Modelling daily ozonio mean

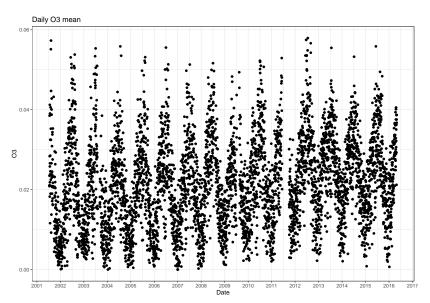
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30/09/2020

Data

Daily data

▶ New York data from 15/07/2001 to 30/04/2016.



Missing data

- ▶ There are 52 time skips in the data, in a total of 473 days.
- ▶ The biggest skips is 108 days in 2011.
- ► The majority of skips are of 1 or 2 days.
- Around 9.5% missing data.
- ► The missing observations are distributed along the time without a clear pattern.

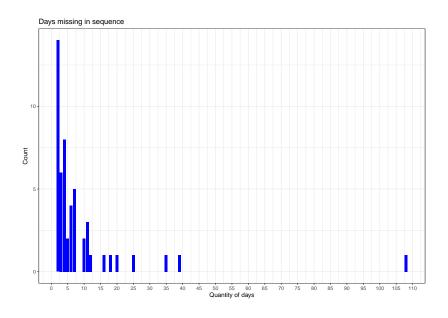
Observations after data skips 0.05 -0.04 -0.03 -8 0.02 -0.01 -

2010

Date

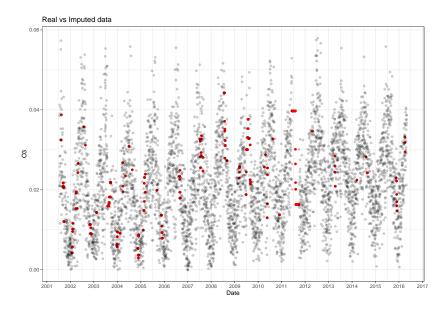
2015

2005

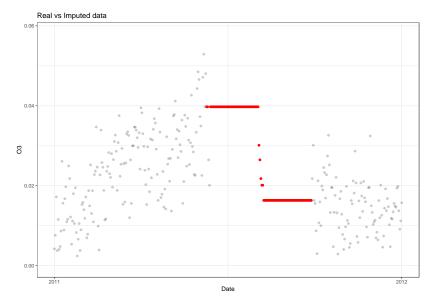


Imputation method

- It was used the kNN method to imputate values on missing observations.
- ► The kNN method needs the parameter k, the number of closest points considered.
- ightharpoonup Starting with k = 7.

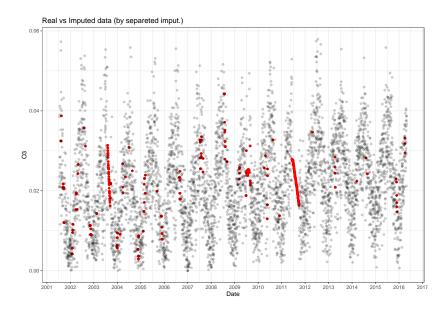


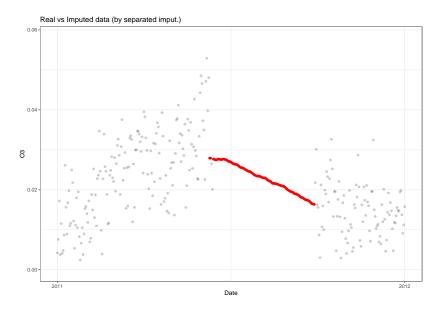
► Method create a bad behavior when the size of the skips is bigger than 7 days.



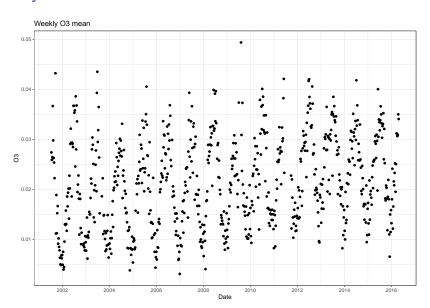
- ➤ To deal with this, the parameter k used for imputation will be different if the size of the skip is minor them 30 days, between 30 days and 100 days, or bigger than 100 days.
- 30 days and 100 days, or bigger than 100 days.

 ▶ k = 7, k = 45, k = 120, respectively.
- We will aggregate closest points by weighted by distance mean.

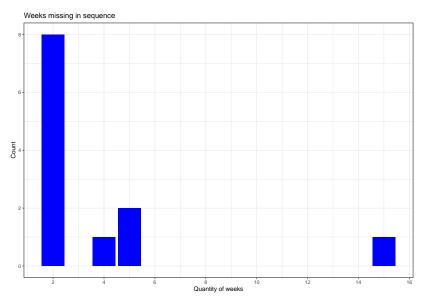


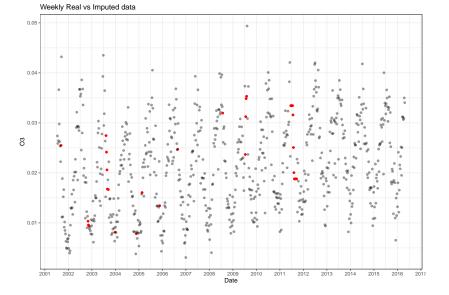


Weekly data



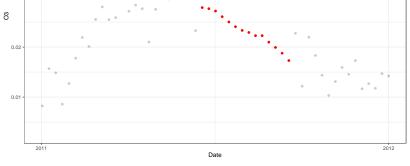
- ▶ If the data is grouped by week, ignoring the missing values when aggregating, it'll have 33 missing observations.
- ► Around 4.3% missing data.





▶ It has the same problem when the sequence of missing data is to big.

Weekly Real vs Imputed data 0.05 0.04 0.03 -

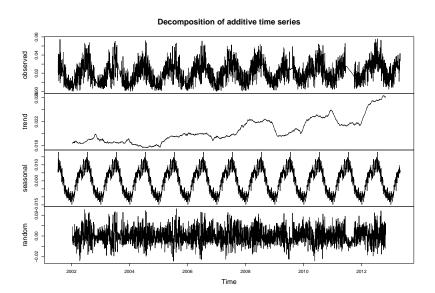




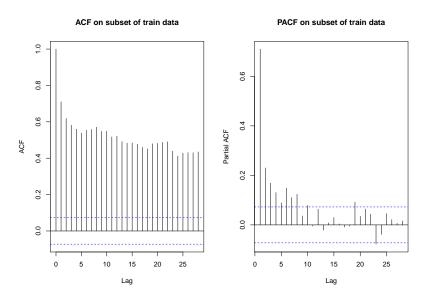
Modelling process

- Metric to be minimized: MAE = $\frac{1}{n} \sum_{n} |y_t \hat{y}_t|$.
- Rolling window of 2 years (730 days).
- Prediction of the next 7 days.
- First: Test if there is tendency with Wald-Wolfowitz runs test.
 - For every 2 years window, the p-value is smaller than 1e-3.
- Second: Fitting of different models and evaluation of MAE error.

Choice of models - trend



Choice of models - ACF and PACF



- ▶ Naive model: the next 7 days are predict as the mean of the last 4 weeks.
- Exponential smoothing forecast.
- Holt model with trend.
- ► ARMA(6,0) model. Auto ARIMA model.

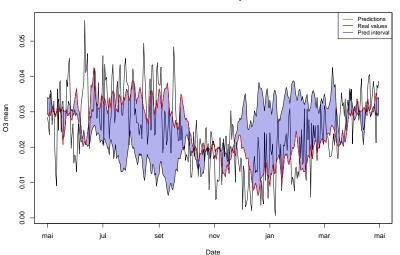
- Process:
 - ► 1. For each model, run a 2 years window, for each:
 - Fit model.
 - Generate predictions of next 7 days.
 - Compute mean of residuals for that window.
 - 2. Compute MAE for model as the mean of residuals.

- Results for train data:
 - Auto ARIMA model: 0.005986731
 - ► Holt model: 0.006142857
 - ► SES model: 0.00617229
 - ► ARMA(6,0) model: 0.006279533
 - Naive model: 0.007498889

Evaluating on test data

MAE: 0.006503142



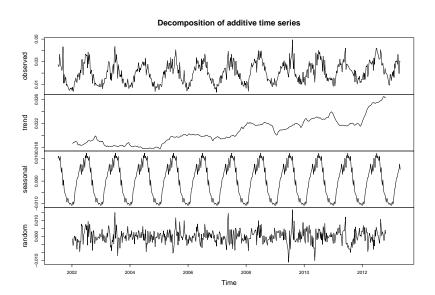




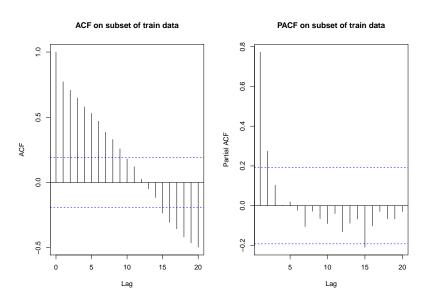
Modelling process

- Metric to be minimized: MAE = $\frac{1}{n} \sum_{n} |y_t \hat{y}_t|$.
- ▶ Rolling window of 2 years (104 weeks), by skiping 4 weeks.
- Prediction of the next 4 weeks.
- First: Test if there is tendency with Wald-Wolfowitz runs test.
 - For every 2 years window, the p-value is smaller than 1e-3.
- Second: Fitting of different models and evaluation of MAE error.

Choice of models - trend



Choice of models - ACF and PACF



- Naive model: the next 4 weeks are predict as the mean of the last 4 weeks.
- Seasonal model: linear regression on seasonal dummies variable, each month is a factor.
- Linear model: linear regression on seasonal dummies and time index.
- ▶ Poly 2 model: linear regression on seasonal dummies and time index with degree 1 and 2.
- Poly 3 model: linear regression on seasonal dummies and time index with degree 1, 2, and 3.
- Holt model with trend.
 Holt Winters model with trend and seasonality (multiplicative and addtive).
- ightharpoonup ARMA(1, 0) model.

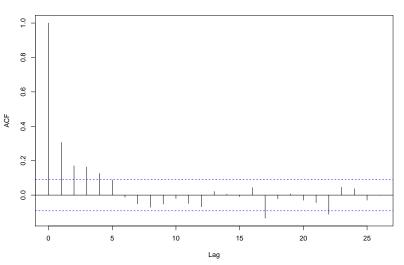
Process:

- ▶ 1. For every 2 years window:
 - Fit all the models.
 - Generate predictions of next 4 weeks.
- 2. With predictions for every week, compute residuals $r_t = y_t \hat{y}_t$.
 - 3. With residuals, compute MAE.

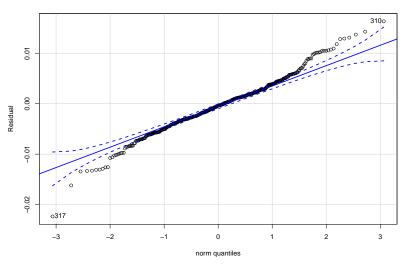
- Results:
 - Sazonal: 0.003771788
 - Linear: 0.003941300
 - Poly 2: 0.004079873
 - Poly 3: 0.004562281
 - Arma(1, 0): 0.004675624
 - ► Holt: 0.004905350
 - HoltWinters additive: 0.005075887
 - ► HoltWinters multiplicative: 0.005126882 Naive: 0.005160706

Residuals



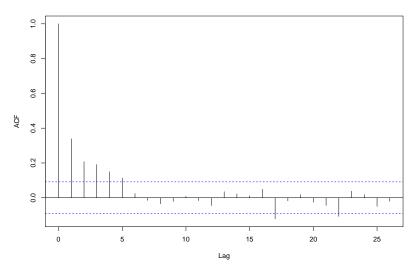


QQPlot of sazonal model residuals

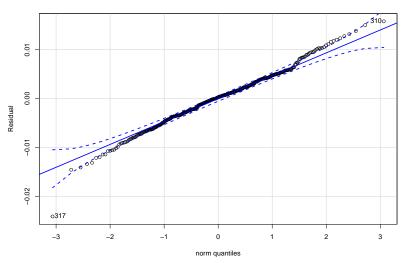


[1] 317 310

Linear model residuals ACF



QQPlot of linear model residuals



[1] 317 310

Evaluating on test data

► MAE: 0.003476891

