



Going beyond conventional web surveys: Opportunities and challenges of using new types of data within the frame of web surveys

HERSS Summer School

Hannover, 11 September 2024

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Acknowledgments:

This project has received funding from the European Research Council (ERC) under the European Union's Horizon 2020 research and innovation programme (grant agreement No. 849165).

I want to thank Oriol Bosch, Patricia Iglesias, and Carlos Ochoa for all their helpful feedback and inputs.

Why is there an interest in using new types of data?

Importance of (web) surveys

- Most frequently used method for collecting data in many disciplines
 - Sociology: 69.7% of the published articles use survey data
 - Political sciences: 41.9% (Saris & Gallhofer, 2007)
- Web surveys: more and more common nowadays
 - 35% spent on research using (mobile) web, *vs* 11% for telephone and 8% for face-to-face (ESOMAR, 2019)
 - With pandemic, switch from other modes to web mode even quicker
- Results potentially used by key actors to take decisions

Problem: surveys suffer from errors

- Both on **representation** and measurement sides (TSE framework)



Final sample \neq target population \rightarrow

Lot of surveys (especially web) use nonprobability-based samples

Even when probability-based sampling is used, response rates have been going down drastically in most countries \rightarrow possible selection bias in who participate

Weighting can sometimes be used but is often not sufficient

Problem: surveys suffer from errors

- Both on representation and **measurement** sides (TSE framework)



To err is human

People do not know everything surveys ask about

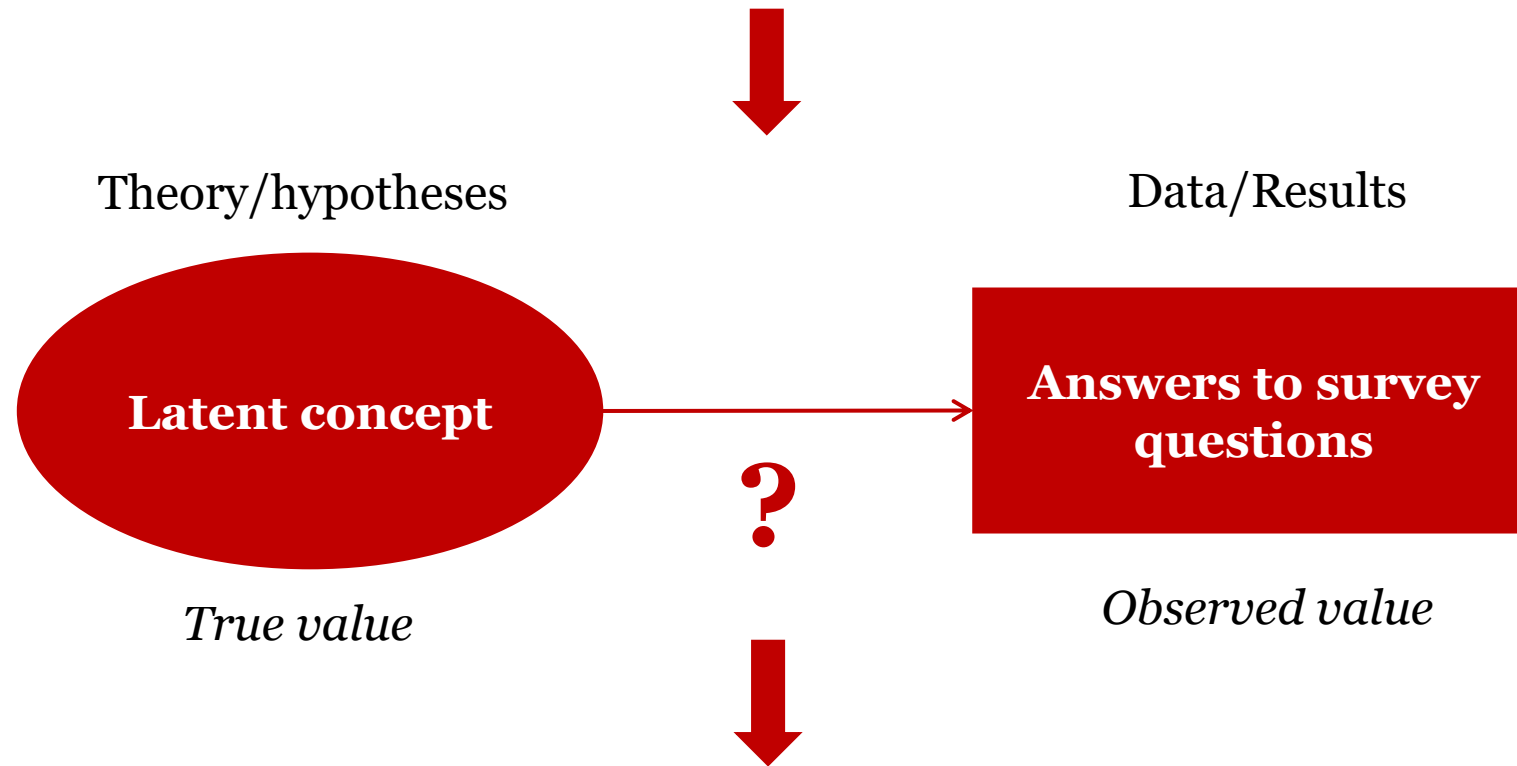
Remembering-self \neq experiencing-self (Kahneman & Riis, 2005)

Lack of effort / satisficing

Social desirability

Problem: surveys suffer from errors

- Both on representation and **measurement** sides (TSE framework)



Average **measurement quality** for 67 ESS questions across up to 41 country-language groups (Poses et al. 2021) = **0.65**

Measurement errors in surveys

- This gap between observed and true values can affect the results substantially
- Crucial to consider measurement errors

Table 6: Estimates of the parameters with and without correction

	Without correction On Allow immigration	With correction On Allow immigration
By		
Better life	-.265*	-.609*
Economic threat	-.133*	.001
Cultural threat	-.154*	-.140*
Total explained (R ²)	.254	.547

 Wrong conclusions

Source: Saris & Revilla, 2016

Overall, need to improve quality of (web) survey data

- But... How?
 - Need for improvement has been clear for decades
 - Lot of knowledge already on survey errors
 - How to reduce + correct for them (see e.g., the work of Willem Saris or Duane Alwin)
 - Lot also known about web surveys (e.g., Couper 2008; Tourangeau et al. 2013)
 - But still large errors, especially on the measurement side
 - **What else can we do?**

How could we enhance or extend
web survey data?

HOW COULD WE ENHANCE?

Main idea that we will discuss

Taking advantage of **new measurement opportunities linked mainly to the growing use of smartphones** to reduce measurement errors in web surveys

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Smartphones are **everywhere**

More people have smartphones than toilets worldwide¹ 🤔

Including in **web surveys**

On average, Millennials answer **79%** of the surveys using smartphones and Boomers **36%** (US Netquest panel 2017/2018; Bosch et al., 2019)

➔ **Create both new challenges and new opportunities**

¹<https://www.globalcitizen.org/en/content/access-denied-toilets-Harpic-Waterorg-RB/>

- Opportunities at different levels

- Phone number: can be used to **contact** respondents

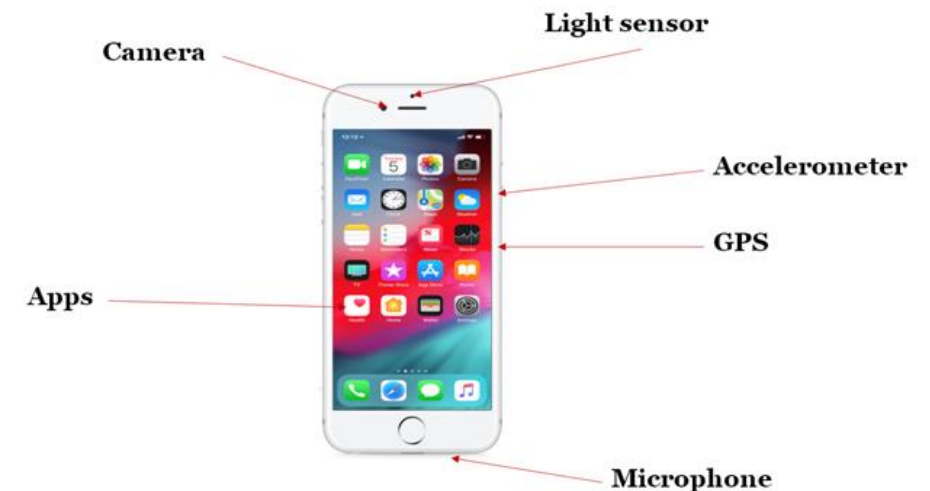
- E.g., through SMS, sending a link to the survey, allows random digit dialing (RDD)

- People take/use their phones everywhere

- Possible to contact them **anytime/anyplace**
- Invitations can be seen very quickly after being sent

- Possibility to collect **new data types**

- Sensors + apps → different types of data



Which new types of data?

- Data collected **passively**
 - Participants only need to accept to share such data and/or set up a tracking app
 - Mainly **digital traces data** = records of activity undertaken through an online information system, including digital footprints left behind by users as they interact with technology (Howison et al., 2011)
 - Ex: browsing history, search queries, social media interactions, GPS data, app usage...
 - Allows studying online behaviors, traveling patterns, etc.

Most of those data can also be collected for PC!

- Data collected **actively**
 - Respondents need to actively provide these data

New data types considered in the WEB DATA OPP project

VISUAL DATA



Screenshots
Photos/videos taken during the survey
Visual files saved on (or accessible from) the device

VOICE DATA



Dictation
Voice recording

4 new types of data

METERED DATA



Obtained through a tracking application (“meter”) installed by the participants on their devices to register at least the URLs of the webpages visited. Usually collected in metered panels.

GEOLOCATION DATA



Obtained through a tracking application installed on participants’ mobile devices to register at least the GPS coordinates

IN-THE-MOMENT SURVEYS triggered by such data

How could these new data types help?

Benefits expected only for some concepts, not all!

Expected **benefits** (Revilla, 2022)

Social desirability bias

	EUR
Banana	1,44 A
1,148 kg x 1,25 EUR/kg	
Freshona/Espinacas	1,15 A
Vemondo/Bebida soja 0%	1,60 B
2 x 0,80	
Vemondo/Tofu ecológico	0,95 B
Mandarina Ebre	2,79 A
Dentalux/Crema dental	0,95 C
Chef Select/Trio de humm	2,19 B
Edulis/Ensalada dúo	1,15 A
Alesto/Mezcla frutos sec	1,89 B
Floralys/Servill 2capas	0,95 C
Favorina/Huevos chocolat	1,49 B
Champiñón	0,65 A
Huevos L suelo	1,79 A
Floralys/Papel higiénico	2,55 C
Total	21,54

Researchers

- Reduce some of the issues related to measurement errors

Information people do not know



Mistakes and satisficing



Expected **benefits** (Revilla, 2022)

Researchers

- Reduce some of the issues related to measurement errors
- Provide data for new concepts (not measured so far)
- Massive amount of data / granular data
- Real time / continuous (passive data)
- Answer new research questions

Nutrition Facts	
Chicken with Mushroom Gravy	
Serving Size: <input type="text" value="1"/> Serving (328g)	
Amount Per Serving	
Calories 398	Calories from Fat 155
% Daily Value*	
Total Fat 17g	26%
Saturated Fat 7.8g	39%
<i>Trans</i> Fat 0.4g	
Polyunsaturated Fat 6g	
Monounsaturated Fat 1.8g	
Cholesterol 152mg	51%
Sodium 730mg	30%
Potassium 569mg	16%
Total Carbohydrates 8.5g	3%
Dietary Fiber 0.9g	4%
Sugars 0.7g	
Protein 50g	
Vitamin A	6.3%
Vitamin C	2.3%
Calcium	1%
Iron	15%

* Percent Daily Values are based on a 2000 calorie diet.

Expected **benefits** (Revilla, 2022)

Researchers

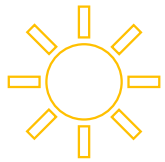
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- Real time / continuous (passive data)
- Answer new research questions

Participants

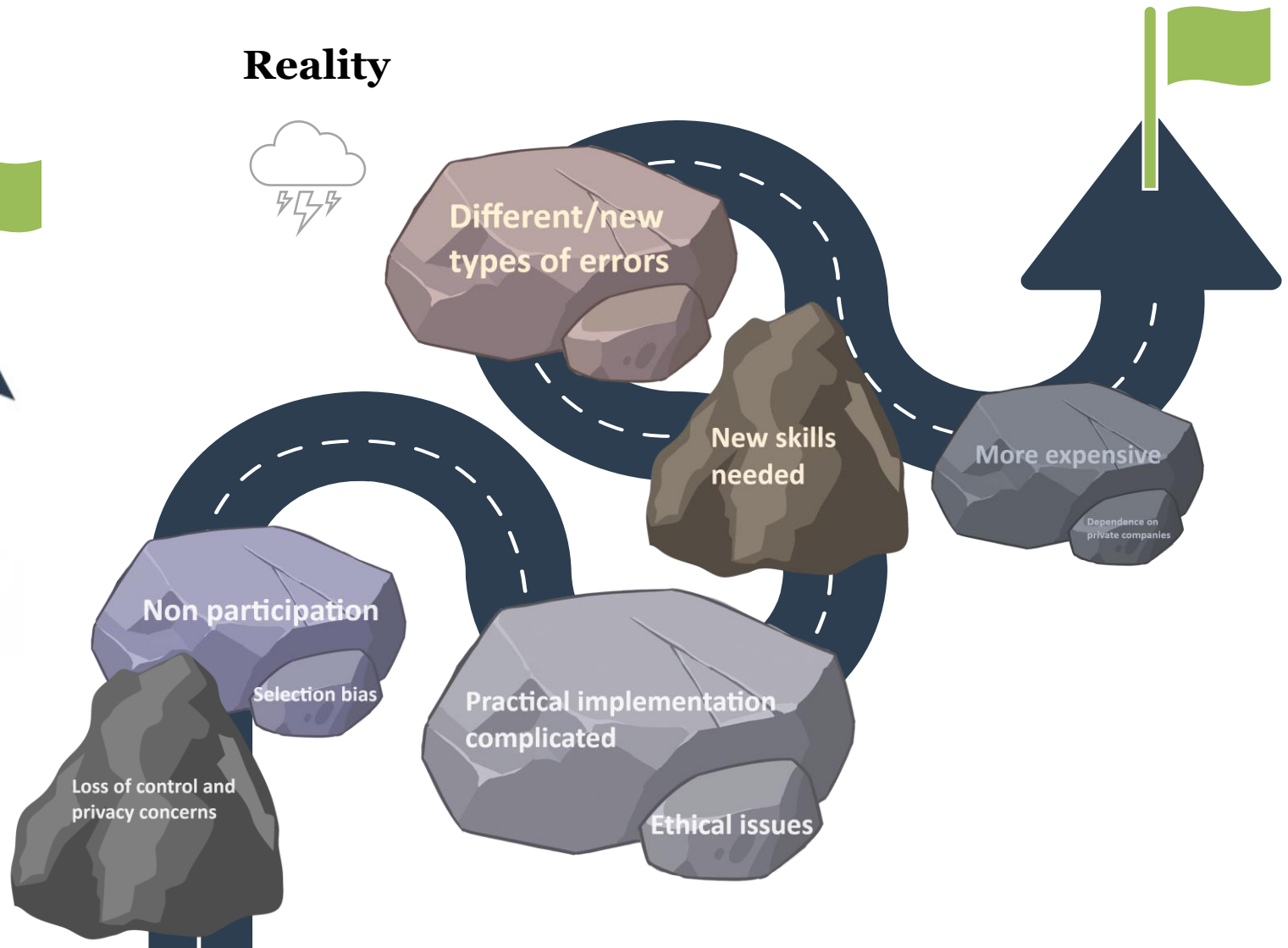
- Reduce time dedicated to provide information
- Reduce efforts
- More enjoyable

But this is not that easy...

What most people think



Reality

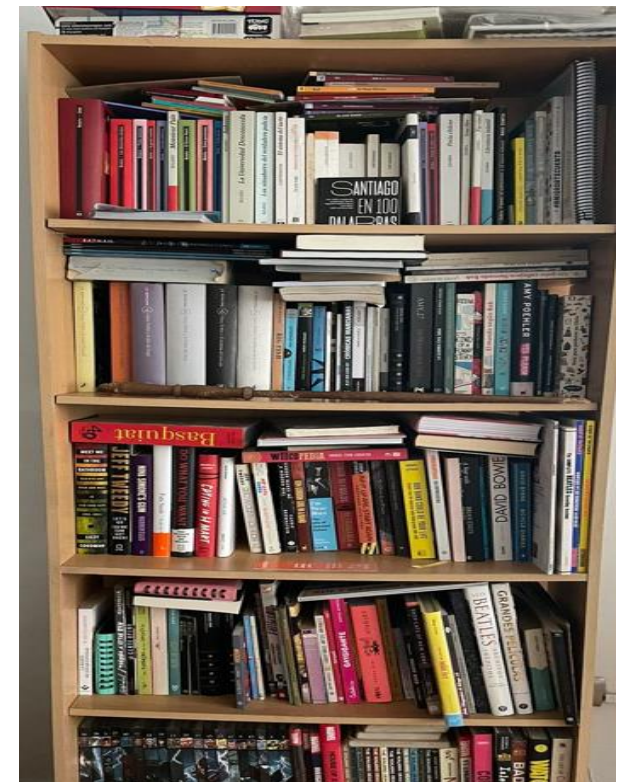


Examples of research with the new types of data

Potential benefits of asking for photos of the books at home

- Number of books often used as indicator of cultural or economic capital
- But people do not know how many books they have
- Social desirability bias expected → over-reporting
- Kind of books also matter (cooking vs history books)
- Asking for photos of the books has the potential to:
 - Provide more accurate information about the number of books
 - While also providing extra information (kind of books, language, storage, etc.)

A picture is worth
a **thousand** words



Lot of challenges

How can the information be extracted from the images?

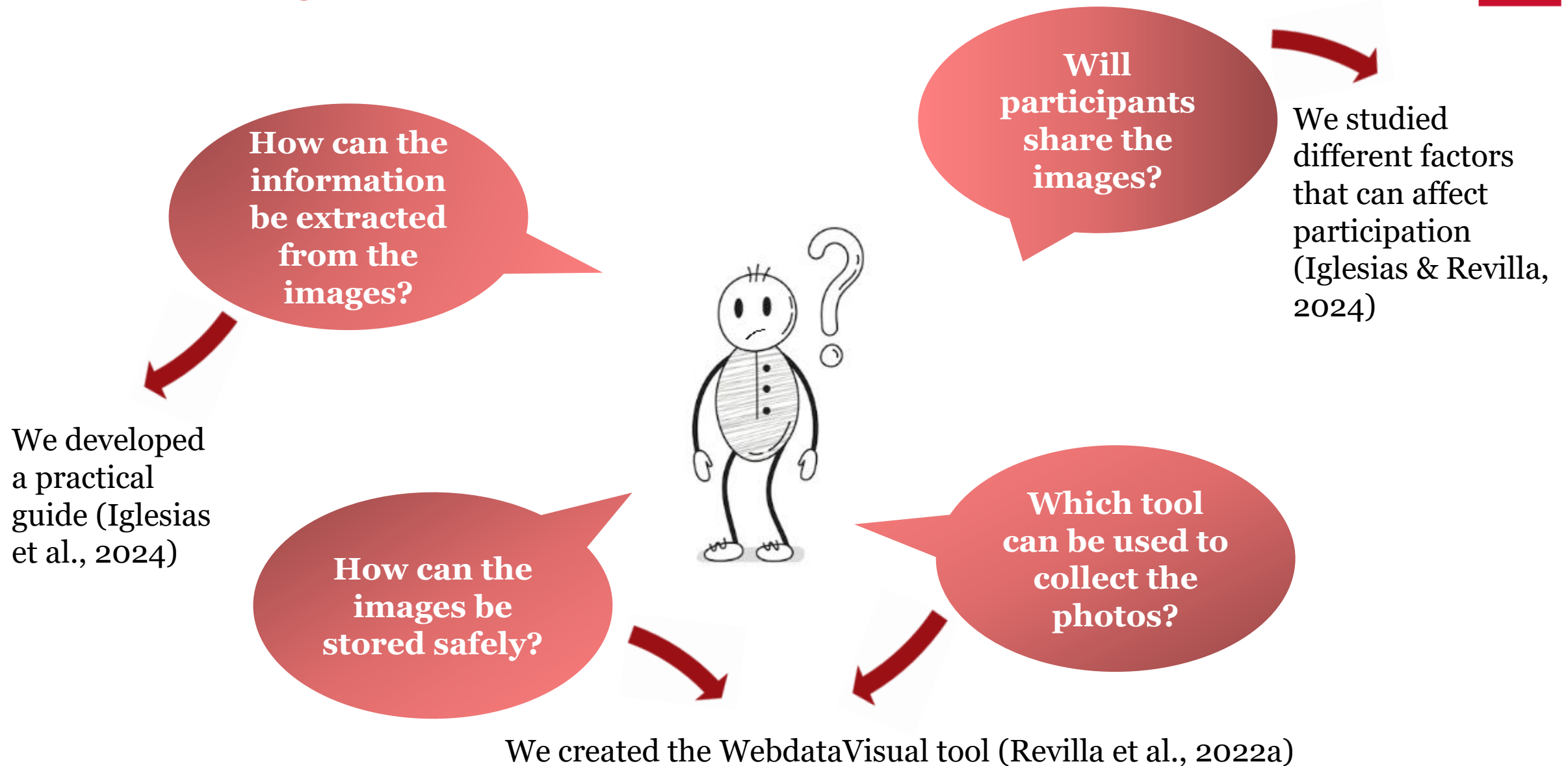
How can the images be stored safely?



Will participants share the images?

Which tool can be used to collect the photos?

Lot of challenges

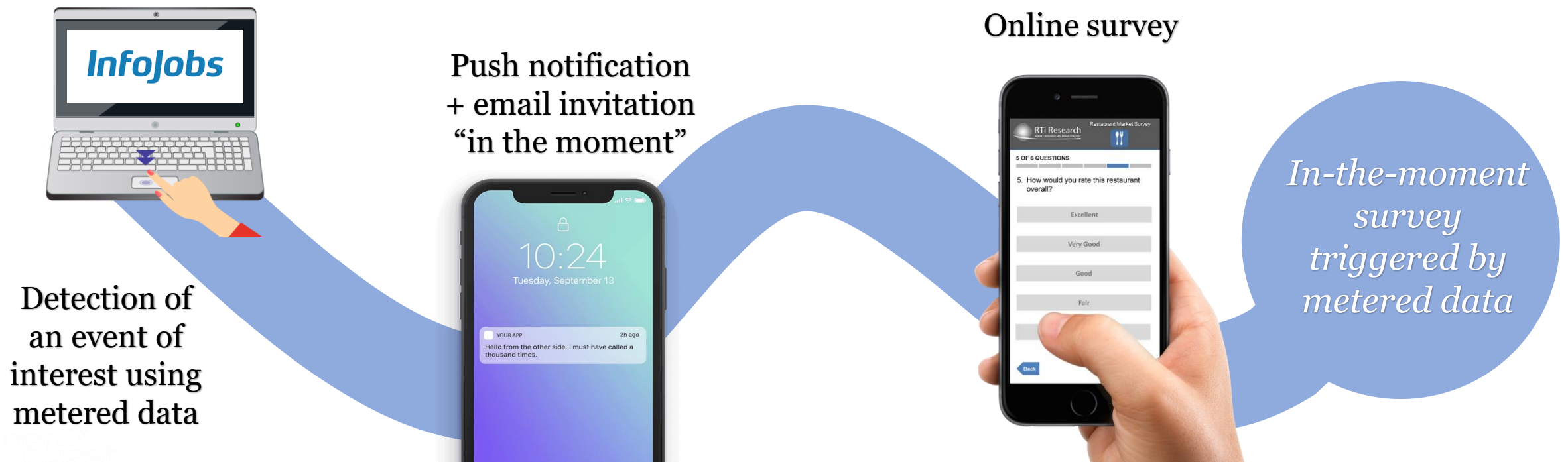


Main learnings from the books-at-home experiment

- Participation was **lower** than expected
 - 66% out of 703 participants asked for photos did not send a photo or left the survey
- Classification was very **challenging**
 - Problem of overlap
 - Problem to differentiate books from CDs, DVDs...
 - Problem to differentiate between illiterate and literate children's books
- Extra-information was available but not systematically
 - Languages of the books, titles, authors, etc, were visible for some books but not all
 - So difficult to use that in quantitative analyses
- At this day, photos can **complement** more than replace

Potential benefits of using an in-the-moment survey

- Reduce the **time** between an event of interest and the questions about this event
- Potential for decreasing **recall errors**, thereby enhancing data quality
- Potential new insights as well



Lot of challenges

Is the quality
of the data
really
improved?

Are panellists
willing to
participate?

Will they see
the invitation
quickly
enough?



How to invite
them really
in-the-
moment?

Lot of challenges

We implemented an experiment (Ochoa, 2023)

Is the quality of the data really improved?

Are panellists willing to participate?

We studied willingness to participate (Ochoa & Revilla, 2022a; Ochoa 2022)



Will they see the invitation quickly enough?

How to invite them really in-the-moment?

We created the WebdataNow tool (Revilla et al., 2022b)

We studied invitation methods (Ochoa & Revilla, 2022b)

Main learnings from the job search experiment

- In-the-moment surveys **can be implemented**: we now have the necessary panels and tools to do so
- However, data collection is **more complex** and take **much more time**
- Participants are **satisfied** with their experience (no issues with intrusiveness)
- **Fewer improvements in data quality** indicators than expected
 - Slight improvement in % DK
 - Some improvement in length of answers to open-ended questions
- **Differences in substantive results** that suggest people might not chose to answer “DK” even if they do not recall (properly)

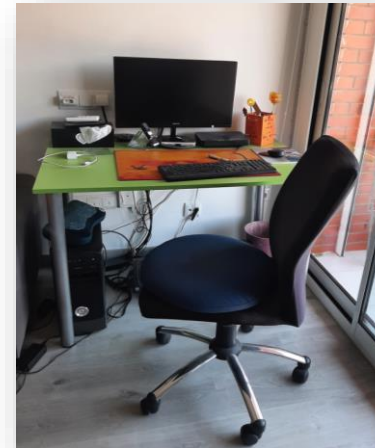
Conclusions

Starting is Difficult, Finishing is Way Harder



Increasing interest in new data types

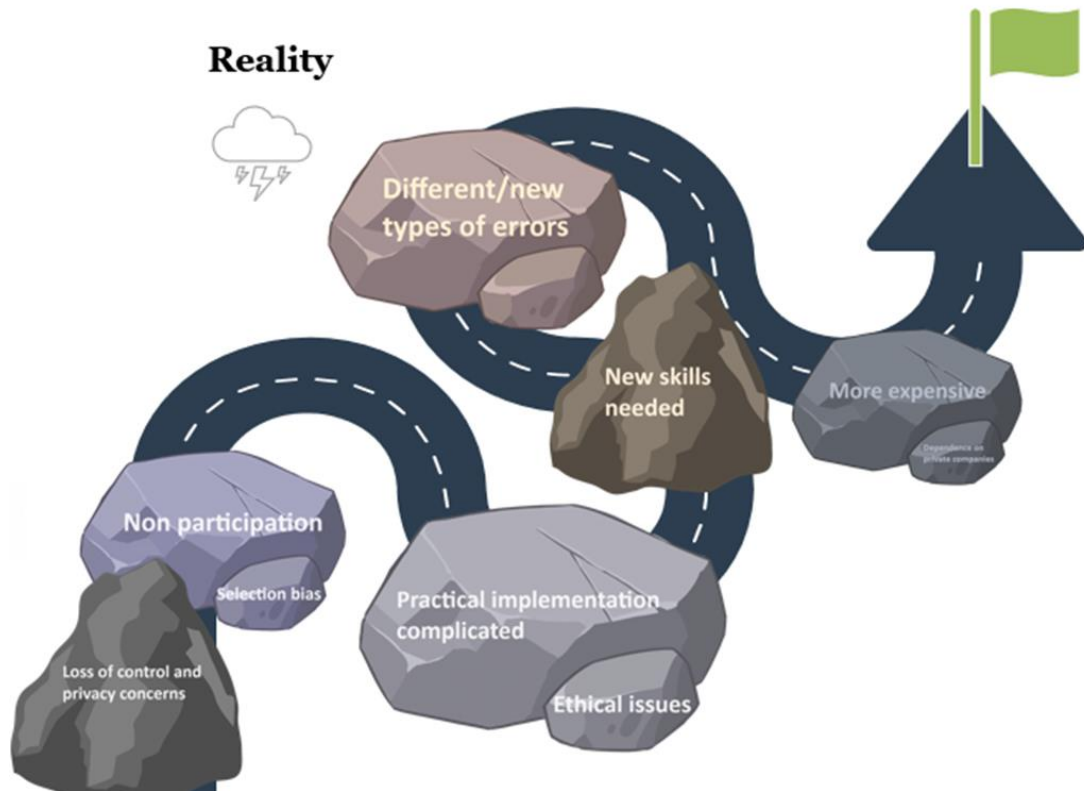
- Issues in conventional surveys
 - Decreasing participation + poor data quality in measures of many concepts
- Push researchers to consider **new data types**
 - Could reduce some types of errors + provide new/more detailed data
- Potentially **broad applications** and new insights



CONCLUSIONS

But this is a complicated road

Remember...



Lot of further research needed



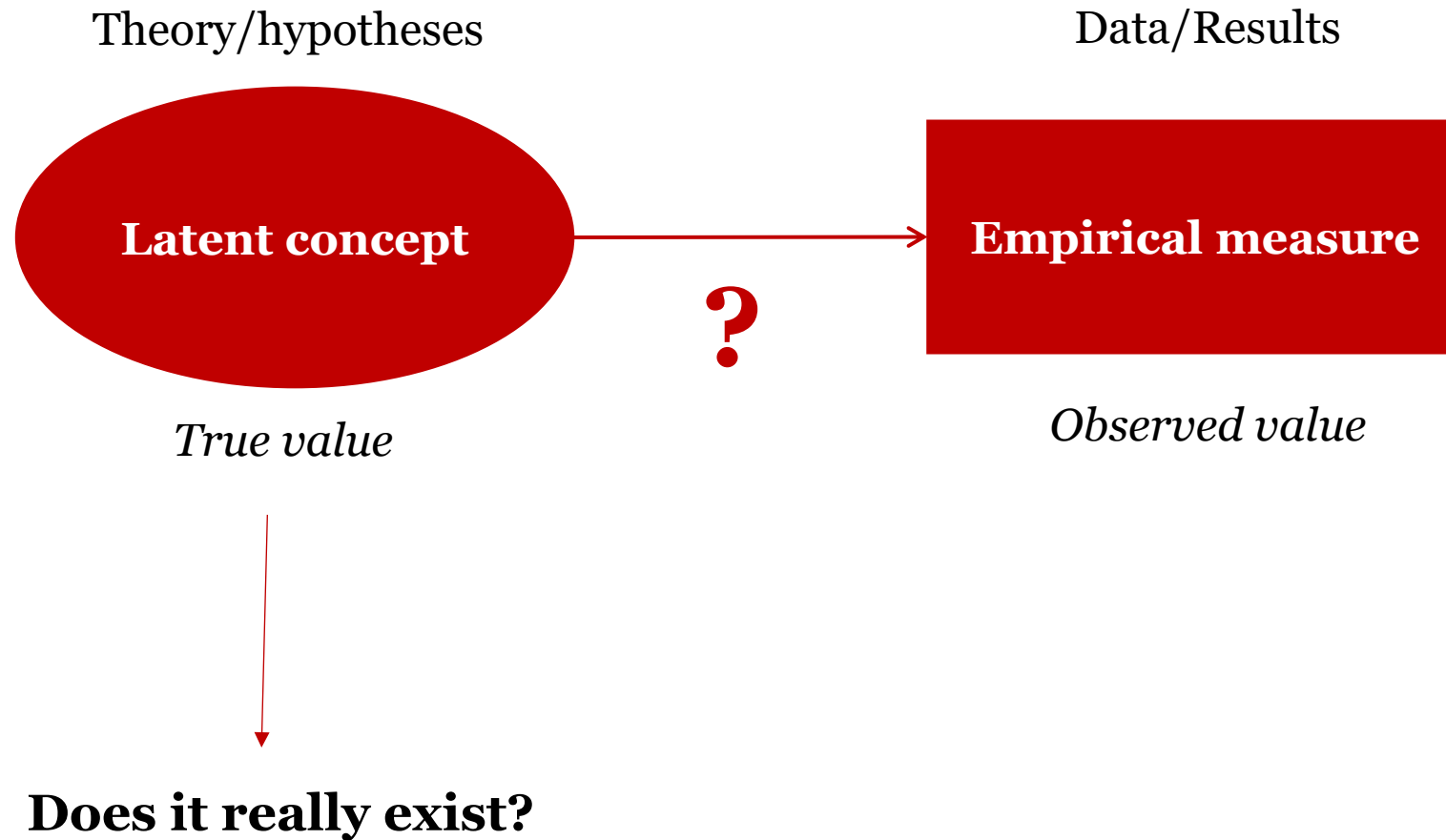
We need to get **similar level of knowledge for the new data types** as we have for conventional questions



Necessary to make informed decisions about their use + to improve data quality

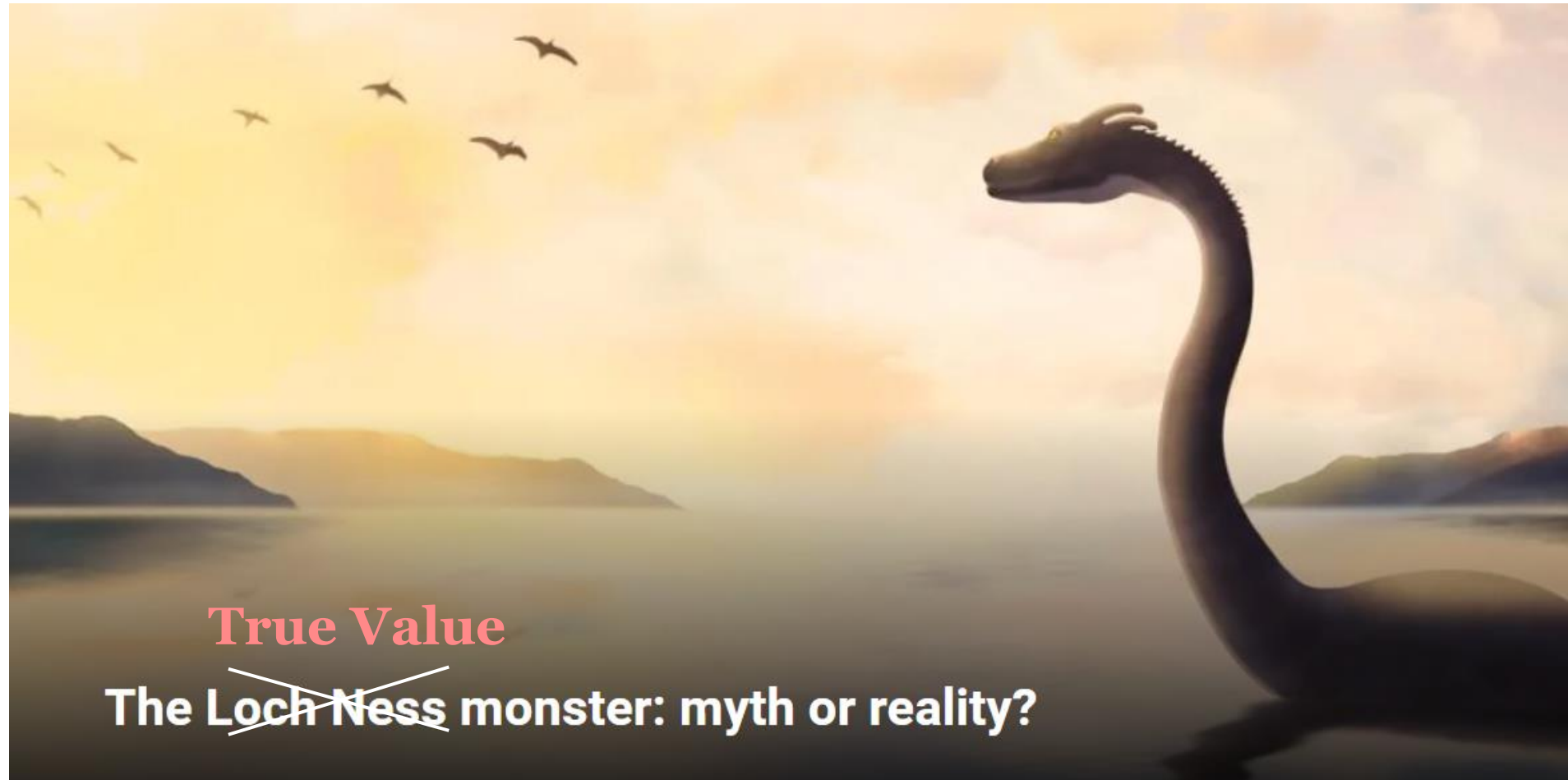
CONCLUSIONS

Even if we improve data quality, errors will remain...



CONCLUSIONS

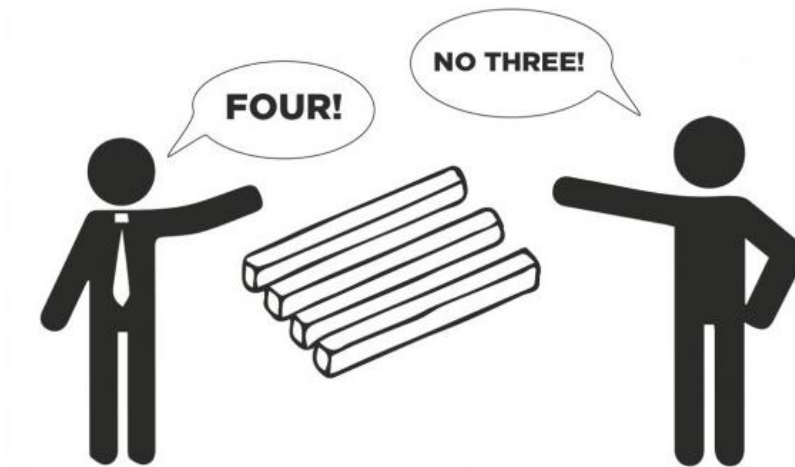
Does it really exist?



CONCLUSIONS

Look from different perspectives

Different types of data provide **different but complementary information**



Combine *several* types of **data!**

Thanks!

Questions?



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<https://www.upf.edu/web/webdataopp>



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References

- Bosch, O.J., Revilla, M., & Paura, E. (2019). Do Millennials differ in terms of survey participation? *International Journal of Market Research*, 61(4): 359-365. <https://doi.org/10.1177/1470785318815567>
- Couper, M. (2008). *Designing Effective Web Surveys*, Cambridge: Cambridge University Press.
- ESOMAR (2019). *Global Market Research Report*. Amsterdam: ESOMAR. ISBN: 978-90-903-2259-9 . Retrieved from (June 2021): <https://www.esomar.org/knowledge-center/library?publication=2926>
- Howison, J., Wiggins, A., & Crowston, K. (2011). Validity issues in the use of social network analysis with digital trace data. *Journal of the Association for Information Systems*, 12(12), 767–797. <https://doi.org/10.17705/1jais.00282>
- Iglesias, P.A., Revilla, M., Heppt, B., Volodina, A., & Lechner, C. (2023). Protocol for a web survey experiment studying the feasibility of asking respondents to capture and submit photos of the books they have at home and the resulting data quality [version 1; peer review: 2 approved, 1 approved with reservations]. *Open Res Europe* 2023, **3**:202 (<https://doi.org/10.12688/openreseurope.16507.1>)
- Iglesias, P.A., Ochoa, C., & Revilla, M. (2024). A practical guide to (successfully) collect and process images through online surveys. *Social Sciences & Humanities Open*, 9, 100792. <https://doi.org/10.1016/j.ssaho.2023.100792>
- Iglesias, P., & Revilla, M. (2024). Skills, availability, willingness, expected participation and burden of sharing visual data within the frame of web surveys. *Quality and Quantity*, 58, 1071–1092. <https://doi.org/10.1007/s11135-023-01670-3>
- Kahneman, D., & Riis, J. (2005). “Living, and thinking about it: Two perspectives on life”. In F. Huppert, B. Keverne, & N. Baylis (Eds.), *The science of well-being*. Oxford, England: Oxford University Press.
- Ochoa, C. (2023). Researching the moment of truth. An experiment comparing in-the-moment and conventional surveys to investigate online job applications. ESRA conference, 20th July 2023. [Presentation \(preliminary results\)](#).

References

- Ochoa, C. (2022). Willingness to participate in geolocation-based research. PLoS ONE 17(12): e0278416. <https://doi.org/10.1371/journal.pone.0278416>
- Ochoa, C., Revilla, M. (2022a). Willingness to participate in in-the-moment surveys triggered by online behaviors. Behavior Research Methods (2022). <https://doi.org/10.3758/s13428-022-01872-x>
- Ochoa, C., Revilla, M. (2022b). Acceptance and coverage of fast invitation methods to in-the-moment surveys. International Journal of Market Research. <https://doi.org/10.1177/14707853221085204> .
- Poses, C., Revilla, M., Asensio, M., Schwarz, H., and Weber, W. (2021). Measurement quality of 67 common social sciences questions across countries/languages: Results of 28 Multitrait-Multimethod experiments implemented in the European Social Survey. *Survey Research Methods*, 15(3): 235-256. <https://doi.org/10.18148/srm/2021.v15i3.7816>
- Revilla, M. (2022). How to enhance web survey data using metered, geolocation, visual and voice data?. *Survey Research Methods*, 16(1): 1-12. <https://doi.org/10.18148/srm/2022.v16i1.8013>.
- Revilla, M., Iglesias, P., Ochoa, C., & Antón, D. (2022a). WebdataVisual: a tool to gather visual data within the frame of web surveys. OSF. <http://doi.org/10.17605/OSF.IO/R7CAX>
- Revilla, M., Ochoa, C., Iglesias, P., Antón, D. (2022b). WebdataNow: a tool to send in-the-moment surveys triggered by passive data. OSF. <http://doi.org/10.17605/OSF.IO/G3MSC>
- Saris, W. E., & Gallhofer, I. (2007). *Design, evaluation, and analysis of questionnaires for survey research*. New York: Wiley
- Saris, W.E., & Revilla, M. (2016). Correction for measurement errors in survey research: necessary and possible. *Social Indicators Research*, 127(3): 1005-1020.
- Tourangeau, R., Conrad, F. & Couper, M.P. (2013). *The Science of Web Surveys*, Oxford: Oxford University Press.