Time Series Decomposition Based on Markov Random Fields

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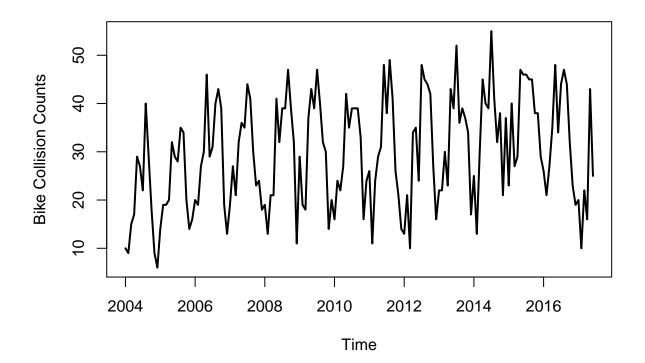
Project Motivation

The project is about time series decomposition based on markov random fields, which is a locally adaptive nonparametric curve fitting method that operates within a fully Bayesian framework. I am going to decomposite a time series into trend part, seasonal part and random part. I am thinking of treating it as an additive model and decompositing it by doing markov random fields fitting twice. I am going to compare the results with some classical parametric techniques, like ARIMA, and machine learning techniques, like RNN/LSTM. Furthermore, I am going to apply the decomposition into forcasting.

Data Description

I have data of bike collision records in downtown Seattle from 01/2004 to 06/2017 and I summarize them by month, so that I have 162 bike collision counts, which is shown as following.

```
load("~/2018-research-project/data/TS.rda")
plot(TS,ylab='Bike Collision Counts',lwd=2)
```



Bayesian Method

The model assumes that each data point is generated independently with some parametric models, like Poisson distributions, and the parameters are follow a markov random fields model, which could, according to the refered paper, provide a combination of local adaptation and global control. For example, supposed $Y_i \sim Poisson(\theta_i)$, where i = 1, 2, ..., n, and $(\theta_{i+1} - \theta_i)|_{\tau_i} \sim N(0, \tau_i^2)$, where i = 1, 2, ..., n - 1, and τ_i are independent and identically distributed.

Challenge

One challenge, for prediction task, is how to use the trend line. I have the following two ideas: first is to use a parametric function to describe the trend line; Second is to choose a window so that I could use the "most recent" data to predict "future".

Reference

[1] Faulkner, James R.; Minin, Vladimir N. Locally Adaptive Smoothing with Markov Random Fields and Shrinkage Priors. Bayesian Anal. 13 (2018), no. 1, 225–252. doi:10.1214/17-BA1050. https://projecteuclid.org/euclid.ba/1487905413