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# Comparison of Component Models in Analysing Dialectal Features

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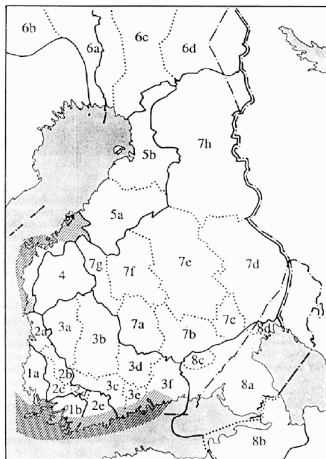




# Background

## Finnish Dialects

- Traditional view, relatively unchanged for the past half century
- Western dialects
  - 1 Southwestern dialects
  - 2 Mid-Southwestern dialects
  - 3 Tavastian dialects
  - 4 Southern Ostrobothnian dialect
  - 5 Central and Northern Ostrobothnian dialects
  - 6 Northernmost dialects
- Eastern dialects
  - 7 Savonian dialects
  - 8 Southeastern dialects





# Corpora

- Traditional Finnish dialectology based largely on morphological and phonological features
- Lauri Kettunen, *Suomen murteet IIIA: murrekartasto* 'Finnish Dialects IIIA: Dialect Atlas', 1940
  - 213 maps
  - Out of print for decades, still widely used
  - Computer corpus (Embleton – Wheeler 1997)
- Lexical variation: *Suomen murteiden sanakirja* 'Dictionary of Finnish Dialects'
  - Ongoing project at the Research Institute for the Languages of Finland: Vol. I 1985, Vol. VIII 2008, Vol. XX c. 2040
  - Here: distribution maps for c. 5 500 articles
  - Used earlier by Hyvönen et al. (2007)



# Objectives

- Earlier work (Hyvönen et al. 2007) indicated that clustering is not ideal for dialect features
- Component analysis works better
  - But which one? So many to choose from
- Compare five different methods to these two corpora
  - Factor Analysis
  - Non-negative Matrix Factorisation
  - Aspect Bernoulli
  - Independent Component Analysis
  - Principal Component Analysis



# Methods Overview

- All these can be thought of as latent variable models
- Aim to find a small number of latent variables / factors / components / aspects that explain the data
- Ideally, factors interpretable in terms of dialect regions
- Each factor can be visualised as a choropleth map, with a colour slide between the extremes



# Methods

## Factor Analysis

- Aims to find a small set of factors which explain the data
- Factors hopefully interpretable
- Here one would expect some correspondence between factors and dialects
- Implementation by Trujillo-Ortiz et al. (2006)



# Methods

## Non-negative Matrix Factorisation

- Aims to find a small set of *non-negative* factors which explain the data in a non-negative way, e.g.
  - Kainuu could be explained using Savonian and Ostrobothnian components with weights 1 and 2 ('1 part Savonian and 2 parts Ostrobothnian')
  - A certain dialect word or feature could be half Tavastian and half Southwestern
- Non-negativity intuitively appealing: what does it mean if a word is '-0.5 Tavastian'?
- Berry et al. (2007)



# Methods

## Aspect Bernoulli

- Designed for binary data
- Interpretation can be given in terms of probabilities
  - E.g. a municipality / dialect word / dialect feature is 83 % Savonian
- Designed to deal with noisy data
  - But as seen later on, too much is too much
- Kaban et al. (2004)





# Methods

## Independent Component Analysis

- Aims to find statistically independent components
- Most often used to separate signals with lots of measurements and a few measurement points
  - E.g. the cocktail party problem: 5 people talking, 5 microphones: separate speech signals
- Here we consider dialect words / features as the signal
- PCA as a pre-processing step: we use only a small number of principal components (otherwise components too localized)
- Hyvärinen et al. (2001)



# Methods

## Principal Component Analysis

- Finds the direction in the data which explains most of the variation
- Each component explains the variation left in the data after the variation explained by previous components has been removed
- Interpretation must be done bearing the previous components in mind
- Hotelling (1933)



# Series I: Dialect Atlas

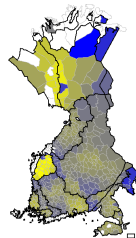
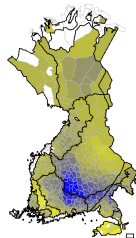
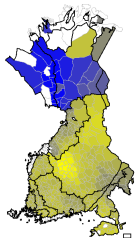
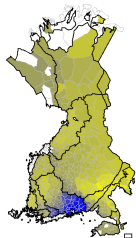
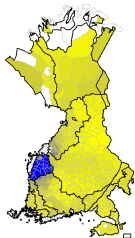
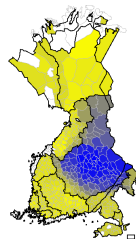
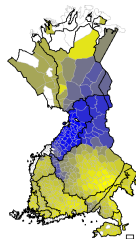
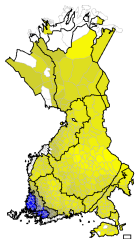
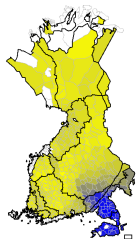
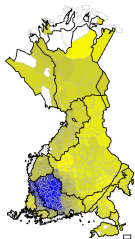
## Overview

- Relatively clean data: no significant gaps
  - Exception: northernmost Finland
- Ten-component run for each method
  - For PCA, first ten components
- Most methods manage this without too many problems
- Some differences



# Series I: Dialect Atlas

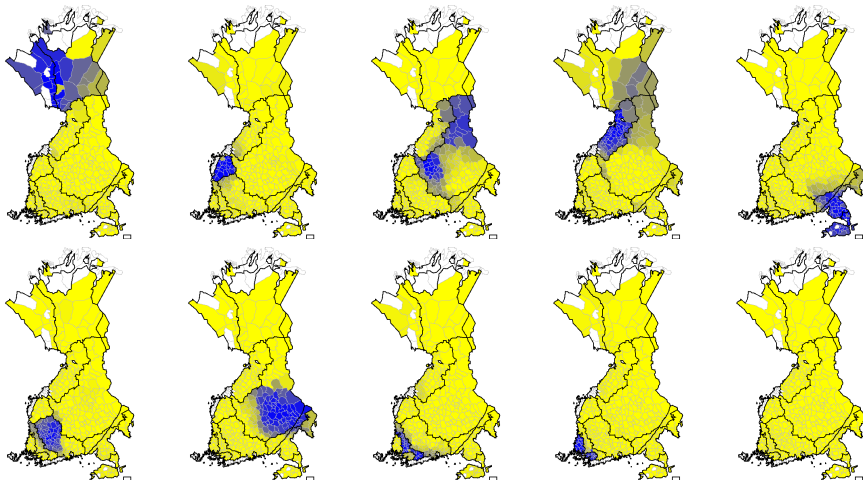
## Factor Analysis





# Series I: Dialect Atlas

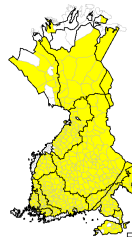
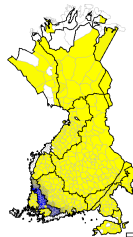
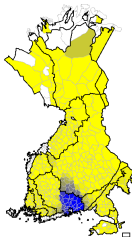
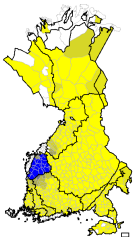
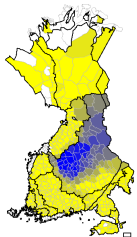
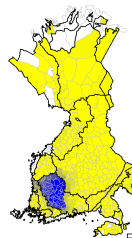
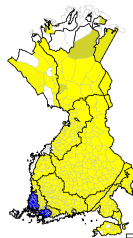
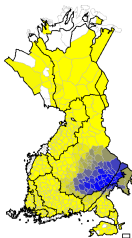
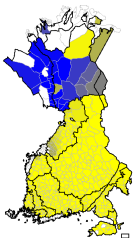
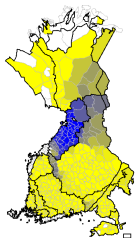
## Non-negative Matrix Factorisation





# Series I: Dialect Atlas

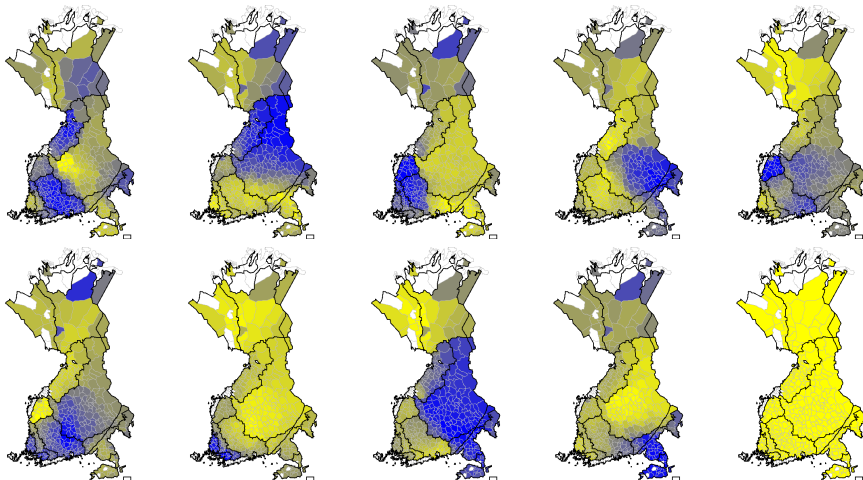
## Aspect Bernoulli





# Series I: Dialect Atlas

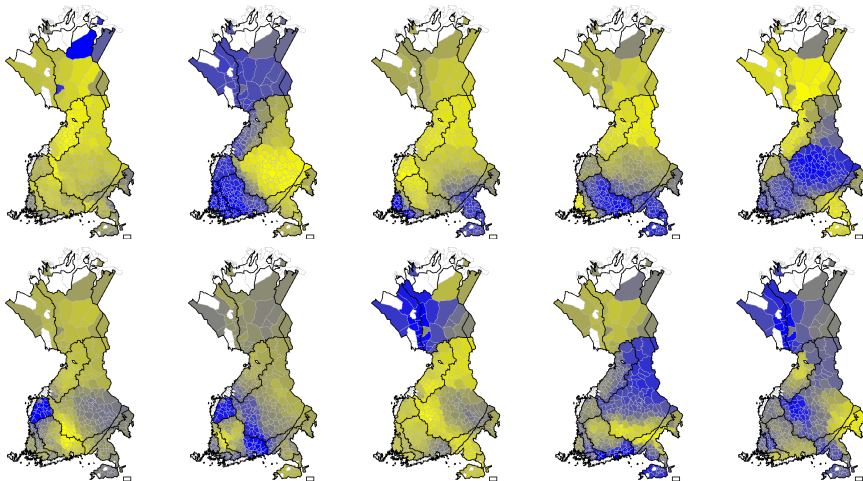
## Independent Component Analysis





# Series I: Dialect Atlas

## Principal Component Analysis







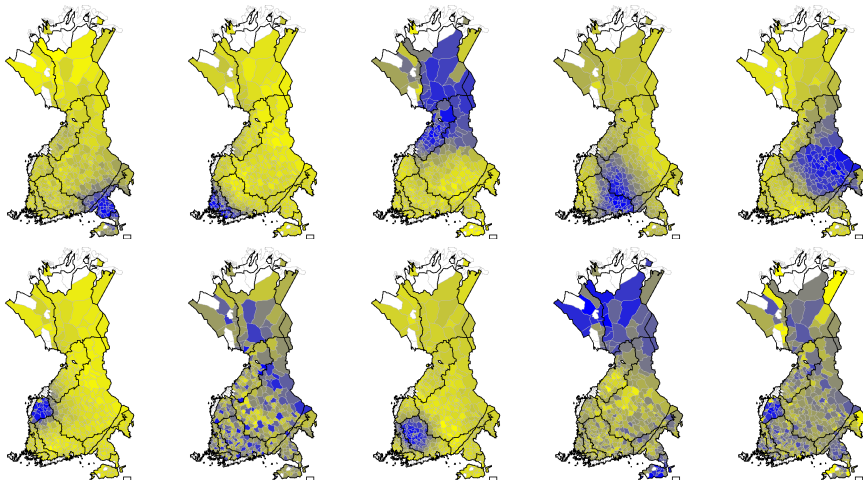
## Series II: Dialect Dictionary Overview

- Much more spotty data
  - Some municipalities thoroughly surveyed
  - Some far less so
- Ten-component run for each method
  - For PCA, first ten components
- Different issues with different methods than in the Dialect Atlas data
- All methods have at least one 'noise' component



# Series II: Dialect Dictionary

## Factor Analysis





# Series II: Dialect Dictionary

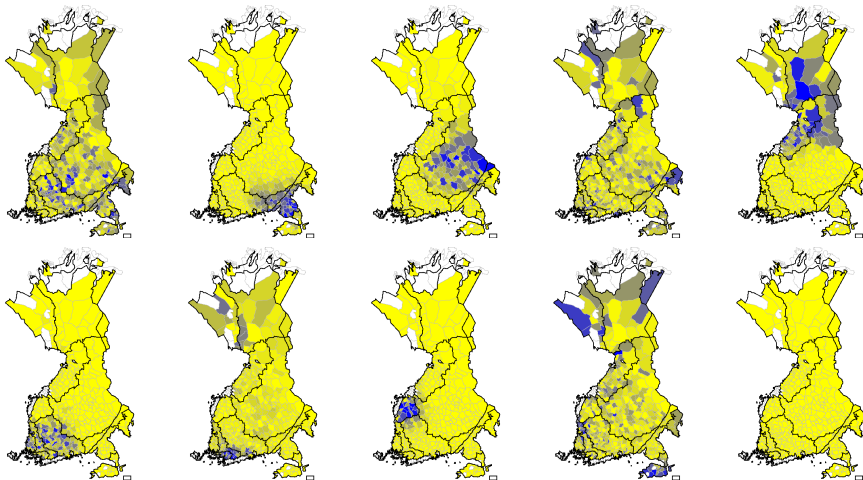
## Non-negative Matrix Factorisation





# Series II: Dialect Dictionary

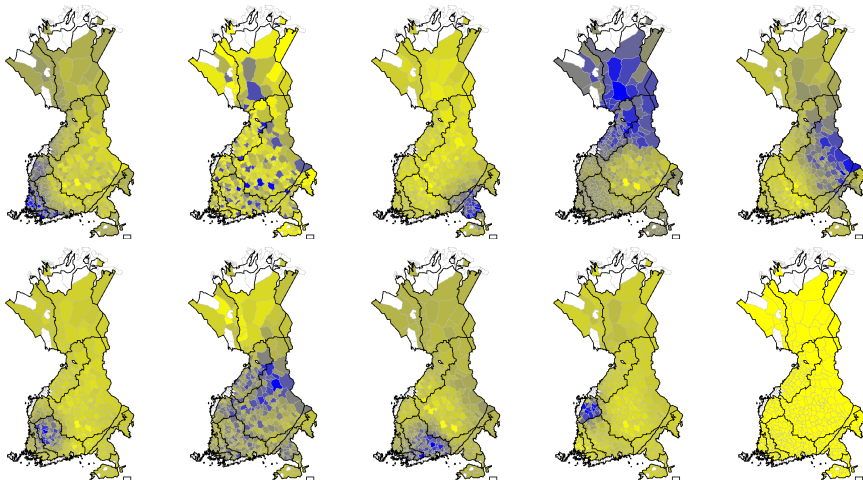
## Aspect Bernoulli





# Series II: Dialect Dictionary

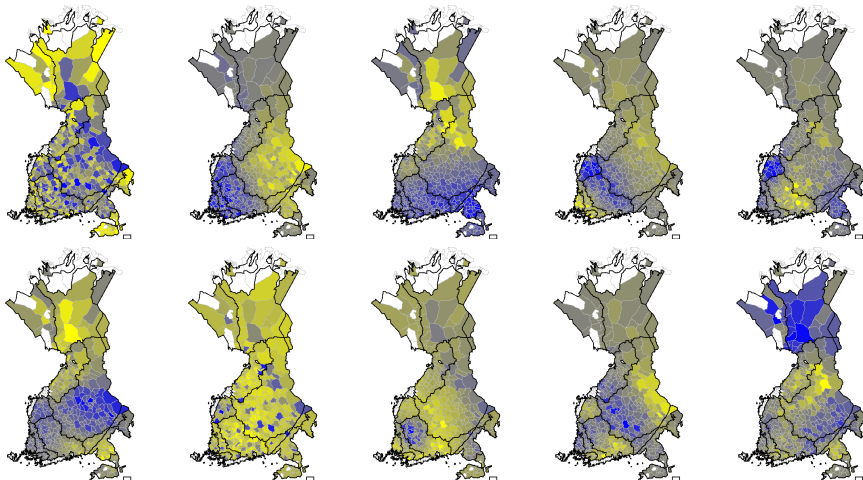
## Independent Component Analysis





# Series II: Dialect Dictionary

## Principal Component Analysis





# Conclusions

- Two different issues with data
  - Amount of variables ( $\approx$  features, words)
  - Amount of noise ( $\approx$  spottiness of data)
- For reasonably clean data, Non-negative Matrix Factorisation and Aspect Bernoulli work well
- For large number of variables Independent Component Analysis works
- Factor Analysis is a good compromise: not the best, but works for both cases
- Principal Component Analysis is very different from the others – use with care



# References

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