

Fall 2010 Graduate Course on
Dynamic Learning

Chapter 2: Autoregressive Models

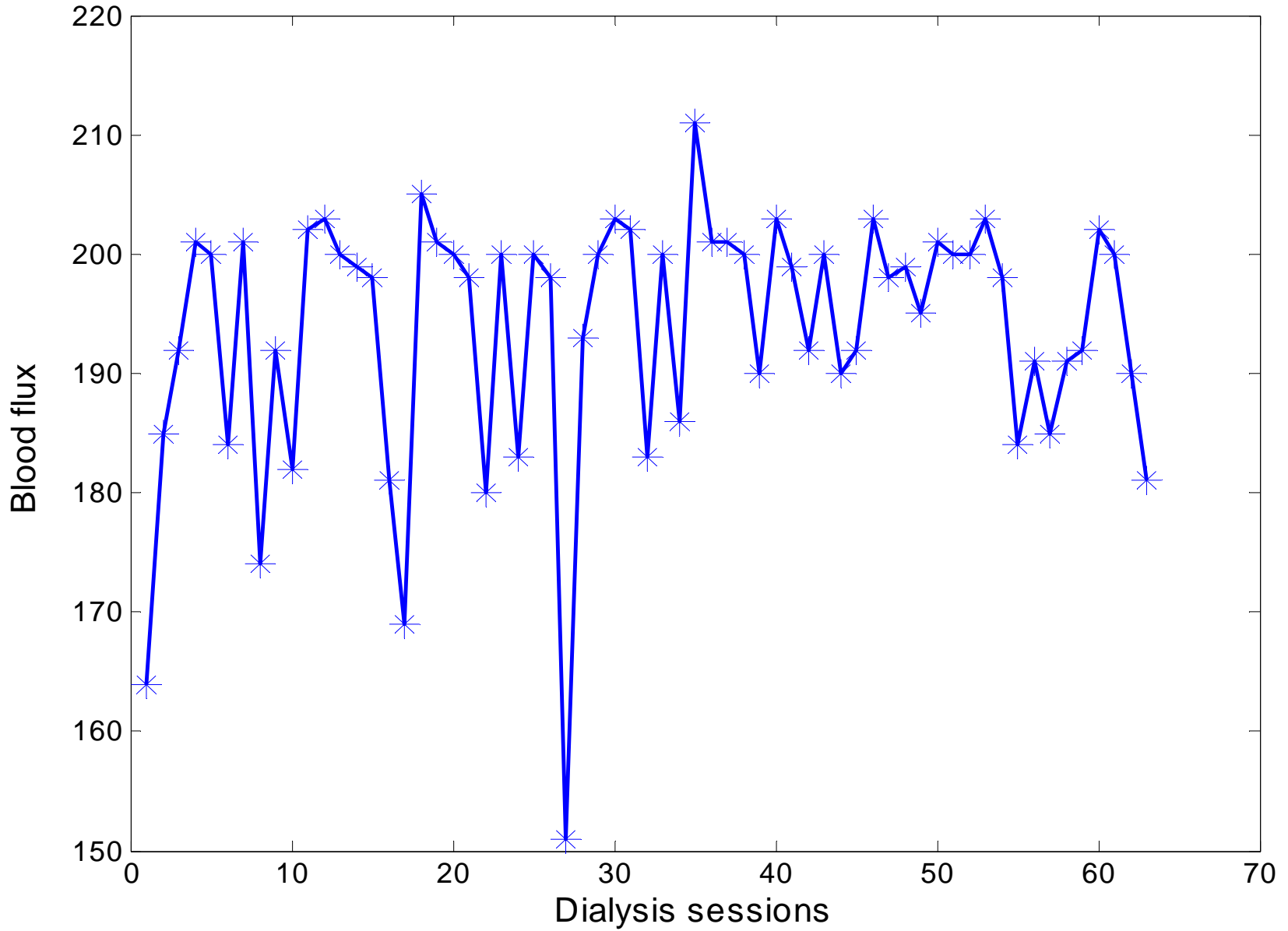
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References

- Slides of Riccardo Bellazzi

time course of blood flux



Time series

- Time series: a collection of observations made sequentially in time
- Many application fields:
 - Economic time series
 - Physical time series
 - Marketing time series
 - Process control
- Characteristics:
 - successive observations are NOT independent
 - The order of observation is crucial

Why time series analysis

- Description
- Explanation
- Prediction
- Control



Understand, then act

Modeling time series

- Time series: data are correlated; data are realizations of stochastic processes
- Stochastic linear discrete input-output models
- Two approaches:
 - Model the data as a function of time (a regression over time)
 - Model the data as a function of its past values: AR MA models
- Often, assumption of stationarity (the mean and variance of the process generating the data do not change over time)

Autoregressive (AR) models

$$y_{k+1} = a_1 y_k + e_k$$

- AR(h) is a regression model that regresses each point on the previous h time points. Example is AR(1)
- Each value is affected by random noise e_k with zero mean and variance σ^2
- Can be learned with linear estimation algorithm

Moving average (MA)

$$y_k = b_1 e_{k-1} + e_k$$

- A different kind of model is the Moving Average model (MA(h))
- It propagates over time the effect of the random fluctuations
- The autocorrelation function may help in choosing proper models
- An iterative estimation process is needed

ARMA

$$y_k = a_1 y_{k-1} + b_1 e_{k-1} + e_k$$

- It can be used to obtain a more parsimonious model, with “difficult” autocorrelation functions

ARMAX

- The system can be driven not only by noise but also by exogenous (control) inputs

$$y_k = a_1 y_{k-1} + b_1 e_{k-1} + c_1 u_k + e_k$$

This is the general ARMAX model

Nonlinear models

- Also non-linear stochastic models have been proposed in the literature
- Examples are NARX models

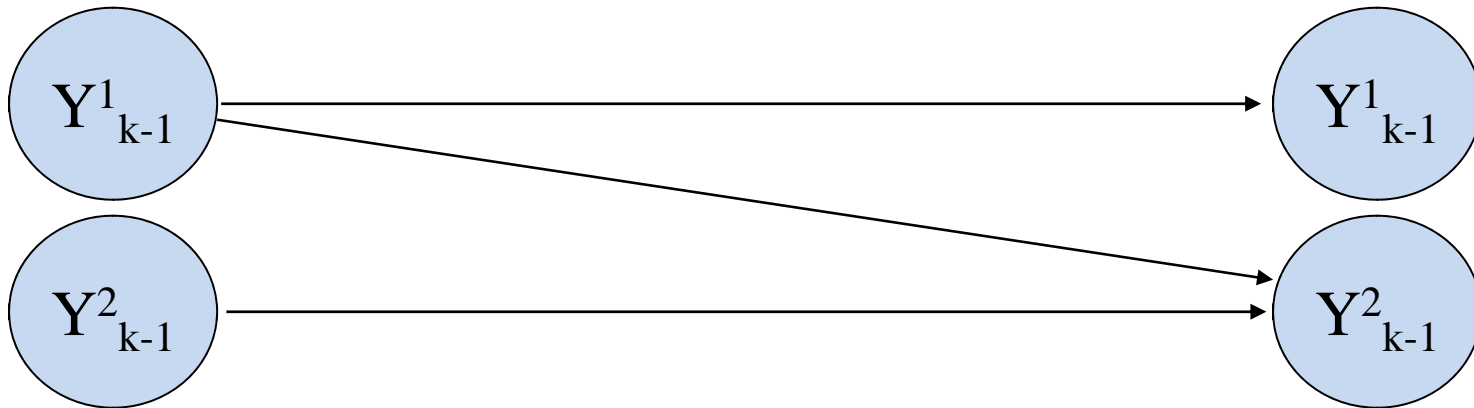
$$y_k = \sum d_k \varphi_k(y_{k-1}) + e_k$$

- NARX models can be easily learned from data with neural nets

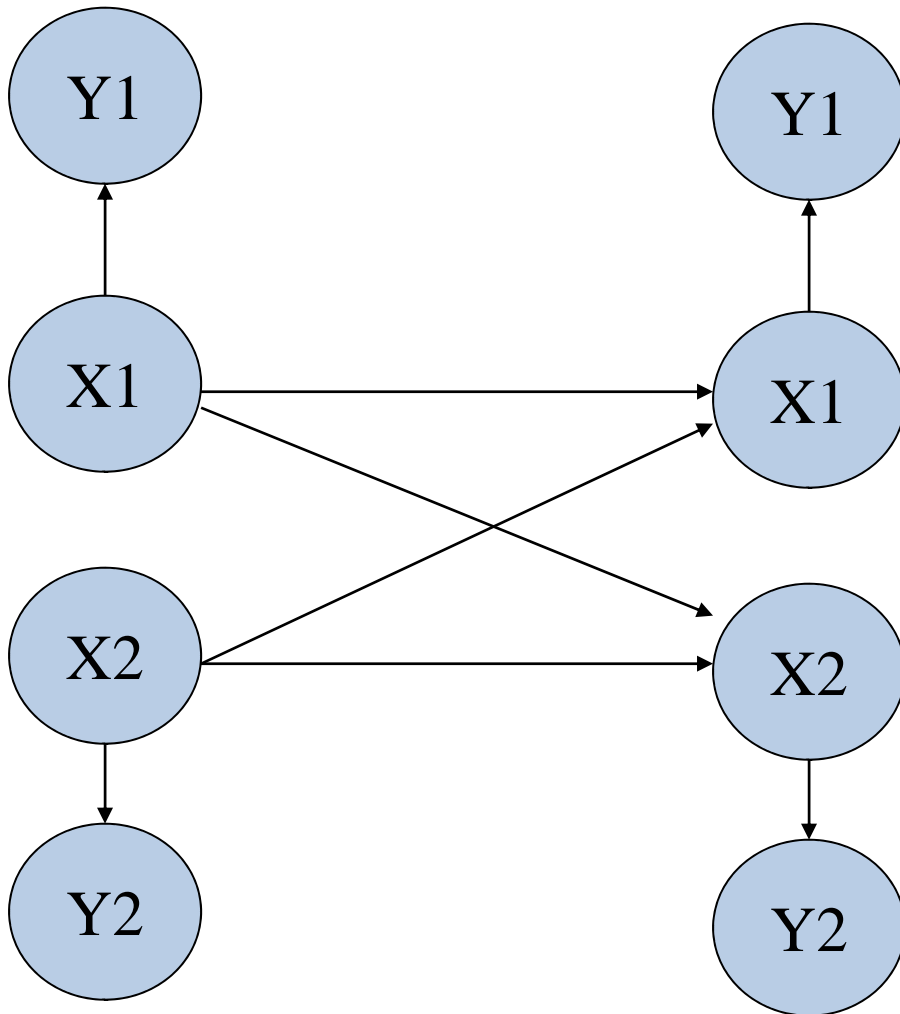
Structural Models

Nonlinear AR models

- Dynamic Bayesian nets



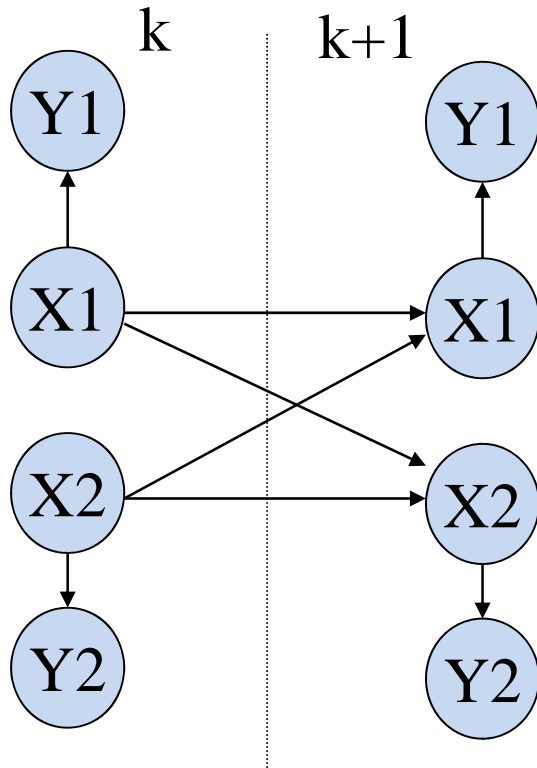
From black-box to structural stochastic models



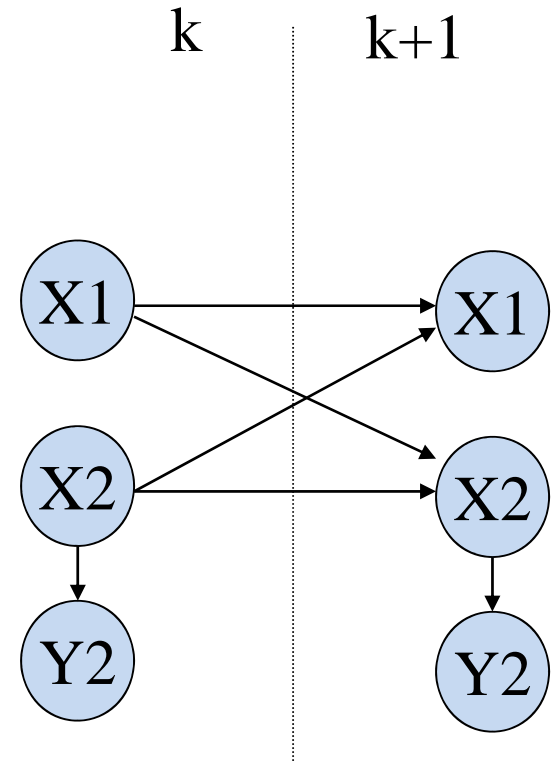
Examples:

- Kalman filters
- Dynamic BNs
- Hidden Markov Models

Observable and partially observable models



Fully observable



Partially observable