Probabilistic Graphical Models

Jerónimo Hernández-González

About me

- Lecturer at UB
- Previously at UPV-EHU, AEPIA, IIIA-CSIC
- PhD in Machine Learning (University of the Basque Country, 2015)
 Learning PGMs for weakly supervised classification
- Research lines:
 - Weakly supervised learning (Crowd learning)
 - ► PGMs (Bayesian networks) for classification
 - Approximate inference in PGMs
 - Applied research: embryo selection, software defect, citizen science, etc.

By the end of the semester...

you will know...

- what probabilistic graphical models (PGMs) are
 - ► The probabilistic approach to machine learning
 - Types of PGMs
- which types of queries we can ask to them l.e., how to make a query...
 - when it can be answered in polynomial time (exact)
 - when it cannot be answered efficiently (approximate)
- how a PGM can be inferred from data

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you will be able to...

- apply the studied algorithms to problems of your interest
- translate PGMs and algorithms into code

Content

- Introduction
- Representation
 - Markov networks (Undirected)
 - Bayesian networks (Directed)
 - Temporal and plate models
- Inference
 - Exact (Variable elimination)
 - Approximate (Belief propagation / Sampling / Variational)
- Learning
 - Parametric learning
 - Structural learning
 - (In)complete data
- Real-world examples

Evaluation

- 30 % Participation
 Dairy work, Problem/Programming assignments, Forums**
- 30 % Presentation
- 40 % Test

**You need to initiate at least 1 debate; you need to engage in at least 2 debates initiated by your classmates

Test

What: Show the acquisition of the subject's contents

How: Development/multiple-choice questions

When: May 17th

- What: (i) Present a state-of-the-art work
 A paper of your choice from PGM conference:
 https://dblp.org/db/conf/pgm/
 - (ii) carry out your own work with PGMs

 Some practical work using some of the techniques seen in class
 - (iii) explain in depth a PGM topic
 - How:
- Choose a classmate and prepare a presentation for your colleagues (∼ 15 min.)
- Notify in the forum's thread your choice before May 17th!
- No topic-paper-work can be presented more than once

When: May 24th

Examples of topics (iii) for your presentation

- Explain, in terms of PGMs, a classical model such a GMMs, HMMs, Kalman Filters, ...
- Log-linear models
- Conditional Random Fields
- Learning Markov Networks
- ▶ Structural learning based on conditional independence tests
- ▶ Structural learning with the K2 algorithm
- k-dependence Bayesian classifier
- Causality

Calendar

Apr. 5 Intro	Apr. 7	May 3	May 5
Apr. 12 No class	Apr. 14 No class	May 10	May 12
Apr. 19	Apr. 21	May 17	May 19 Test
Apr. 26	Apr. 28	May 24	May 26 Presentations

Bibliography

- Koller, D., and Friedman, N. (2009). Probabilistic Graphical Models: Principles and Techniques. The MIT Press.
- ► Murphy, K.P. (2012). Machine Learning. A Probabilistic Perspective. The MIT press.
- Castillo, E., Gutiérrez, J.M., and Hadi, A.S. (1997). Expert Systems and Probabilistic Network Models. Monographs in Computer Science, Springer.

Coursera

- Daphne Köller (Stanford University) teaches a PGM course in Coursera
- Specialization (3 courses) (Video lectures, Questions, Programming assignments) https://www.coursera.org/specializations/ probabilistic-graphical-models
- ▶ A lot of her material is used in these lessons
- You might be interested in enrolling (partially free)

Strongly recommended

About you

Please, answer this survey:

https://forms.gle/ZWiXjwgcWfVb32rt7



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