

Statistics 5350/7110

Time Series Analysis

Lecture 10

Decomposing Time Series

Professor Robert Stine

Admin Issues

- Questions
- Assignments
 - Q2 due today
- Quick review
 - Scatterplot smoothing
 - Moving averages, kernel smoothing
 - Polynomials, smoothing splines, lowess and loess
 - Cross-validation and the choice of bandwidth
 - Global constant controls amount of smoothing

Today's Topics

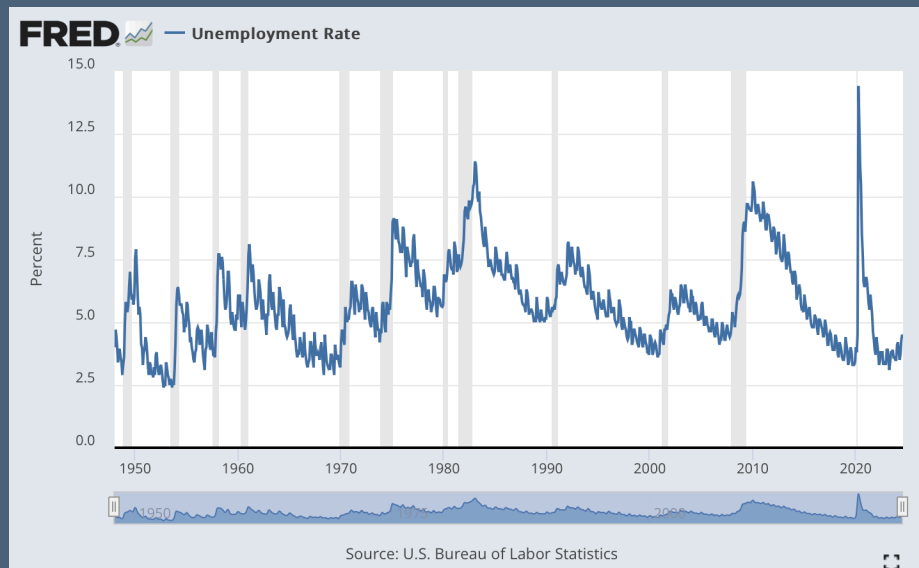
Text, §3.3
FPP, Chap 6

- Decomposing a time series
 - Official statistics often remove seasonal patterns from data
 - Rational: Don't want users confusing a seasonal pattern with "substantive" changes
e.g. Unemployment in US goes up in the summer when schools let out
 - Typically descriptive or treated as "preprocessing"
- Techniques
 - Median polish (robust alternative to anova)
 - Generalized additive model (GAM)
 - Black boxes with many tunable attributes, relying on judgement
 - Loess-based STL (a descendant of X-11 developed at US Census)
- Examples with R
 - Data preparation from on-line source (FRED)
 - Unemployment data
 - Climate (southern oscillation index) shown in Rmd file.

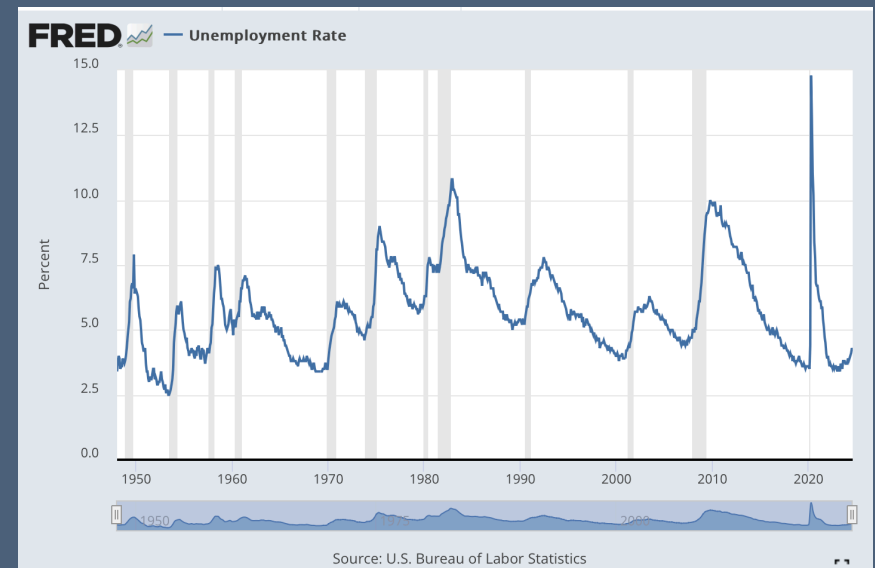
Seasonal Adjustment

- Monthly US unemployment
 - Shaded intervals in FRED graph locate recessions in the US
 - Covid spike
 - How do they go from the raw values to the seasonally adjusted values?

Raw



Seasonally Adjusted



Time Series Decomposition

- Represent a time series X_t as sum of three components (or product)

$$X_t = T_t + S_t + N_t$$

- Trend... slowly varying, smooth
- Seasonal... periodic
- Noise, remainder, irregular... everything that's not trend or seasonal. Often a stationary process.
- Additive or multiplicative
 - Distinguish from relationship between mean and variation
 - Logs convert multiplicative to additive
 - Substantive insight: How you think seasonal factors work
Example: Does travel in the summer increase by additive amount or percentage gain?
- Questions
 - Estimate the trend first or the seasonal first? Simultaneously?
 - How to estimate either?

Decompose with Median Polish

- Tabular arrangement

- Monthly: Organize time series into table with 12 columns, one for each month. Rows represent years.
- Estimate row (annual) and column (month) effects using medians
- Like a two-way analysis of variance, with medians replacing means

Tabular arrangement
anticipates STL algorithm

$$X_{rc} = \mu + \alpha_r + \beta_c + e_{ij} \Rightarrow X_{year, month} = \mu + \alpha_{year} + \beta_{month} + e_{y,m}$$

- Ambiguous (need to impose constraints)

- Neat aspects

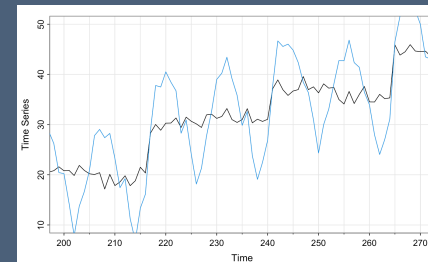
- Simple, intuitive, and very fast
- Robust to outliers by using median rather than mean
- Anticipates more sophisticated methods with better trend estimates
- Handy diagnostic plot to check for additivity

Iterative calculation,
unlike sweeping out means

Tukey was famous for clever
diagnostic plots (e.g. one
degree of freedom test for
non-additivity)

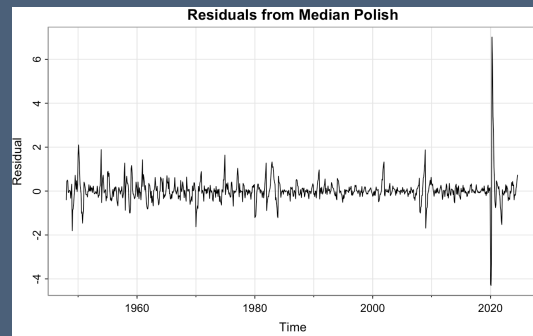
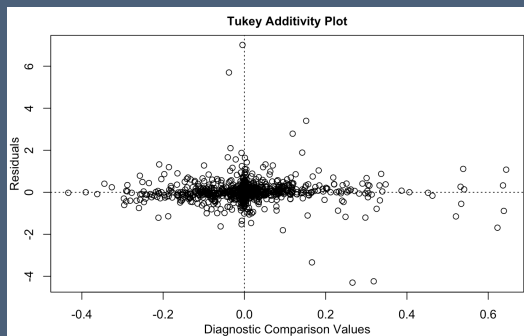
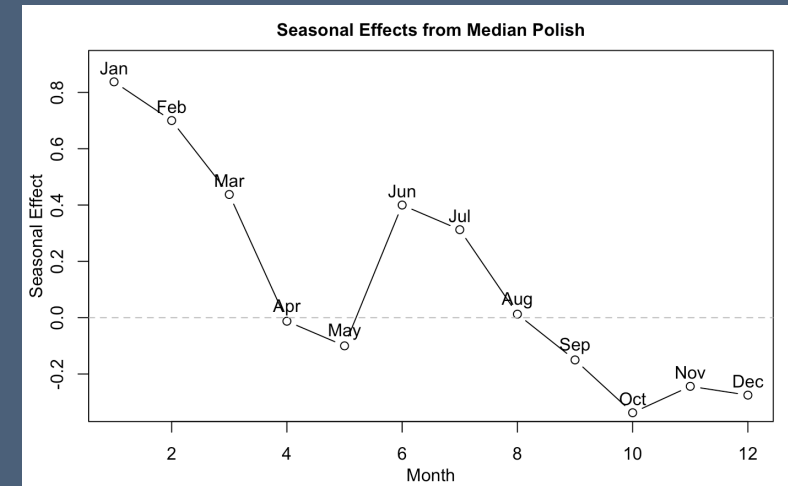
- Weakness

- Not such a great model for overall trend
(piecewise constant), particularly if trend is strong



Decomposition by Median Polish

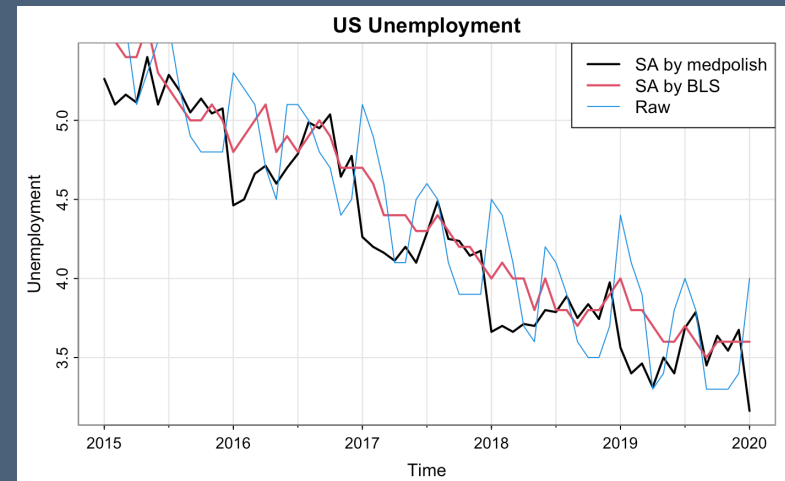
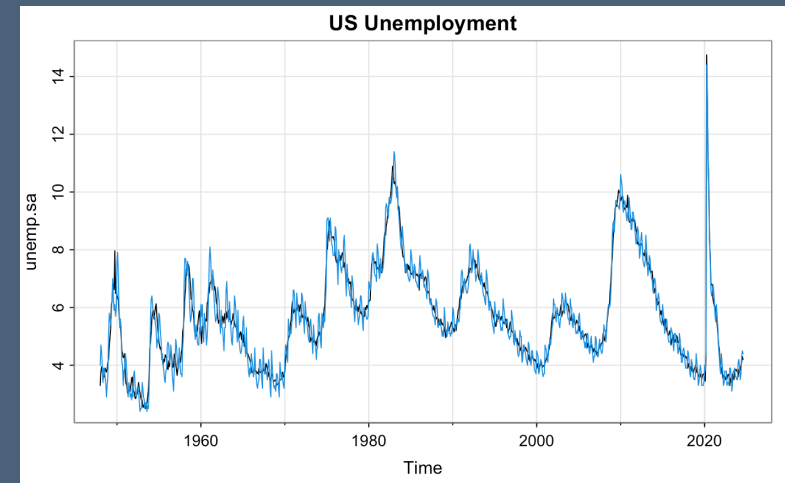
- Apply to Unemployment
 - Arrange data into a table with 77 rows and 12 columns
 - Deal with missing values (incomplete in 2024)
- Seasonality
 - Column medians estimate seasonality
 - Peaks in January and June (post holiday, summer vacation)
- Diagnostic analysis
 - Diagnostic shows no evidence of multiplicative effect
 - Outliers associated with Covid easy to spot



This whiplash of residuals in 2020 is a by-product of the limited trend fit by a median polish, namely one value for each year (piecewise constant).

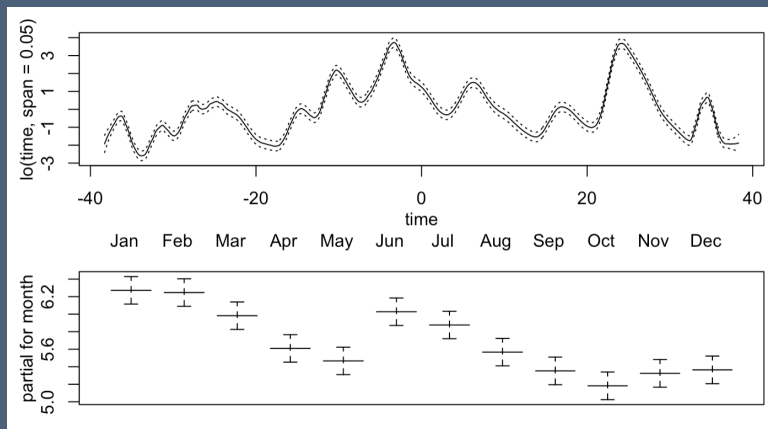
Seasonally Adjusted Unemployment

- Seasonally adjusted time series
 - Use fit from median polish
 - Reconstruct data without seasonal terms
 - Hard to see effects over full time span
- Zoom in
 - Time series for 2015 - 2020
 - Include official seasonally adjusted unemployment rate
- Comparison
 - Suppress seasonal jumps
 - Medpolish trend inadequate (e.g. 2018 flat)



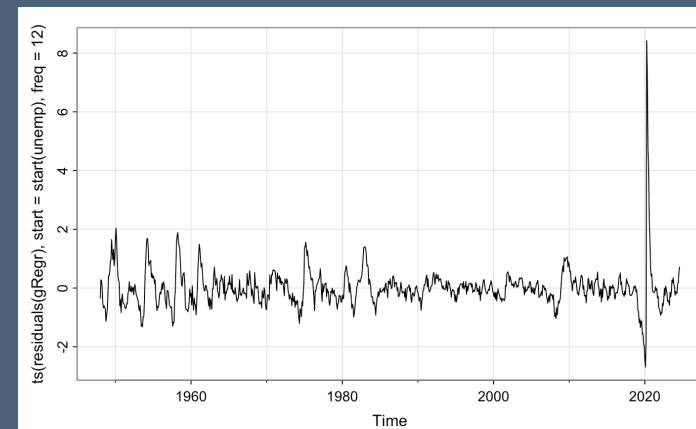
Decomposition by Regression

- Improve on the median polish
 - Allow a slowly varying smooth trend rather than piecewise constant
 - Possible smooths from splines, moving averages, or loess.
- Generalized additive model
 - Combine smooth features with standard regressors
 - Estimate the trend with loess
 - Estimate the seasonal component from monthly dummy variables
 - Provides standard errors for estimates (but these are not reliable due to autocorrelation)



GAM fit results:
Explore various spans

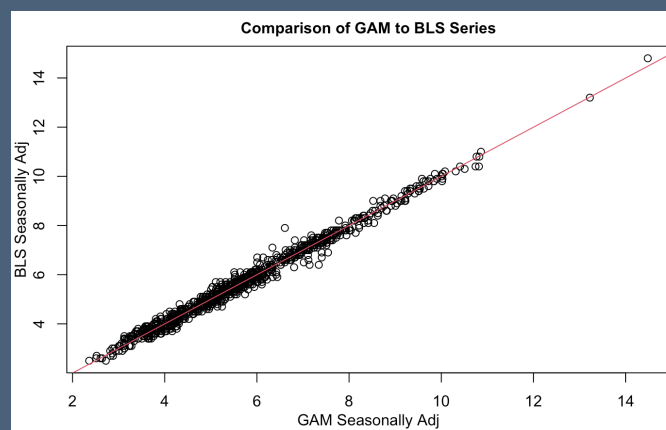
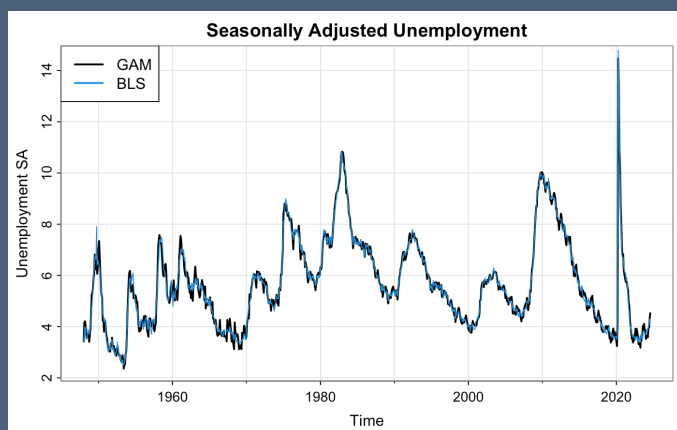
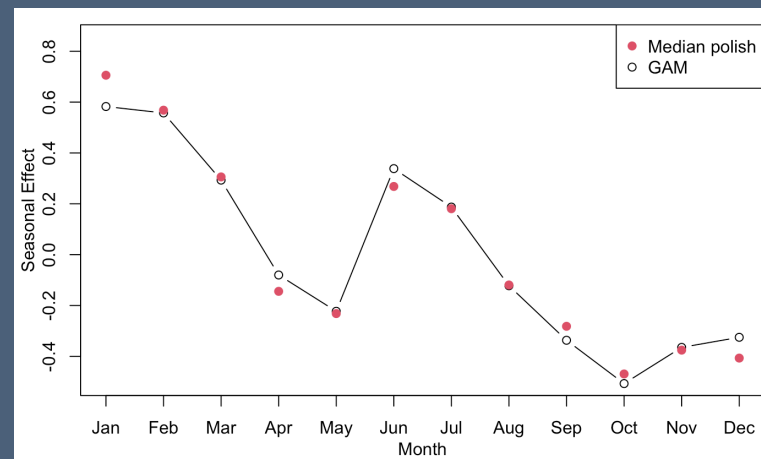
Seasonality resembles
median polish result



Whiplash persists
since this model
leaks the future into
the past

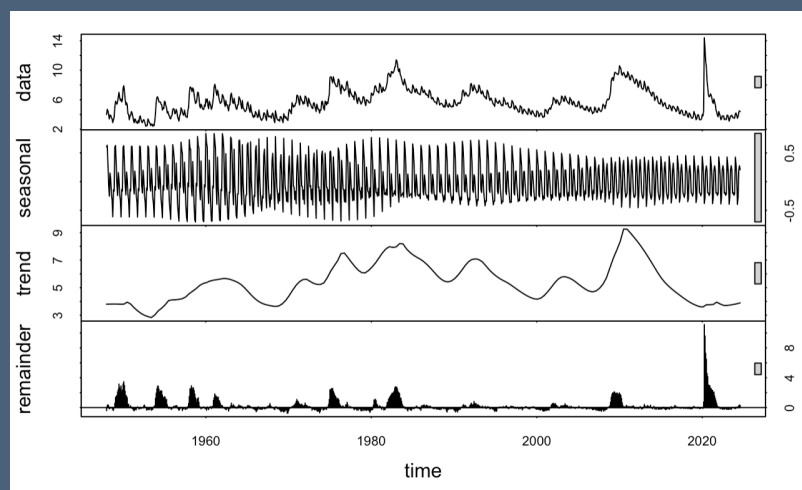
Comparisons

- Seasonal terms similar to those from median polish
 - Both scaled so average seasonal effect is zero.
 - GAM also provides std. error (albeit not reliable in this application)
- Similar to BLS seasonal adjustment

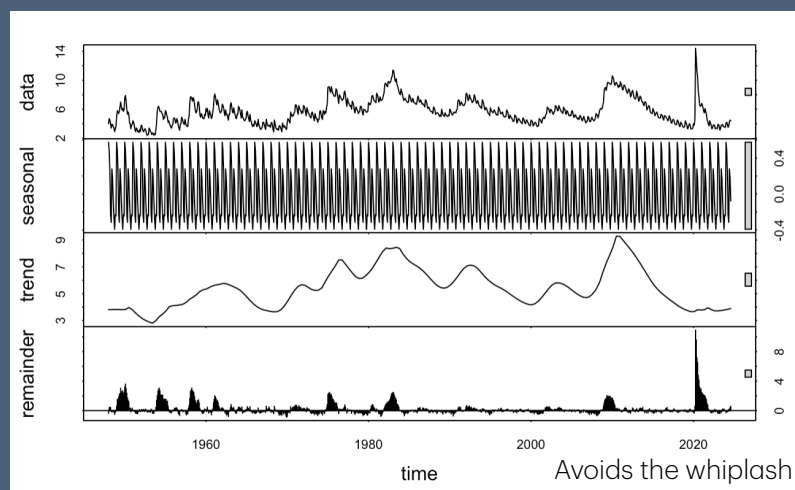


Decomposition using STL

- STL automatic decomposition
 - Estimates trend and seasonal using loess
 - Seasonal component can be allowed to vary over time
- Seasonal patterns
 - Fixed periodic effect, as in dummy variable model
 - Alternatively using loess fit to sub-series (e.g. sequence of January values)

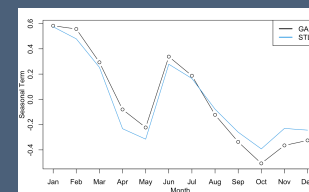


Smooth



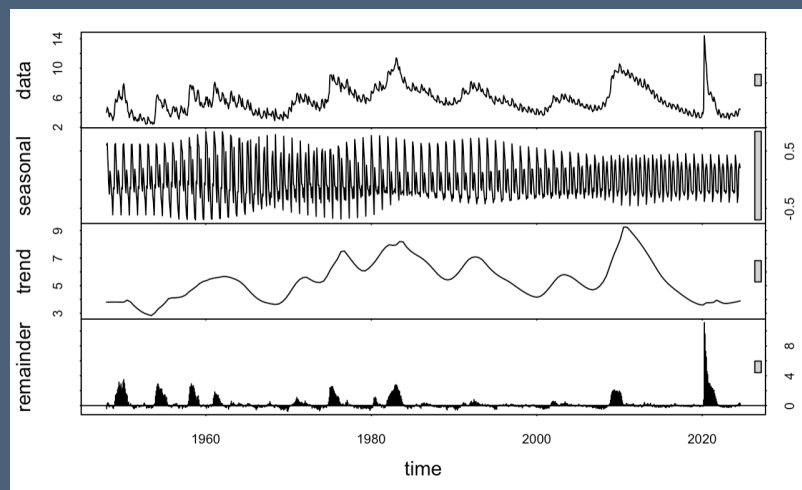
Periodic

Compare to GAM estimates

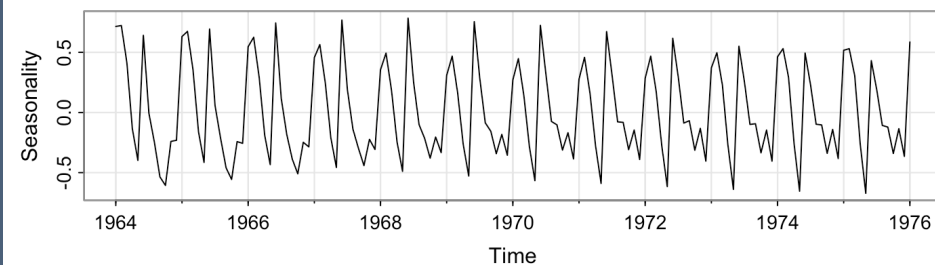
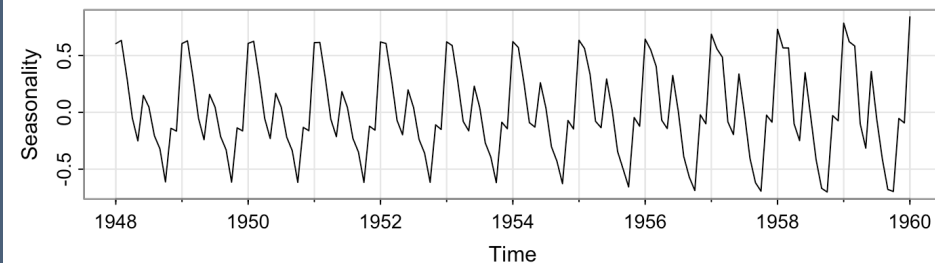


Decomposition using STL

- Flexible seasonal patterns
 - Loess fit to sub-series (e.g. sequence of January values)
 - Allow subtle changes to seasonality
- Example
 - Pre-1960: Peak seasonal effect in January
 - Post-1964: Peak seasonal effect shifts to June



Smooth



Discussion

- Ambiguity between “trend” and “seasonal”
 - What’s a trend? What’s seasonal?
 - Is that a trend or a seasonal variation?
- Seasonal rigidity
 - Periodic seasonal modeling is rigid, assuming constant effect over the years
 - Effect might be proportional, suggesting multiplicative model
 - Employment example shows could be important to allow drift
- Inference
 - Are these changes in seasonality large, relative to random variation?
 - Easy to lose sight of confidence intervals
 - Provided by a GAM model, though dubious here due to autocorrelation
- Revisions
 - Decomposition methods use data from the future to seasonally adjust data in the past!
 - With more recent data, often change impression of what happened previously

What next?

- ARMA models for time series
 - autoregressive, moving average (maybe better autoregressive, moving sum)
 - a.k.a., Box-Jenkins models
 - You've already seen them in their simplest guise...
Combine autoregression with a moving average of errors