

Climate Data Visualization - Atmospheric CO_2 Concentration / Temperature / Precipitation

Wolfgang Vollmer

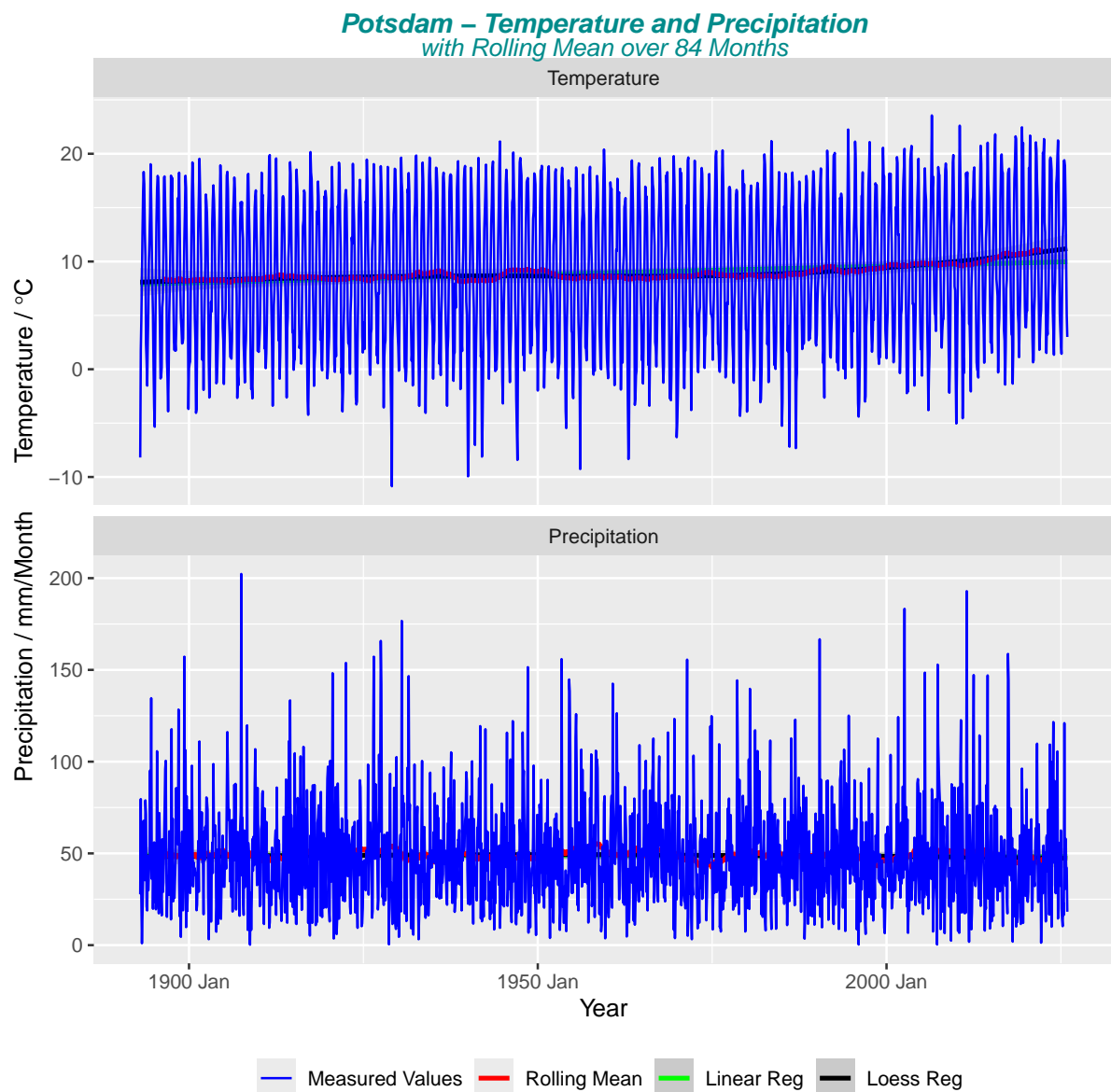
2026-01-08

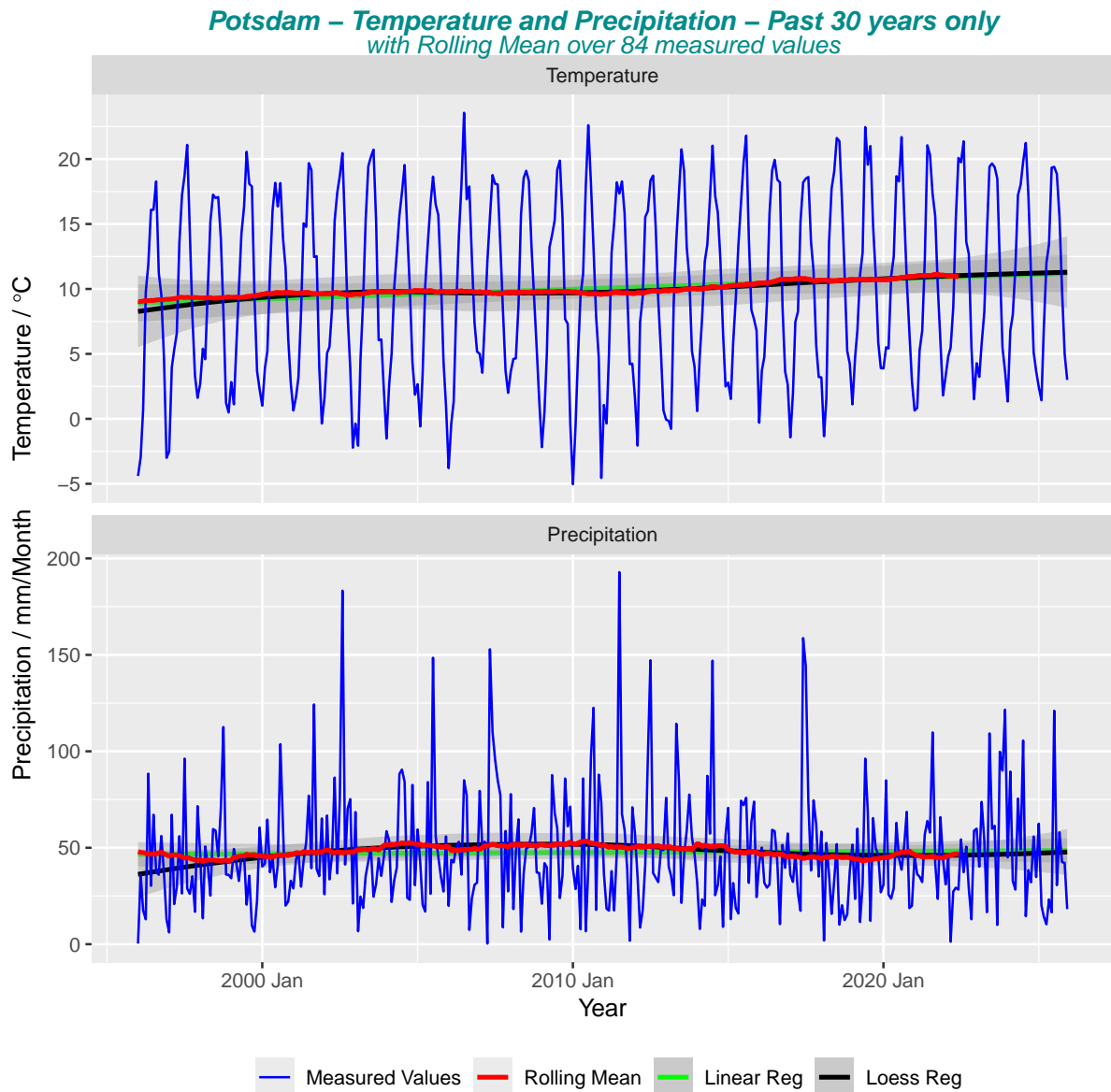
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1 Potsdam - Visualization of Temperature, Precipitation Data 1893 - 2025

1.1 Monthly Time Plots with Rolling Mean

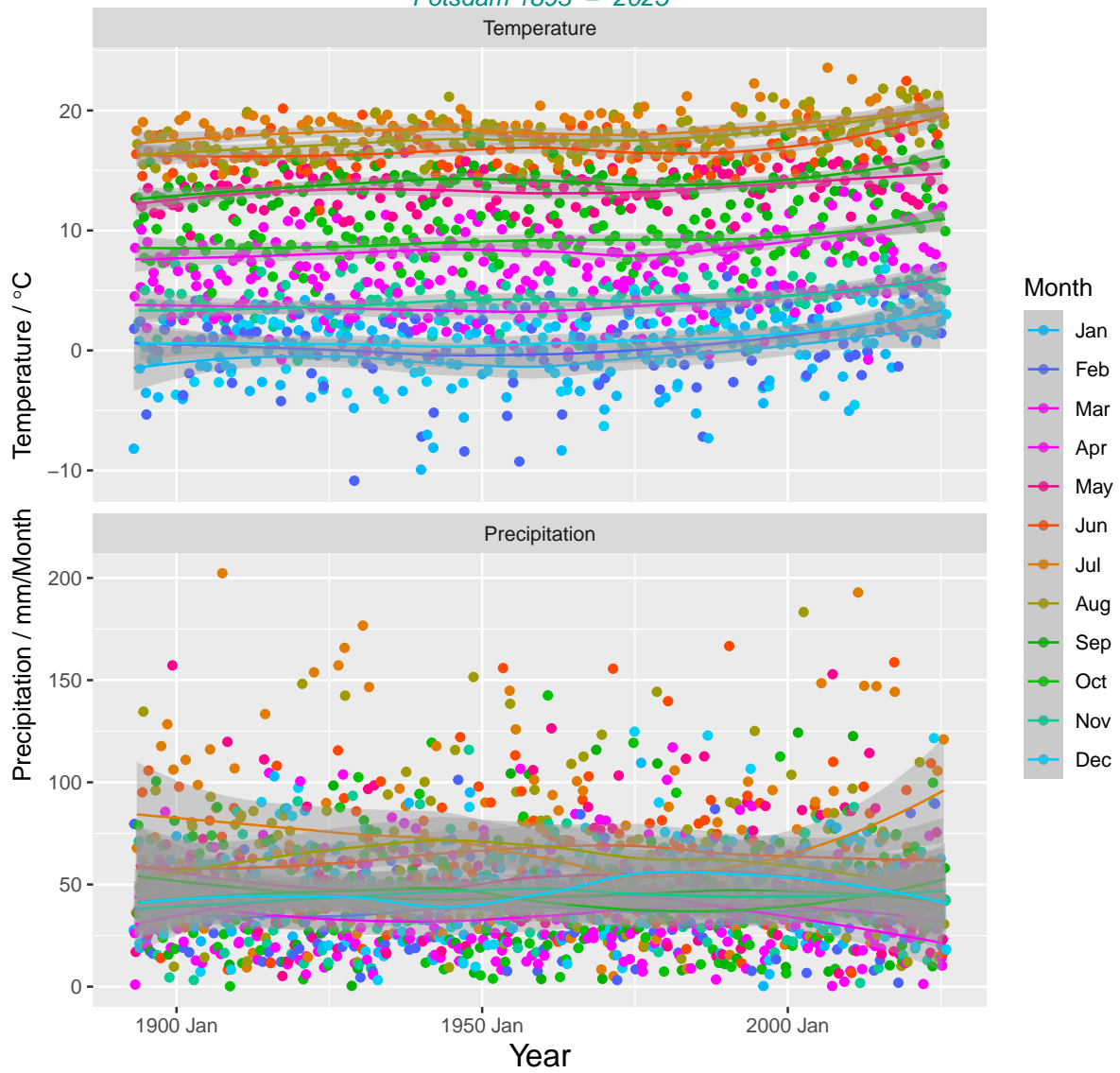




1.2 Annual seasonal plots with monthly breakdown

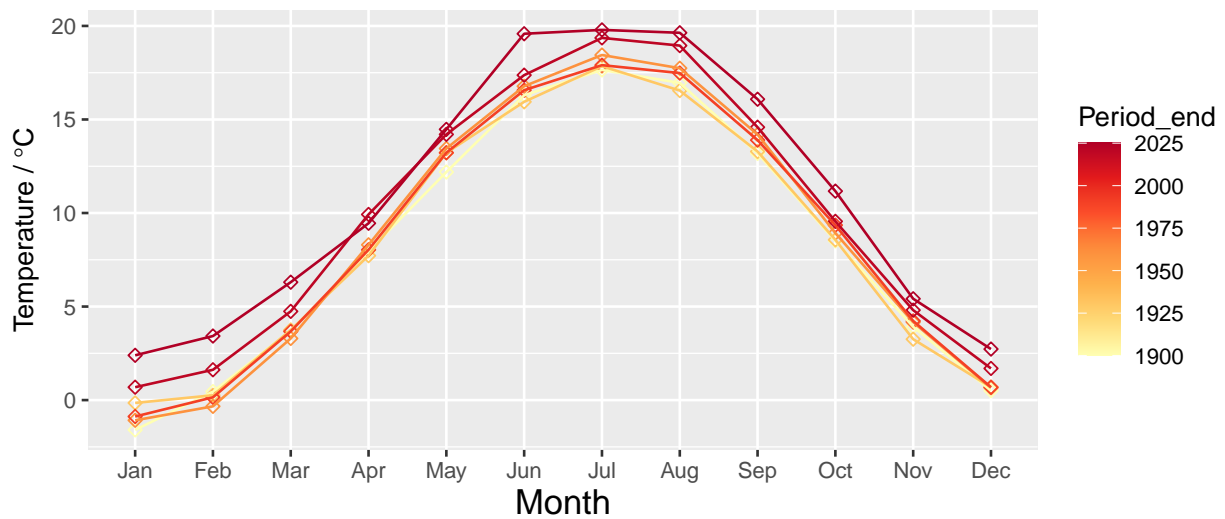
The seasonal charts show the monthly seasonal patterns, where available.

Monthly Data with Local Polynomial Regression Fitting Potsdam 1893 – 2025

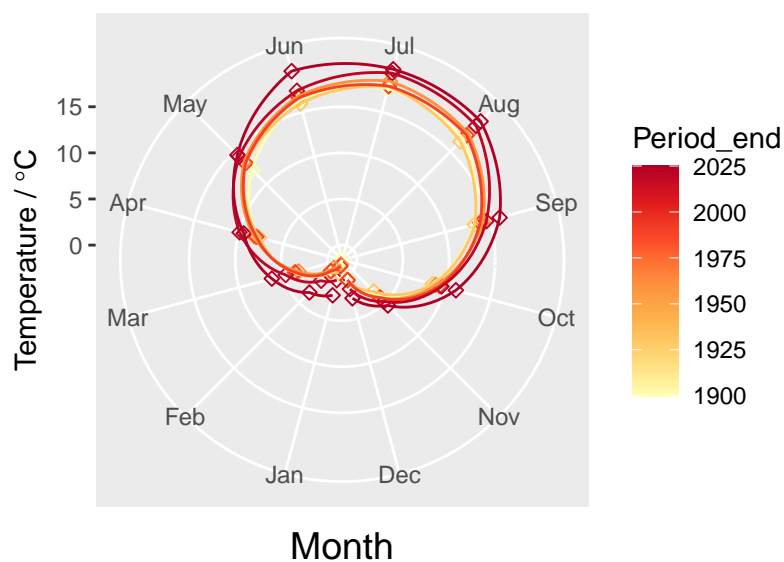


1.2.1 30-year period plots with monthly breakdown - Cartesian and Polar Coordinates

Temperature – Monthly Variations of 30-Year Periods
Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025



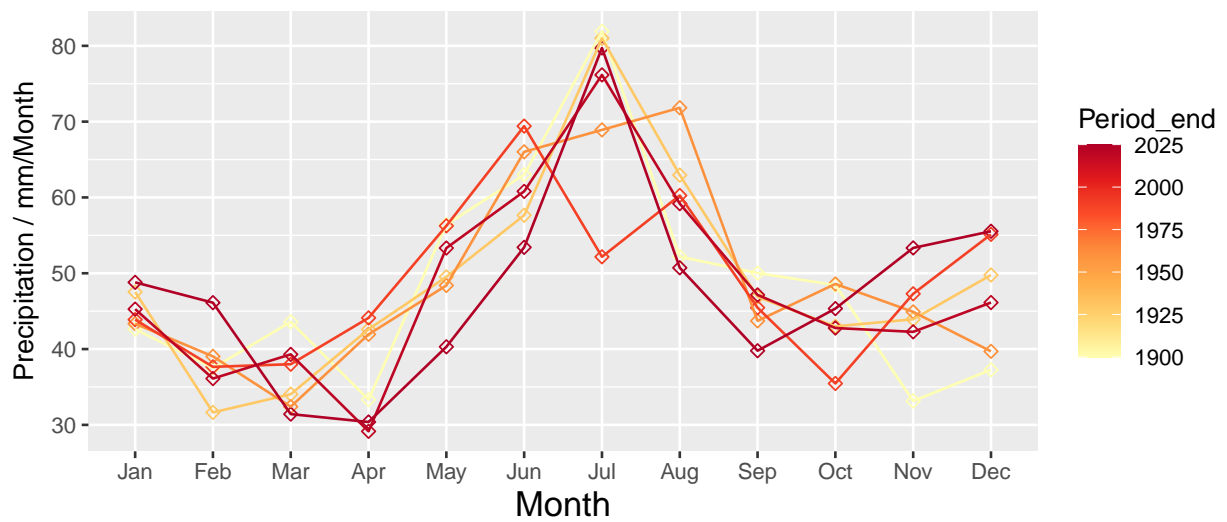
Temperature – Monthly Variations of 30-Year Periods
Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025



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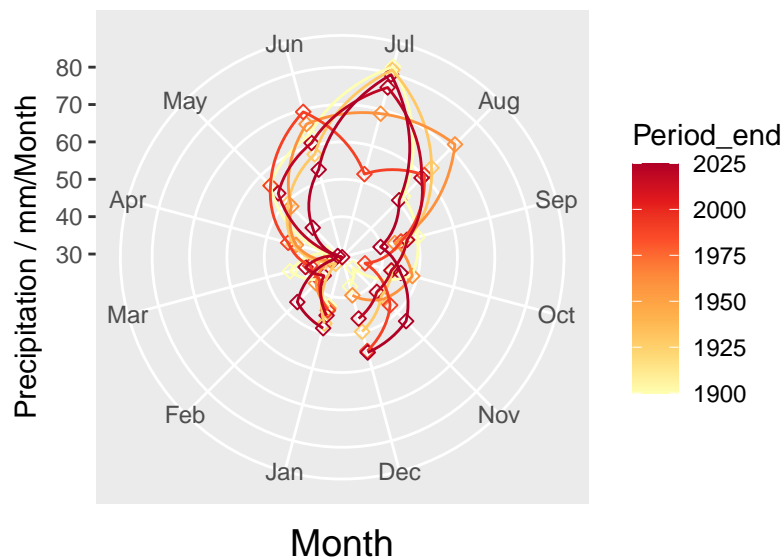
Precipitation – Monthly Variations of 30-Year Periods

Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025



Precipitation – Monthly Variations of 30-Year Periods

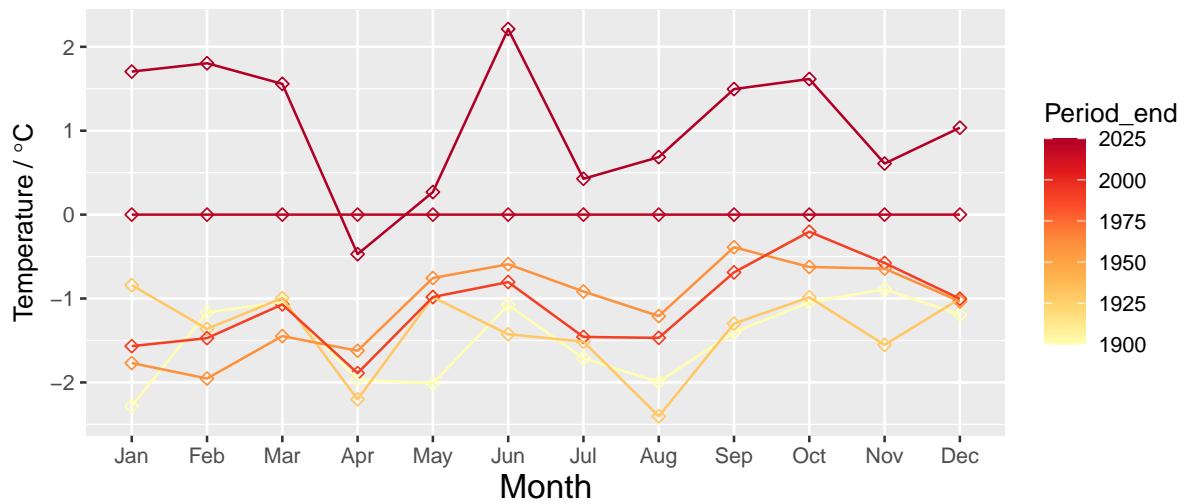
Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025



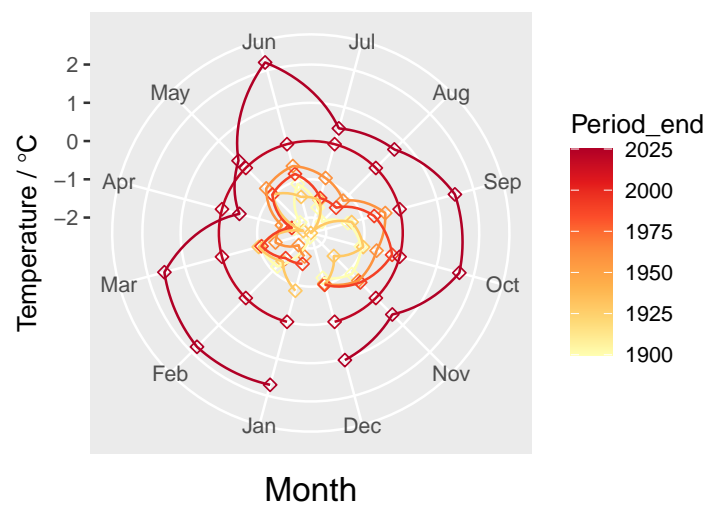
#>
#> _

1.2.2 Plot Monthly Delta to Reference Period - Cartesian and Polar Coordinates

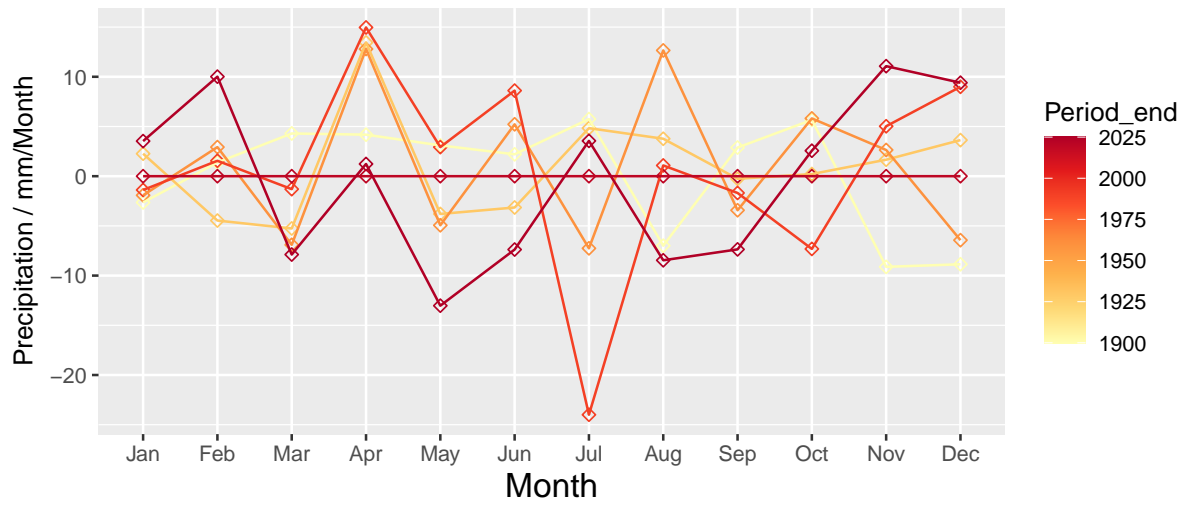
Temperature – Monthly Variations of 30–Year Periods (Delta to Reference) *Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025*



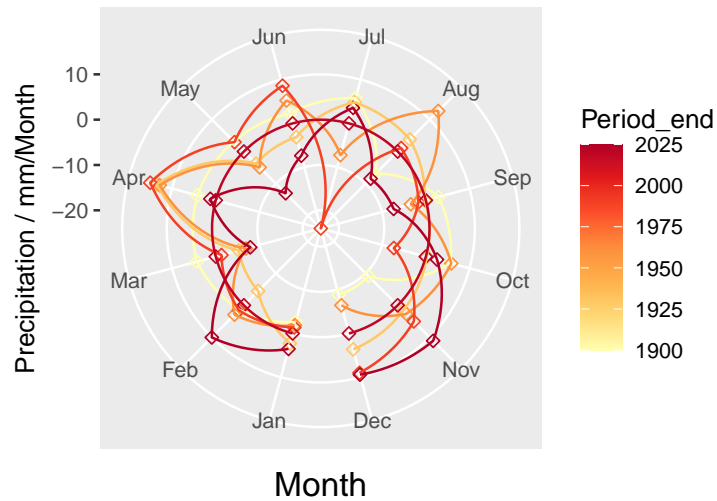
Temperature – Monthly Variations of 30–Year Periods (Delta to Reference) *Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025*



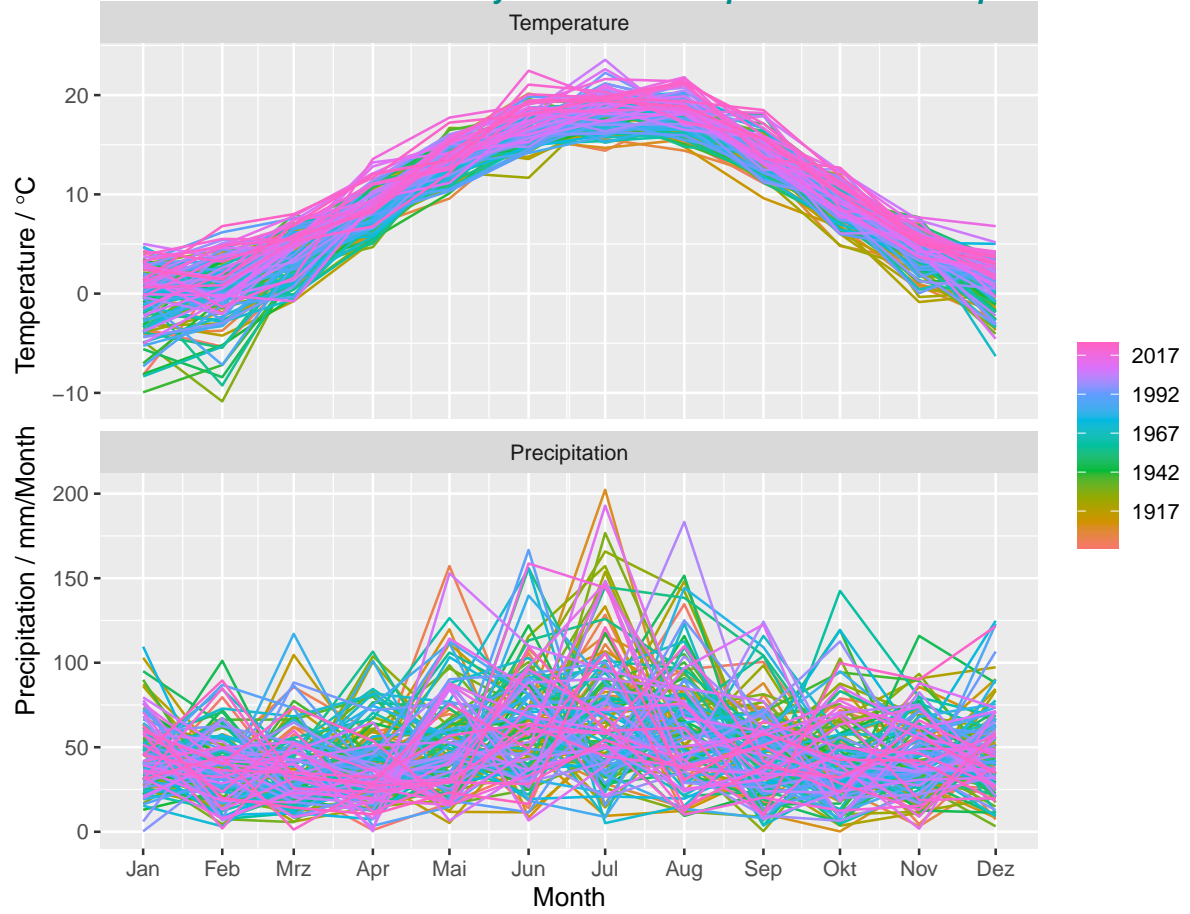
Precipitation – Monthly Variations of 30–Year Periods (Delta to Reference)
 Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025

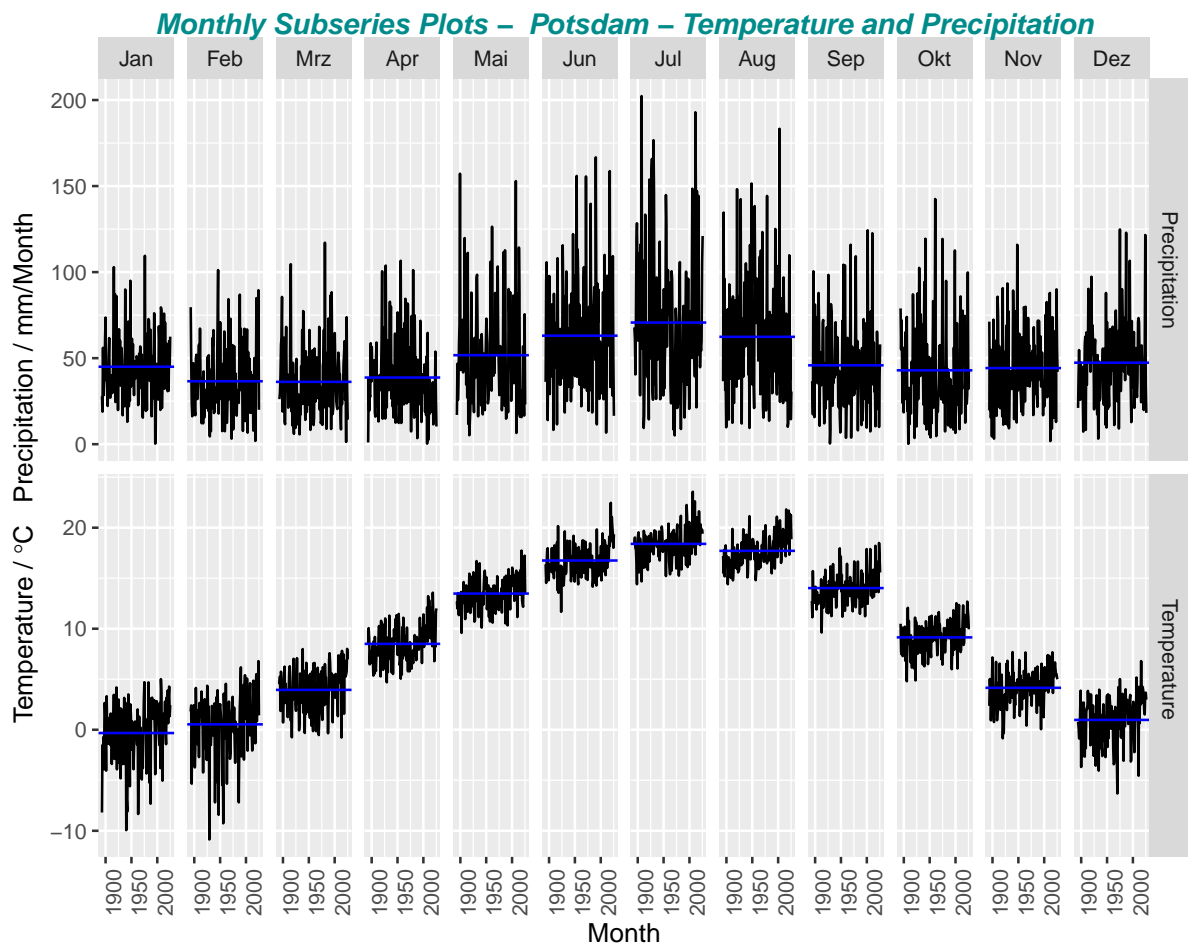


Precipitation – Monthly Variations of 30–Year Periods (Delta to Reference)
 Periods: First 1893–1900 / Reference 1991–2020 / Last 2021–2025



Annual Seasonal Plots – Monthly Potsdam – Temperature and Precipitation

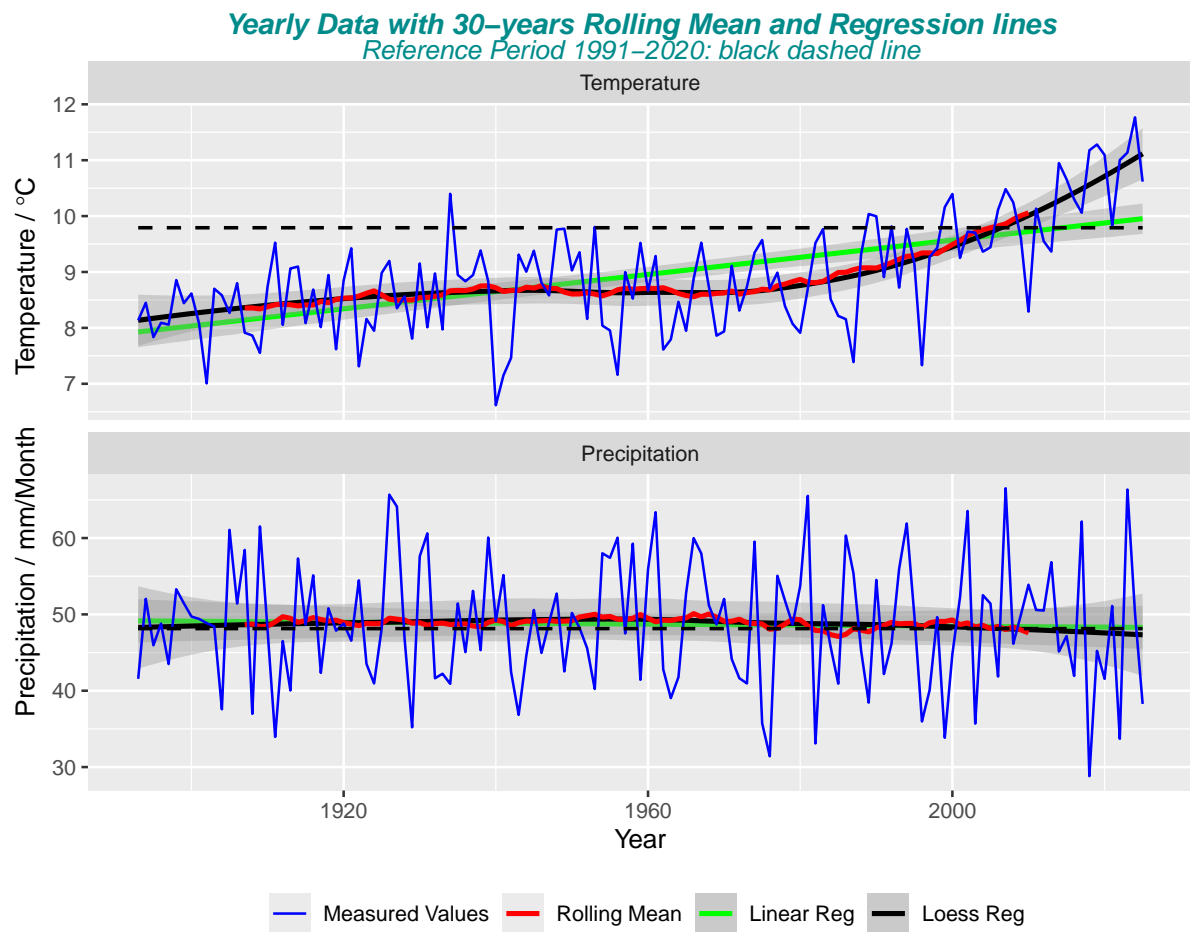




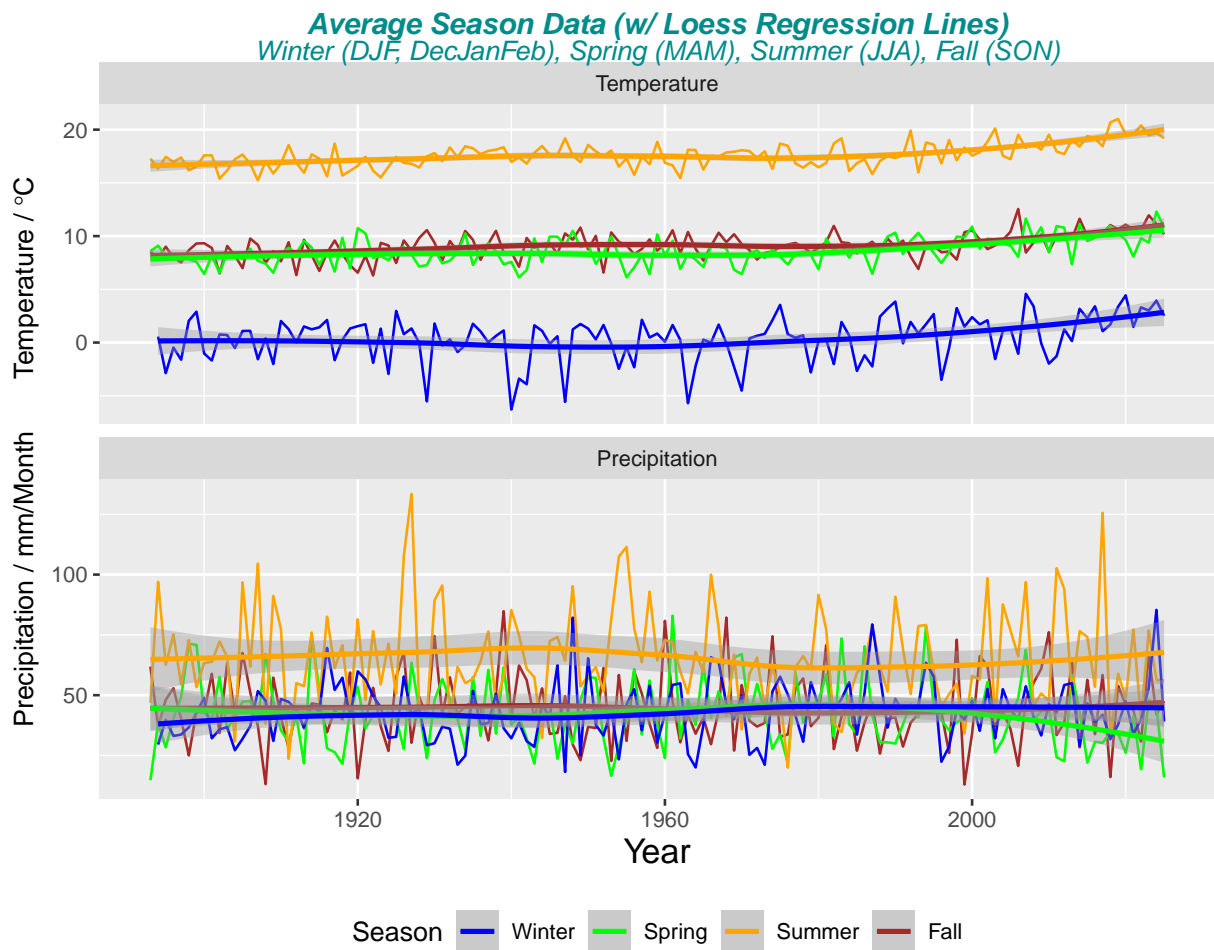
The blue horizontal lines within the seasonal subseries plot indicate the means for each month.

1.3 Annual Potsdam - Temperature and Precipitation

1.3.1 Annual Time Plot of Temperature, Precipitation



1.3.2 Annual Seasonal Plot of Temperature, Precipitation



2 Trend and Seasonal Analysis

2.1 Time Series Decomposition - Trend and Seasonal Components

An *additive model* would be used when the variations around the trend do not vary with the level of the time series whereas a *multiplicative model* would be appropriate if the trend is proportional to the level of the time series.

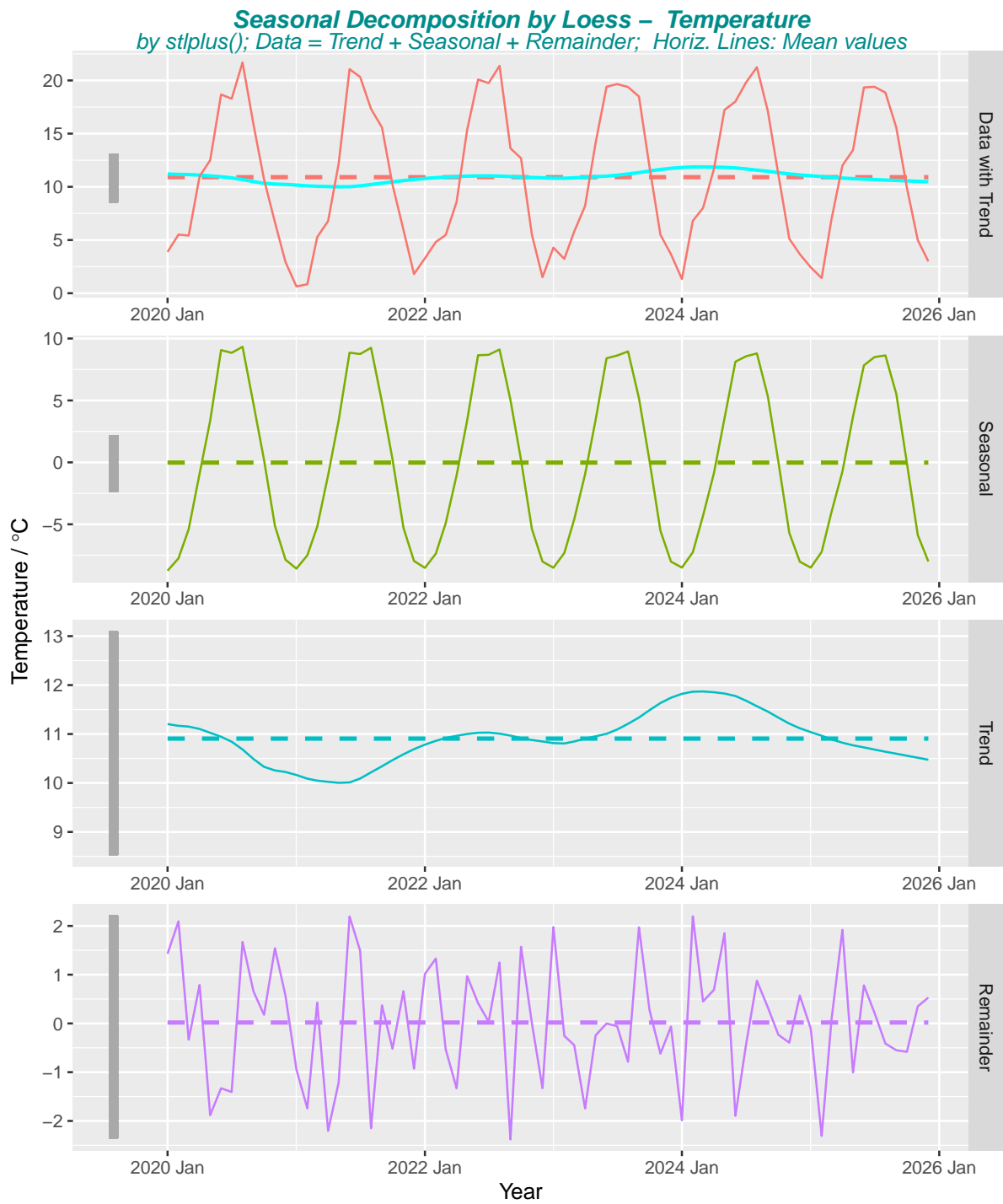
Time series using an

- additive model: $y_t = T_t + C_t + S_t + \epsilon_t$
- multiplicative model: $y_t = T_t * C_t * S_t * \epsilon_t$

Trend / Cycle / Seasonal / Noise component

Cyclical components is often grouped into the Trend component

For *Seasonal decomposition of time series by Loess (stlplus)* uses in general an additive error model, it only provides facilities for additive decompositions. It is possible to obtain a multiplicative decomposition by first taking logs of the data.

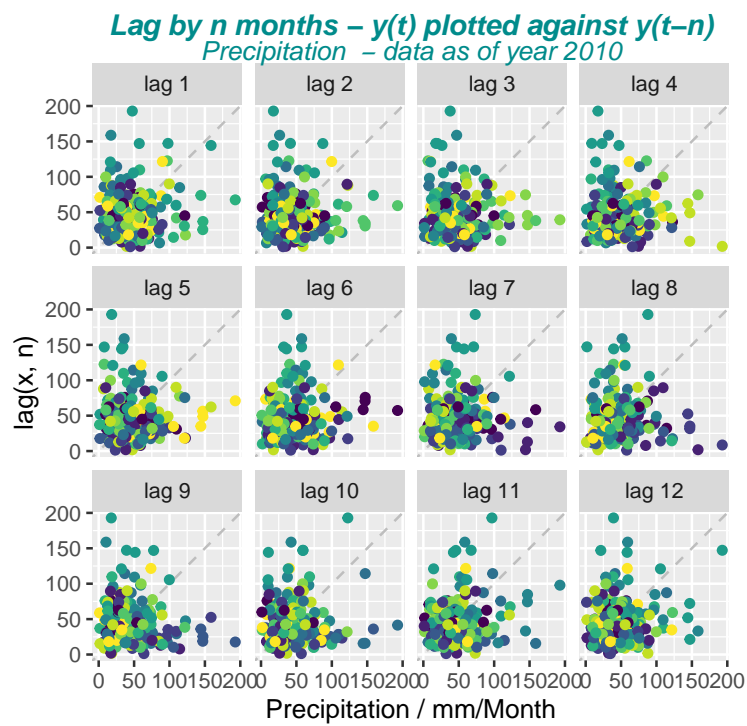
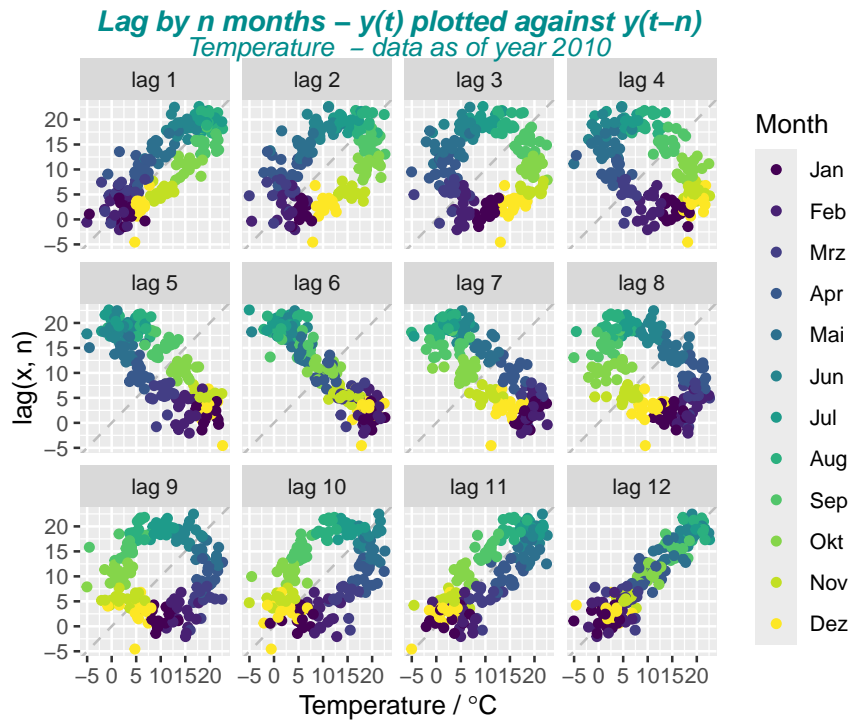




2.2 Periodicities - Season Frequency

2.2.1 Lag Plot - Differences

Lagged scatterplots, where the horizontal axis shows lagged ($k = 1, \dots, 12$) values of the time series. Each graph shows y_t plotted against y_{t-k} for different values of k . For seasonal data the relationship is strongly positive at a lag $k = 12$, reflecting the strong seasonality of the data. The strongly negative relationship is evident in the case of lag $k = 6$.



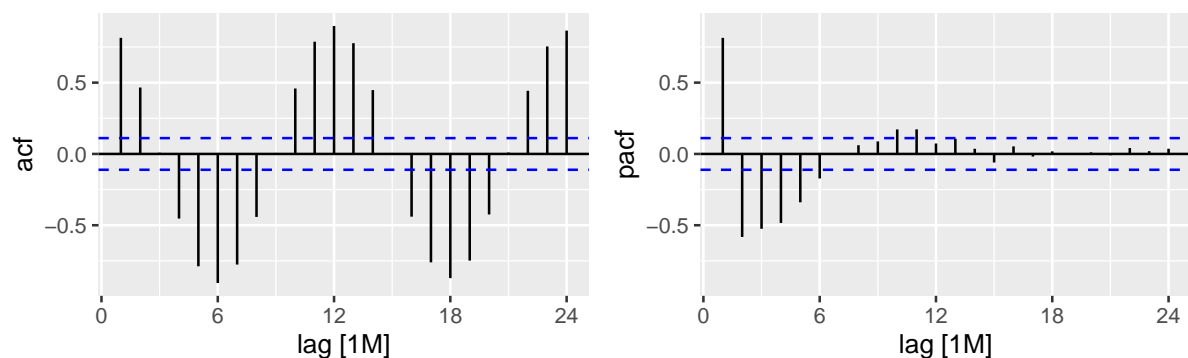
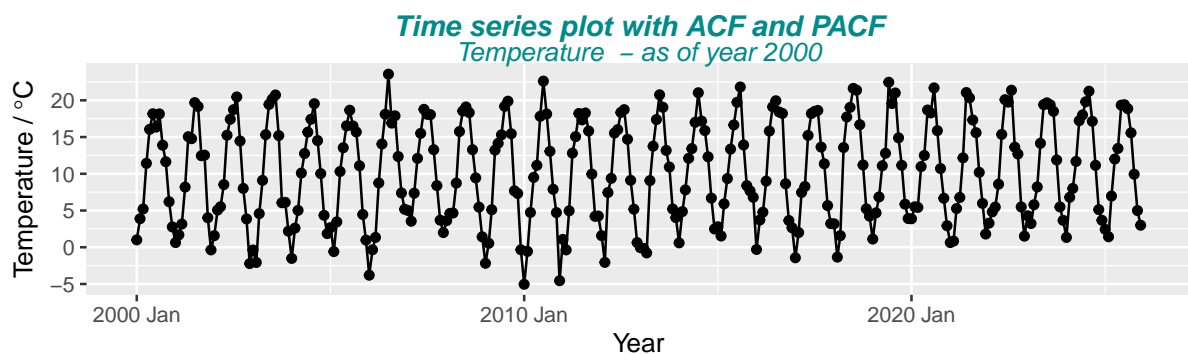
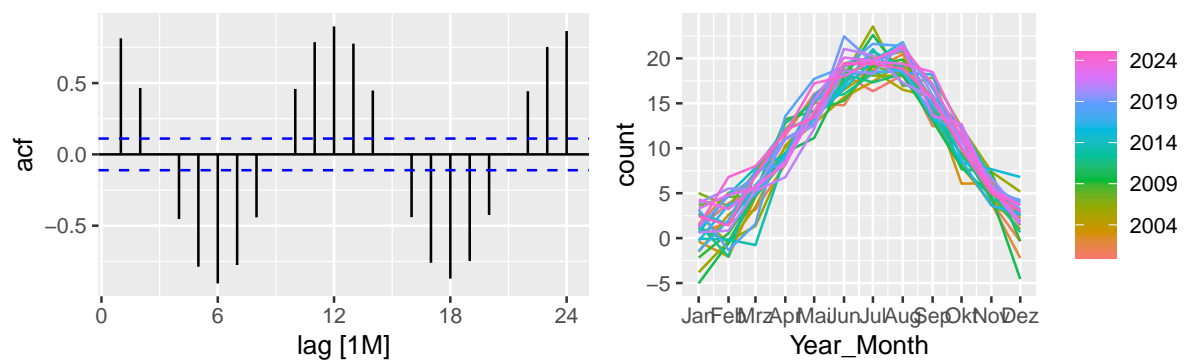
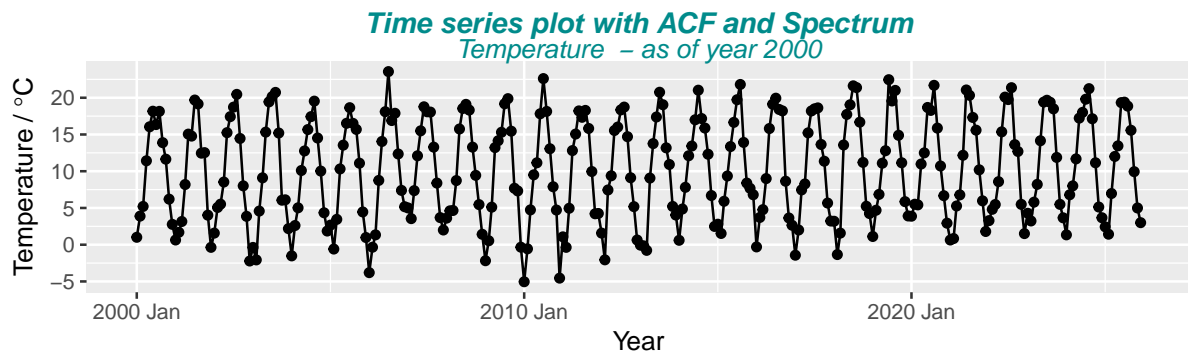
2.2.2 Periodogram - Spectral Density Estimation of a Time Series

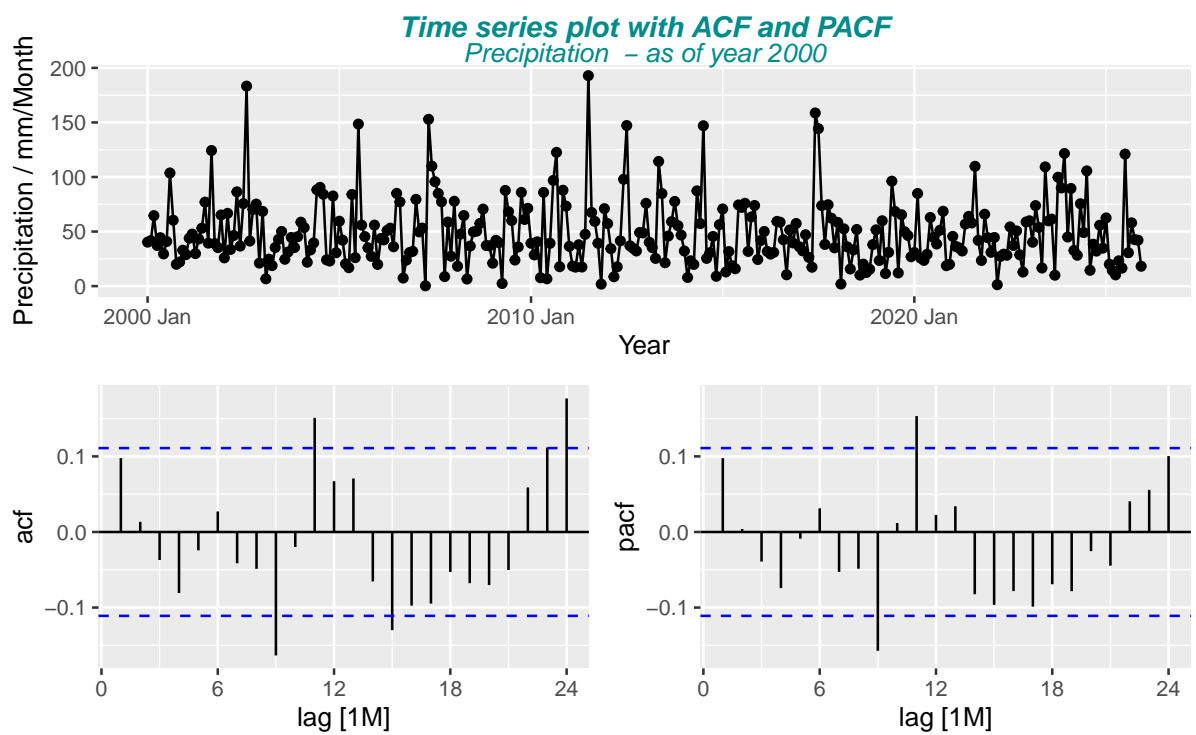
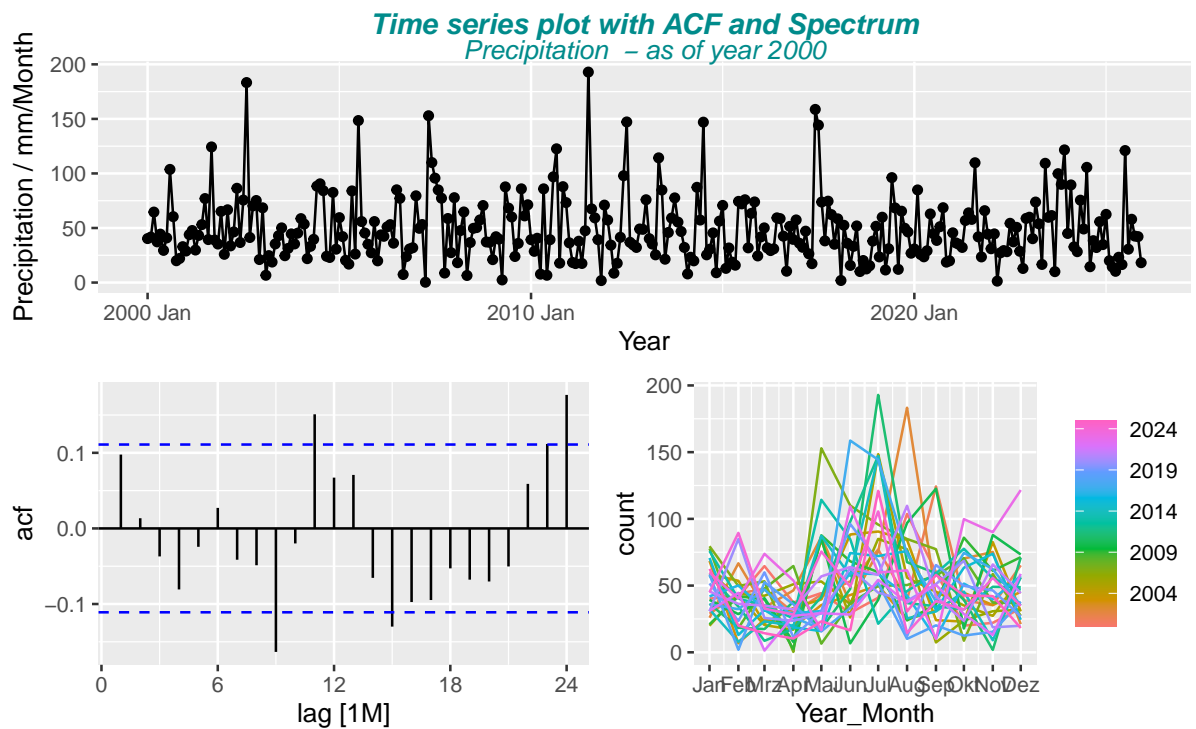
The spectral density characterizes the frequency content of the signal. One purpose of estimating the spectral density is to detect any periodicities in the data, by observing peaks at the frequencies corresponding to these periodicities.

At frequency $\lambda = 1/12$ there is a significant peak \Rightarrow This pattern repeats every full frequency = every 12 months / every year

The remaining peaks are random and therefore cannot be assigned significantly.

Note: The blue dashed lines in the (P)ACF plots ((Partial) Autocorrelation Function) indicate white noise series limits. In that case 95% of the spikes lie within the dashed lines.





3 Forecasting - Estimate/Train the model

3.1 Forecasting with ETS and ARIMA model

Exponential Smoothing (ETS) and **AutoRegressive Integrated Moving Average Forecasting Models ARIMA** models are the two most widely used approaches to time series forecasting, and provide complementary approaches to the problem.

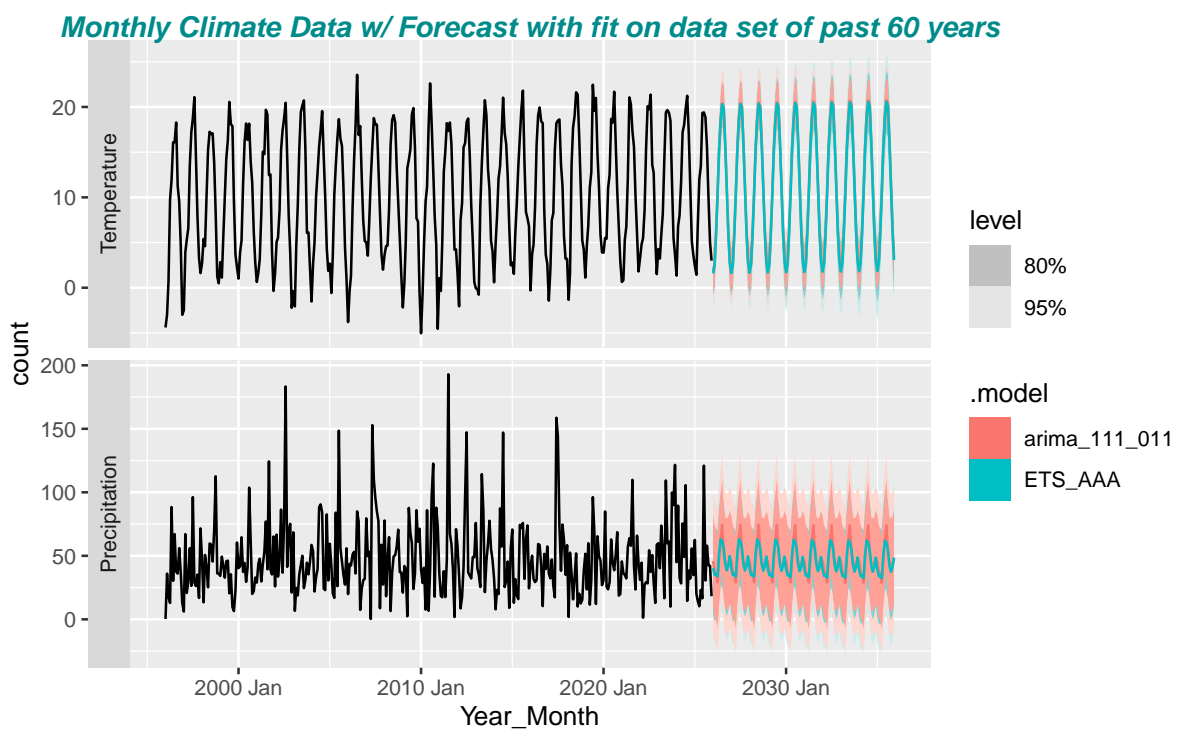
Forecasts produced using **ETS** methods are weighted averages of past observations, with the weights decaying exponentially as the observations get older.

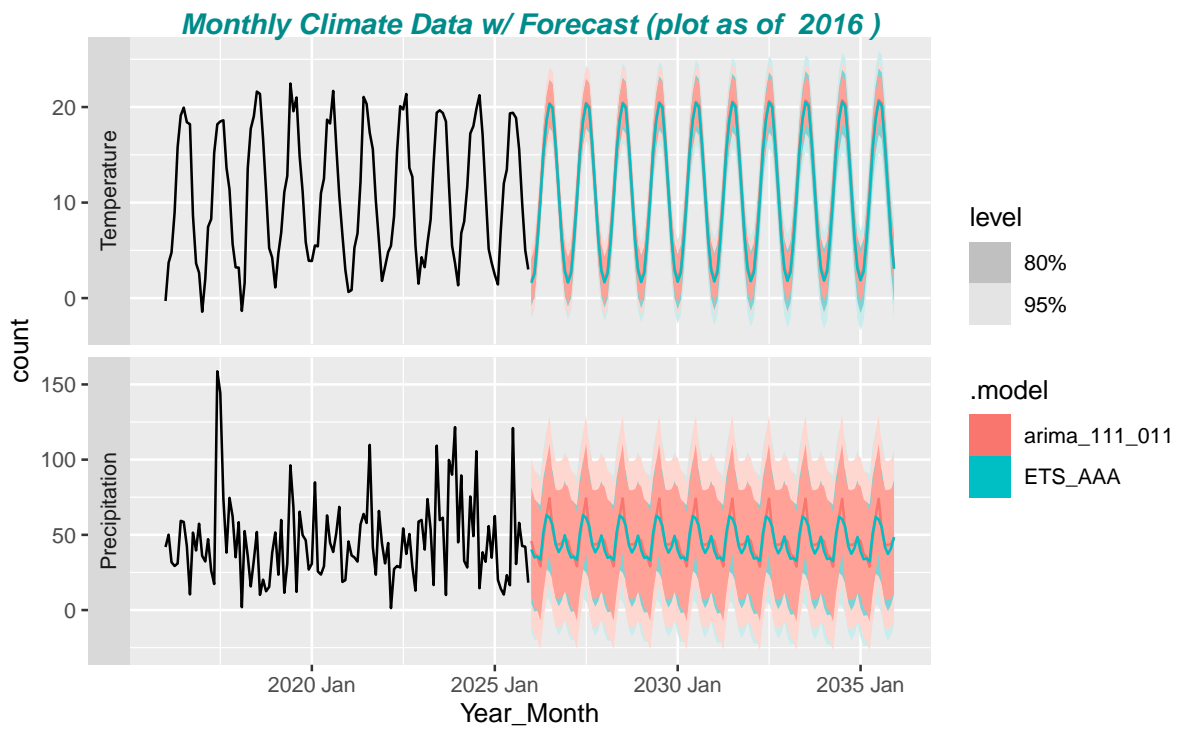
Here a $ETS(A, A|A)$ **model** with additive (“A”) *Error term*, *Trend term* and *Seasonal term* was chosen.

While exponential smoothing models are based on a description of the trend and seasonality in the data, **ARIMA** models aim to describe the autocorrelations in the data.

Here a $ARIMA(111)(011)_{12}$ **model** with autoregressive, differencing, and moving average terms of (111) in the ordinary and 011 in the seasonal term with a seasonal period 12 (12 months/year)

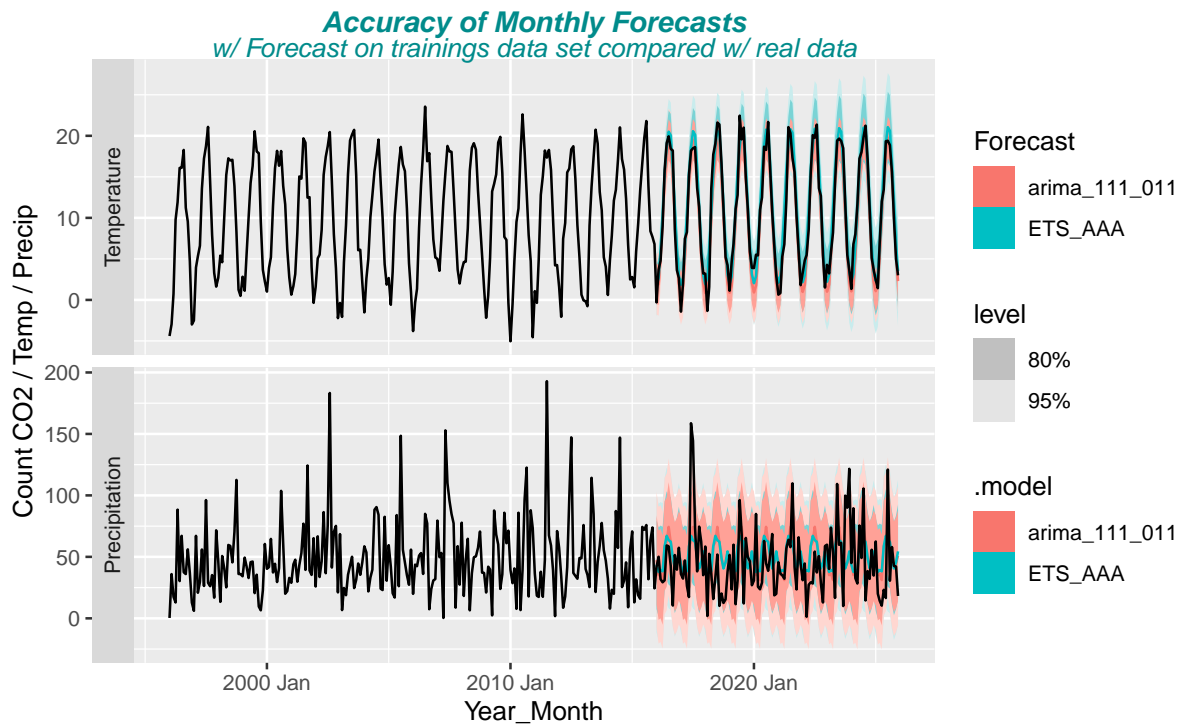
```
#> # A mable: 2 x 4
#> # Key:      City, Measure [2]
#>   City      Measure      ETS_AAA      arima_111_011
#>   <chr>    <fct>        <model>      <model>
#> 1 Potsdam Temperature <ETS(A,A,A)> <ARIMA(1,1,1)(0,1,1)[12]>
#> 2 Potsdam Precipitation <ETS(A,A,A)> <ARIMA(1,1,1)(0,1,1)[12]>
```





3.2 Forecast Accuracy Evaluation

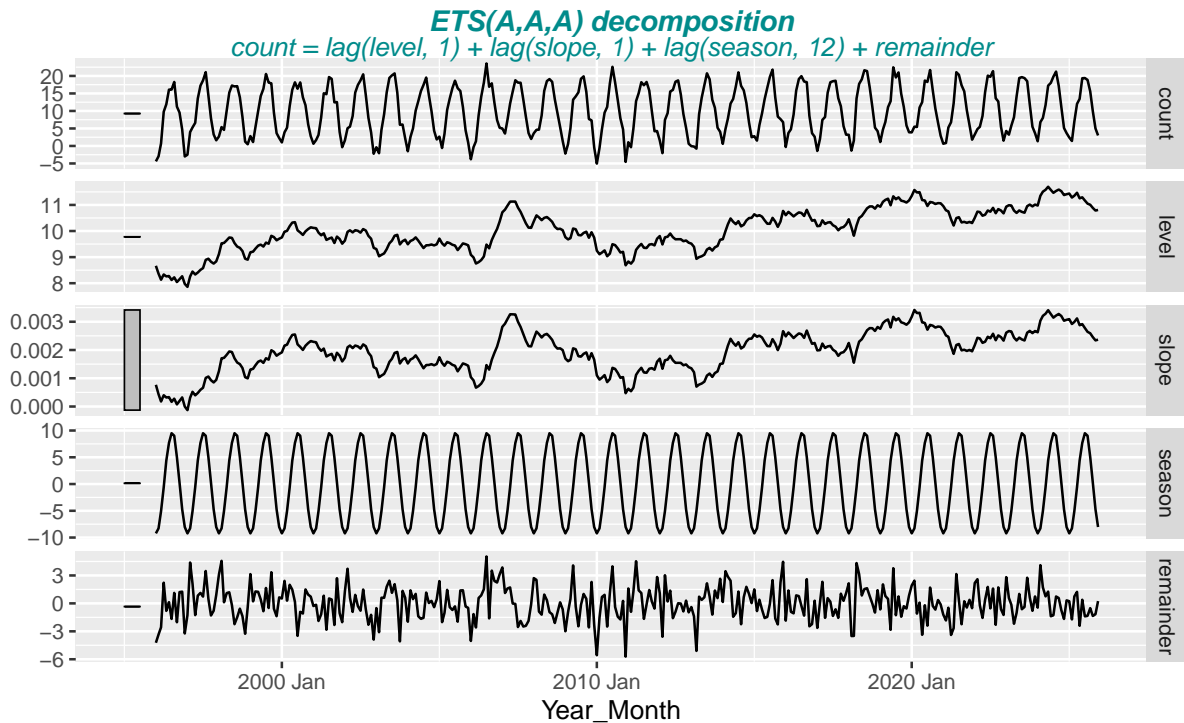
- Forecast Accuracy Evaluation w/ training data “data_train” & test data “data_test”
 - “data” : complete dataset includes the forecasted (future) data range on top of data_train
 - “data_train” = “data” - forecast_range (“data_test”)
 - * data used to train the model (~80% of “data”)
 - “data_test” = “data” - “data_train”
 - * ~ 20% of “data”
 - e.g. for last_year = 2025:
 - * data_train is selected from 1966 - 2015
 - * data_test is selected from 2016 - 2025



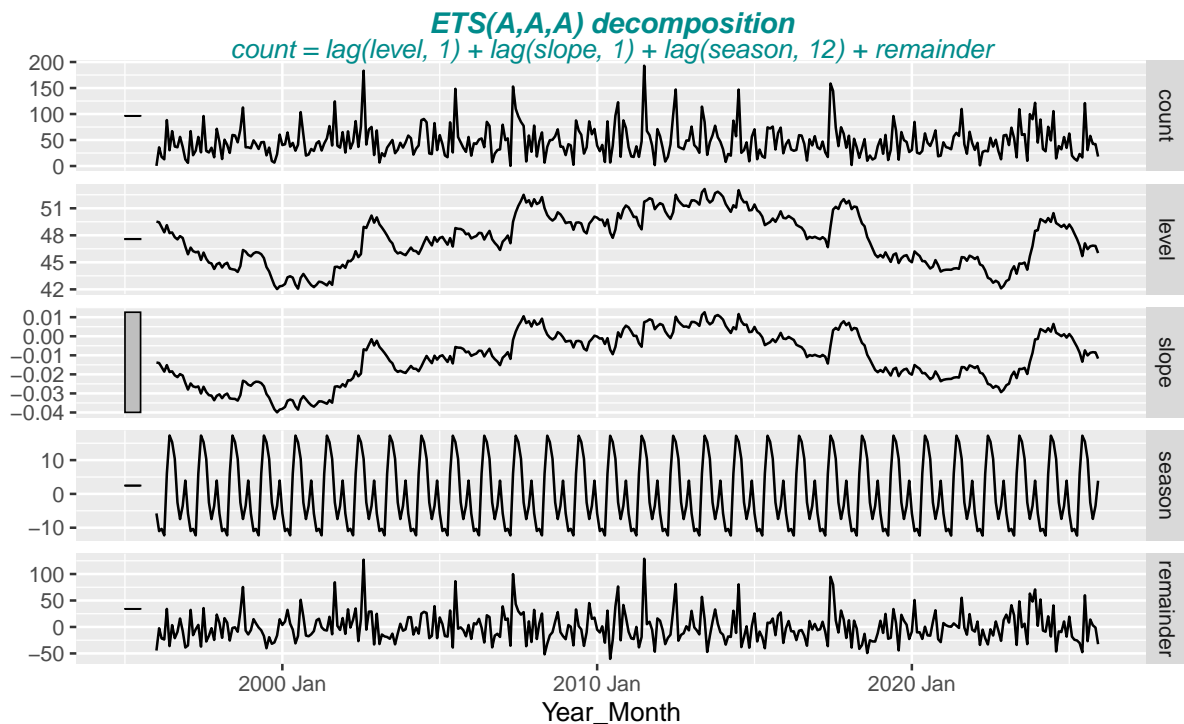
3.2.1 components(fit_ets) - plot of the decomposition of the fitted ETS model

- Note: compare Time series decomposition, for ETS model is valid:
– $\text{count} = \text{lag}(\text{level}, 1) + \text{lag}(\text{slope}, 1) + \text{lag}(\text{season}, 12) + \text{remainder}$

```
#> [1] "Temperature"
```

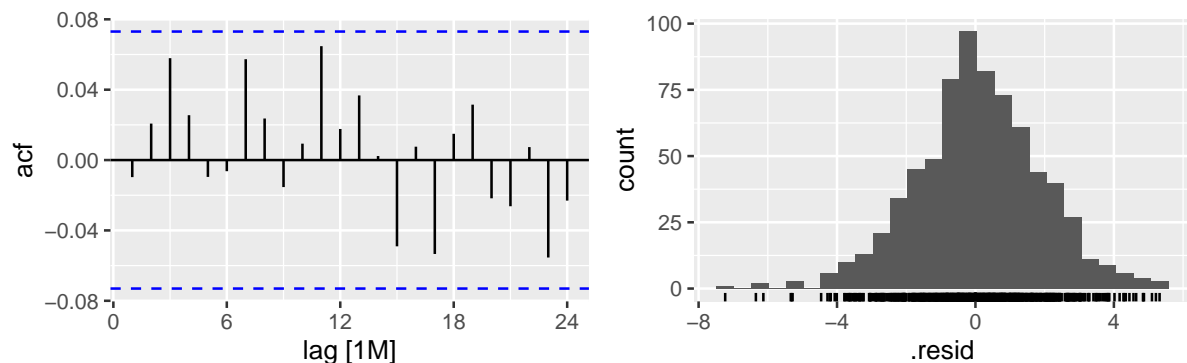
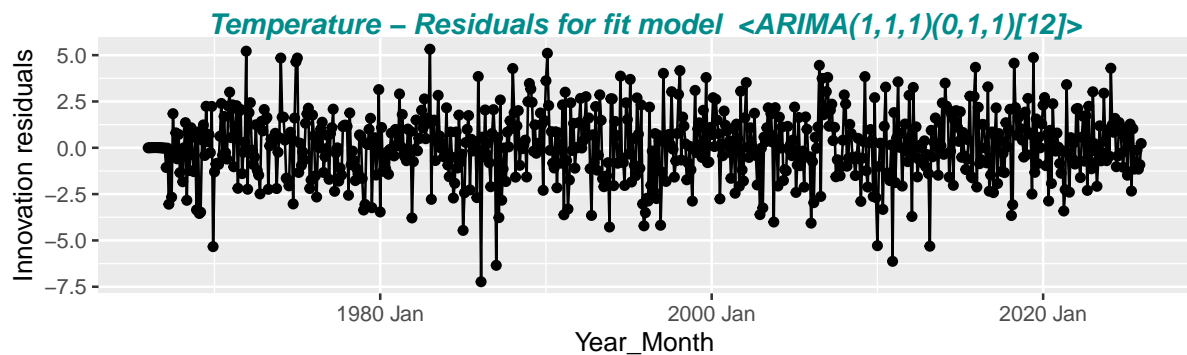
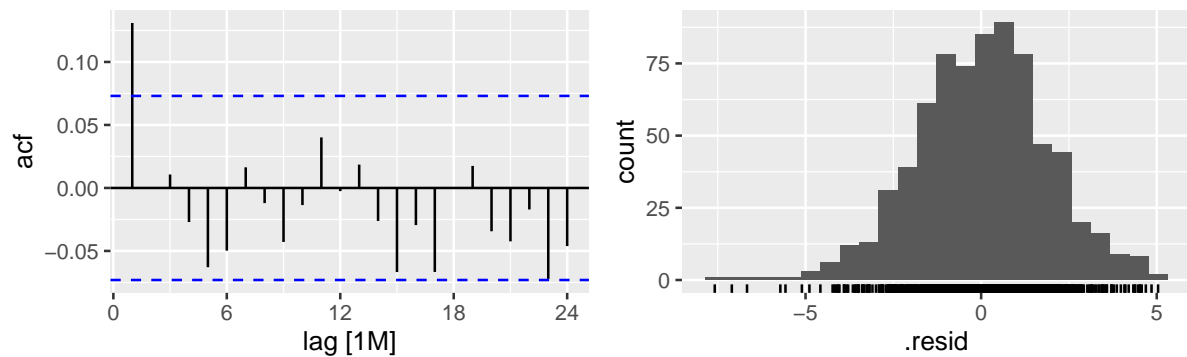
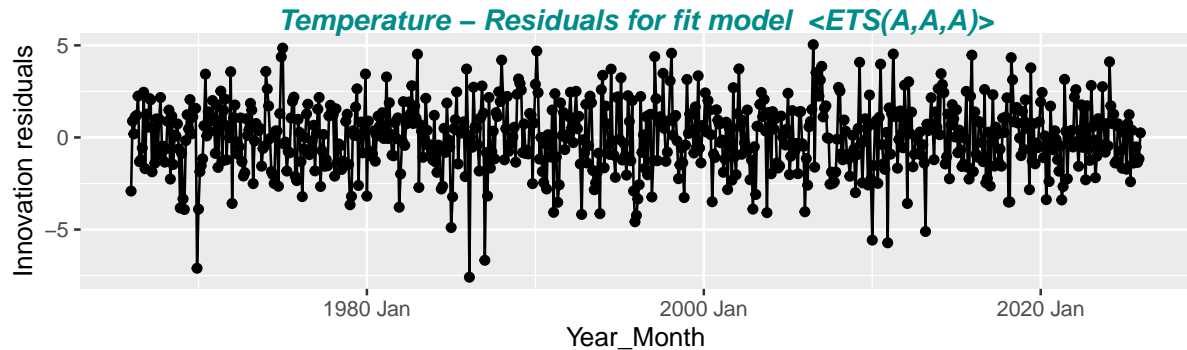


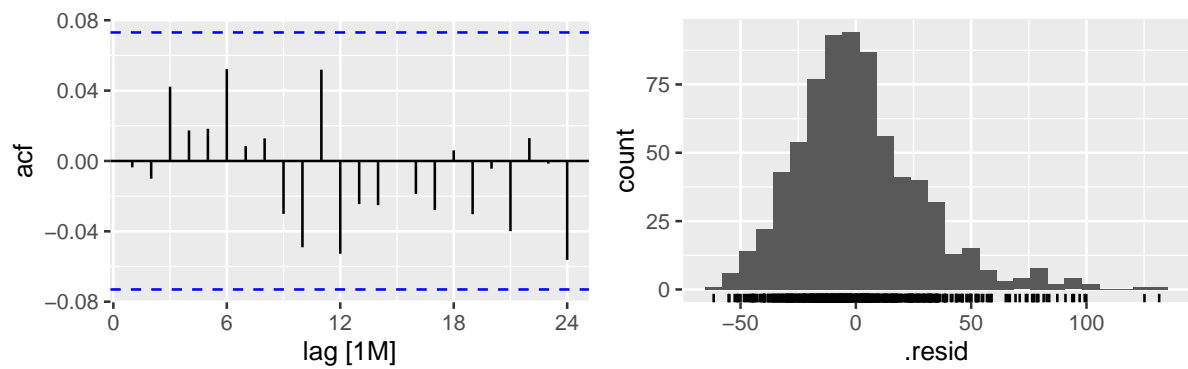
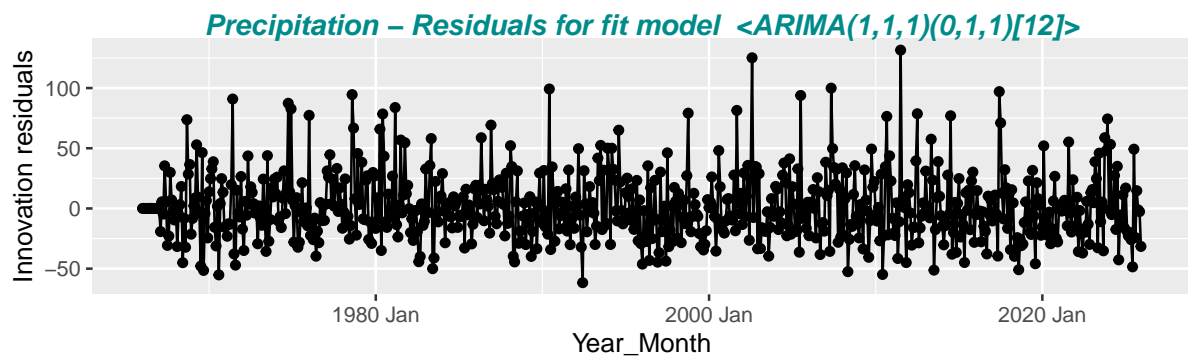
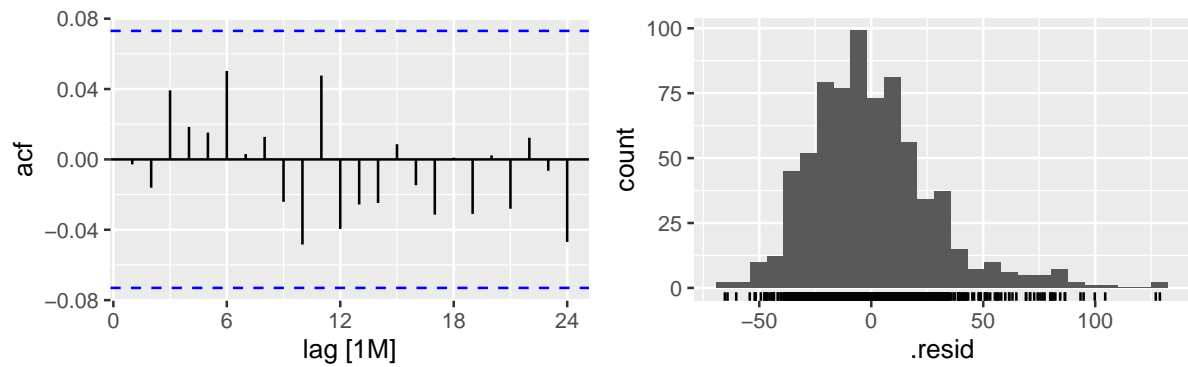
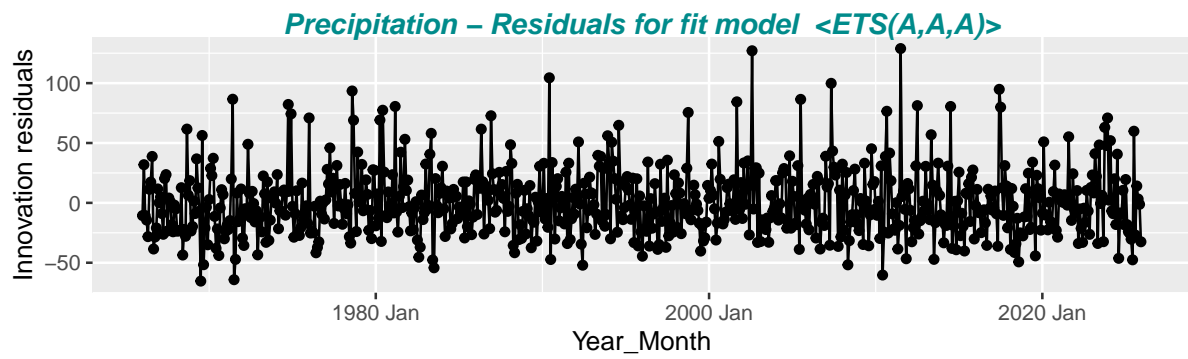
```
#> [1] "Precipitation"
```



3.2.2 gg_tsresiduals(fit) - plot of innovation residuals, acf and histogram

- gg_tsresiduals(fit) (Ch 7.3 Evaluating the regression model)
 - TS of innovation residuals, acf plot, histogram of residuals | PACF (plot_type='partial')
 - innovation residuals should have constant variance (“homoscedasticity”)
 - histogram of the innovation residuals: should be normally distributed





4 Forecast Tables

4.1 Yearly mean values of past time periods

Table 1: Mean values for the given time periods; Units: Temperature (degree C), Precipitation (mm/Month), CO2 (ppm)

Period_Time	Temperature	Precipitation
1893-1900	8.3	48.3
1901-1930	8.4	49.2
1931-1960	8.7	49.1
1961-1990	8.7	48.8
1991-2020	9.8	48.1
2021-2025	10.9	47.9

4.2 Yearly mean forecast values for the next 25 years

Table 2: Mean Yearly ARIMA and ETS Forecast values (next 25 years); Units: Temperature (degree C), Precipitation (mm/Month), CO2 (ppm)

City	Measure	Year	ETS_AAA	arima_111_011
Potsdam	Temperature	2026	10.82	11.00
Potsdam	Temperature	2030	10.94	11.16
Potsdam	Temperature	2035	11.08	11.37
Potsdam	Temperature	2040	11.22	11.57
Potsdam	Temperature	2045	11.36	11.78
Potsdam	Temperature	2050	11.51	11.98
Potsdam	Precipitation	2026	45.96	47.20
Potsdam	Precipitation	2030	45.39	47.04
Potsdam	Precipitation	2035	44.69	46.83
Potsdam	Precipitation	2040	43.98	46.62
Potsdam	Precipitation	2045	43.28	46.42
Potsdam	Precipitation	2050	42.58	46.21

Table 3: Forecast increase/decrease over the next 25 years; Units: Temperature (degree C), Precipitation (mm/Month), CO2 (ppm)

Measure	Year.x	Year.y	ETS.x	ARIMA.x	ETS.y	ARIMA.y	Delta_ET	Delta_ARIMA
Temperature	2026	2050	10.82	11.0	11.51	11.98	0.68	0.99
Precipitation	2026	2050	45.96	47.2	42.58	46.21	-3.38	-0.99

Table 4: Forecast increase/decrease over the next 25 years; Units: Temperature (degree C), Precipitation (mm/Month), CO2 (ppm)

Measure	Month	Year.x	Year.y	Mean.x_ET	Mean.x_ARIMA	Mean.y_ET	Mean.y_ARIMA	Delta_ET	Delta_ARIMA
Temperature	Jan	2026	2050	1.59	1.79	2.27	2.78	0.68	0.99
Temperature	Feb	2026	2050	2.53	2.74	3.21	3.73	0.68	0.99
Temperature	Mar	2026	2050	5.76	6.07	6.44	7.06	0.68	0.99
Temperature	Apr	2026	2050	10.23	10.54	10.91	11.52	0.68	0.99
Temperature	May	2026	2050	15.20	15.43	15.88	16.42	0.68	0.99
Temperature	Jun	2026	2050	18.52	18.70	19.20	19.69	0.68	0.99

Measure	Month	Year.x	Year.y	Mean.x_ET	Mean.x_ARIMA	Mean.y_ET	Mean.y_ARIMA	Delta_ET	Delta_ARIMA
Temperature	Jul	2026	2050	20.32	20.39	21.00	21.38	0.68	0.99
Temperature	Aug	2026	2050	19.86	20.02	20.54	21.01	0.68	0.99
Temperature	Sep	2026	2050	15.86	15.92	16.54	16.91	0.68	0.99
Temperature	Oct	2026	2050	11.12	11.16	11.80	12.15	0.68	0.99
Temperature	Nov	2026	2050	6.11	6.17	6.79	7.15	0.68	0.99
Temperature	Dec	2026	2050	2.80	3.04	3.48	4.02	0.68	0.99
Precipitation	Jan	2026	2050	40.29	45.91	36.91	45.05	-3.38	-0.86
Precipitation	Feb	2026	2050	35.01	37.60	31.63	36.60	-3.38	-1.00
Precipitation	Mar	2026	2050	35.64	35.75	32.26	34.75	-3.38	-1.00
Precipitation	Apr	2026	2050	33.63	29.15	30.26	28.15	-3.38	-1.00
Precipitation	May	2026	2050	51.92	47.75	48.54	46.75	-3.38	-1.00
Precipitation	Jun	2026	2050	63.19	62.63	59.81	61.63	-3.38	-1.00
Precipitation	Jul	2026	2050	61.45	74.30	58.07	73.30	-3.38	-1.00
Precipitation	Aug	2026	2050	56.29	53.21	52.91	52.21	-3.38	-1.00
Precipitation	Sep	2026	2050	43.40	43.98	40.02	42.98	-3.38	-1.00
Precipitation	Oct	2026	2050	38.51	43.52	35.13	42.52	-3.38	-1.00
Precipitation	Nov	2026	2050	42.36	44.46	38.98	43.46	-3.38	-1.00
Precipitation	Dec	2026	2050	49.79	48.11	46.41	47.11	-3.38	-1.00

5 Backup

5.1 Potsdam - Average Yearly and Seasonal Data

Table 5: Annual paste(“Temperature /”, degree * C) (first and last 10 years)

City	Measure	Year	Winter_avg	Spring_avg	Summer_avg	Fall_avg	Year_avg
Potsdam	Temperature	1893	NA	8.6	17.3	8.5	8.1
Potsdam	Temperature	1894	0.6	9.1	16.4	8.0	8.4
Potsdam	Temperature	1895	-2.9	8.1	17.4	9.0	7.8
Potsdam	Temperature	1896	-0.5	7.9	17.0	8.0	8.1
Potsdam	Temperature	1897	-1.6	8.1	17.4	7.5	8.1
Potsdam	Temperature	1898	2.0	7.8	16.2	8.7	8.9
Potsdam	Temperature	1899	2.9	7.6	16.5	9.3	8.4
Potsdam	Temperature	1900	-1.0	6.4	17.6	9.3	8.6
Potsdam	Temperature	1901	-1.7	8.1	17.6	8.9	8.1
Potsdam	Temperature	1902	0.8	6.5	15.4	6.5	7.0
Potsdam	Temperature	2016	3.4	9.9	19.2	10.2	10.3
Potsdam	Temperature	2017	1.1	10.3	18.4	10.2	10.1
Potsdam	Temperature	2018	1.7	11.0	20.7	11.0	11.2
Potsdam	Temperature	2019	3.4	10.2	21.0	10.6	11.3
Potsdam	Temperature	2020	4.4	9.6	19.6	11.1	11.1
Potsdam	Temperature	2021	1.5	8.1	19.6	10.6	9.8
Potsdam	Temperature	2022	3.3	9.8	20.4	10.6	11.0
Potsdam	Temperature	2023	3.0	9.4	19.5	12.0	11.1
Potsdam	Temperature	2024	3.9	12.3	19.7	11.1	11.8
Potsdam	Temperature	2025	2.5	10.8	19.2	10.2	10.6

Table 6: Annual Precipitation / mm/Month (first and last 10 years)

City	Measure	Year	Winter_avg	Spring_avg	Summer_avg	Fall_avg	Year_avg
Potsdam	Precipitation	1893	NA	14.7	46.9	61.8	41.6
Potsdam	Precipitation	1894	29.5	38.3	97.0	37.0	52.0
Potsdam	Precipitation	1895	39.8	28.4	62.6	48.1	46.0
Potsdam	Precipitation	1896	33.1	40.3	75.2	52.9	48.8
Potsdam	Precipitation	1897	33.7	49.1	54.4	40.2	43.5
Potsdam	Precipitation	1898	36.7	71.3	72.9	25.0	53.3
Potsdam	Precipitation	1899	42.9	70.8	53.3	43.1	51.5
Potsdam	Precipitation	1900	48.9	42.7	63.3	42.8	49.7
Potsdam	Precipitation	1901	32.1	40.5	63.8	58.9	49.4
Potsdam	Precipitation	1902	35.8	57.6	72.2	34.3	48.9
Potsdam	Precipitation	2016	38.8	30.7	53.5	33.8	41.9
Potsdam	Precipitation	2017	41.9	30.2	125.6	58.4	62.2
Potsdam	Precipitation	2018	31.8	34.7	31.7	16.0	28.8
Potsdam	Precipitation	2019	37.7	34.1	58.9	53.9	45.2
Potsdam	Precipitation	2020	47.5	26.1	48.7	46.2	41.6
Potsdam	Precipitation	2021	33.9	41.4	77.2	43.8	51.1
Potsdam	Precipitation	2022	40.0	19.3	40.0	30.7	33.7
Potsdam	Precipitation	2023	53.0	48.1	76.8	66.6	66.3
Potsdam	Precipitation	2024	85.4	45.5	56.4	42.1	50.1
Potsdam	Precipitation	2025	39.1	15.9	56.0	47.6	38.3

Table 7: Monthly Means over all Years (Temperature / degree C and Monthly Precipitation / mm)

City	Month	Temperature	Precipitation
Potsdam	Jan	-0.3	45.0
Potsdam	Feb	0.5	36.6
Potsdam	Mar	3.9	36.2
Potsdam	Apr	8.5	38.7
Potsdam	May	13.5	51.7
Potsdam	Jun	16.8	63.1
Potsdam	Jul	18.4	70.7
Potsdam	Aug	17.7	62.4
Potsdam	Sep	14.0	45.8
Potsdam	Oct	9.1	42.9
Potsdam	Nov	4.2	44.2
Potsdam	Dec	1.0	47.4

5.2 Potsdam - Head and tail of data

```
#> # A tibble: 6 x 5 [1M]
#> # Key:      City, Measure [1]
#> # Groups:   City, Measure [1]
#>   City      Measure      Year_Month Period_Time count
#>   <chr>    <fct>          <moth> <chr>      <dbl>
#> 1 Potsdam Temperature  1893 Jan 1893-1900  -8.19
#> 2 Potsdam Temperature  1893 Feb 1893-1900   1.81
#> 3 Potsdam Temperature  1893 Mrz 1893-1900   4.51
#> 4 Potsdam Temperature  2025 Okt 2021-2025   9.94
#> 5 Potsdam Temperature  2025 Nov 2021-2025    5
#> 6 Potsdam Temperature  2025 Dez 2021-2025    3
#> # A tibble: 6 x 5 [1M]
#> # Key:      City, Measure [1]
#> # Groups:   City, Measure [1]
#>   City      Measure      Year_Month Period_Time count
#>   <chr>    <fct>          <moth> <chr>      <dbl>
#> 1 Potsdam Precipitation  1893 Jan 1893-1900   27.7
#> 2 Potsdam Precipitation  1893 Feb 1893-1900   79.6
#> 3 Potsdam Precipitation  1893 Mrz 1893-1900   26.2
#> 4 Potsdam Precipitation  2025 Okt 2021-2025   42.6
#> 5 Potsdam Precipitation  2025 Nov 2021-2025   42.1
#> 6 Potsdam Precipitation  2025 Dez 2021-2025   18.2
```

5.3 Data Sources

5.3.1 Temperatures and Precipitation

- Basel / Davos: **Federal Office of Meteorology and Climatology MeteoSwiss**
 - <https://www.meteoswiss.admin.ch/home/climate/swiss-climate-in-detail/homogeneous-data-series-since-1864.html>
- Cottbus/ Giessen/ Hohenpeissenberg/ Mannheim/ Potsdam: **DWD Archiv Monats- und Tageswerte**
 - <https://www.dwd.de/DE/leistungen/klimadatendeutschland/klarchivtagmonat.html>
 - *Monatswerte historisch und aktuell*
 - File: produkt_klima_monat_xy.txt
 - * column MO_TT (Temperature; Monatsmittel der Lufttemperatur in 2m Höhe in °C and MO_RR (Precipitation; Monatssumme der Niederschlagshoehe in mm))
- England **Met Office - National Meteorological Service for the UK**
 - <https://www.metoffice.gov.uk/hadobs/hadcet/data/download.html>
 - Monthly_HadCET_mean.txt, 1659 to date

5.3.2 CO2 Concentrations

- **National Oceanic & Atmospheric Administration - Earth System Research Laboratory**
 - *NOAA ESRL* <https://www.esrl.noaa.gov/gmd/ccgg/trends/global.html>
 - Data file: *Mauna Loa CO2* monthly mean data
 - <https://www.esrl.noaa.gov/gmd/ccgg/trends/data.html>

5.4 R code

- Source code (maybe not yet the latest version) and output files are stored on GitHub repository <https://github.com/WoVollmer/R-TimesSeriesAnalysis/tree/master/Climate>
- Partially based on *c't Magazin* articles by *Andreas Krause*:
 - #3/2014 p.188 <http://www.ct.de/1403188> & #6/2014 p.180 <http://www.ct.de/1406180>
- *Forecasting: Principles and Practice (3rd ed)* <https://otexts.com/fpp3>
 - Rob J Hyndman and George Athanasopoulos; Monash University, Australia