

Project in Probabilistic Models  
Spring 2011: Introductory lecture

Antti Honkela

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# Project in Probabilistic Models

Course code: 582637  
Credit points: 2 cr  
Teacher: Dr Antti Honkela (& Prof Petri Myllymäki)  
Contact email: `antti.honkela@cs.helsinki.fi`  
Office hours: Please make an appointment by email  
Prerequisites: 582636 Probabilistic Models

# Graphical model structure learning

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- ▶ Why do we want to learn the structure
  - ▶ Scientific discovery
  - ▶ More efficient density modelling
- ▶ Potential challenges
  - ▶ Uncertainty about the correct structure (weak links, limited data, ...)
  - ▶ Learning of correlation instead of causation, equivalent structures

# The project task

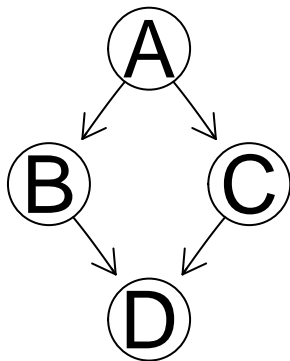
- ▶ Infer the structure and corresponding distributions of a discrete graphical model
- ▶ Data: 2500 samples of 21 variables (all discrete with 3 values)
- ▶ The network connectivity has been extracted from a real network, but the probability model is synthetic
- ▶ Required outputs:
  - ▶ Ranked list of all possible arcs in the model
  - ▶ Normalised probability distribution over a set of 1500 test vectors



## Ranked arc list

- ▶ Return a list of all potential 420 arcs in the model in ranked order with ones you believe to be active in the beginning
- ▶ Example:

A B  
A C  
B D  
C D  
A D  
B A  
B C  
C A  
C B  
D A  
D B  
D C





## Test set probabilities

- ▶ Return a normalised list of probabilities (must sum to 1) for the test vectors (in order)

- ▶ Example:

3.941543e-01

1.637665e-02

3.199843e-01

1.524029e-02

7.287055e-04

1.392376e-03

3.395783e-09

2.521234e-01

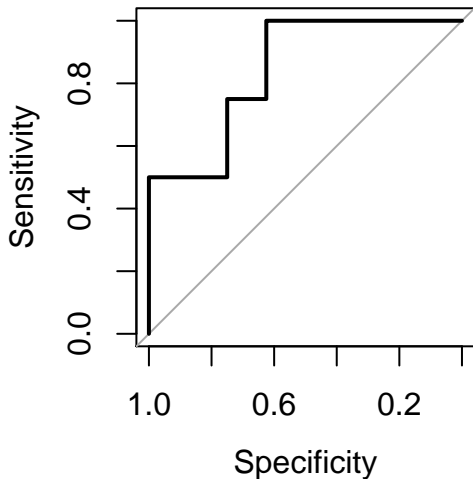
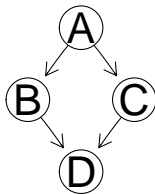
# Evaluation of the predictions

- ▶ Ranked arc list
  - ▶ Evaluated using the area under ROC curve
- ▶ Predicted probabilities
  - ▶ Evaluated using the Kullback–Leibler divergence between the true distribution and the prediction

$$D_{KL} = \sum_i p_i \log \frac{p_i}{q_i}$$

## Area under ROC example

A B  
A C  
A D  
D B  
C D  
B C  
B D  
B A  
C A  
C B  
D A  
D C



Here:  $AUC=0.8438$

# Scoreboard

- ▶ All the scores will be published *anonymously* on a scoreboard together with brief descriptions of the methods used
- ▶ You will receive an email notification with your own score
- ▶ *Positions on the scoreboard will not be used as a criterion for course grading!*

## Return instructions

- ▶ There are four deadlines during the course (always on Tuesdays)
  - ▶ 5 April
  - ▶ 12 April
  - ▶ 19 April
  - ▶ 26 April (final DL)
- ▶ You must return all your submissions to the course Moodle area
- ▶ The return consists of
  - ▶ Predictions as specified above
  - ▶ 1 line public summary of the methods you have used for the score board
  - ▶ 1/2 page diary of your progress

# Final return instructions

- ▶ The final return (26 April) consists of
  - ▶ Your final predictions
  - ▶ 1 line summary of the methods
  - ▶ A written report of the project containing introduction, methods, results and discussion
    - ▶ The weekly diary entries will be included in the report
  - ▶ All source code used

## Using existing software

- ▶ Using existing software in your project *is permitted* if the software is freely available for academic use
  - ▶ Use of commercial packages *is not allowed*
- ▶ Using own code is rewarded in grading but not required
- ▶ *Remember to give proper credit to packages you use!*

## Return logistics: Moodle

- ▶ All returns must be made to Moodle  
`https://moodle.helsinki.fi`
- ▶ You must log in using your University (non-CS) account
- ▶ Please register to the course “Project in Probabilistic Models, spring 2011”
  - ▶ The course registration key is “structure”
- ▶ For more instructions, please see “Student guide” on Moodle home page



## Schedule of the meetings for the rest of the course

- ▶ Course meetings on Thursdays at 16-18
- ▶ Mandatory attendance on feedback sessions starting 7 April

24 March Q+A session

31 March **No meeting**

7 April First feedback session

14 April Second feedback session

21 April **Easter holiday, no meeting**

28 April Final session

# Grading

- ▶ The grading will be based on your returned reports and presentations given during course sessions
- ▶ The following will positively influence your grade:
  - ▶ Effort put to the problem, innovativeness
  - ▶ Good presentations of your work during the course
  - ▶ Being able to improve your performance during the course and learn from previous results
  - ▶ Use of own software
- ▶ Score board positions *will not be used* in grading!

## Final warning

- ▶ In case you are tempted: the test data *do not come from the same distribution* as the training data. Using them in training the model *is not recommended!*

# Questions?

- ▶ Any questions?