

## Introduction

### Collaborators



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- 1. Technological advances More efficient refrigerators, tv's, etc.
- 2. Changing energy behaviour This requires understanding energy behaviour

### Special characteristics of psychological research

- Distinction between independent and dependent variables often not clear: everything affects everything
- Causal relations are hard. Often variables cause each other (A  $\rightarrow$  B and B  $\rightarrow$  A)
- Many variables involved. Often many overlapping theories involved.
- Technical background of psychologists more limited than of, e.g., biologists. Results should be interpretable for non-statisticians.

Solution: use psychological network models.

### From latent variable models to network models



(Cramer et al., 2010)

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(Cramer et al., 2010)

### Example: General Anxiety Disorder & Major Depression



(Borsboom & Cramer, 2013)

In essence, psychological network models are:

A set of nodes and edges, depicting random variables and their relations.

Gaussian graphical models are a specific case of psychological network models.

#### Networks

### • Network of things

### Psychological networks

• Network of variables

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- Examples: social networks, traffic networks

- Network of variables
- Examples: correlation networks, depression symptom models

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- Examples: correlation networks, depression symptom models
- Nodes represent random variables

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### Psychological networks

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- Examples: correlation networks, depression symptom models
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- Edges are estimated statistical relations, subject to uncertainty
- Main goal: explain (co)occurrence of certain nodes
- Conceptually improvement to latent variable models

(Table: Sacha Epskamp)

### $X \sim N(\mu, \Sigma), \quad X: n \times p$

Example: grades for n = 44 students on p = 5 grades. Correlations:

	mechanics	vectors	algebra	analysis	statistics
mechanics	1.00	0.32	0.41	0.38	0.27
vectors	0.32	1.00	0.43	0.23	0.22
algebra	0.41	0.43	1.00	0.58	0.55
analysis	0.38	0.23	0.58	1.00	0.51
statistics	0.27	0.22	0.55	0.51	1.00

### $X \sim N(\mu, \Sigma), \quad X: n \times p$

Example: grades for n = 88 students on p = 5 grades. Correlations (from  $\hat{\Sigma}$ ) and partial correlations (from  $\hat{\Sigma}^{-1}$ ):



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Example: grades for n = 88 students on p = 5 grades. Correlations (from  $\hat{\Sigma}$ ) and partial correlations (from  $\hat{\Sigma}^{-1}$ ):



### Psychological networks: fewer edges

 $I \text{ nodes} \rightarrow I(I-1)/2 \text{ potential edges. Quickly too many.}$ 

Graphical lasso. Estimate  $\Sigma^{-1}$  subject to  $L_1$ -penalties. Commonly used is the Graphical lasso (Friedman, Hastie, Tibshirani, Biostatistics, 2008) approach with the extended BIC to select tuning parameter.



### Which nodes matter most?



 $W_{ij}$ : matrix with weights ( $\in [-1, 1]$ ) in the network.  $SD_{ij}$ : shortest distance from node *i* to node *j*.

- Node strength/centrality:  $S_i = \sum_j W_{ij}$
- Closeness:  $CL_i = (\sum_j SD_{ij})^{-1}$
- Betweenness:  $B_i = \#(\text{paths through node } i)/\#(\text{all paths})$

### Comparing networks



#### Below median score for mechanics Above median score for mechanics

### **Comparing networks**



Below median score for mechanics Above median score for mechanics

• Structural Hamming Distance.

Count number of edges that (dis)appeared/changed sign. Here: 5 out of 10.

Network Comparison Test (Van Borculo et al.)
 Statistical permutation test, similar to Mantel's test (1967).
 Here: p = .62.

In this talk, only undirected GGM.

Network models also possible for:

- Directed graphs, when visualising temporal dynamics (e.g. VAR-models; Bringmann et al.) or causal models (DAGs).
- Ordinal data: polychoric rather than Pearson correlations.
- Binary data: Ising-models.

Furthermore, all estimates (e.g. edge weights) can be equipped by bootstrap CI's (Epskamp et al, 2017).

## **Environmental Applications**

### **Buurkracht: Introduction**

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- Community energy initiative for promoting sustainable energy behaviour.
- Research on energy behaviour usually focused on individual.
- Community efforts might be more effective.
- Combination of various psychological and sociological theories at play.

### **Buurkracht: Introduction**

# **Buurkracht**<sup>\*</sup>

- Community energy initiative for promoting sustainable energy behaviour.
- Research on energy behaviour usually focused on individual.
- Community efforts might be more effective.
- Combination of various psychological and sociological theories at play.
- $N_1 = 334$  initiative participants,  $N_2 = 360$  right-door-neighbours.
- Data on 22 variables, and 65 items, 7-point Likert scales.

**Personal factors** Values. Environmental self-identity. Personal importance of sustainable energy behaviour. Outcome efficacy.

Social context Need to belong Need to be unique. Neighbourhood identification. Neighbourhood homogeneity. Neighbourhood interaction. Neighbourhood environmental identity. Neighbourhood importance of sustainable energy behaviour.

Opinions on energy companies and the government Group-based anger.

Sustainable energy intentions and behaviour Household sustainable energy intentions. Collective sustainable energy intentions. Collective social intentions. Self-reported sustainable energy behaviour.

Initiative membership

### Buurkracht: Findings



- Altruistic values
- Biospheric Values
- Collective energy intentions
- Collective social intentions
- Egoistic Values
- Environmental Self identity
- Group-based anger
- Hedonic Values
- Homogeneity in neighbourhood
- Individual energy intentions
- Initiative outcome efficacy
- Interaction in neighbourhood
- Interaction with neighbourhood
- Membership
- Need to be Unique
- Need to belong
- Neighbourhood energy norms
- Neighbourhood environmental Identity
- Neighbourhood identification
- Past energy behavior
- Personal Energy Norms
   Personal outcome efficacy

### **Buurkracht: Findings**



#### Factors related to the social context

- 5: Neighbourhood environmental Identity
- 6: Neighbourhood identification
- 7: Interaction with neighbourhood
- 8: Neighbourhood energy norms
- 9: Homogeneity in neighbourhood
- I0: Interaction in neighbourhood
- 13: Need to belong
- 15: Need to be Unique
- 20: Initiative outcome efficacy

#### Membership

o 22: Membership

#### **Opinions on energy companies**

• 21: Group based anger

#### Personal factors

- 11: Personal outcome efficacy
- 12: Environmental Self identity
- 14: Personal Energy Norms
- 16: Altruistic values
- 17: Biospheric Values
- 18: Egoistic Values
- 19: Hedonic Values

#### Sustainable energy intentions

- I: Collective energy intentions
- 2: Collective social intentions
- 3: Individual energy intentions
- 4: Past energy behavior

### **Buurkracht: Findings**



Structural Hamming Distance: 12

**ESS: Introduction** 





- Open Data, www.europeansocialsurvey.org
- "Public Attitudes to Climate Change, Energy Security, and Energy Preferences"
- N = 38,437 participants from 18 countries (880 to 2,766 per country).



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Research questions:

- 1. What, and how strong, are the relations between the variables?
- 2. Are these relations the same across countries?

The module "Public Attitudes to Climate Change, Energy Security, and Energy Preferences" includes 32 items in the areas:

- 1. Beliefs on climate change
- 2. Concerns about climate change and energy security
- 3. Personal norms, efficacy and trust
- 4. Energy preferences.

### ESS: findings



#### Policy support

- PS1: Increasing taxes on fossil fuels
- 9 PS2: Subsidise renewable energy
- PS3: Ban least energy efficient appliances

#### **Climate change beliefs**

- CCB1: Belief in climate change
- CCB2: Climate change cause
- OCCB3: Impact of climate change

#### Climate change salience

· CCS: Thought about climate change before today

#### Climate concern

· CC: Climate change worry

#### Efficacy beliefs

- EB1: Could use less energy
- EB2: Peoples impact on climate change
- · EB3: Likeliness of people limiting energy use
- EB4: Likeliness of poopulariting analysis
- · EB5: My impact on climate change

#### Energy demand measures

- · EDM1: Buy energy efficient appliances
- EDM2: Reduce your energy use

#### Energy security concern

- ESC1: Power cuts
- ESC2: Expensive
- ESC3: Energy imports
- o ESC4: Fossil fuels
- ESC5: Natural disasters
   ESC6: Insufficient power
- eSC7: Technical failurer
- ESC7: Tecnnical failures
   ESC8: Terrorist attacks

#### Energy supply source preference

- · ESP1: Coal
- ESP2: Natural gas
- ESP3: Hydroelectric power
- e ESP4: Nuclear power
- ESP5: Sun or solar power
- o ESP6: Wind power o ESP7: Biomass energy

Personal norms

9 PN: Personal responsibility to reduce climate change

### ESS: findings



No major between-network differences. *K*-means clustering yields 1 cluster.

### ESS: findings





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### Detailed look at five variables



Steinhorst & Matthies, 2016; Thøgersen & Noblet, 2012:

Engaging in **Energy Behaviour** 

Support for **Energy Policies** 

Desire for consistency:

Stronger relation between related concepts

### Detailed look at five variables



#### Behaviour

B1: Buying an energy–efficient appliance B2: Energy–saving behaviour

Policy support P1: Taxes on fossil fuels P2: Subsidies for renewable energy P3: Ban on inefficient appliances

### Country comparison

Number of cluster: 4





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Strong relation between the various correlations per country.

Conclusions

Psychological networks are useful for:

- Exploratory analyses of (relatively) high-dimensional data
- Validation of scales of various items
- Testing hypotheses on the absence/presence, direction or strength of relations
- $\cdot\,$  Testing hypotheses on the co-occurrence of relations

With these models, we explored and confirmed theories on energy-related behaviour and found differences and similarities between countries.

Papers in preparation on this topic:

- N. Bhushan, F. Mohnert, C. J. Albers, L. Jans, L. Steg. The value of Gaussian graphical models to explore relationships between environmental psychological constructs.
- M. Verschoor, C. J. Albers, L. Steg. A network model of the environmental module in the European Social Survey 2016
- T. Bouman, M. Verschoor, L. Steg, G. Böhm, S. D. Fisher, W. Poortinga, L. Whitmarsh, C. J. Albers. *General personal factors predicting energy-saving behaviours and climate policy support*
- E. J. Sharpe, M. Verschoor, L. Steg, G. Perlaviciute, C. J. Albers. *Similarity encourages* consistency across energy-saving behaviour and energy policy support in Europe and beyond

### Further reading - Environmental Psychology

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### Further reading - Psychological Networks

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- Epskamp, S., Borsboom, D., & Fried, E. I. (2017). *Estimating psychological networks and their accuracy: a tutorial paper*. Behavioural Research.