

Climate change evidence

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1 Data processing

1.1 NASA/GISS data

Analyze global land-ocean data. No need of pre-processing, data already clean.

```
df_temp_anomaly <- read.csv("data/NASA/global_temperature_anomaly.csv",
                             sep = ",", header = TRUE, skip = 2)

# Have a look at first 5 row of the dataset
kable(head(df_temp_anomaly))
```

Year	No_Smoothing	Lowess.5.
1880	-0.16	-0.09
1881	-0.08	-0.12
1882	-0.10	-0.16
1883	-0.17	-0.20
1884	-0.28	-0.23
1885	-0.33	-0.26

1.2 Berkley Earth data

Have a look at different data per continent.

1.2.1 Analysis of one continent file

Do the processing of one single file, i.e. data processing of one continent to have a better idea of what to do

```
# Define columns data frame
cols_month <- c("year", "month", "monthly_anomaly",
               "monthly_uncertainty", "annual_anomaly",
               "annual_uncertainty", "five_year_anomaly",
               "five_year_uncertainty", "ten_year_anomaly",
               "ten_year_uncertainty", "twenty_year_anomaly", "twenty_year_uncertainty")

path <- "data/Berkley Earth/continent_monthly_avg_temp/europe-TAVG-Trend.txt"

# Read data
df_europe_month <- read_delim(path,
                              delim = " ",
                              col_names = cols_month,
                              # Specify all columns as character otherwise error show up
                              col_types = "cccccccccccc",
                              skip = 70,
                              na = c("", "NA", "NaN"))

# Only keep first 4 columns
df_europe_month <- df_europe_month[, 1:4]

# Convert all columns to numeric
df_europe_month <- data.frame(sapply(df_europe_month, as.numeric))

# Get name of zone from the path name
zone_name_path <- unlist(strsplit(path, split = "/"))[4]

# Get name of the zone as a list
zone_name <- head(unlist(strsplit(zone_name_path, split = "-")), -2)

# Collapse name into a single string
zone_name <- paste(zone_name, collapse = '_')

# Add name of the zone as a column to the dataset
df_europe_month$zone <- zone_name

# Create a single date column
df_europe_month$date = as.yearmon(paste(df_europe_month$year, df_europe_month$month), "%Y %m")

# Delete column year and month
df_europe_month <- subset(df_europe_month, select = -c(year, month))

dim(df_europe_month)

## [1] 3252    4

str(df_europe_month)

## 'data.frame':    3252 obs. of  4 variables:
```

```
## $ monthly_anomaly      : num  0.119 0.886 1.855 0.566 -1.065 ...
## $ monthly_uncertainty: num  1.75 2.28 2.59 1.56 1.47 ...
## $ zone                 : chr  "europe" "europe" "europe" "europe" ...
## $ date                 : 'yearmon' num  Jan 1750 Feb 1750 Mar 1750 Apr 1750 ...
```

```
kable(head(df_europe_month))
```

monthly_anomaly	monthly_uncertainty	zone	date
0.119	1.749	europe	Jan 1750
0.886	2.276	europe	Feb 1750
1.855	2.588	europe	Mar 1750
0.566	1.558	europe	Apr 1750
-1.065	1.472	europe	May 1750
-0.099	1.525	europe	Jun 1750

```
kable(summary(df_europe_month))
```

monthly_anomaly	monthly_uncertainty	zone	date
Min. :-6.3520	Min. : 0.057	Length:3252	Min. :1750
1st Qu.: -0.8530	1st Qu.: 0.210	Class :character	1st Qu.:1818
Median :-0.0540	Median : 0.523	Mode :character	Median :1885
Mean :-0.1167	Mean : 1.237	NA	Mean :1885
3rd Qu.: 0.6930	3rd Qu.: 1.814	NA	3rd Qu.:1953
Max. : 5.4490	Max. :10.288	NA	Max. :2021
NA's :1	NA's :1	NA	NA

1.2.2 Aggregate all data

Using a for loop, read all the files per single continent contained in the same folder and apply the necessary cleaning

```
# Get a list with all the files inside the folder zone_monthly_average_temp
files <- list.files(path="data/Berkley Earth/continent_monthly_avg_temp",
  pattern="*.txt",
  full.names=TRUE,
  recursive=FALSE)

# Create empty data frame with correct columns to store all the data
df_continent_avg_temp <- data.frame(zone = character(),
  year = character(),
  anomaly = double(),
  lowess = double(),
  uncertainty = double(),
  stringsAsFactors=FALSE)

# Loop true all the files in the folder and read them
for (file in files){

  # Read and create a temporary dataset
```

```

df_year_temp <- read_delim(file,
  delim = " ",
  col_names = cols_month,
  # Specify all columns as character otherwise error show up
  col_types = "cccccccccccc",
  skip = 70,
  na = c("", "NA", "NaN"))

# Only keep first 4 columns
df_year_temp <- df_year_temp[, 1:4]

# Convert all columns to numeric
df_year_temp <- data.frame(sapply(df_year_temp, as.numeric))

# CALCULATE AVERAGE PER YEAR GIVEN MONTHLY DATA

# Aggregate data per year and calculate mean
df_year_temp <- aggregate(monthly_anomaly~year, transform(df_year_temp, year = df_year_temp$year), mean)

# Change name of the column
colnames(df_year_temp)[2] <- "yearly_anomaly"

# Compute LOWESS smoother which uses locally-weighted polynomial regression
line_trend <- lowess(as.numeric(df_year_temp$year), df_year_temp$yearly_anomaly, f = 1/5)

# Add lowess to year temp dataframe as new column
df_year_temp$lowess <- line_trend$y

# GET ZONE NAME FOR PATH

# Unlist path name
zone_name_path <- unlist(strsplit(file, split = "/"))[4]

# Get name of the zone as a list
zone_name <- head(unlist(strsplit(zone_name_path, split = "-")), -2)

# Collapse name into a single string
zone_name <- paste(zone_name, collapse = '_')

# Add name of the zone as a column to the temporary dataset
df_year_temp$zone <- zone_name

# Update dataset
df_continent_avg_temp <- rbind(df_continent_avg_temp, df_year_temp)
}

```

2 Descriptive Statistics

2.1 NASA/GISS

```
# Get dimension of the database
dim(df_temp_anomaly)
```

```
## [1] 141 3
```

```
# Get information about columns
str(df_temp_anomaly)
```

```
## 'data.frame': 141 obs. of 3 variables:
## $ Year : int 1880 1881 1882 1883 1884 1885 1886 1887 1888 1889 ...
## $ No_Smoothing: num -0.16 -0.08 -0.1 -0.17 -0.28 -0.33 -0.31 -0.36 -0.17 -0.11 ...
## $ Lowess.5. : num -0.09 -0.12 -0.16 -0.2 -0.23 -0.26 -0.27 -0.27 -0.26 -0.25 ...
```

```
# Get summary information about columns
kable(summary(df_temp_anomaly))
```

Year	No_Smoothing	Lowess.5.
Min. :1880	Min. :-0.48000	Min. :-0.41000
1st Qu.:1915	1st Qu.: -0.20000	1st Qu.: -0.22000
Median :1950	Median :-0.07000	Median :-0.04000
Mean :1950	Mean : 0.04858	Mean : 0.04858
3rd Qu.:1985	3rd Qu.: 0.23000	3rd Qu.: 0.22000
Max. :2020	Max. : 1.02000	Max. : 1.01000

```
# Get summary information about columns
kable(summary(df_temp_anomaly))
```

Year	No_Smoothing	Lowess.5.
Min. :1880	Min. :-0.48000	Min. :-0.41000
1st Qu.:1915	1st Qu.: -0.20000	1st Qu.: -0.22000
Median :1950	Median :-0.07000	Median :-0.04000
Mean :1950	Mean : 0.04858	Mean : 0.04858
3rd Qu.:1985	3rd Qu.: 0.23000	3rd Qu.: 0.22000
Max. :2020	Max. : 1.02000	Max. : 1.01000

2.2 Berkley Earth

```
dim(df_continent_avg_temp)
```

```
## [1] 1158 4
```

```
kable(summary(df_continent_avg_temp))
```

year	yearly__anomaly	lowess	zone
Min. :1750	Min. :-1.684333	Min. :-0.769483	Length:1158
1st Qu.:1887	1st Qu.: -0.402314	1st Qu.: -0.369790	Class :character
Median :1935	Median :-0.096292	Median :-0.120499	Mode :character
Mean :1928	Mean :-0.009384	Mean :-0.006192	NA
3rd Qu.:1979	3rd Qu.: 0.337562	3rd Qu.: 0.176794	NA
Max. :2020	Max. : 2.163750	Max. : 1.750654	NA

```
str(df_continent_avg_temp)
```

```
## 'data.frame': 1158 obs. of 4 variables:
## $ year : num 1880 1881 1882 1883 1884 ...
## $ yearly_anomaly: num -0.282 -0.206 -0.487 -0.497 -0.767 ...
## $ lowess : num -0.498 -0.49 -0.481 -0.472 -0.462 ...
## $ zone : chr "africa" "africa" "africa" "africa" ...
```

```
kable(head(df_continent_avg_temp))
```

year	yearly__anomaly	lowess	zone
1880	-0.2818000	-0.4982481	africa
1881	-0.2058000	-0.4901990	africa
1882	-0.4874167	-0.4814832	africa
1883	-0.4966364	-0.4721923	africa
1884	-0.7671429	-0.4624419	africa
1885	-0.4330000	-0.4523444	africa

```
kable(describe(df_continent_avg_temp)[1:7])
```

	vars	n	mean	sd	median	trimmed	mad
year	1	1158	1927.5077720	62.6804564	1935.0000000	1932.4202586	68.1996000
yearly__anomaly	2	1158	-0.0093837	0.6044417	-0.0962917	-0.0401278	0.5353421
lowess	3	1158	-0.0061921	0.4912716	-0.1204991	-0.0676041	0.4048334
zone*	4	1158	4.1908463	1.8243581	4.0000000	4.2381466	1.4826000

```
kable(describe(df_continent_avg_temp)[7:13])
```

	mad	min	max	range	skew	kurtosis	se
year	68.1996000	1750.0000000	2020.0000000	270.0000000	-0.6111347	-0.2285290	1.8419501
yearly__anomaly	0.5353421	-1.6843333	2.163750	3.848083	0.4930823	0.3923491	0.0177623
lowess	0.4048334	-0.7694827	1.750654	2.520136	1.1026295	0.7690255	0.0144367
zone*	1.4826000	1.0000000	7.000000	6.000000	-0.1638085	-0.8302282	0.0536112

2.2.1 Descriptive statistics per continent

Using describeBy() function of psych package.

```
describeBy(df_continent_avg_temp, df_continent_avg_temp$zone)
```

```
##
## Descriptive statistics by group
## group: africa
##      vars    n   mean    sd median trimmed   mad    min    max
## year      1 141 1950.00 40.85 1950.00 1950.00 51.89 1880.00 2020.00
## yearly_anomaly 2 141   0.11  0.49   0.02   0.07  0.41  -0.81   1.44
## lowess      3 141   0.10  0.44  -0.02   0.05  0.44  -0.50   1.15
## zone*      4 141   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##      range skew kurtosis   se
## year      140.00 0.00   -1.23 3.44
## yearly_anomaly 2.25 0.68   -0.21 0.04
## lowess      1.65 0.88   -0.29 0.04
## zone*      0.00 NaN     NaN 0.00
## -----
## group: antarctica
##      vars    n   mean    sd median trimmed   mad    min    max
## year      1 65 1988.00 18.91 1988.00 1988.00 23.72 1956.00 2020.00
## yearly_anomaly 2 65   0.36  0.55   0.41   0.37  0.48  -0.96   1.41
## lowess      3 65   0.35  0.36   0.32   0.36  0.29  -0.39   1.16
## zone*      4 65   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##      range skew kurtosis   se
## year      64.00 0.00   -1.26 2.35
## yearly_anomaly 2.37 -0.12   -0.36 0.07
## lowess      1.55 0.01   -0.57 0.04
## zone*      0.00 NaN     NaN 0.00
## -----
## group: asia
##      vars    n   mean    sd median trimmed   mad    min    max
## year      1 182 1929.50 52.68 1929.50 1929.50 67.46 1839.00 2020.00
## yearly_anomaly 2 182   0.00  0.61  -0.14  -0.06  0.44  -1.55   2.11
## lowess      3 182   0.01  0.55  -0.11  -0.07  0.34  -0.77   1.68
## zone*      4 182   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##      range skew kurtosis   se
## year     181.00 0.00   -1.22 3.91
## yearly_anomaly 3.66 0.92   1.00 0.05
## lowess      2.45 1.35   1.18 0.04
## zone*      0.00 NaN     NaN 0.00
## -----
## group: europe
##      vars    n   mean    sd median trimmed   mad    min    max
## year      1 271 1885.00 78.38 1885.00 1885.00 100.82 1750.00 2020.00
## yearly_anomaly 2 271  -0.12  0.69  -0.15  -0.16  0.63  -1.63   2.16
## lowess      3 271  -0.09  0.51  -0.31  -0.19  0.23  -0.62   1.75
## zone*      4 271   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##      range skew kurtosis   se
## year     270.00 0.00   -1.21 4.76
## yearly_anomaly 3.79 0.56   0.33 0.04
## lowess      2.37 1.77   2.72 0.03
```

```
## zone*          0.00 NaN      NaN 0.00
## -----
## group: north_america
##          vars  n    mean    sd median trimmed  mad    min    max
## year          1 198 1921.50 57.30 1921.50 1921.50 73.39 1823.00 2020.00
## yearly_anomaly 2 198  -0.08  0.65  -0.17  -0.12  0.62  -1.68   1.94
## lowess         3 198  -0.10  0.50  -0.18  -0.17  0.50  -0.66   1.26
## zone*         4 198   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##          range skew kurtosis  se
## year          197.00 0.00   -1.22 4.07
## yearly_anomaly  3.63 0.59    0.25 0.05
## lowess         1.93 1.00    0.26 0.04
## zone*         0.00 NaN      NaN 0.00
## -----
## group: oceania
##          vars  n    mean    sd median trimmed  mad    min    max
## year          1 145 1948.00 42.00 1948.00 1948.00 53.37 1876.00 2020.00
## yearly_anomaly 2 145   0.06  0.48  -0.04   0.03  0.47  -0.88   1.42
## lowess         3 145   0.06  0.39  -0.16   0.00  0.15  -0.28   1.15
## zone*         4 145   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##          range skew kurtosis  se
## year          144.00 0.00   -1.22 3.49
## yearly_anomaly  2.29 0.60   -0.17 0.04
## lowess         1.44 1.19    0.19 0.03
## zone*         0.00 NaN      NaN 0.00
## -----
## group: south_america
##          vars  n    mean    sd median trimmed  mad    min    max
## year          1 156 1942.07 45.91 1942.50 1942.50 57.82 1856.00 2020.00
## yearly_anomaly 2 156  -0.07  0.50  -0.13  -0.09  0.48  -1.60   1.24
## lowess         3 156  -0.08  0.45  -0.15  -0.12  0.40  -0.73   1.05
## zone*         4 156   1.00  0.00   1.00   1.00  0.00   1.00   1.00
##          range skew kurtosis  se
## year          164.00 -0.05   -1.15 3.68
## yearly_anomaly  2.84 0.32   -0.04 0.04
## lowess         1.78 0.73   -0.43 0.04
## zone*         0.00 NaN      NaN 0.00
```

Manual approach.

```
# Name of continents
continents <- c("africa", "europe", "asia", "oceania",
               "north_america", "south_america", "antarctica")

# Create empty vectors
min_year <- c()
min_temp_anom <- c()
max_temp_anom <- c()
mean_temp_anom <- c()
var_temp_anom <- c()

# Loop through all the continents
for (continent in continents){
```



```

# Filter the continent
df_continent <- df_continent_avg_temp[df_continent_avg_temp$zone == continent, ]

# Get descriptive statistics per continent
min_year <- c(min_year, min(df_continent$year))
min_temp_anom <- c(min_temp_anom, min(df_continent$yearly_anomaly))
max_temp_anom <- c(max_temp_anom, max(df_continent$yearly_anomaly))
mean_temp_anom <- c(mean_temp_anom, mean(df_continent$yearly_anomaly))
var_temp_anom <- c(var_temp_anom, var(df_continent$yearly_anomaly))
}

df_summary_continent <- data.frame(continents, min_year, min_temp_anom,
                                   max_temp_anom, mean_temp_anom, var_temp_anom)

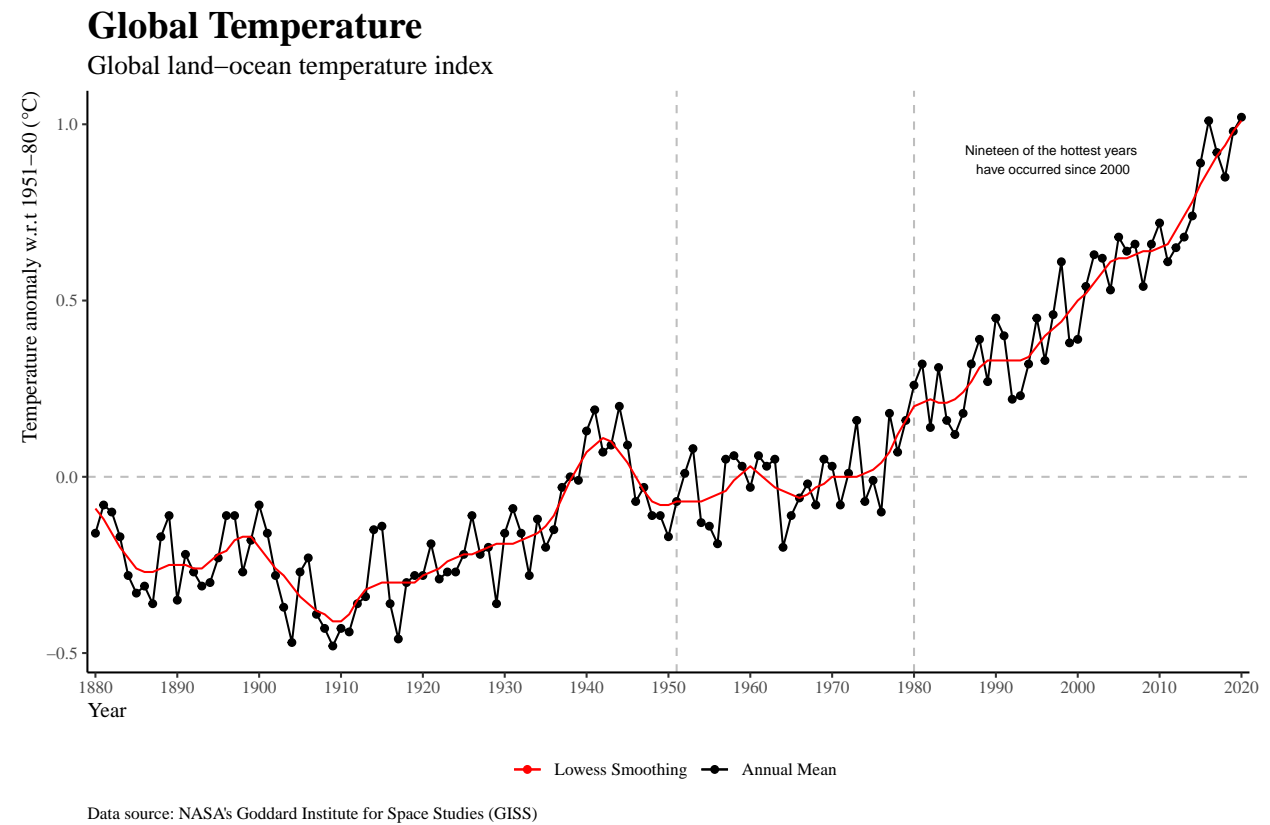
kable(df_summary_continent)

```

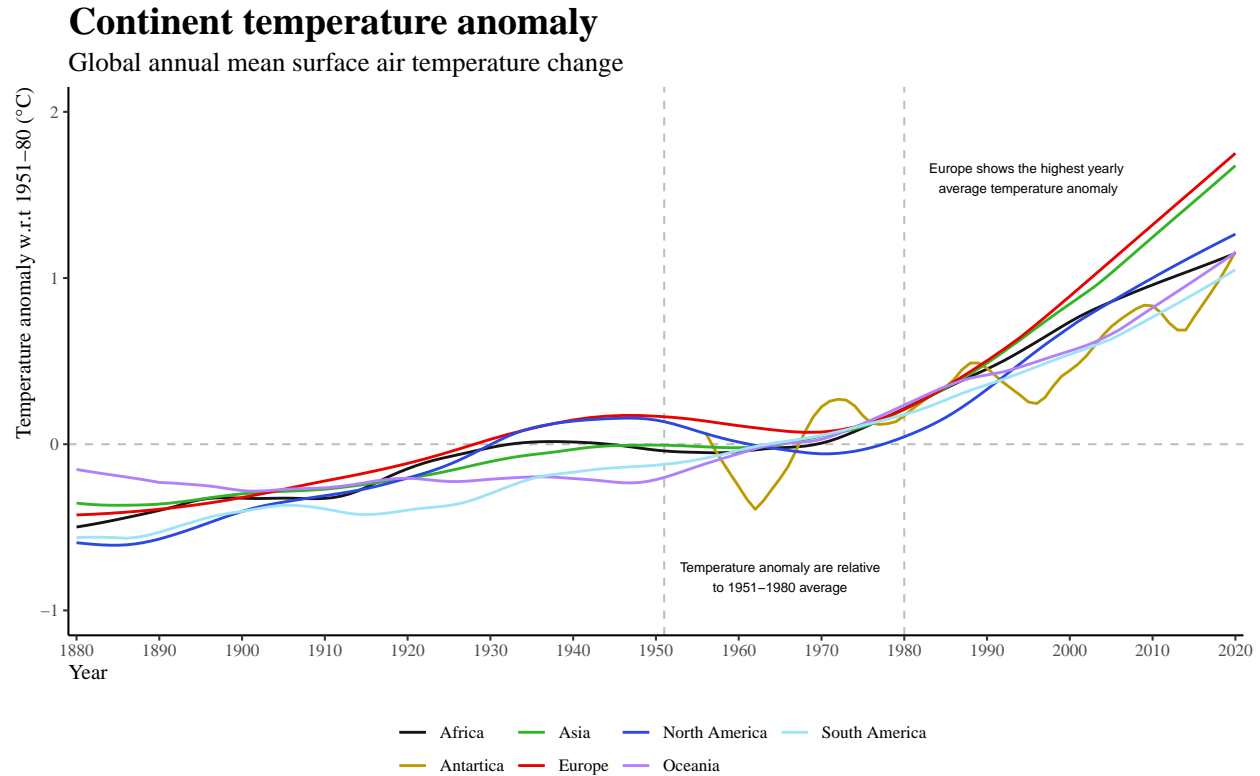
continents	min_year	min_temp_anom	max_temp_anom	mean_temp_anom	var_temp_anom
africa	1880	-0.8090000	1.443167	0.1101983	0.2410926
europa	1750	-1.6291667	2.163750	-0.1166967	0.4793055
asia	1839	-1.5506000	2.111417	0.0031615	0.3752881
oceania	1876	-0.8770833	1.416000	0.0612455	0.2329804
north_america	1823	-1.6843333	1.943500	-0.0828213	0.4250174
south_america	1856	-1.6050000	1.237333	-0.0720385	0.2505224
antarctica	1956	-0.9570000	1.409417	0.3600167	0.2979611

3 Data Visualization

3.1 Global temperature



3.2 Temperature anomaly trend for each continent



Data source: Berkley Earth.

3.3 Temperature records months and regions

```
# Order data by yearly anomaly
df_top_year <- df_continent_avg_temp[order(-df_continent_avg_temp$yearly_anomaly), ]
kable(head(df_top_year, 10))
```

	year	yearly_anomaly	lowess	zone
659	2020	2.163750	1.750654	europe
388	2020	2.111417	1.676391	asia
853	2016	1.943500	1.162198	north_america
658	2019	1.846000	1.707456	europe
653	2014	1.817583	1.492195	europe
654	2015	1.802917	1.535166	europe
657	2018	1.757583	1.664258	europe
847	2010	1.729000	1.000779	north_america
387	2019	1.688917	1.632638	asia
385	2017	1.685833	1.545925	asia

```
df_eu <- df_continent_avg_temp[df_continent_avg_temp$zone == 'europe', ]
write.csv(df_eu, "data/Berkley Earth/df_yearly_anomaly_EU.csv", row.names = TRUE)
```