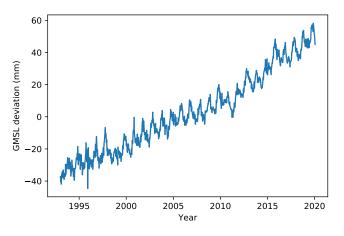


Time Series and Sequence Learning

Classical regression in a time series context

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What do you see in the data?

Data from https://climate.nasa.gov/vital-signs/sea-level/

Linear trend model

$$Y_{t} = \Theta_{0} + \Theta_{1} U_{t} + \frac{E_{t}}{E_{t}}$$
 Error term

 $U_{t} = \text{"fine when observation #t was recorded"}$
 $U_{t} = 1993 \frac{4}{365}, \dots, 2020 \frac{20}{365}$

White noise

We model the errors as independent random variables

$$\varepsilon_t \stackrel{\text{iid}}{\sim} \mathcal{N}(0, \sigma_{\varepsilon}^2), \qquad \qquad t = 1, 2, \ldots$$

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In a time series context, such a sequence of random variables is referred to as (Gaussian) white noise.

Linear trend model, cont'd

Simple linear trend model: $y_t = \theta_0 + \theta_1 u_t + \varepsilon_t$

The model parameters
$$\theta = (\theta_0, \theta_1)$$
 are estimated using CLS (= maximum likelihood)
$$\hat{\theta} = \underset{\theta}{\text{arg min}} \sum_{t=1}^{n} (\gamma_t - \{\theta_0 + \theta_1 u_t\})^2$$

Linear trend model, cont'd

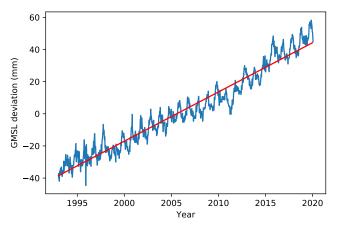
$$\hat{\Theta} = (\Phi^{\mathsf{T}}\Phi)^{\mathsf{T}}\Phi^{\mathsf{T}}\mathbf{y}$$

$$\mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \qquad \Phi = \begin{bmatrix} 1 & u_1 \\ \vdots & \vdots \\ 1 & u_n \end{bmatrix}$$

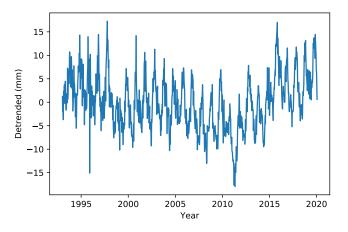
theta 0 is from the original formula Theta1 need to calc the formula

$$\hat{\theta}_{0} = \overline{y} - \hat{\theta}_{1} \overline{u} \qquad (\overline{u})$$

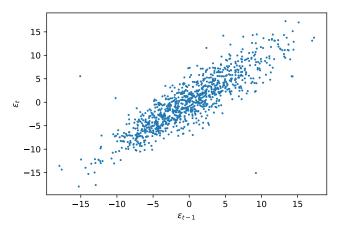
$$\hat{\theta}_{1} = \underbrace{\sum_{i=1}^{n} (y_{i} - \overline{y})(u_{i} - \overline{u})}_{\sum_{k=1}^{n} (u_{k} - \overline{u})^{2}}, \quad \overline{u} = \frac{1}{n} \sum_{k=1}^{n} u_{k}$$



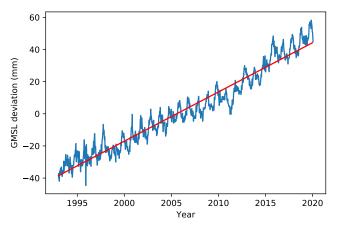
Linear trend fitted with OLS



Residuals – Do they match the model?



Scatter plot of residuals at lag 1



Linear trend fitted with OLS

Classical regression in a time series context

Classical regression is insufficient!

Classical regression in a time series context

Classical regression is insufficient!

Why?

- Time series data have temporal dependencies
- Forecasting = extrapolation

ex) The number of infected individuals tomorrow is **statistically dependent** on the number of infected individuals today