



< Previous	✓	✓	✓	✓	✓	✓	✓	✓	✓		✓	Next >
------------	---	---	---	---	---	---	---	---	---	--	---	--------

10. Local Linear regression model

Bookmark this page

One of the model explored in the paper is called **Local Linear Regression** . This method does not assume any functional form of the underlying non-linear relationship (polynomial, exponential, etc.). The idea of the approach is regardless of the non-linear relationship, it can always be approximately by a linear function for a small neighborhood near any data point (i.e., locally). Combining all local linear approximations together, we can obtain a prediction of the overall non-linear relationship.

More formally: for any data point \boldsymbol{x}_0 , we want to locally fit a linear model around \boldsymbol{x}_0 . We only want to fit this model in a small neighborhood \boldsymbol{D} containing data point \boldsymbol{x}_0 . The neighborhood \boldsymbol{D} is defined as:

$$\boldsymbol{D} = \{\boldsymbol{x} : \|\boldsymbol{x} - \boldsymbol{x}_0\| \leq h\}$$

within the neighborhood, we can use weighted ordinary least square to estimate the coefficients for the linear model:

$$\hat{\beta}_{\boldsymbol{x}_0} = \arg \min_{\beta} \sum_{\boldsymbol{x}_i \in \boldsymbol{D}} w_i (y_i - \beta^T (\boldsymbol{x}_i - \boldsymbol{x}_0))^2$$

w_i is the weight value for each data point \boldsymbol{x}_i . Usually higher weight is assign to data points closer to \boldsymbol{x}_0 , and smaller weight when the data point is further from \boldsymbol{x}_0 . The weight w_i is usually determined by some kernel function: $w_i = K\left(\frac{\boldsymbol{x}_i - \boldsymbol{x}_0}{h}\right)$

In the paper, the authors use the following model:

- Fit one model for each day of the year (shared across years) and grid point \boldsymbol{g}
- The neighborhood \boldsymbol{D} is defined as the ± 56 days around target day of the year
- For each day of the year, they fit the following model

$$\hat{\beta}_{\boldsymbol{g}} = \arg \min_{\beta} \sum_{t \in \boldsymbol{D}} w_{t,\boldsymbol{g}} (y_{t,\boldsymbol{g}} - b_{t,\boldsymbol{g}} - \beta^T \boldsymbol{x}_{t,\boldsymbol{g}})^2$$

- Weighting scheme: $w_{t,\boldsymbol{g}} = 1$ or $w_{t,\boldsymbol{g}} = 1/\text{var}(\text{prediction}_t)$
- Offsets: $b_{t,\boldsymbol{g}} = 0$ or $b_{t,\boldsymbol{g}}$ set up to take out the seasonality

For each local linear model, we still need to determine what features to use to predict the weather variables. The paper used a backward regression approach to select the features:

- Start by using all features and remove features one by one
- Drop the feature that reduce the prediction accuracy the least, if below a tolerance threshold
- Prediction accuracy is determined by leave-one-year-out cross validation

Discussion

Topic: Module 4: Time Series:Introduction to Time Series Analysis 3 / 10. Local Linear regression model

Hide Discussion

Add a Post



edX

- [About](#)
- [Affiliates](#)
- [edX for Business](#)
- [Open edX](#)
- [Careers](#)
- [News](#)

Legal

- [Terms of Service & Honor Code](#)
- [Privacy Policy](#)
- [Accessibility Policy](#)
- [Trademark Policy](#)
- [Sitemap](#)

Connect

- [Blog](#)
- [Contact Us](#)
- [Help Center](#)
- [Media Kit](#)
- [Donate](#)



© 2021 edX Inc. All rights reserved.
深圳市恒宇博科技有限公司 [粤ICP备17044299号-2](#)