

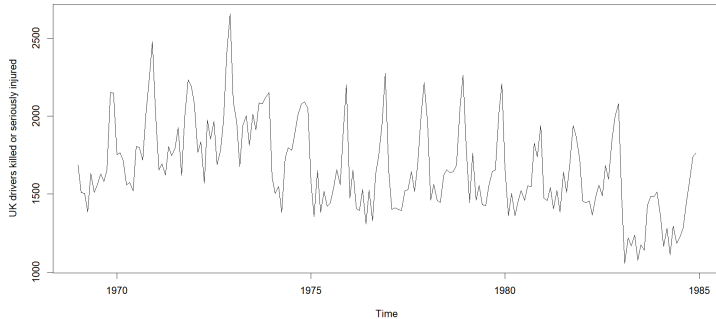
# Improving A Bayesian Procedure to Detect Breakpoints in Time Series Data

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# Introduction



# What is a Time Series?

## Definition

**Time series** are observations collected over regular intervals of time.



# What are some common types of time series models?

- **Auto-regressive (AR)** model: each output value depends linearly upon previous values and an independent error term

AR(1)

$$x_t = \phi x_{t-1} + \epsilon_t$$

- **Moving average (MA)** model: output value depends linearly upon previous error terms

MA(1)

$$x_t = \epsilon_t + \theta \epsilon_{t-1}$$

# What are some common types of time series models?

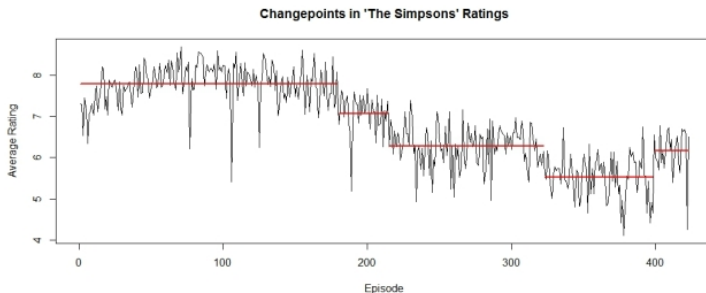
- **Auto-regressive moving average (ARMA)** model: output value depends on a mixed polynomial of previous values and error terms
- **Auto-regressive integrated moving average (ARIMA)** model: output value depends on a mixed polynomial of previous values and error terms, as well as the order of integration (the number of times the original time series is differenced to achieve stationarity)

# What is a Breakpoint?

## Definition

**Breakpoints** (also known as Change Points) are points in time where the model changes.

- Breakpoints are significant changes in time series data



# How do we find Breakpoints?

- 1 **Expert Opinion** : breakpoints are approximated by experts in specific field based on historical knowledge.
- 2 **Reverse Order Cusum (ROC)** : reverses data set and uses historical knowledge to group data points together with boundaries being locations of breakpoints

# What is Bai-Perron Test?

## Definition

**Bai-Perron Test** is a general algorithm to find an optimal breakpoint set.

- 1 a frequentist approach
- 2 checks almost every single location for a breakpoint and returns the optimal set
- 3 requires a user to specify the number of breakpoints



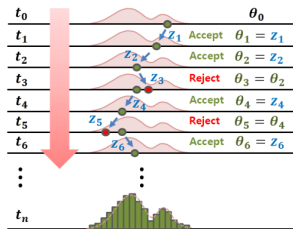
# What is the difference between Bayesian and Frequentist approach in Statistics?

- In **Frequentist** Statistics, the parameter is fixed and the data is random. This is the standard type of statistics learned in schools.
- In **Bayesian** Statistics, the parameter is random and the data is fixed. With this method, you apply prior information to the data that is being explored. The posterior is the result.

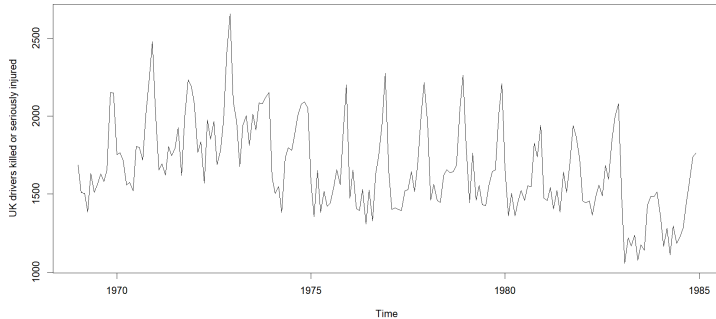
$$f(\theta|x) = \frac{f(\theta|x)\pi(\theta)}{\int f(\theta|x)\pi(\theta)d\theta}$$

# What is Markov Chain Monte Carlo (MCMC)?

- 1 A class of algorithms for sampling from a probability distribution
- 2 Generates and records a random sample sequence from one sample to another where each proposed sample is accepted or rejected by the algorithm; this process is repeated until a stationary distribution sample is found

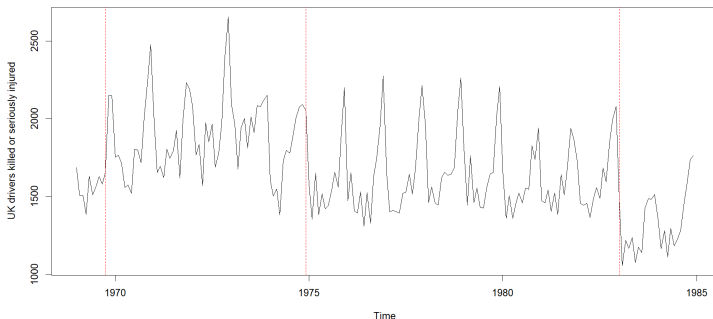


# Remember...



# Example

## Story time



**Figure:** Breakpoints, represented by the red lines, can be found in Oct. 1969, Dec. 1974, and Jan. 1983.

# The Project

**2018 REU Goal** : Kathryn Haglich, Sarah Neitzel, and Amy Pitts wanted to develop a better quantitative method for locating breakpoints in time series data.

# The Project

## Bayesian Adaptive Auto-Regression (BAAR)

**BAAR** is a Bayesian method used to find the location and number of breakpoints in a time series.

- 1 needs to have an input stating breakpoint places (Bai-Perron)
- 2 A new breakpoint set is proposed at each step of the MCMC
- 3 **Metropolis-Hastings** (a MCMC method for receiving a sequence of random samples from the probability distribution when direct sampling is difficult) ratio determines the set's acceptance
- 4 Hence, a distribution of possible breakpoints locations can be obtained

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# General Plan

**Goal** : to improve and expand BAAR (Bayesian Adaptive Auto-Regression) technique developed by previous REU group.

# General Plan

## Our four goals

- 1 Find a better starting point
- 2 Stress Models beyond even-spaced breaks.
- 3 Evaluate and alter code as needed to enhance the algorithm
- 4 Explore other models beyond evenly-spaced AR models

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# Moving Forward

- 1 Familiarize ourselves with the original code and develop more randomized breakpoints.
- 2 Understand and optimise math equations behind previous REU research
- 3 Research log likelihood equations for other models, explore math for MA models and start to develop algorithms
- 4 Explore alternative starting points