### methodological challenges and theoretical opportunities of collecting large personal networks in large samples

gert stulp gertstulp.com



# Disclaimer

# LARGE NETWORKS



# LARGE SAMPLES









Year



one kind of social interaction, informal conversations with networks of relatives, friends, and neighbours, was important for historical change in bedroom behavior

**WATKINS 1995** 

# Social Influence & Fertility

### **Spatial Analysis of the Causes of Fertility Decline** in Prussia

### JOSHUA R. GOLDSTEIN SEBASTIAN KLÜSENER

in the dependent variable unoders in the Ausonate change in th



### historical evidence | convenience samples | qualitative studies

### **Does Fertility Behavior Spread among Friends?**

### Nicoletta Balbo<sup>a</sup> and Nicola Barban<sup>b</sup>

American Sociological Review 2014, Vol. 79(3) 412-431 © American Sociological Association 2014 DOI: 10.1177/0003122414531596 http://asr.sagepub.com



### **Channels of social influence on reproduction**

LAURA BERNARDI Max Planck Institute for Demographic Research

> social learning social contagion social pressure social support



# quantifying social influences on fertility behaviour using personal network data

# PART I

# PART II





# PART III



### Ī RT P А







# PART III



# Bigger Is Better (?)



### weak ties

### structure characteristics









Number of alters







Social Networks

Volume 32, Issue 2, May 2010, Pages 105-111

### Does the online collection of ego-centered network data reduce data quality? An experimental comparison

Uwe Matzat <sup>A</sup> <sup>⊠</sup>, Chris Snijders







# 

# (+H)NSI

### Graphical Ego-centered Network Survey Interface





journal homepage: www.elsevier.com/locate/socnet

### GENSI: A new graphical tool to collect ego-centered network data

### Tobias H. Stark<sup>a,\*</sup>, Jon A. Krosnick<sup>b</sup>

SEVIER

<sup>a</sup> Utrecht University/ICS, Padualaan 14, 3584 CH Utrecht, The Netherlands <sup>b</sup> Stanford University, 450 Serra Mall, Stanford, CA 94305, United States



Contents lists available at ScienceDirect

### Social Networks







# 

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### compard to standard survey-methods,

people who used GENSI:

- enjoyed the survey more
- thought the survey was more interesting

Contents lists available at ScienceDirect



- said they were more willing to participate in a future survey

# 

### Graphical Ego-centered Network Survey Interface





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### Tobias H. Stark<sup>a,\*</sup>, Jon A. Krosnick<sup>b</sup>

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> "A practical limitation for future research with GENSI is that the tool is only suitable for small ego-centered networks. When the number of alters exceeds seven or eight, it gets visually challenging to see all circles in a network."

Contents lists available at ScienceDirect

### Social Networks







# GENSI

# Collecting large personal networks in representative sample of Dutch women, using GENSI

# Methodology



### Longitudinal Internet Studies for the Social sciences

True probability sample of households drawn from the population register. Respondents participate in monthly Internet surveys. Extensive background information available on respondent High retention rates (e.g., 70 %)



Incentive: 12.50 euro Period of 1 month (~ march)

```
All women between 18 - 40 asked (N = 1322)
N = 758 responded (57%); age: 29 (± 6)
```

# Methodology



### Detailed fertility intentions

Alters (25)



Sex Age Education Relationship type Closeness Frequency of con Frequency of othe

	Number and age of children
	Friend
	Wants children
9	Does not want children
	Help with children
ntact F2F	Talk about children
ner contact	Relationship with other alters

# GENSI: Name Generator

Please list 25 names of individuals 18 years or older with whom you have had contact in the last year. This can be face-to-face contact, but also contact via phone, internet, or email. You know these people and these people also know you from your name or face (think of friends, family, acquaintances, et cetera). You could reach out to these people if you would have to. Please name your partner in case you have one.



# GENSI: 5 response options



Heel hecht	Hecht	

### How close are you to these people?

# GENSI: Alter-Alter-ties

Als het gaat om ANNE

Met wie heeft ANNE contact? Met contact bedoelen we alle vormen van contact, zoals face-to-face contact, contact via (mobiele) telefoon, post, email, sms, en andere manieren van online en offline communicatie.

Selecteer de personen die contact met elkaar hebben door met de muis op het bolletje te klikken. Er zal een lijn ontstaan die aangeeft dat de personen contact met elkaar hebben. Druk nogmaals op het bolletje om de lijn weer te laten verdwijnen, als de personen geen contact met elkaar hebben.



### Screencastify L





### THE TASK coming up with 25 names, answering 16 questions about all alters, evaluating 300 alter-alter ties

### THE RESULT 50% within 21 minutes

### 97% hardly any missing values

![](_page_23_Picture_0.jpeg)

![](_page_23_Figure_1.jpeg)

Alter name #

 $\mathsf{N}=\mathbf{654}$ 

### Listing the last alters took about twice as long as listing the first alters, but still only about 10 seconds

![](_page_24_Figure_1.jpeg)

![](_page_24_Figure_2.jpeg)

![](_page_25_Picture_0.jpeg)

![](_page_25_Figure_1.jpeg)

![](_page_26_Figure_0.jpeg)

Responding to all alter-question took 15 minutes

Naming 25 alters took 3.5 minutes

Listing all alter-alter-ties took 3.5 minutes

# Conclusion

### Collecting large personal networks feasible Not too time-consuming Little missing data Data quality?

![](_page_28_Figure_0.jpeg)

![](_page_28_Figure_1.jpeg)

![](_page_29_Figure_0.jpeg)

### 89% of all possible ties were reported

![](_page_29_Figure_2.jpeg)

### 87% of all possible ties were reported

# Conclusion

### Collecting large personal networks feasible Not too time-consuming Little missing data Data quality?

# GENSI useful for large(r) networks Improved user experience?

### 65% of the respondents enjoyed filling in the survey, whereas 10% did not enjoy it so much

![](_page_31_Figure_1.jpeg)

![](_page_31_Figure_2.jpeg)

# Conclusion

### Collecting large personal networks feasible Not too time-consuming Little missing data Data quality?

# GENSI useful for large(r) networks Improved user experience?

## Valuable data

# (+H)NSI

### **Programmed in JavaScript**

### Pros

- "Light" (1 Mb)
- Works with any device with browser
- Can be implemented in other surveys
- Free

### www.tobiasstark.nl/GENSI www.gertstulp.nl/GENSI

### Cons

- Not ideal for mobile phones (currently)
- Answers can't be saved for later use
- Being able to "go back" requires considerable coding

# Alternatives

![](_page_34_Picture_1.jpeg)

# OpenEddi

# **JENILE**

### https://www.networkcanvas.com/

### https://github.com/jfaganUK/openeddi3

https://www.gentle.eu/

![](_page_34_Picture_7.jpeg)

# PART I

![](_page_35_Figure_1.jpeg)

![](_page_35_Figure_2.jpeg)

![](_page_35_Picture_3.jpeg)

### Marie Stadel

![](_page_35_Picture_5.jpeg)

# PART III

![](_page_35_Figure_7.jpeg)

# Balancing Bias and Burden

scientific interest

weak ties network structure network composition

![](_page_36_Figure_3.jpeg)

### respondent burden

time boredom poor(er) response

# Quantifying Bias

### evaluating two strategies to reduce burden by lowering number of alters

# 1. dropping alters 5 3 4 9 6 8

![](_page_37_Figure_3.jpeg)

# Quantifying Bias

network structure Density Proportion of Isolates Maximum Degree **Degree Centralisation Betweenness Centralisation** Mean Betweenness Centrality Maximum Betweenness Centrality **Closeness Centralisation** Mean Closeness Centrality Maximum Closeness Centrality

network composition Average and SD of: Alter age Closeness Frequency of F2F contact Frequency of other contact Education **Proportion of:** Female Alters **Friends** Kin

![](_page_39_Picture_0.jpeg)

### https://socialsciencemethods.shinyapps.io/BalancingBiasAndBurden

![](_page_39_Picture_2.jpeg)

# Conclusions

### Lowering number alters increases bias 15-20 'sufficient' for most measures

### Randomly sampling alters superior to dropping alters More consistent, less bias

### More bias in structural versus compositional measures Huge variation

# Practical Guide

# A potentially useful strategy:

Eliciting large number of alters Alter-alter-ties for random sample 2) Alter attributes for smaller subsample 3)

Results can serve as guide for novel data collection https://socialsciencemethods.shinyapps.io/BalancingBiasAndBurden Carefully examine outcome Amount of bias versus time gains Time gains through different type of questions

# Practical Guide

# A potentially useful strategy:

Eliciting large number of alters Alter-alter-ties for random sample Alter attributes for smaller subsample 3)

Results can serve as guide for novel data collection https://socialsciencemethods.shinyapps.io/BalancingBiasAndBurden **Carefully examine outcome** Amount of bias versus time gains Time gains through different type of questions

![](_page_42_Picture_5.jpeg)

### **Results May Vary** "representative" survey experience paid well

![](_page_42_Picture_7.jpeg)

# PART I

# PART II

![](_page_43_Figure_2.jpeg)

![](_page_43_Picture_3.jpeg)

# PART III

![](_page_43_Picture_5.jpeg)

### Vera Buijs

![](_page_43_Picture_7.jpeg)

# Friends, Family, Family Friends

### friends

# family

### family of choice close seen often long-term

![](_page_44_Picture_4.jpeg)

# "Friends"

inconsistent concept

people vary in use "residual category"

### close people you want to see often

high-quality relation

### role relation

### mutual agreement role-related norms

![](_page_45_Picture_7.jpeg)

### predicting who is considerd a friend among kin and non-kin using three measures of tie strength:

closeness frequency of f2f contact frequency of other forms of contact

### Personal characteristics (e.g. age of respondent)

Alters (25 names)

Origin of the relationship ("What is your relationship with <name> or how do you know him/her?")

> Relationship characteristics (e.g. closeness to alter, per alter)

Friendship ("Which of these people do you consider a friend?")

# SETUP

![](_page_47_Figure_0.jpeg)

### 701 respondents reporting on 17,525 alters classified 7,331 as friends

### on average 10 friends (SD = 5)

### Friend certainly not orthogonal to family

High-school 1100 College 1806 Primary school 514 Partner 489 Social activity 1717 Partner's friends 903 Mutual acquaintance 1295 Neighbourhood 717 Other 98 Work 2571 Sibling 1190 Kin 2485 Parent 1226 In-law 1324 13% 0

25

16%

15%

![](_page_48_Figure_3.jpeg)

![](_page_49_Figure_0.jpeg)

![](_page_49_Figure_1.jpeg)

### Closeness strong predictor of friendship particularly in non-family, not close people also considered friends

# Frequency of face-to-face contact weaker predictor, different effect in family versus non-family

![](_page_50_Figure_1.jpeg)

# Frequency of other forms of contact consistently predicts friendship, but much weaker than closeness

![](_page_51_Figure_1.jpeg)

# Prediction

### Prediction accuracy of friendship based on measures of tie strength: [closeness, frequency of f2f contact, frequency of other forms of contact]

Family

![](_page_52_Picture_3.jpeg)

![](_page_52_Picture_4.jpeg)

# Prediction

Prediction accuracy of friendship based on measures of tie strength: [closeness, frequency of f2f contact, frequency of other forms of contact]

Family

![](_page_53_Picture_3.jpeg)

![](_page_53_Picture_5.jpeg)

# Prediction

Prediction accuracy of friendship based on measures of tie strength: [closeness, frequency of f2f contact, frequency of other forms of contact]

Family

![](_page_54_Picture_3.jpeg)

![](_page_54_Picture_5.jpeg)

![](_page_55_Picture_0.jpeg)

people vary in use "residual category" inconsistent concept

![](_page_55_Picture_3.jpeg)

### close people you want to see often

high-quality relation

> role relation

mutual agreement role-related norms

### Kitts & Leal 2021 [GO READ!]

![](_page_55_Picture_9.jpeg)

![](_page_55_Picture_10.jpeg)

# Asking for a friend...

### when using name generators:

- asking for friends might give you in-laws
- asking for family might give you friends
- people vary in use, some unpredictable • asking for close, frequently seen people might not give friends some predictable (e.g. age, sex)

### probably too vague a concept to be used in scientific research

Claude Fischer (1982)

### when used as classification:

 friend not orthogonal to family, neighbours, colleagues

![](_page_56_Picture_9.jpeg)

# PART I

# PART II

![](_page_57_Figure_2.jpeg)

![](_page_57_Picture_3.jpeg)

# PART III

![](_page_57_Figure_5.jpeg)

# Shrinking kin-networks

![](_page_58_Figure_1.jpeg)

![](_page_58_Picture_4.jpeg)

![](_page_59_Figure_0.jpeg)

![](_page_60_Figure_0.jpeg)

![](_page_60_Figure_3.jpeg)

### methodological challenges and theoretical opportunities of collecting large personal networks in large samples

- Sinlp, G. Collecting large personal networks in a representative sample of Dutch women. Social Networks Baris, YL & Sinlp, G.
- Family, and Family Friends: Predicting Friendships of Dutch Women Stadel, M & Stulp, G. (2021).

Balancing Bias and Burden in Personal Network Studies.

• Stulp, G & Barrett, L.

Do data from large personal networks support cultural evolutionary ideas about kin and fertility?

gert stulp gertstulp.com

![](_page_61_Picture_7.jpeg)