



Adventures in Bayesian Structural Time Series

Part 1: Introduction

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⊠ Structural time series



- ⊠ Structural time series
- ⊠ Bayesian approach to structural time series



- ⊠ Structural time series
- ⊠ Bayesian approach to structural time series
- ⊠ Implementation via `bsts` in R



- ⊠ “Predicting the Present with Bayesian Structural Time Series” Steven L. Scott and Hal Varian



Useful Resources

- ❖ “Predicting the Present with Bayesian Structural Time Series” Steven L. Scott and Hal Varian
- ❖ “An Introduction to State Space Time Series Analysis”
Commandeur and Koopman



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- ❖ **bsts** documentation
- ❖ Adventures in BSTS GitHub



Time Series Review

⊠ Stochastic process indexed by time $\{X_t, t \in \mathbb{T}\}$

$$E[X_t] = \mu$$

$$\text{Cov}(X_t, X_{t+k}) = \gamma(k)$$



Time Series Review

- ⊠ Stochastic process indexed by time $\{X_t, t \in \mathbb{T}\}$
- ⊠ (weak) Stationarity

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⊠ Not i.i.d.



AR(1)

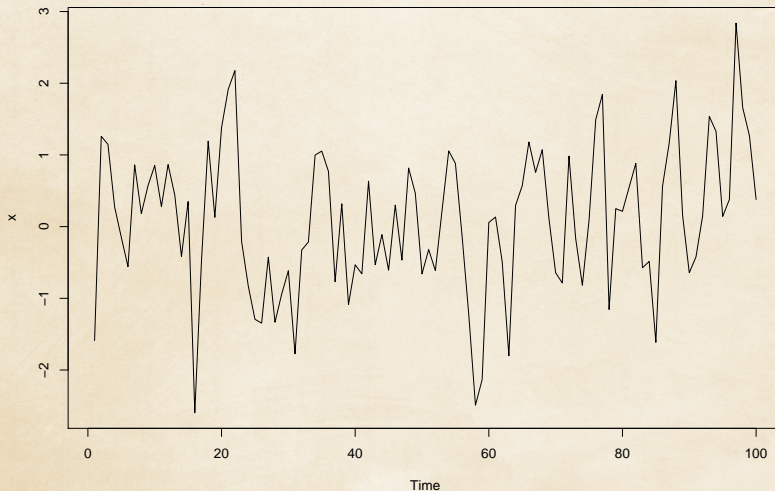
$$X_t = \phi X_{t-1} + e_t$$

e_t i.i.d. $(0, \sigma^2)$

$$|\phi| < 1$$



Simulated AR(1)





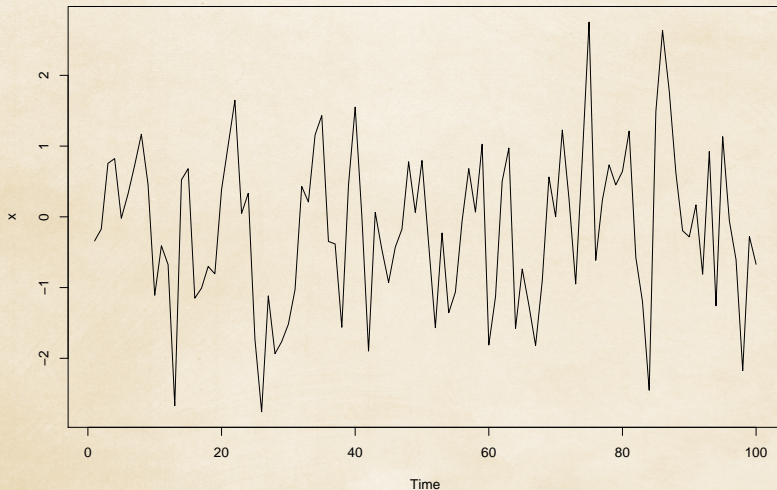
MA(1)

$$X_t = \theta e_{t-1} + e_t$$

e_t i.i.d. $(0, \sigma^2)$



Simulated MA(1)





ARMA Models

ARMA(p,q)

$$\phi(B)X_t = \theta(B)e_t$$

$$\phi(B)X_t = X_t - \phi_1 X_{t-1} - \cdots - \phi_p X_{t-p}$$

$$\theta(B)e_t = e_t + \theta_1 e_{t-1} + \cdots + \theta_q e_{t-q}$$



Your quest...