (probably)
- (Participant)

#### The drive behind our commentary: open science principles

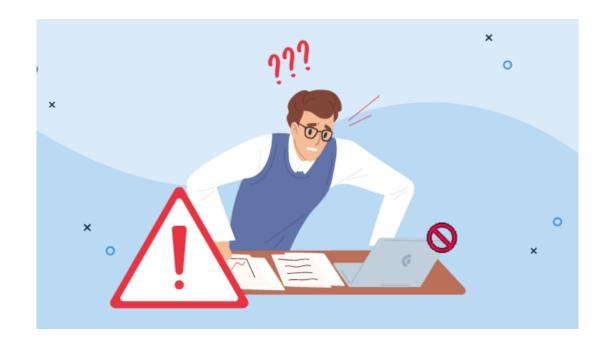
- Genuine scientific interest in the phenomenon + open data (kudos to the authors) that should serve a purpose: be explored by other researchers
- ❖ Data is complex and statistics may be blind to some complexities (if not modeled or when it can't be modeled; too many variables and few datapoints)



- Clarify the nature of an effect that was being cited as generalizable and given for granted
- Discuss limitations and possible errors with no ethical stigma, especially for ECRs
- But also: a bit worried about possible negative effects on us, but also on the first author of the original paper (an ECR like us!)



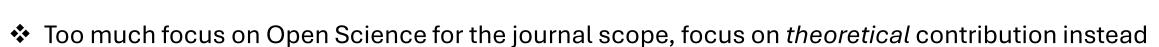
- We are afraid of making mistakes
- We are terrified of other people's finding out possible mistakes in our work = stain in our career
- We should be free of making genuine mistakes and acknowledge limitations



- Published science is not carved in stone but should be submitted to a continuous process of collective correction
- Making mistakes is **not** committing a fraud

#### The peer review process

- 8 months, 3 rounds: is it common for commentaries?
- "Too harsh", "too focused on picking on flaws", "should highlight more the positive aspects of the original work"...
- "It appears the original work offers no insights, this should be evidenced more strongly"
- Discussing construct validity is not relevant
- Too many supplementary figures



- Decision to eliminate "collaborative open science" from the title
- Gradual limitation of OS first in Introduction and Discussion, then only in Discussion in a dedicated small paragraph



## Did we approach it in the best way?

- We emailed the authors when we had a full manuscript and reproducible scripts; criticism: we should have contacted them earlier
- Informally, one of the senior authors knew we were working on a commentary (not informed by us and without us knowing beforehand); the other authors did not know until they read our email
- Availability on our part to modify before submission any sections in which we may have incorrectly reported their work, request to check fair reporting; denied, perceived by the authors as a request for thorough review of the work in little time
- New stimuli & naming data not shared: seen as an accusation, would have shared if contacted earlier
- Unfortunate timing: the first author was on maternity leave and rightfully had other priorities; however, none of the coauthors intervened to provide any feedback either



- Criticism that this is not collaborative open science
- Ultimately, the commentary and our emails were not well received, and our work was dismissed as not worthy of their time because of other priorities (especially by a senior author emailing us without including their coauthors: "my co-authors and I are all tied up with more pressing matters")

#### The power and perils of commentaries as ECRs



- Contribute to clarify & deepen understanding of existing findings
- Offer new ideas for better constructs and measures
- Promote a new approach in facing uncertainty in a collective way
- Boost the usefulness of open data (= real open science)



- Marked in the black book (or Deathnote) of someone
  You can even be blocked on X/Twitter! Don't tell anyone about this
- Risk of negative reactions due to a possible sense of personal attack (we tend to identify with our papers)

#### I want to hear from you now



- What would you have done differently from us?
- What is your experience with commentaries?
- Have you ever been the target of a commentary?
- Can we make commentaries less adversarial and more collegial (our attempt aimed at this direction)?

#### **Readings & Resources**

- Cirillo et al. (preprint; v1 Apr 21; v2 Dec 23): https://doi.org/10.31234/osf.io/cjy24
- Cirillo et al. (published, Cognition, Jul 22):
  <a href="https://doi.org/10.1016/j.cognition.2022.105213">https://doi.org/10.1016/j.cognition.2022.105213</a>





- Original paper OSF repository: <a href="https://osf.io/f6gu3/">https://osf.io/f6gu3/</a>
- Gastaldon & Calignano (preprint; v1 Jul 24; v2 Nov 24): https://doi.org/10.31234/osf.io/v9ufk
- Gastaldon & Calignano (published, Cognition, Feb 25): https://doi.org/10.1016/j.cognition.2025.106099
- Commentary OSF repository: <a href="https://osf.io/yeqgp/">https://osf.io/yeqgp/</a>

MultiPic: A standardized set of 750 drawings with norms for six European languages

MultiPic: paper

(https://doi.org/10.1080/17470218.2017.1310261) & database (https://www.bcbl.eu/databases/multipic)

