

# WEB DATA OPP workshop

*Barcelona, Spain*

18-19 March 2024



Universitat  
Pompeu Fabra  
*Barcelona*

**RECSM**

Research and Expertise Centre  
for Survey Methodology



INSTITUT  
BARCELONA  
ESTUDIS  
INTERNACIONALS

## Context

- Workshop organized in the frame of the WEB DATA OPP project
  - Funded by an ERC starting grant
  - Started in 2020 – Will end in 2026
- Why organizing this workshop?
  - To disseminate our results
  - To learn about others' research
  - To share experiences in particular regarding the challenges faced
  - To get feedback that can help us in the next months
  - To generate new collaborations?



# Acknowledgements



**WEB DATA OPP team**

Oriol Bosch

Patricia Iglesias

Carlos Ochoa

Ksenija Ivanovic

Tracy Silva



All of you!



Maria Paula Acuña



Mick Couper

## Some practical points

- Necessary that you all sign the assistance list **every day**
- Wi-Fi: free university Wi-Fi (network guest@upf) or Eduroam
- Coffee breaks & lunch: next to Auditori Mercè Rodoreda → be careful with personal belongings
- Dinner meeting points:
  - 19h30 entrance hotel H10 Marina
  - 20h restaurant Gastrobar Pipa (Carrer de Llull, 159)

## Some practical points

- Regarding the presentations:
  - Remember that each presenter has **15-16 minutes to present**, followed by 4-5 minutes for questions
  - If you want the session chair to indicate you when X minutes are left, please let them know
  - Please check during one of the breaks that your slides have been correctly uploaded on the computer and that everything works as expected

# Overview of the WEB DATA OPP project

*18 March 2024*

**Melanie Revilla** | IBEI

***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). Part of the research done has also been supported by the Spanish Ministry of Science and Innovation under the "R+D+i projects" programme (grant number PID2019-106867RB-I00 /AEI/10.13039/501100011033 (2020-2024)); the BBVA foundation under their grant scheme to scientific research teams in economy and digital society, 2019; and GESIS.*

# Project's research questions



1. Can we **improve** web survey **data quality** by using new measurement opportunities?
2. Can we **replace** part of the web survey data by using new measurement opportunities?
3. Can we achieve a **more complete picture** of the reality by **combining** new measurement opportunities with web survey data?

# Which new opportunities?



## Growing use of (mobile) Internet

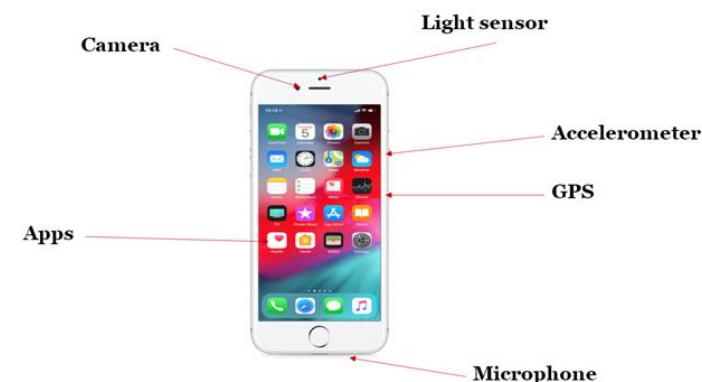
More and more of people's life happens **online**

**+30m** Average daily time<sup>1</sup> spent online by each internet user from 2016 to 2021

More and more of the online activity is done through **smartphones**

**83%** of the world population have smartphones<sup>2</sup>

Smartphones have sensors + apps  
→ possible to collect many different types of new data



# New data types considered

## VISUAL DATA



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

## VOICE DATA



Dictation  
Voice recording

Most of those data can also be collected for PCs

## 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 they help?

## Main expected benefits (Revilla, 2022)

### Researchers

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

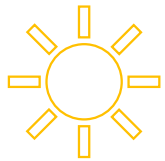
### Participants

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

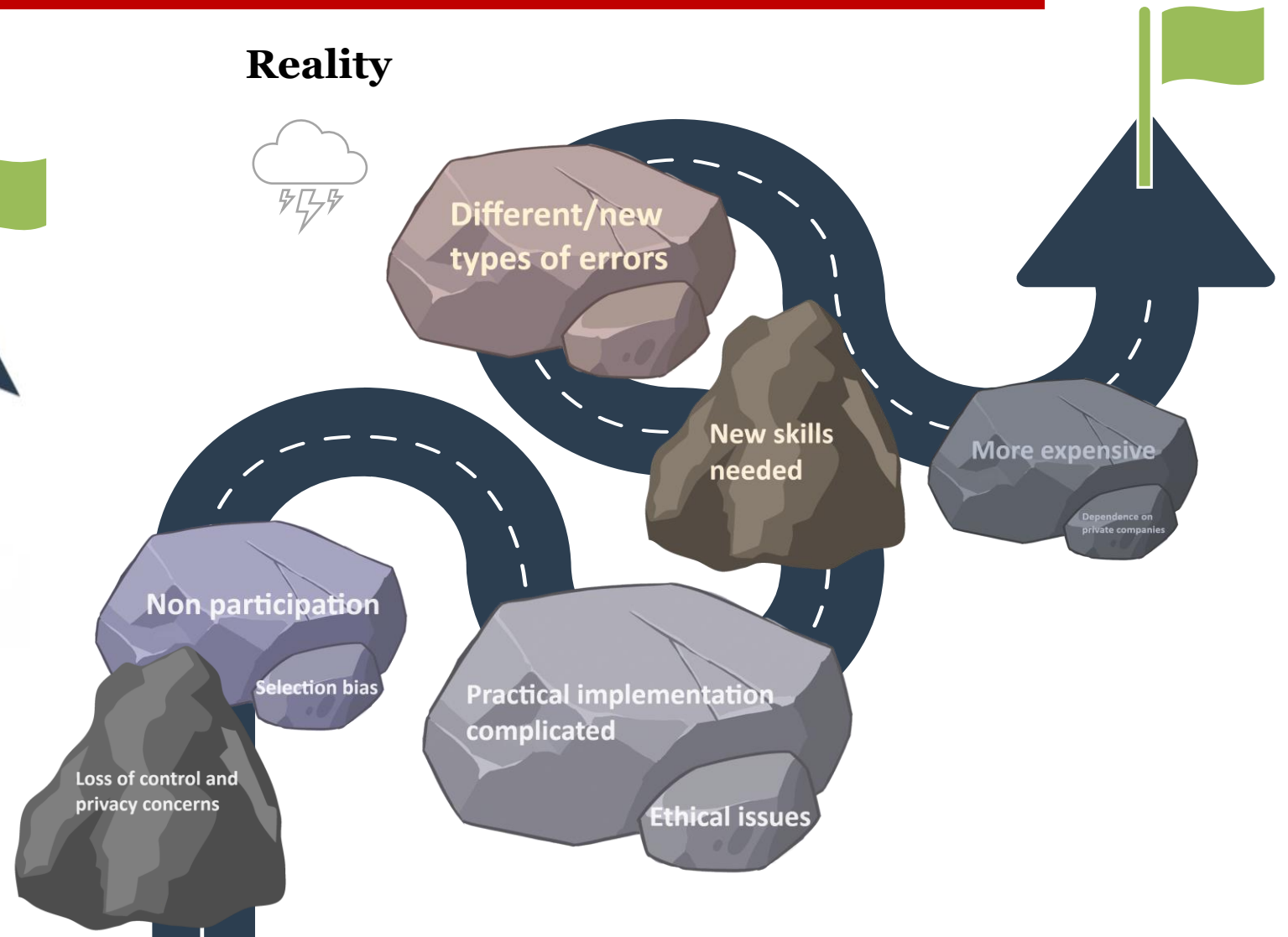
# But this is not that easy...

Our goal = get more knowledge that will help better use such data

## What most people think



## Reality



# Summary of the research we did or are doing

# Visual data

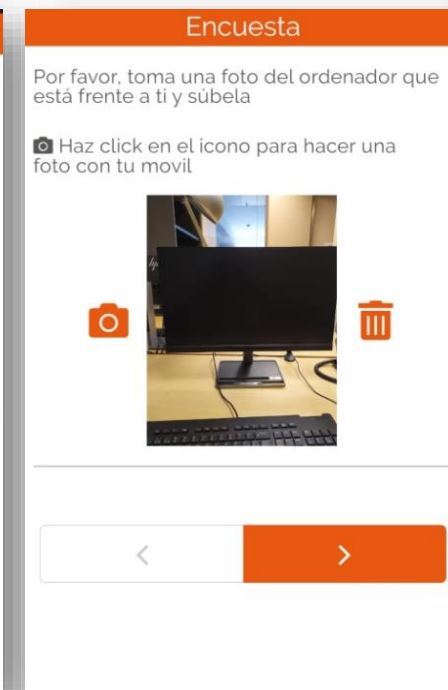
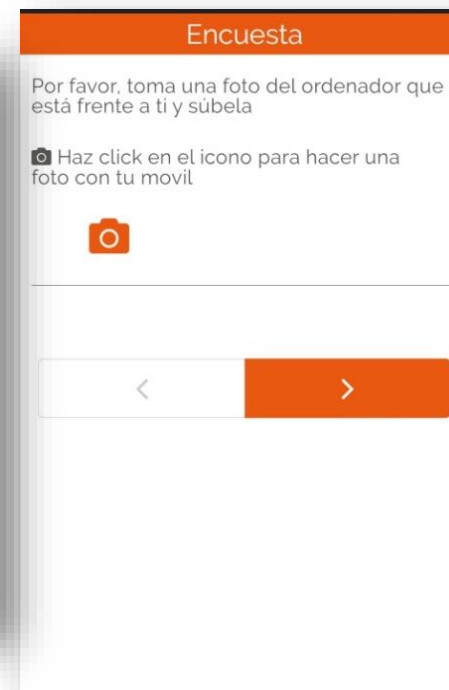
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*WebdataVisual* (Revilla et al., 2022a): allows collecting visual data produced during the survey (using the device camera or screenshots) or sharing already saved visual data.

*Try to disentangle the mechanisms behind non-response*

*Prepare a guide about collecting images in surveys*

*Implement an experiment*



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Previous studies suggested only  $\approx 50\%$  of participants sent valid images when asked in surveys. Reasons for this unclear. We tried to disentangle the role of skills + availability + willingness + burden.

Results: Availability seems to be the most limiting factor for participation (Iglesias & Revilla, 2023).

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## WHAT WE DID OR ARE DOING

# Visual data

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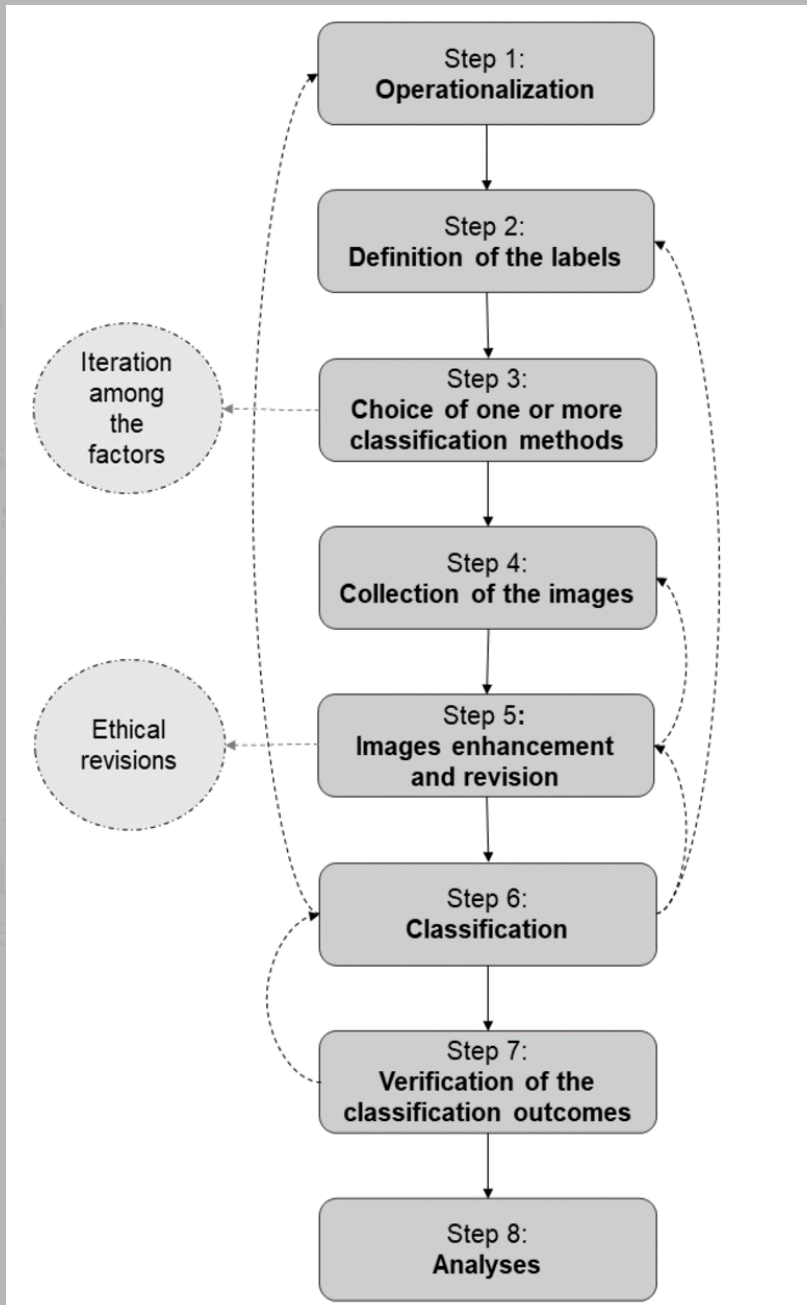
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*WebdataVisual survey (using the*

*Previous studies surveys. Reason: willingness + budget. Results: Availability + Revilla, 2023).*

*Quality of the information needed information prepared a guide (al., 2024).*



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Asked participants to share photos of the books they have in their main residence + to report the number of books in their residence, the proportions in different languages and the way they store such books. Compare participation, quality, evaluation, of the different methods (Iglesias, under review). Data also used for substantive research.

# Voice data

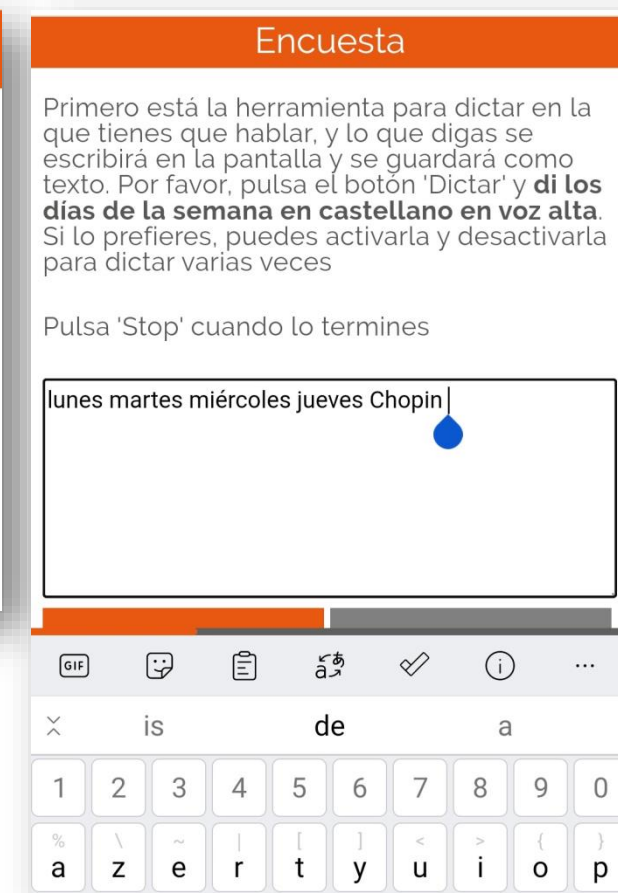
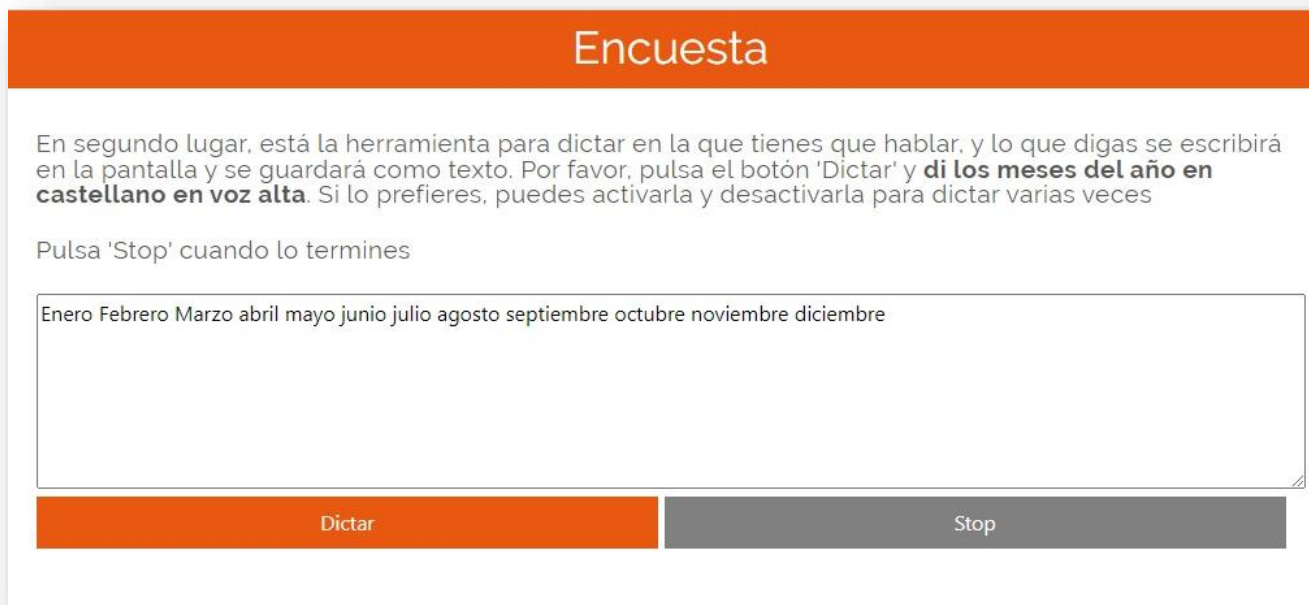
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*Compare different ways to propose voice inputs*

*Try to understand the problems & challenges*

*Propose additional incentive*



More information available at: <https://www.upf.edu/web/webdataopp/tools>

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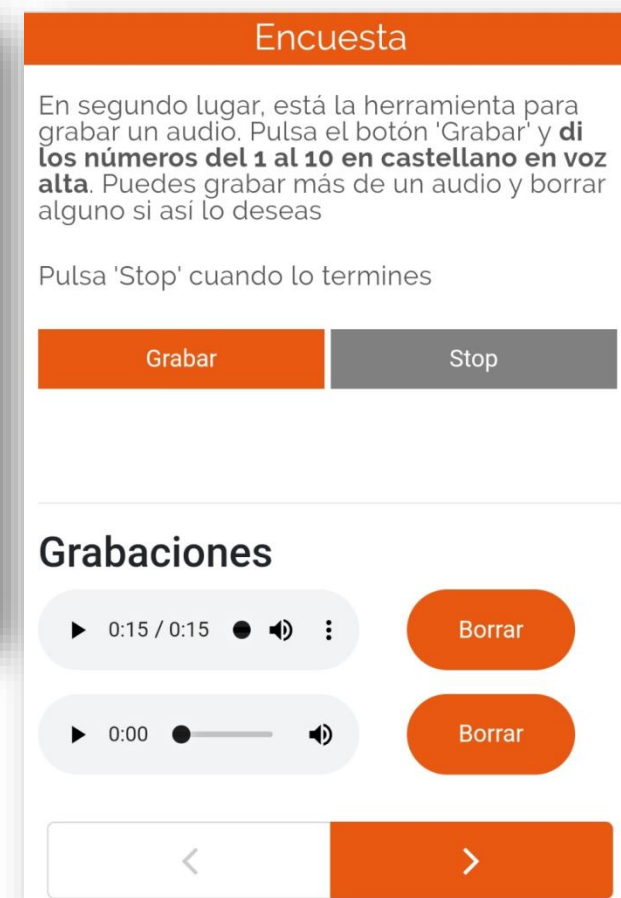
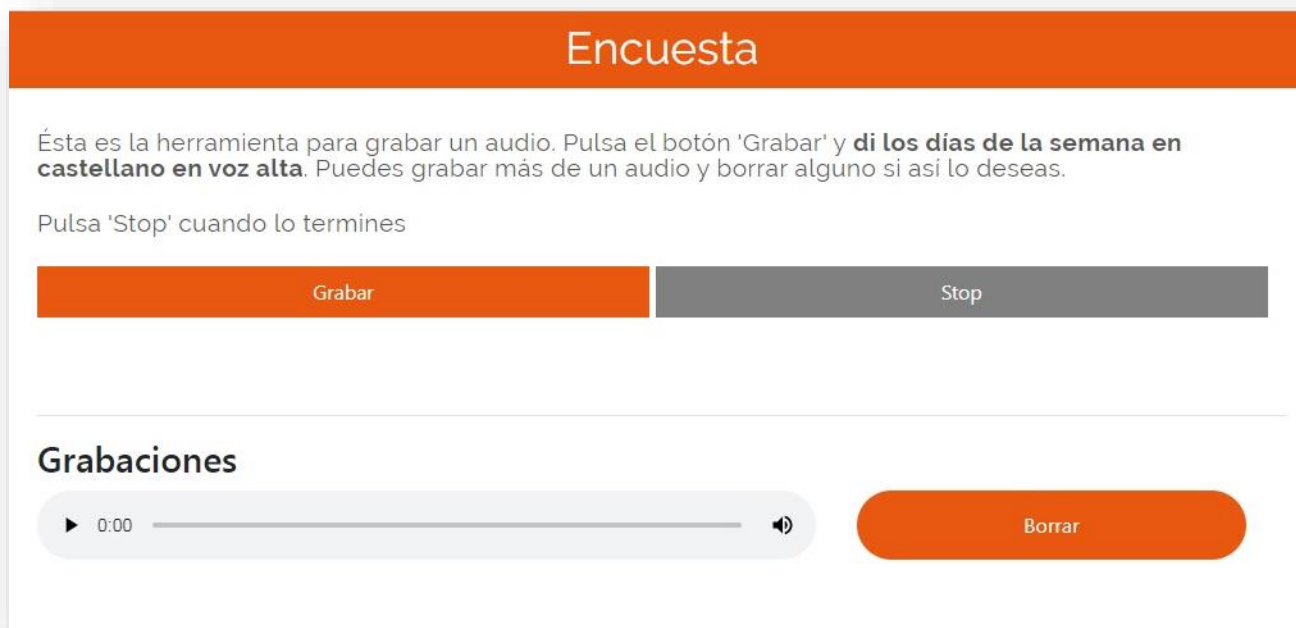
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Ask participants of opt-in online panel to answer 2 experimental questions through dictation or voice recordings, either providing a choice or pushing them to voice inputs. Results: in all experimental groups, participation is lower, even if text was always possible. Quality seems to be slightly higher (Revilla & Couper, under review, a).

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*Try to understand the problems & challenges*

Study reasons for not using voice inputs and challenges faced when using them + whether panelists dis/liked and found it easy/difficult to use the voice input tools. Results: Context and difficulty in expressing orally are the most reported issues. Most participants found it easy; less liked it (Revilla & Couper, under review, b).

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*Propose additional incentive*

Focus on push-to voice recording. Try to propose an extra incentive conditional on using voice to answer the 2 experimental questions. Distinguish 3 groups depending on the likelihood of using voice (Höhne et al., pre-registration)



# Metered data

*Create a  
framework*

Total error framework for metered data = adaptation of the total survey error (TSE) framework to metered data → explain all possible errors (Bosch & Revilla, 2022a)

*Apply the  
framework*

*Estimate the size  
of tracking  
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*Study the  
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# Metered data

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Total error from framework to

survey error (TSE) & Revilla, 2022a)

Error components	Specific error causes
Specification error	<ul style="list-style-type: none"><li>- Measuring concepts from which not enough data is available</li><li>- Inferring attitudes</li><li>- Defining valid information</li></ul>
Measurement error	<ul style="list-style-type: none"><li>- Non-trackable target</li><li>- Meter not installed</li><li>- Uninstalling the meter</li><li>- New non-tracked device</li><li>- Technology limitations</li><li>- Technology errors</li><li>- Hidden behaviours</li><li>- Shared device</li><li>- Social desirability</li><li>- Extraction error</li></ul>
Processing error	<ul style="list-style-type: none"><li>- Coding error</li><li>- Aggregation at the domain level</li><li>- Data anonymization</li></ul>
Coverage error	<ul style="list-style-type: none"><li>- Non-trackable individuals</li></ul>
Sampling error	<ul style="list-style-type: none"><li>- Same error causes than for surveys</li></ul>
Missing data error	<ul style="list-style-type: none"><li>- Noncontact</li><li>- Non-consent</li><li>- Non-trackable target</li><li>- Meter not installed</li><li>- Uninstalling the meter</li><li>- New non-tracked device</li><li>- Technology limitations</li><li>- Technology error</li><li>- Hidden behaviour</li></ul>

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Key problem for metered data: tracking undercoverage = only part of the participants' online behaviors are tracked, for instance because they do not install the meter in all the devices/browsers they use to go online (Bosch et al., 2023).

Combined survey+metered data (TRI-POL) → 76% participants with at least 1 device undercovered.

Simulations → Tracking undercoverage bias *both* univariate and multivariate estimates

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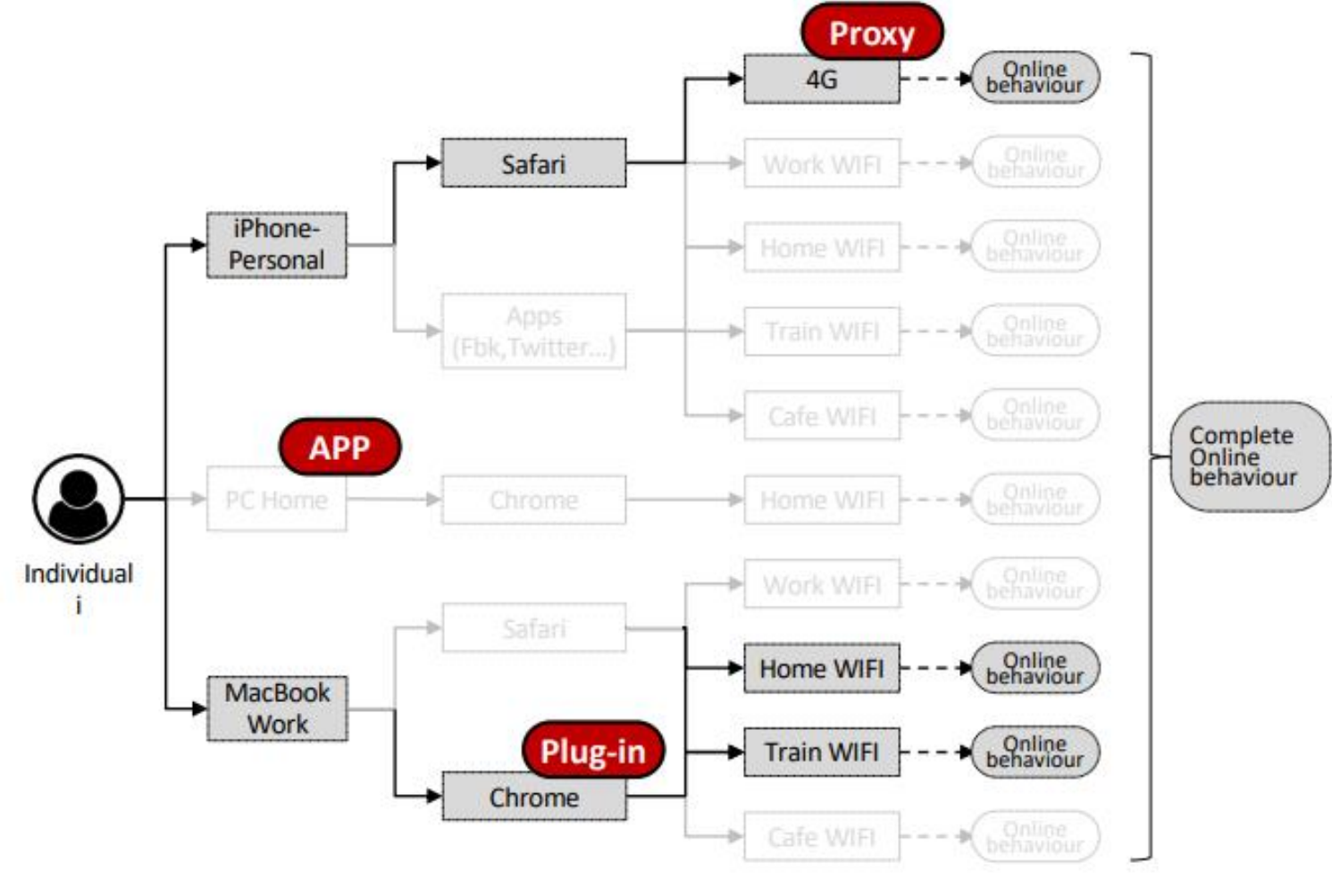
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*Study the reliability & validity of metered measures*

Most research using metered data assume the measurement is perfect. Clearly not true, but to what extent? Focus on “online news media exposure” (Bosch, 2023).

Study validity & reliability + look at the impact of different design choices (e.g., which list of media news domain was used, how many media on those lists, how many days of metered data collection, etc.).

# Metered data

*Estimate  
measurement  
quality*

To learn more about measurement errors, an MTMM experiment was implemented (Netquest metered panel, Spain, June 2023) → allowed to estimate the reliability and measurement validity of measures based on conventional questions vs metered data (Bosch et al., 2023b). Potentially important errors when using metered data.

*Detect cheating in  
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*Predict  
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Different way to combine metered data with surveys: use metered data to detect cheating in political knowledge questions (Bosch & Revilla, 2024). Tested using same survey as previous one. Web tracking can be an option to catch people cheating, but it is still imperfect, mainly when there is no in-app data.

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Plan to study whether we can predict pregnancy and parenthood (of “young children”) using metered data (URLs, app use and search terms). Will use metered data and profiling data collected by Netquest. In progress (Joshua Claassen).

**COMING SOON**

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*Study job search*

Provide new insights about how people search for jobs online using metered data. In progress (Carlos Ochoa).

# In-the-moment surveys (ITMS)

*Create a tool*

*WebdataNow* (Revilla et al., 2022c): allows implementing ITMS triggered by metered or geolocation data.

*Acceptance and coverage of fast invitation methods*

*Study the willingness and its determinants*

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Participation in ITMS requires that invited units see the survey notification quickly enough. Thus, information about acceptance and coverage of different invitation methods (e.g., SMS or app-notification) and which one participants see first is crucial. Results: Invitation through a mobile app is the best option. Using several methods is recommended (Ochoa & Revilla, 2022a).

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### *Study the willingness and its determinants*

2 conjoint experiments: ITMS triggered by 1) metered data and 2) geolocation data. Results: Willingness to participate in ITMS is high. Main limitation = willingness to share the passive data needed to trigger the surveys. Different attributes (incentive, survey length, etc.) affect the willingness to participate (Ochoa & Revilla, 2022b; Ochoa, 2022).

### *Implement 2 experiments*

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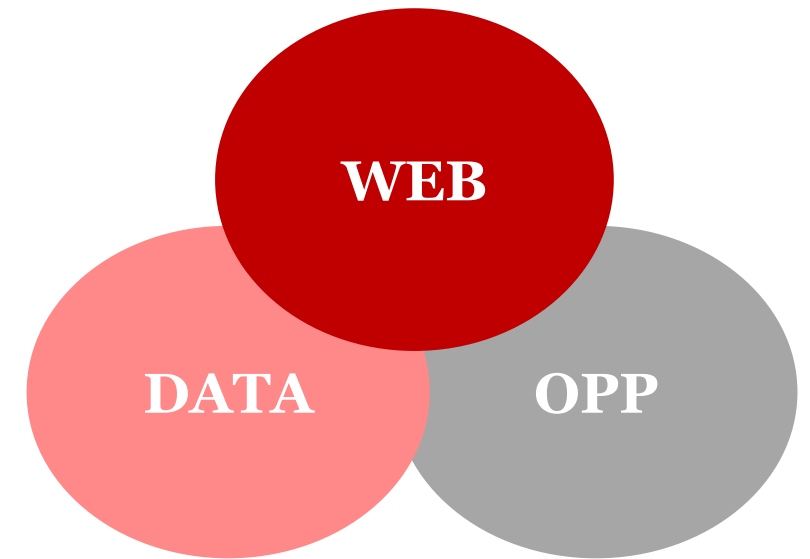
### *Implement 2 experiments*

Experiment 1: ITMS triggered by online job application. Long fieldwork + practical difficulties; but some improvements + new insights (Ochoa, 2023). Data also used for substantive research.

Experiment 2: ITMS triggered by a visit to a beach in Spain. Planned for summer 2024.

## Summing up

- Very little had been done before regarding the 4 types of data that the project is focusing on
- Contribution at different levels
  - New tools
  - Framework
  - Practical guidelines
  - New empirical evidences
  - Cases studies regarding different topics
  - Data have also been used for substantive research

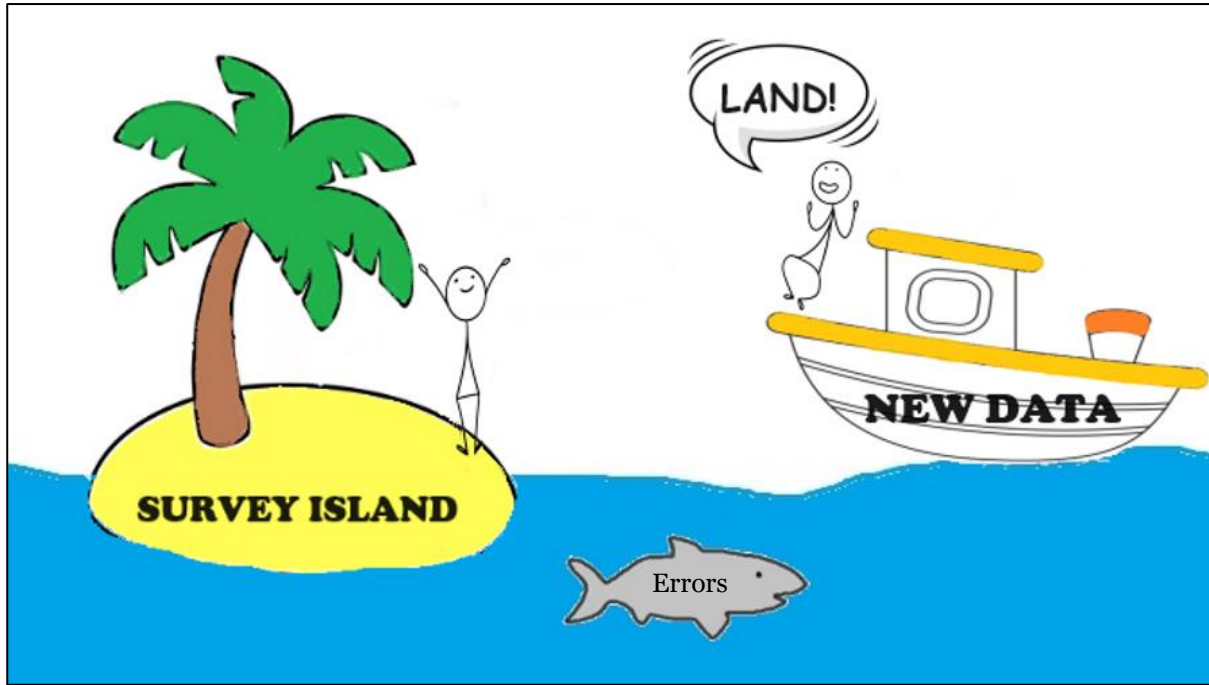
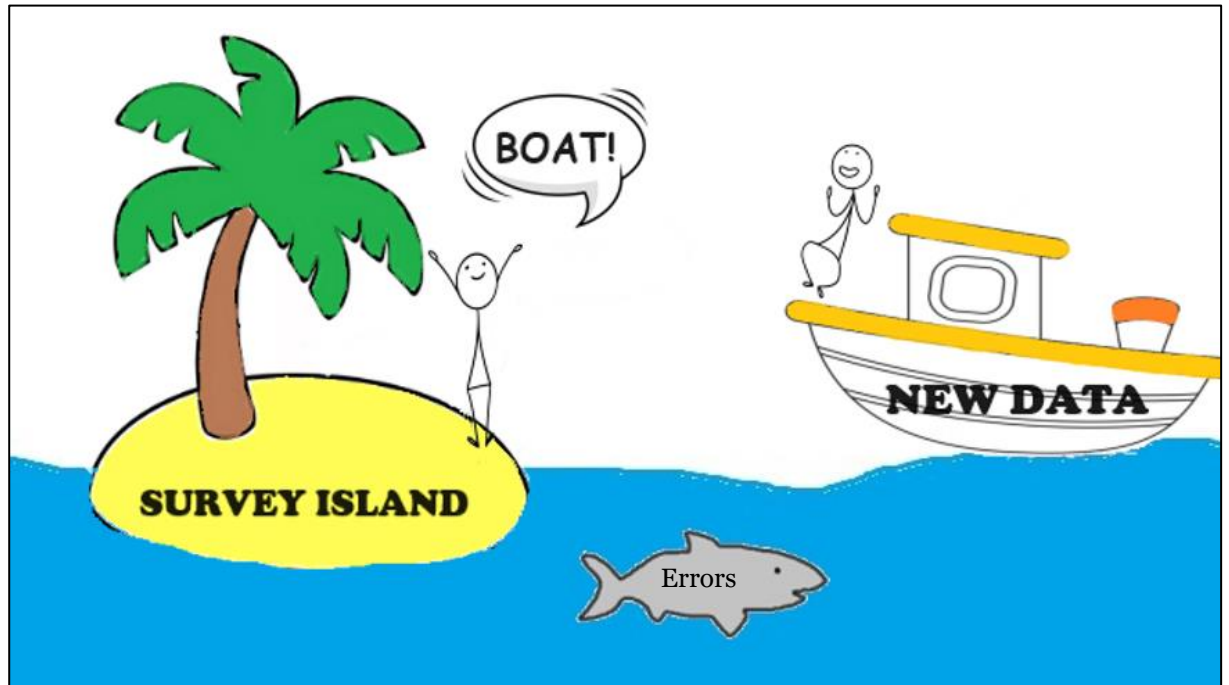
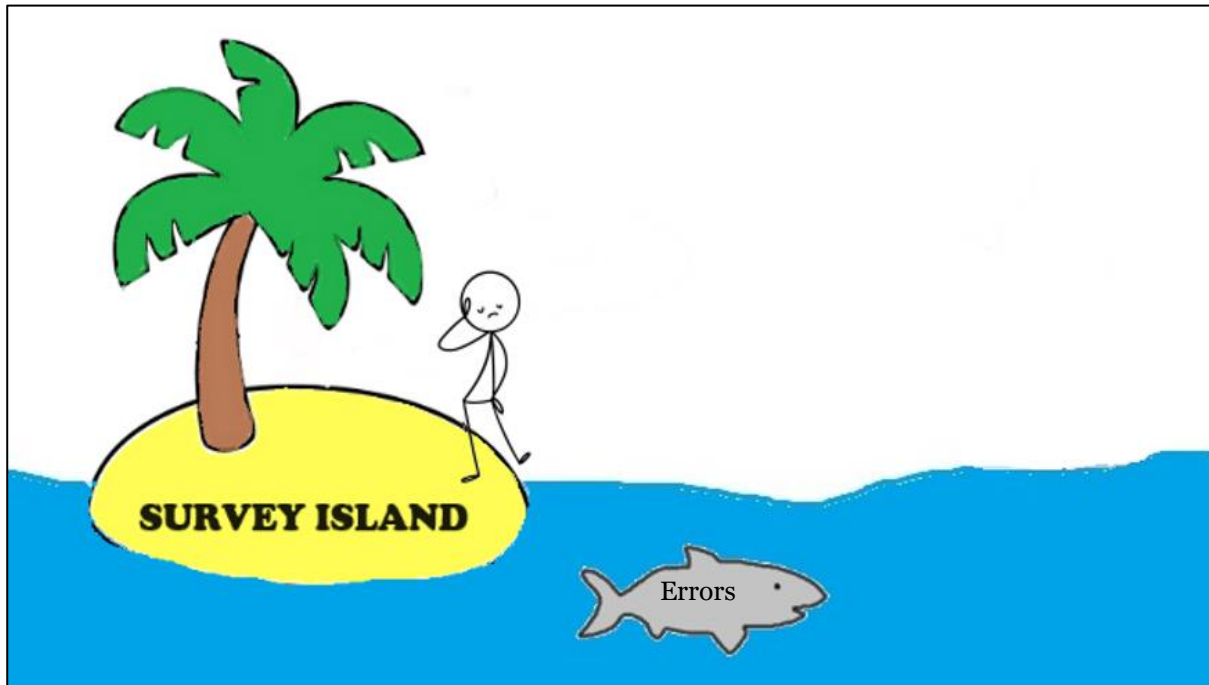


# Conclusions



**Starting is Difficult, Finishing is Way Harder**



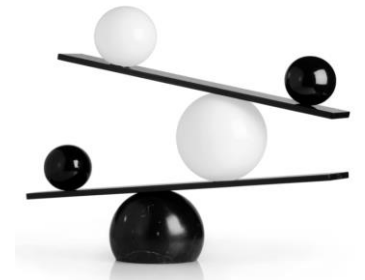


## Much more research needed

- 1 Learn more about the **errors** of these data in different contexts



- 2 Understand better **when** it is worth using new data types



# Much more research needed

3 Understand better **how** to use these data

**Replace** conventional data?



Errors will always be present

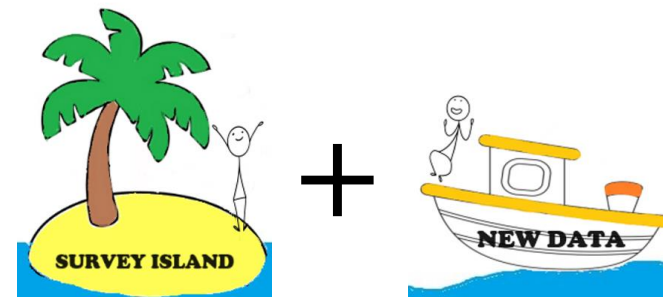


Crucial to **identify** them and think about their **consequences**

**Combine** with conventional surveys?



Provide **different** but **complementary information**



## CONCLUSIONS

# Maybe we just need more time...

In the 1950s people predicted flying cars by 2000



**YOUR PERSONAL "FLYING CARPET"** Step into it, press a button, and off you go to market, to a friend's home, or to your job. Take off and land anywhere; no parking problems. Plug in to any electric outlet for recharging. They're working on it!

My parents' car in 2000...





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Paris Prepares to Launch Flying Taxis during 2024 Olympics



eVTOLs Expected to Take Center Stage at Paris 2024 Olympic Games (Credit: Volocopter)

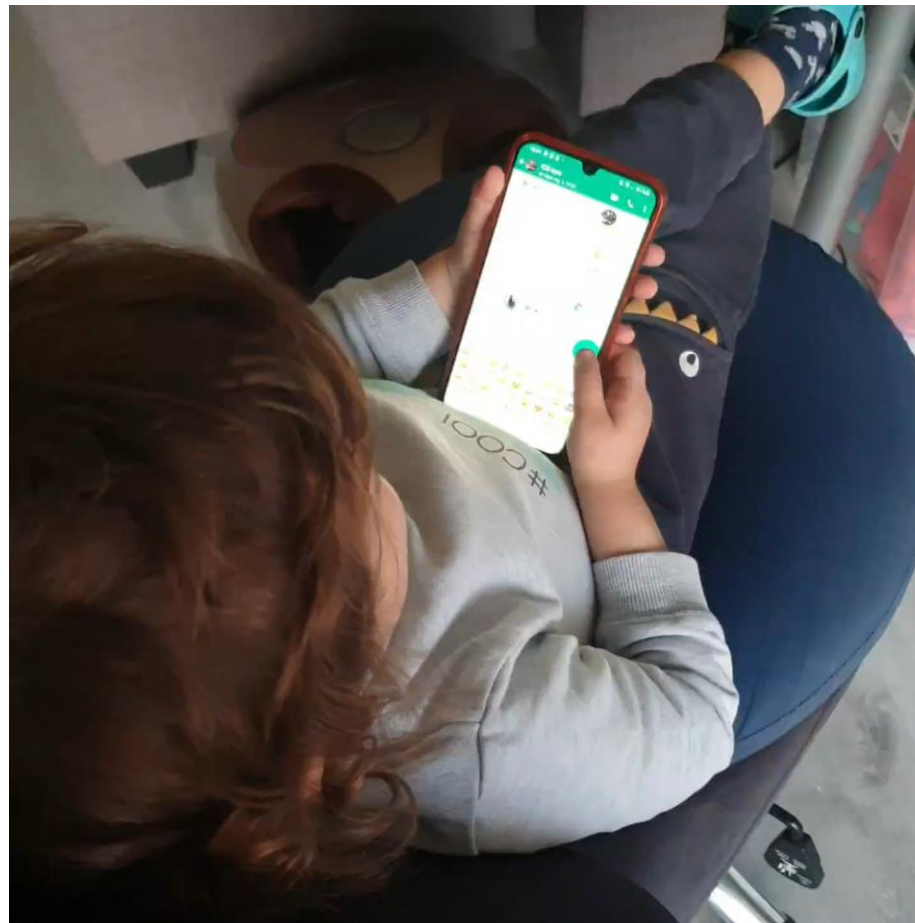
## CONCLUSIONS

# Maybe we just need more time...

What my 2-year-old son can do with pencils



But he can send voice messages...



# Thanks!

## *Questions?*

Melanie Revilla | IBEI



[mrevilla@ibei.org](mailto:mrevilla@ibei.org)



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



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INTERNACIONALS





# References

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