

Knock, knock! Who's there? A Latent Class Typology of Household Photovoltaic Investors

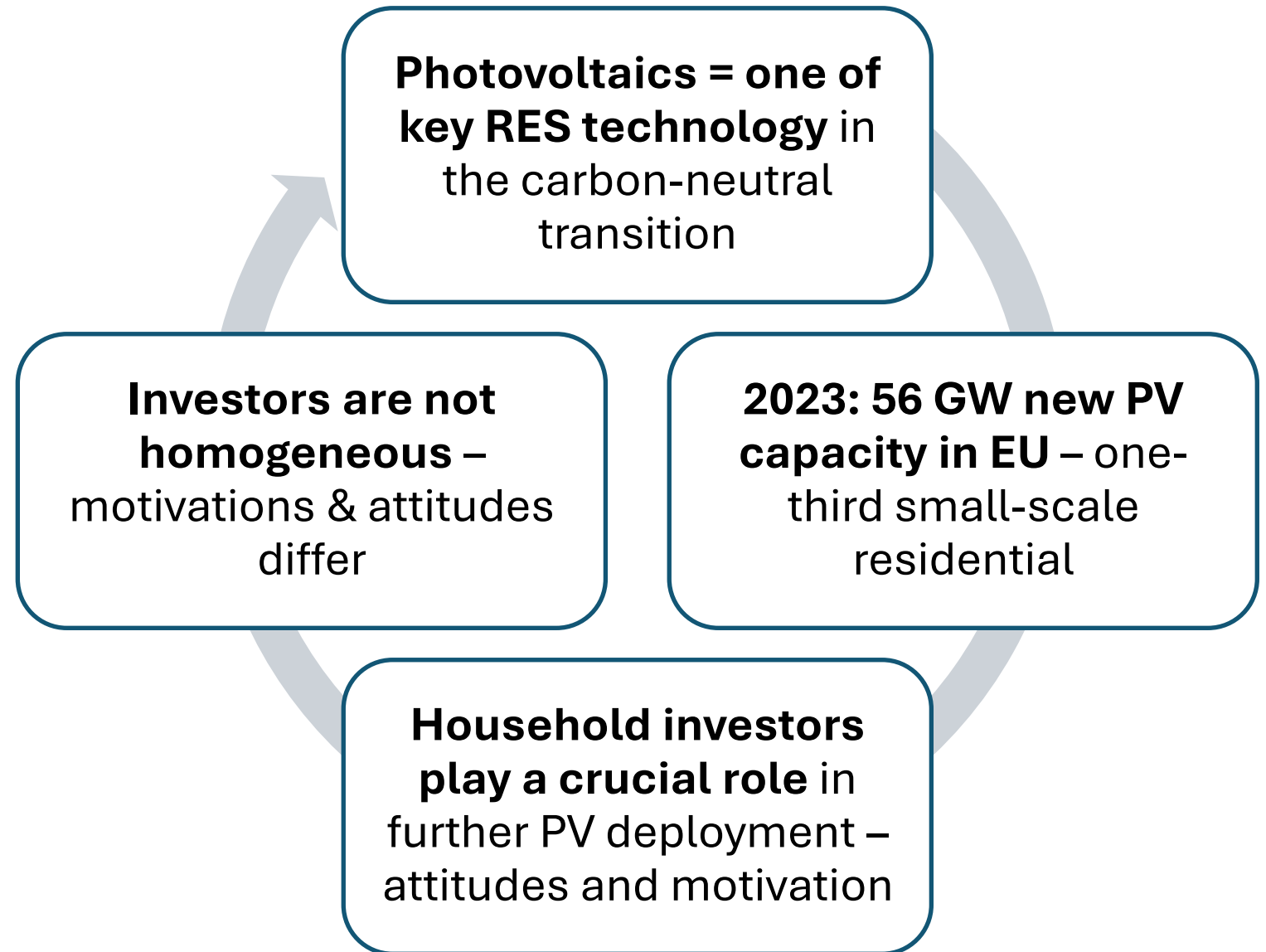
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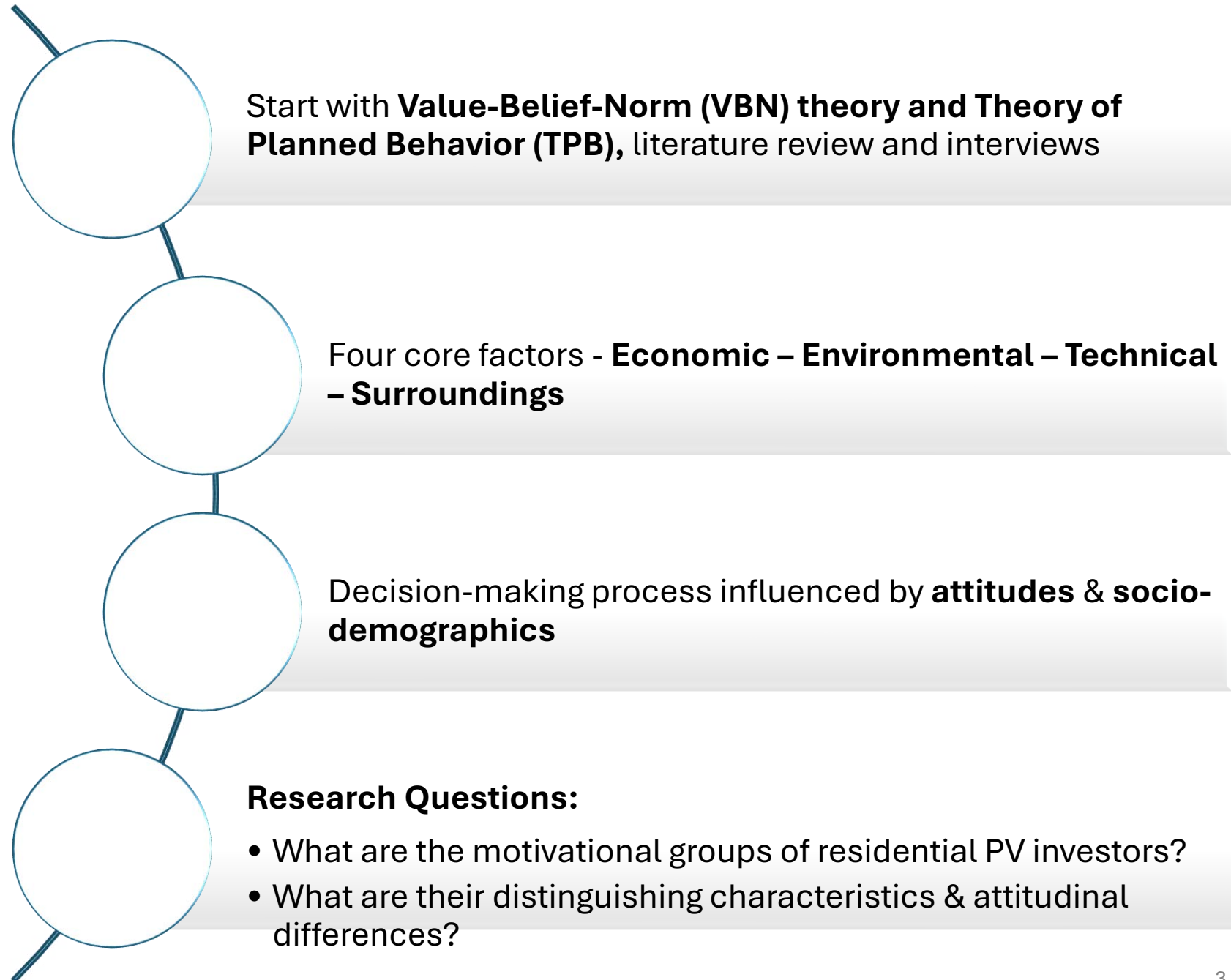
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Context



Theoretical framework



Data



2,565 households (Czechia, 2022 survey)



Questionnaire: 4 factors of motivation, attitudes, technical characteristics of the instalation, sociodemographics



Latent Class Analysis (poLCA in R) – 40 models tested, 3 pre-selected (fit criteria, interpretability)



Final solution: **6 investor classes**

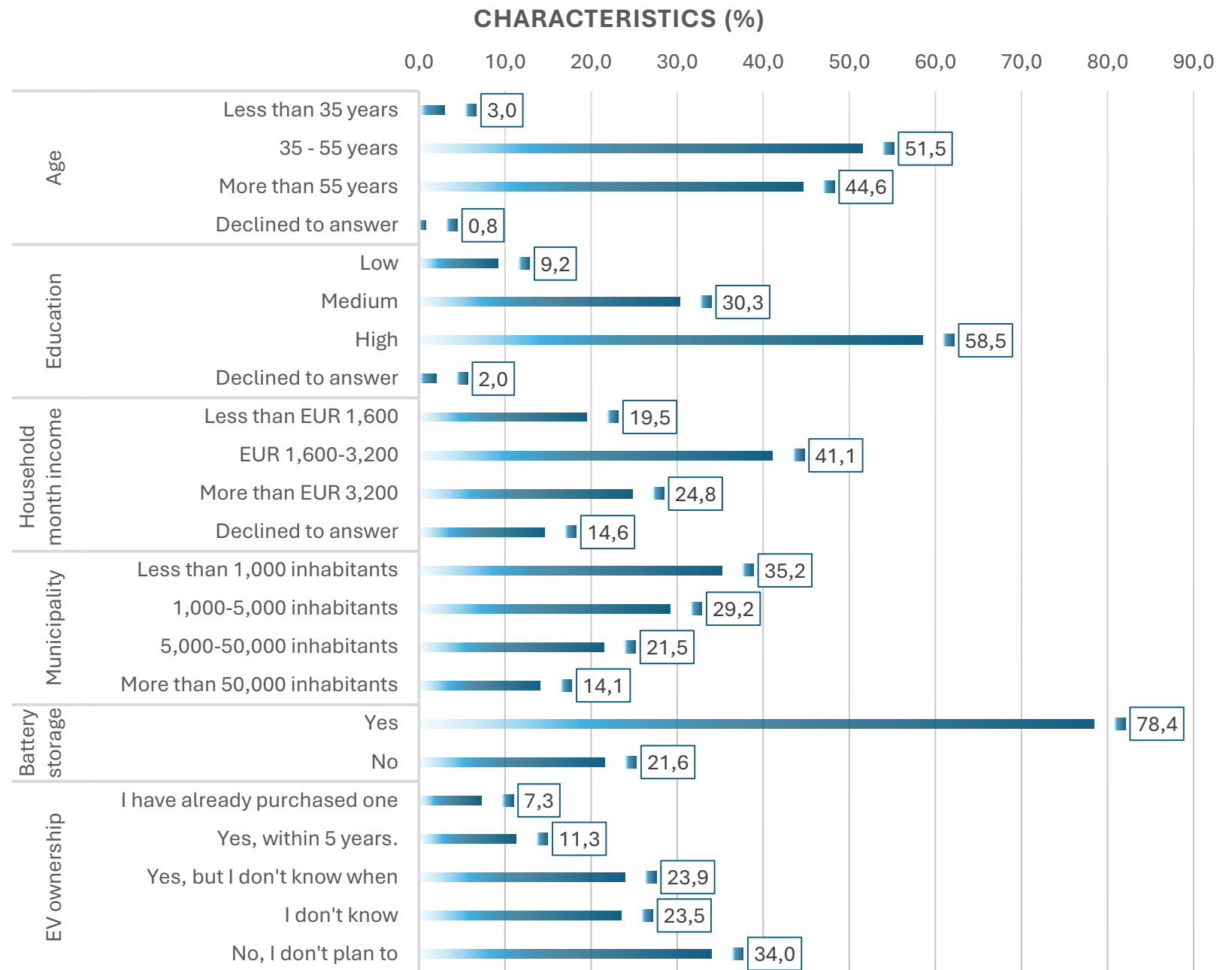


Methodological challenges addressed:

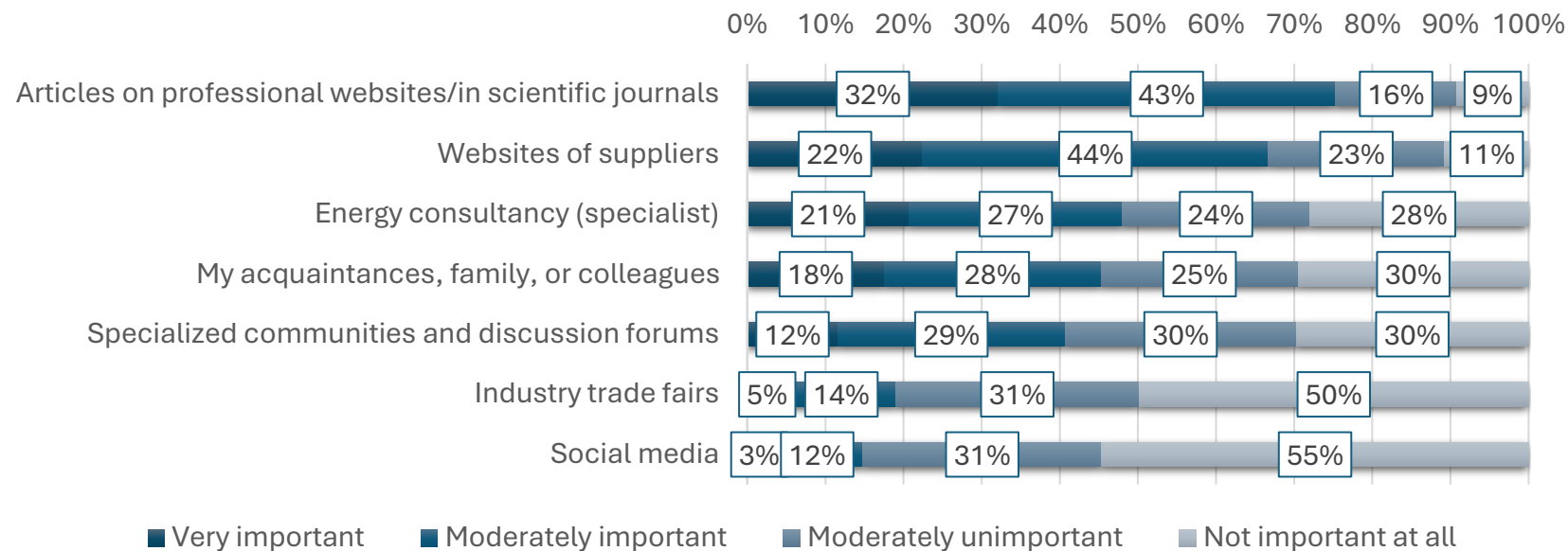
Response bias

“Yes-to-all” effect → normalized values

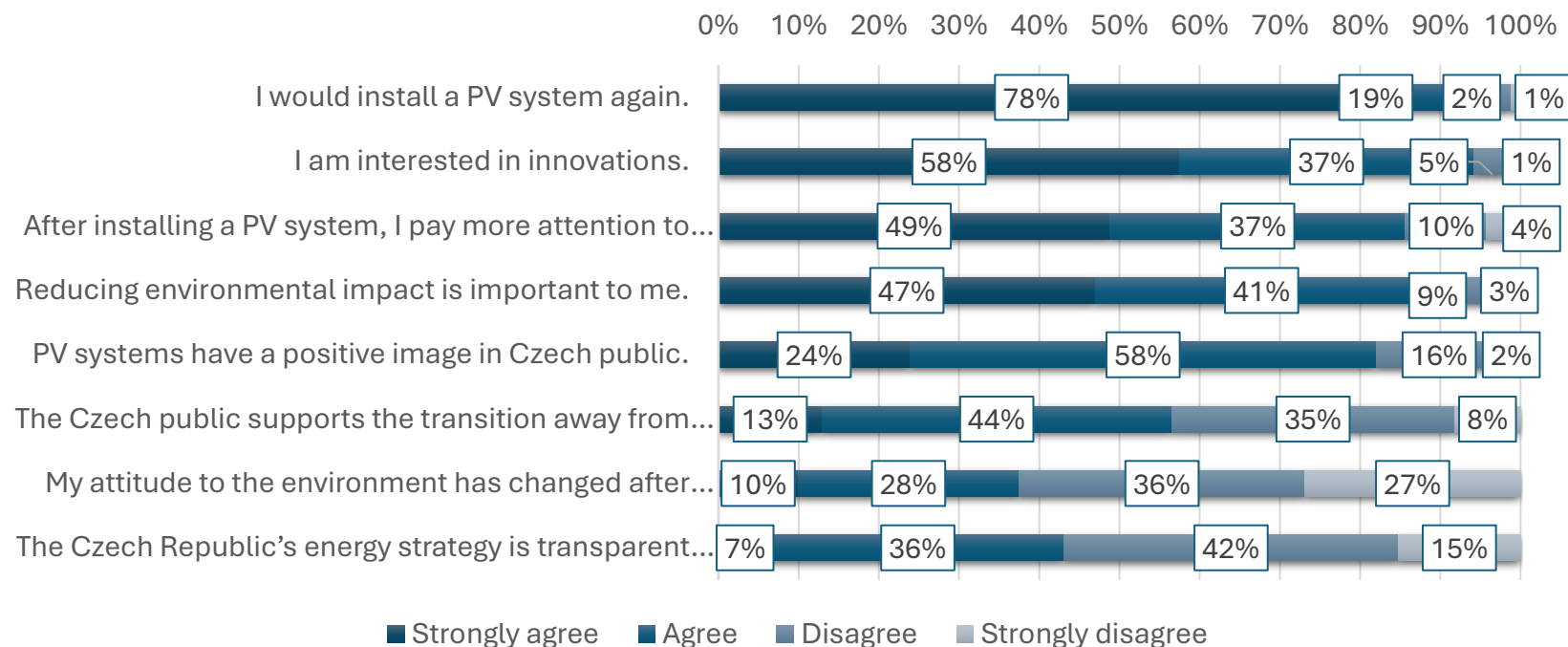
Sample



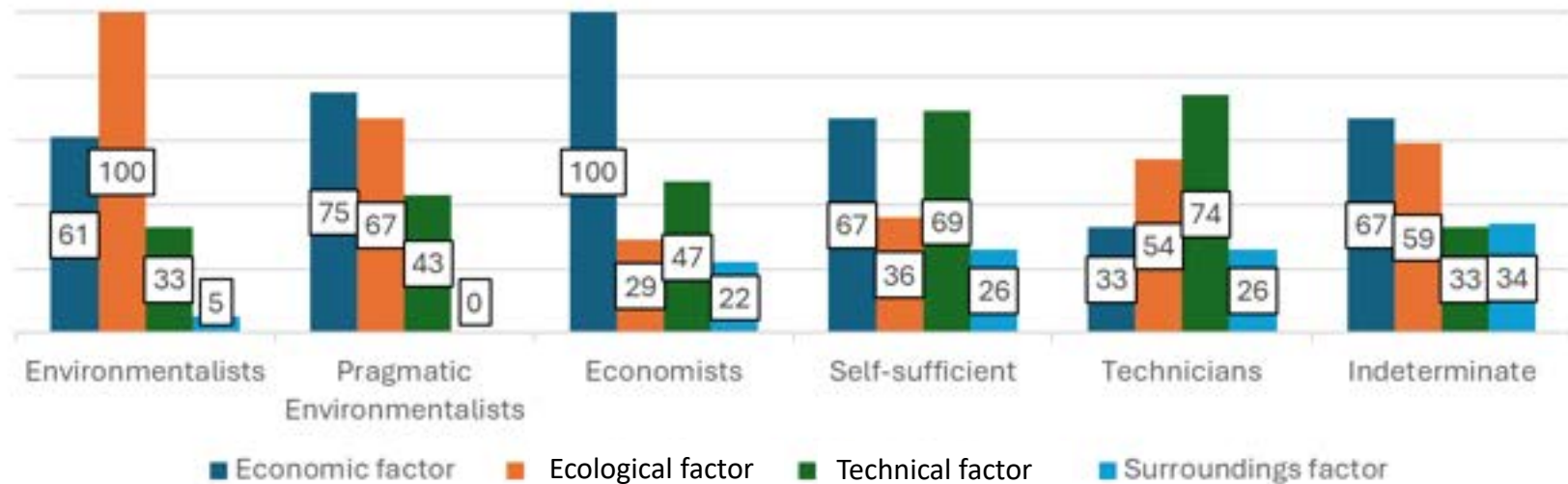
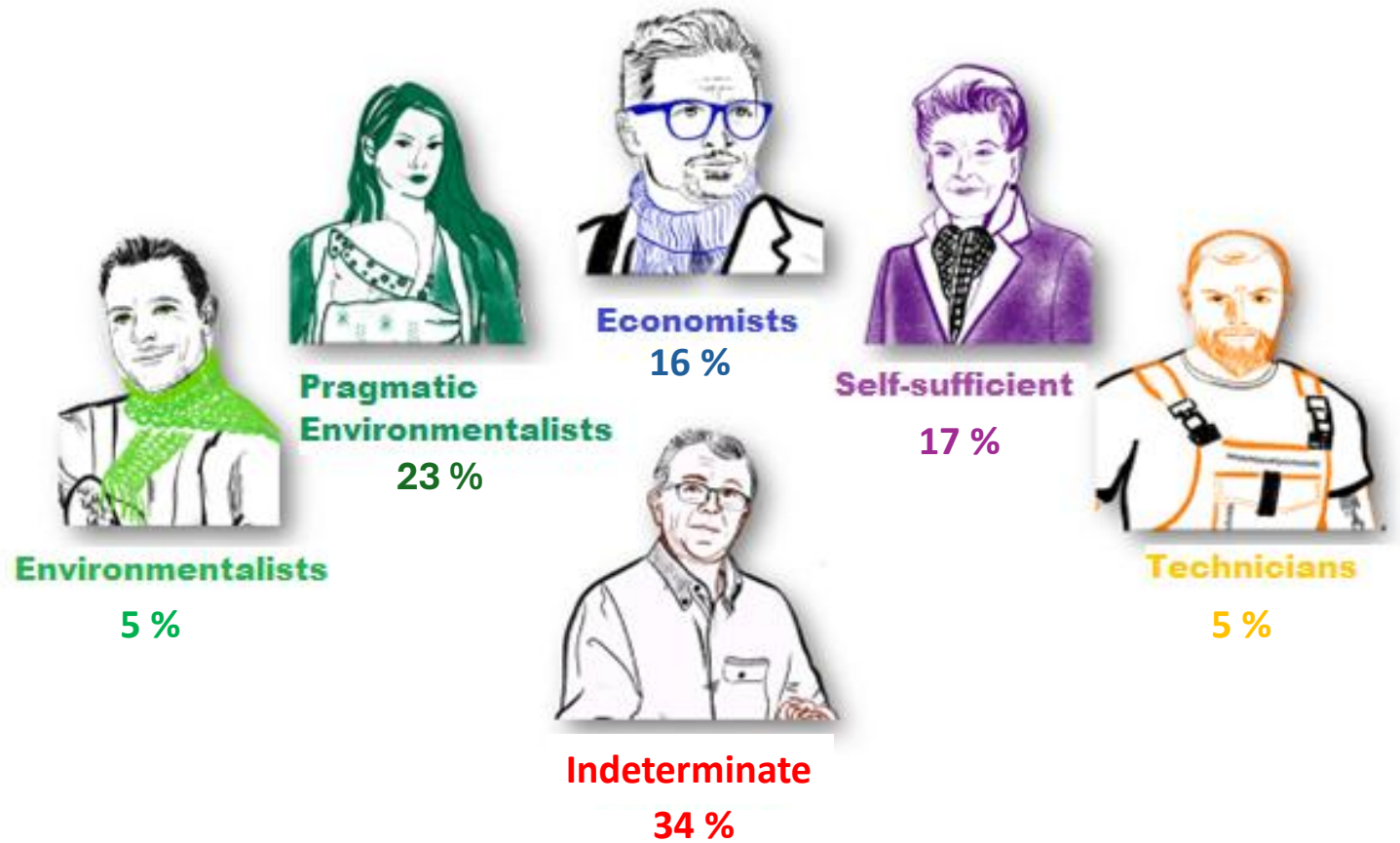
Sources of information



Attitudes



Six classes of investors



Environmentalists (5 %)



- Main motive: **ecology** (100% see as very important)
- Younger (<45), higher education & income, big cities (>50k)
- Strong interest in innovation, **often own/plan EVs**
- 98% would reinvest in PV
- Rely on **online sources** for information
- Stable, deeply rooted environmental attitudes

Pragmatic Environmentalists (23 %)



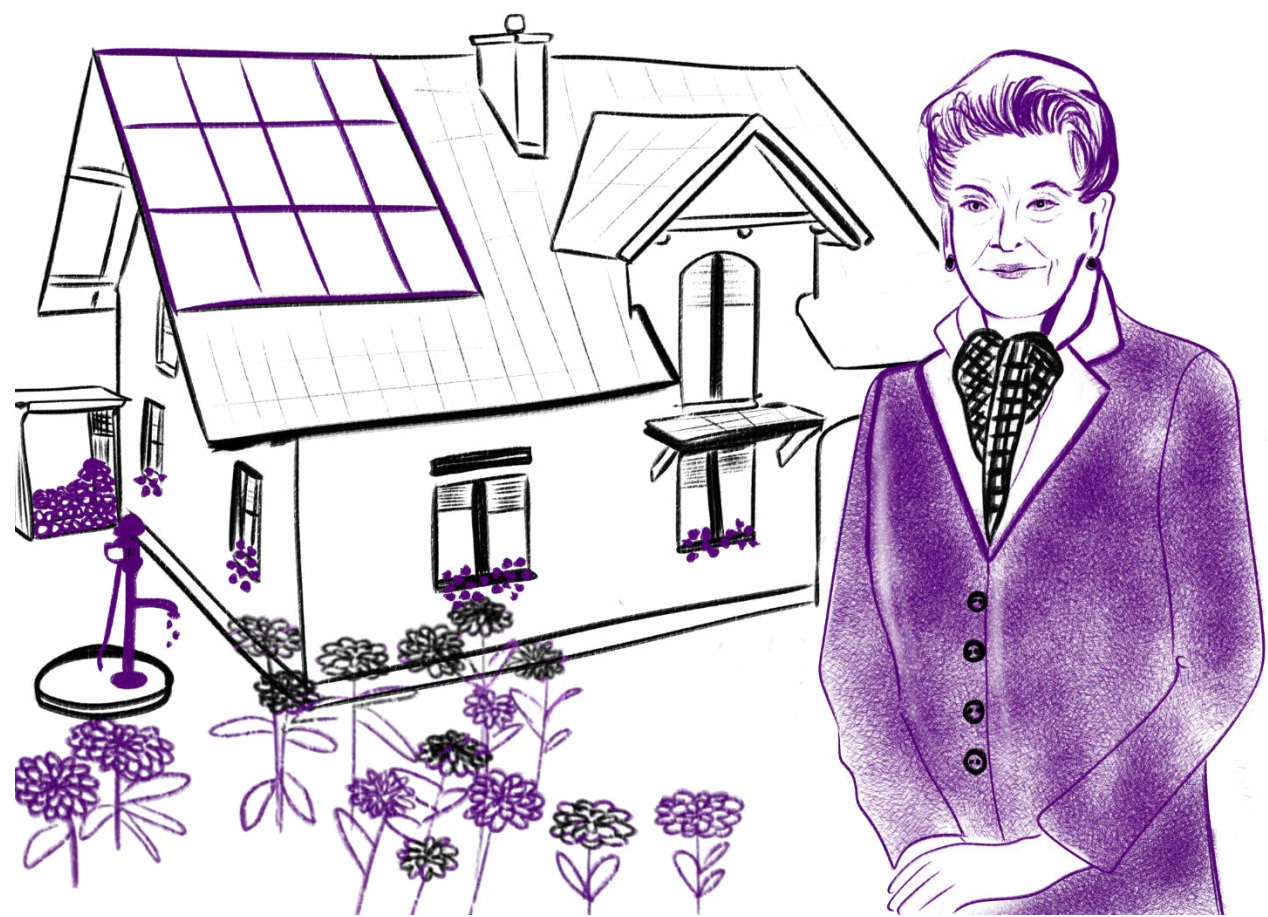
- Balanced motive: **ecology + economics**
- Younger (under 45), low education less common
- High trust in **public support for RES** (65%)
- Strong interest in innovation
- Prefer **professional literature & forums**
- Likely to reinvest, but less than Environmentalists

Economists (16 %)



- Motive: **financial benefit only**
- Mid-income, mid-sized cities
- Skeptical of politics & public support for RES
- Least EV ownership & plans, fewest batteries
- Information: mainly **expert online sources**
- Hold the **most fixed and strong opinions**

Self-sufficient (17 %)



- Motive: **independence from the grid** (economics + technical + surroundings)
- Aged 56–65, higher incomes
- Strong **preference for batteries**
- EV only if it “pays off”
- Skeptical of politics, do not rely on sources
- Less likely to reinvest compared to others

Technicians (5 %)



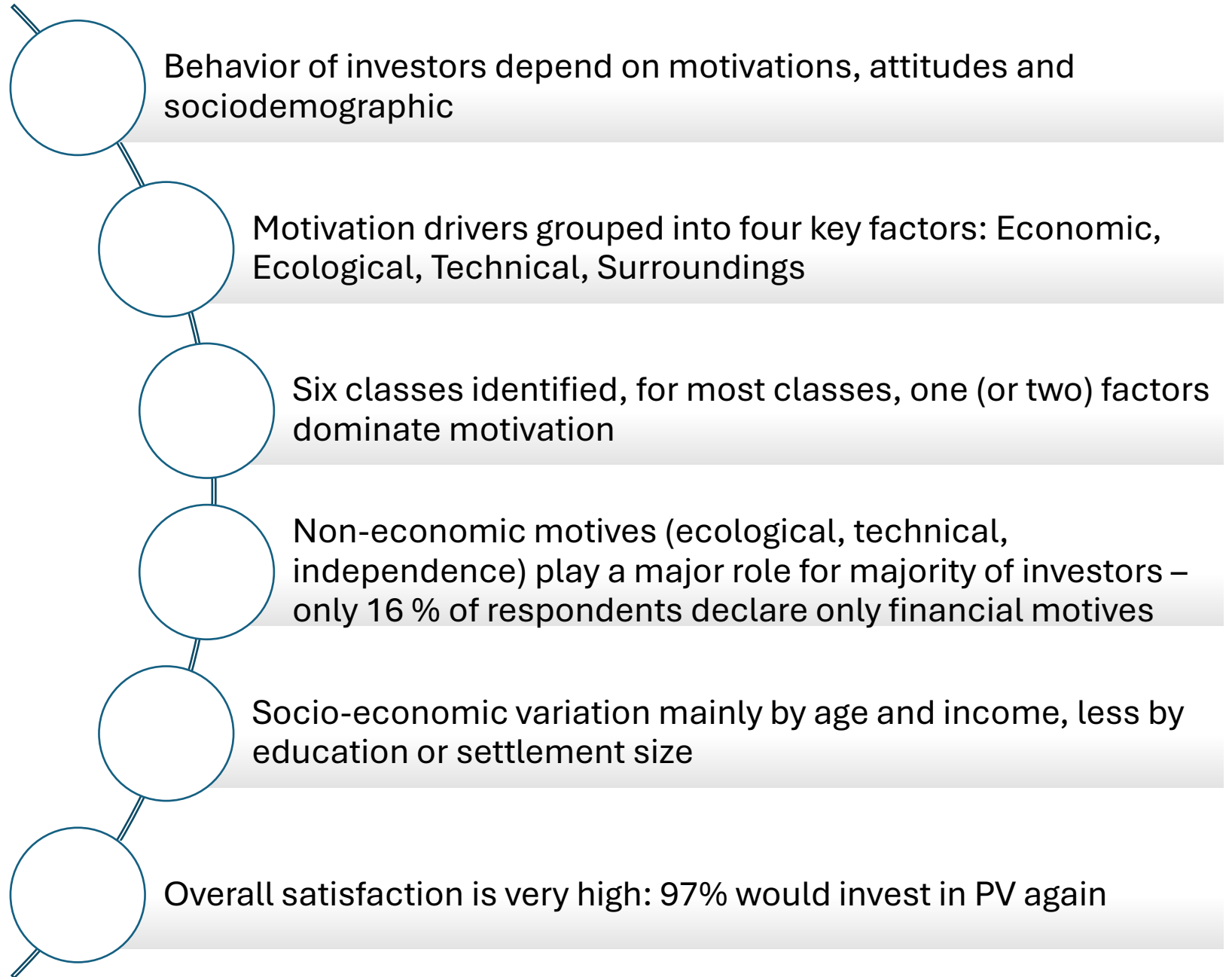
- Motive: **technical aspects**, least focus on economics
- Higher incomes, more men, smaller towns
- High **battery adoption**
- EVs more often considered for future
- Skeptical of policies & public support for RES
- 8% would not install again their PV (highest share)

Indeterminate (34 %)



- **Mixed motives**, economics slightly dominant
- Oldest group (65+), mid & lower incomes
- Often without batteries, PV part of renovation projects
- Strongest **trust in state policy** (55%)
- Most likely to change attitudes after installation (53%)
- Believe in **positive PV image** (90%)

Summary



Limitations and Following research

Limitations

- Sample only includes **current PV adopters**, not non-adopters or those unable to invest (e.g., tenants in apartment buildings)
- February 2022
- LCA method: assumptions of independence between classes + challenges of interpretation
- Results depend strongly on **data preprocessing** and normalization strategies

Future Research

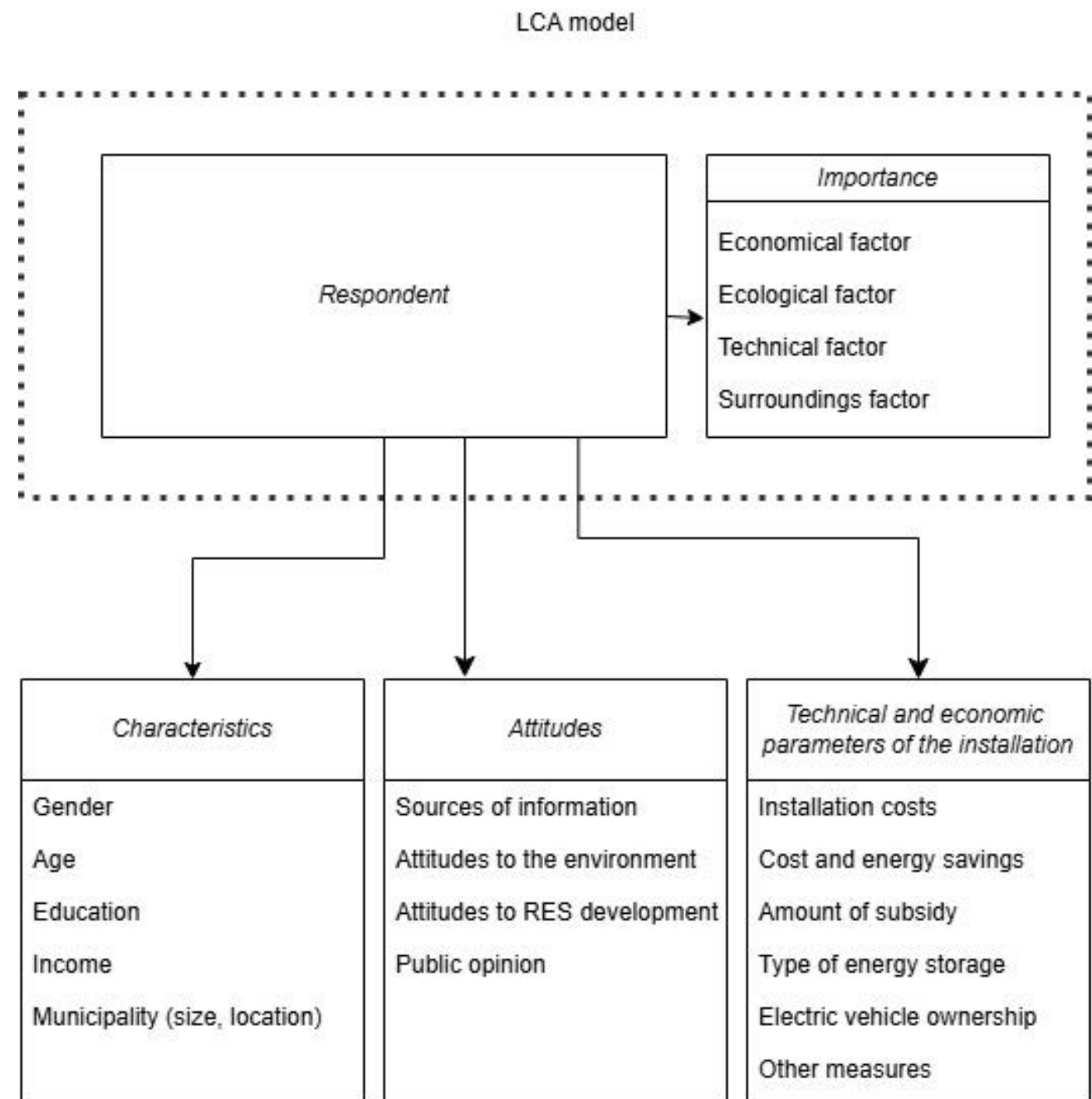
- Explore **non-adopters** and their motivations, barriers, and perceptions
- Apply LCA to other **low-carbon technologies** (heat pumps, EVs, building renovation)
- Study **energy communities and collective models** in energy transition
- Further refine methods to address **response bias (“yes-people”)** and improve interpretability
- Assess how **tailored policy narratives** can increase acceptance and effectiveness

Q&A

Thank you for your
attention!



Model



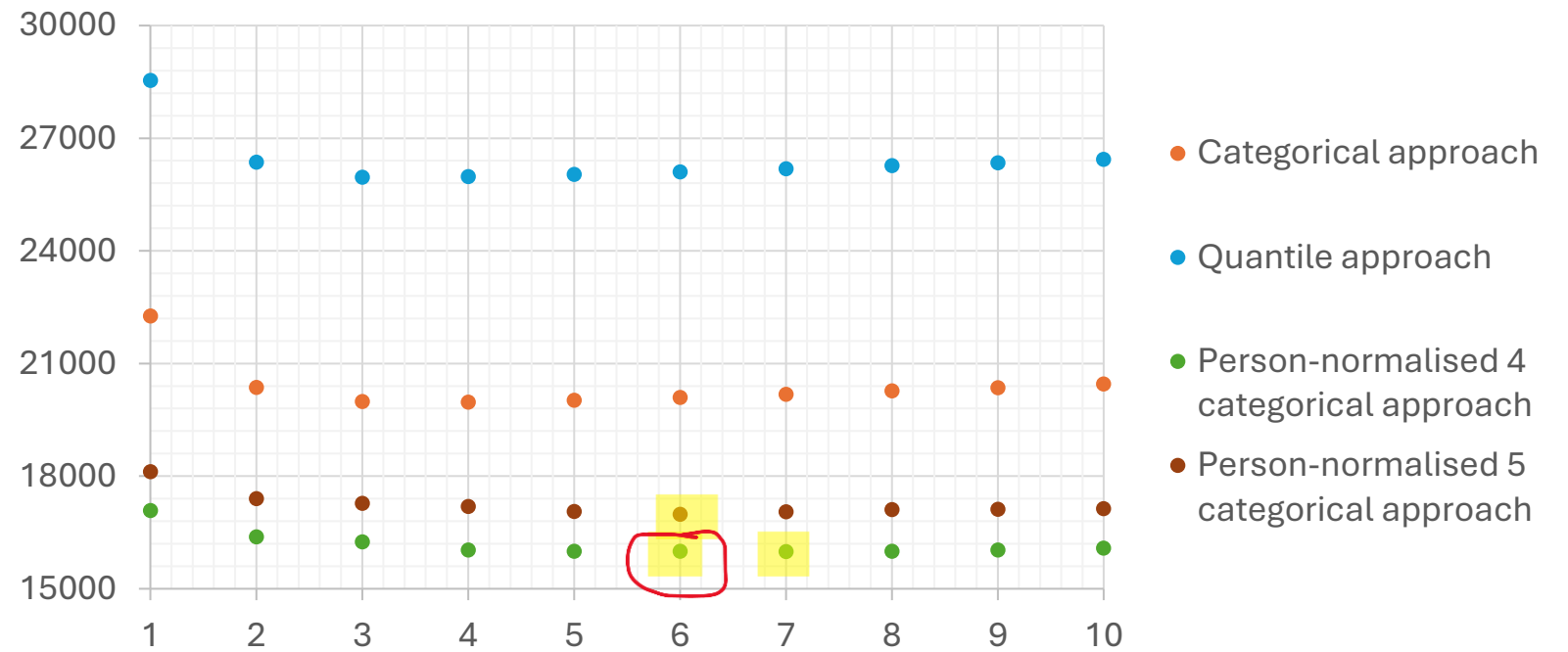
Data
preparation

Discrete
requirements
of R

Yes-person
bias

- Categorical approach: Continuous factor importance values were categorized into four discrete levels, where category 1 represented the range $[0, 0.25)$, category 2 represented $[0.25, 0.5)$, and so on.
- Quantile approach: Data were divided into four quantiles, where each category represents a quartile of respondents. Category 1 included values up to the first quartile, category 2 up to the second quartile, and so on.
- Person-normalised 4 categorical approach and (4) Person-normalised 5 categorical approach: To minimize response bias, values of factors scores were standardized by subtracting the mean value of each respondent. The standardized data were then categorized into four (Person-normalised 4 categorical approach) and five (Person-normalised 5 categorical approach) levels of the same range, respectively.

BIC criteria



Entropy

