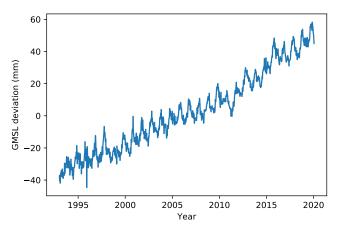


Time Series and Sequence Learning

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

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

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

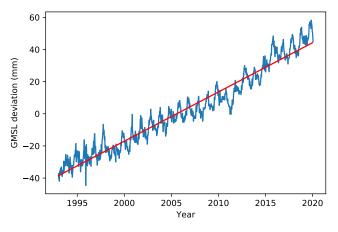
Simple linear trend model: $y_t = \theta_0 + \theta_1 u_t + \varepsilon_t$

The model paremeters
$$\theta = (\theta_0, \theta_1)$$
 are estimated using CLS (= maximum likelihood) $\hat{\theta} = \arg\min_{\theta} \sum_{t=1}^{n} (\gamma_t - [\theta_0 + \theta_t u_t])^2$

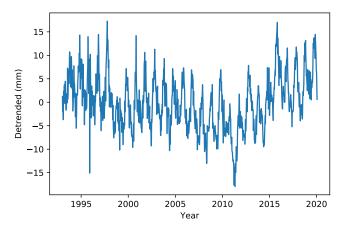
Linear trend model, cont'd

$$\hat{\Theta}_{0} = \overline{y} - \hat{\Theta}_{1} \overline{u} \qquad , \quad \overline{y} = \frac{1}{n} \sum_{k=1}^{n} \gamma_{k}$$

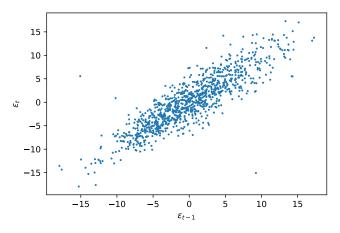
$$\hat{\Theta}_{1} = \underbrace{\sum_{k=1}^{n} (\gamma_{k} - \overline{y})(u_{k} - \overline{u})}_{\sum_{k=1}^{n} (u_{k} - \overline{u})^{2}} , \quad \overline{u} = \frac{1}{n} \sum_{k=1}^{n} u_{k}$$



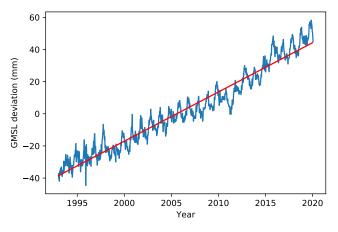
Linear trend fitted with OLS



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