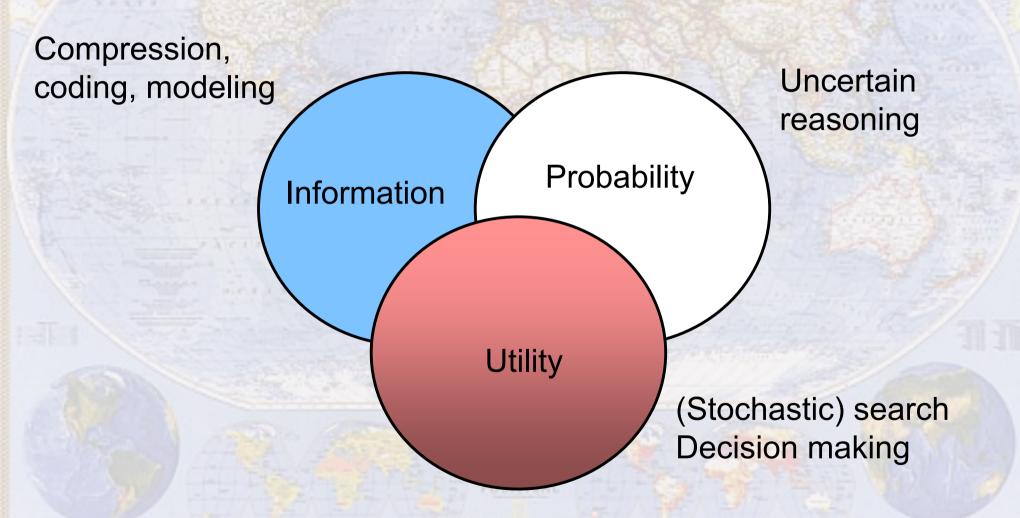
# 581287 Three concepts: Probability Spring 2007

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http://www.cs.helsinki.fi/group/cosco/Teaching/Probability/

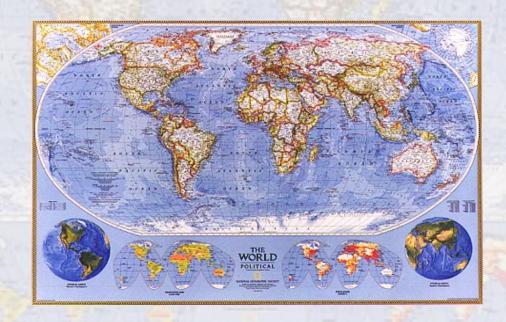


# Three concepts

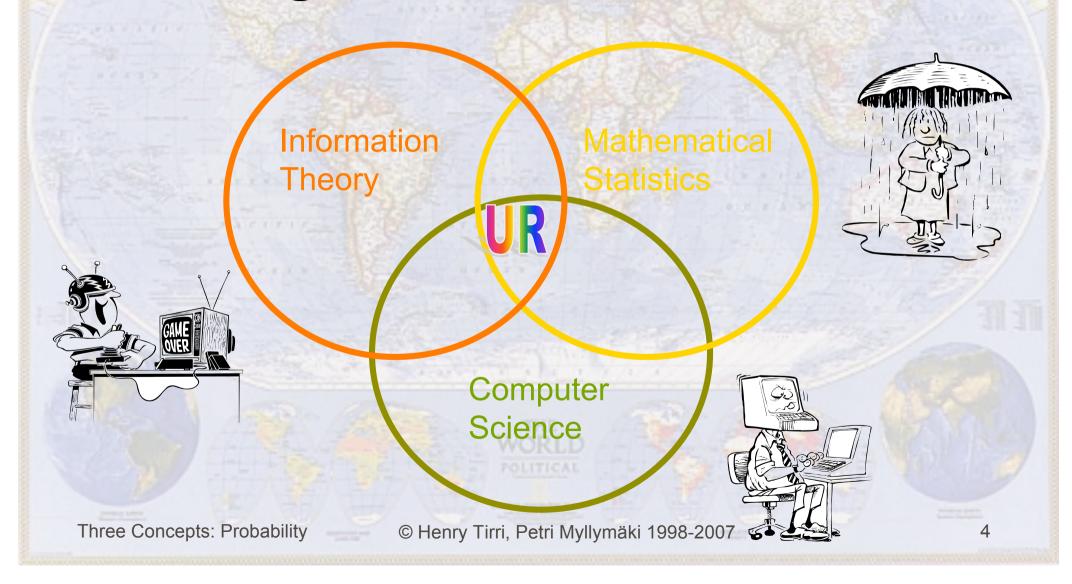


# On modeling

In building intelligent systems, statistics and in the rest of the world ...



# Our motivation: Uncertain reasoning



# Do I really need this stuff?

- "Machine learning"
- "Data mining"
- "Intelligent Systems"
- "Neural networks"
- "Pattern recognition"
- "Computational Intelligence"



or



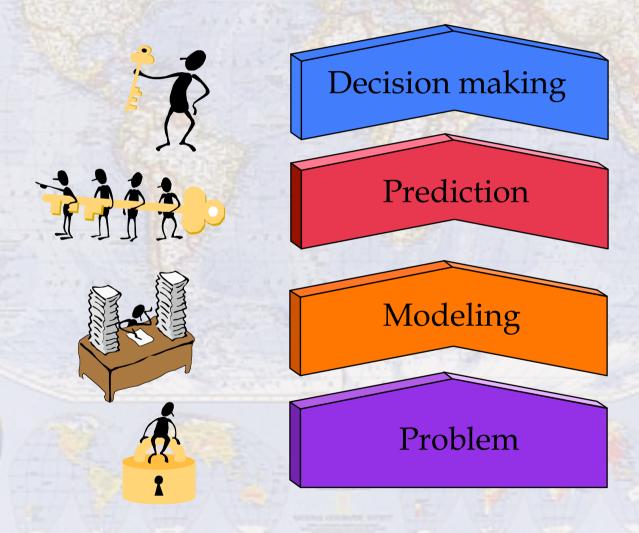
#### ...and more?

- "Statistics"
- "Robotics"
- "Expert system design"
- "Medical informatics"
- "User interface design"
- "Artificial life" (Evolutionary computation)

**-**/....



# Modeling framework



#### What does this mean?

- Problem: there is a need to model some part of the universe and make decisions based on the model
- Modeling: build the best model possible from a priori knowledge and data available
- Prediction: use the model to predict properties of interest
- Decision making: decide actions based on the predictions

# For example

- Problem: online troubleshooting of software/hardware
- Modeling: build a latent variable (Bayes)
  model of the problems user encounters
  based on knowledge about the software
  and symptom data
- Prediction: use the model to predict the underlying problem given symptoms
- Decision making: propose actions to remove the problem (or to find more symptoms)

# Microsoft Technical support



# Bayesian email spam filters

- SpamBayes, OPFile, Outclass, bayespam, bogofilter, ifile, PASP, spamoracle, Spam Assassin Annoyance Filter, BSpam, Spam Bully, Death2Spam, InBoxer, ...
- Software:
  - > http://spambayes.sourceforge.net/related.html
- Background:
  - > http://spambayes.sourceforge.net/background.html

# Real questions are ...

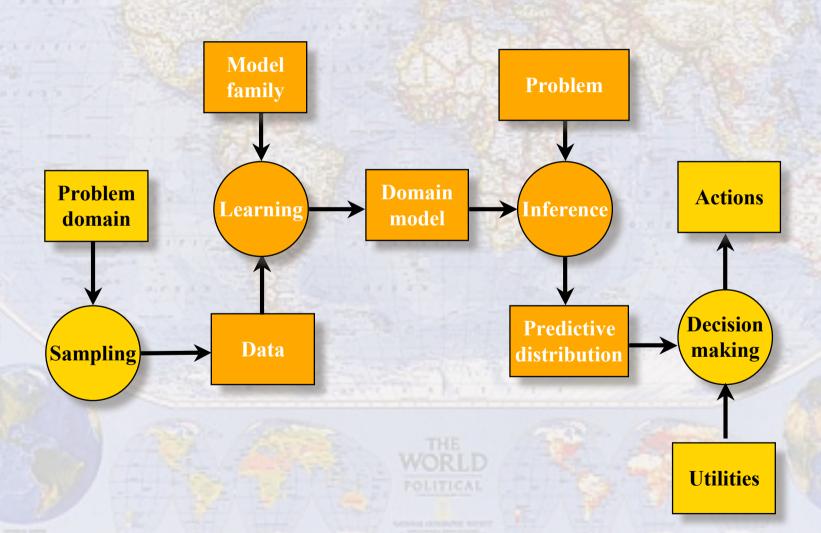
- Infinite number of models what models do we consider?
  - Model is always chosen from a set of possible models!
- How do we compare models (i.e., measure that one model is better than another one) given some data?
- How do we find good models?

### ...and more

- How do we use the models to predict unobserved quantities of interest?
- What actions do we choose given the predictions?



# General "rational agent" framework



#### Choice of models



- Simple models vs. complex models
- Linear models vs. non-linear models
- Parametric models vs. non-parametric models
- Flat models vs. structural models



#### Paradise lost

- What is simple and what is complex?\*
  - > The Catastrophic
    - ✓ Intuition: small, gradual changes in causes give rise to small, gradual changes in effects
  - > The Chaotic
    - ✓ Intuition: Deterministic rules of behavior give rise to completely predictable events
  - > The Lawless
    - ✓ Intuition: All real-world truths are logical outcome of following a set of rules
  - > The Irreducible
    - ✓ Intuition: Complicated systems can always be understood by breaking them down into simpler parts
  - > The Emergent
    - ✓ Intuition: Surprising behavior results only from complicated, hard-tounderstand interactions among a system's component parts

\* John. L. Casti, Complexification

# What is complex?

- what is complex is a totally nontrivial question
- one intuition: a complex model has more effective parameters
  - > example: 100 factor model of IQ is more complex than 10 factor model!

$$P(D|M) = \frac{(l-1)!}{(N+l-1)!} \prod_{k=1}^{l} |c_k|! \prod_{j=1}^{m} \prod_{k=1}^{l} \left( \frac{n_j - 1!}{(|c_k| + n_j - 1)!} \prod_{i=1}^{n_j} F_k(a_{ji})! \right)$$



# The Occam's razor principle

#### The problem:

- > You are given the following sequence: -1, 3, 7, 11
- > Question: What are the two next numbers?

#### Solution 1:

- > Answer: 15 and 19
- > Explanation: add 4 to the previous number

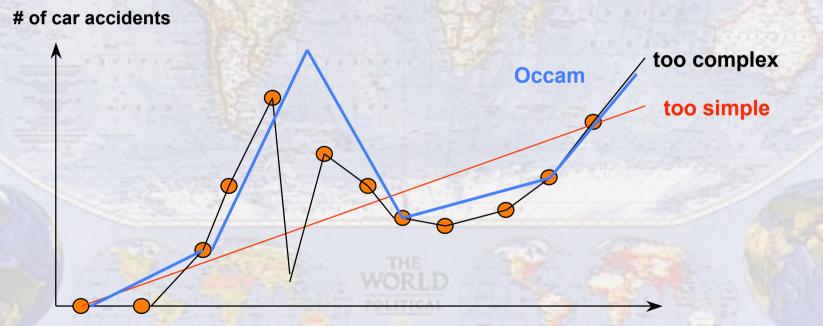
#### Solution 2:

- > Answer: -19.9 and 1043.8
- Explanation: if the previous number is x, the next one is  $-x^3/11 + 9/11x^2 + 23/11$
- "Of two competing hypotheses both conforming to our observations, choose the simpler one."



# Occam's Razor in Modeling

 there is a trade-off between the model complexity and fit to the data



# Simpler models are better than complex models

- interpretation: they are easier to understand
- computation: predictions are typically easier to compute (not necessarily!)
- universality: they can be applied in more domains (more accurate predictions)
- "models should be only as complex as the data justifies"
- Bayesian model selection: automatic Occam's razor for model complexity regularization

# The source of great confusion

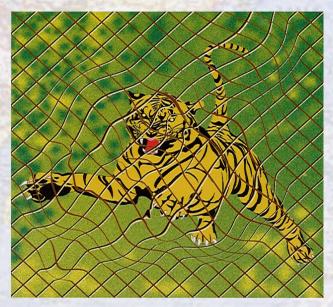
- Descriptive models ("Statistical modeling")
  - > describe objects (e.g., data) as they are
  - >typically exploratory structures
- Predictive models ("Predictive inference")
  - >models that are able to predict unknown objects (e.g., future data)
  - > models of the underlying process

### Some viewpoints

- "prediction is our business"
- why the best fit to data is not the best predictor
  - >data can be erroneous perfect fit is too "specialized" and models the errors also!
  - >a sample can only "identify" up to a certain level of complexity
- intuitive goal: minimize model complexity + prediction error it keeps you honest!

#### Alternatives

- Probabilistic inference
  - >Statistical inference
  - > Bayesian inference
- Fuzzy inference
- Dempster-Shafer inference
- Non-monotonic logic



# All models have probabilistic counterparts

#### **Probability distributions**

$$P(x) = K2^{-L(x)}$$

P(x) = K2Kraft inequality  $P(y | H, x) = \frac{1}{7}e^{-\beta ER(y|H,x)}$ 

(Shannon-Fano) codes

$$-\log P(x)$$

Non-probabilistic models with error function ER(y|H,x)