

# Computing extreme indices for GWLs projections

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from The IPCC Sixth Assessment Report (WG1- *Chapter 1* and *Annex VI*) :

## **Definition of Climatic-impact drivers (CIDs)**

**Climatic impact-drivers (CIDs)** are physical climate system conditions (e.g., means, events, extremes) that affect an element of society or ecosystems. In contrast to the term “hazards”, it provides a more value-neutral characterization of climatic changes that may be relevant for understanding potential impacts, without pre-judging



from The IPCC Sixth Assessment Report (WG1- Chapter 1 and Annex VI) :

## Definition of Climatic-impact drivers (CIDs)

**‘Extremes’** are a category of CID, corresponding to unusual events with respect to the range of observed values of the variable.

➤ **Climatic impact-drivers may not be related only to extremes;**

**‘Climatic impact-driver indices’** are numerically computable indices using one or a combination of climate variables designed to measure the intensity of the climatic impact-driver, or the probability of exceedance of a threshold.

➤ **Indices are, in principle, computable from observations, reanalyses or model simulations,** although it is important to consider scale in comparing across datasets. For example, an extreme precipitation event has a lower magnitude across a large grid cell than it would at a single station within that grid cell.

# A complete set of extreme indices has been identified in the *IPCC Sixth Assessment Report*:

Table AVI.1 | Table listing extreme indices used in Chapter 11.

Extreme	Label	Index Name	Units	Variable
Temperature	TXx	Monthly maximum value of daily maximum temperature	°C	Maximum temperature
	TXn	Monthly minimum value of daily maximum temperature	°C	Maximum temperature
	TNn	Monthly minimum value of daily minimum temperature	°C	Minimum temperature
	TNx	Monthly maximum value of daily minimum temperature	°C	Minimum temperature
	TX90p	Percentage of days when daily maximum temperature is greater than the 90th percentile	%	Maximum temperature
	TX10p	Percentage of days when daily maximum temperature is less than the 10th percentile	%	Maximum temperature
	TN90p	Percentage of days when daily minimum temperature is greater than the 90th percentile	%	Minimum temperature
	TN10p	Percentage of days when daily minimum temperature is less than the 10th percentile	%	Minimum temperature
	ID	Number of icing days: annual count of days when TX (daily maximum temperature) <0°C	Days	Maximum temperature
	FD	Number of frost days: annual count of days when TN (daily minimum temperature) <0°C	Days	Minimum temperature
	WSDI	Warm spell duration index: annual count of days with at least six consecutive days when TX >90th percentile	Days	Maximum temperature
	CSDI	Cold spell duration index: annual count of days with at least six consecutive days when TN <10th percentile	Days	Minimum temperature
	SU	Number of summer days: annual count of days when TX (daily maximum temperature) >25°C	Days	Maximum temperature
	TR	Number of tropical nights: annual count of days when TN (daily minimum temperature) >20°C	Days	Minimum temperature
	DTR	Daily temperature range: monthly mean difference between TX and TN	°C	Maximum and minimum temperature
	GSL	Growing season length: annual (1 Jan to 31 Dec in Northern Hemisphere (NH), 1 July to 30 June in Southern Hemisphere (SH)) count between first span of at least six days with daily mean temperature TG >5°C and first span after July 1 (Jan 1 in SH) of six days with TG <5°C	Days	Mean temperature
	20TXx	One-in-20 year return value of monthly maximum value of daily maximum temperature	°C	Maximum temperature
	20TXn	One-in-20 year return value of monthly minimum value of daily maximum temperature	°C	Maximum temperature
	20TNn	One-in-20 year return value of monthly minimum value of daily minimum temperature	°C	Minimum temperature
	20TNx	One-in-20 year return value of monthly maximum value of daily minimum temperature	°C	Minimum temperature
Precipitation	Rx1day	Maximum one-day precipitation	mm	Precipitation
	Rx5day	Maximum five-day precipitation	mm	Precipitation
	R5mm	Annual count of days when precipitation is greater than or equal to 5 mm	Days	Precipitation
	R10mm	Annual count of days when precipitation is greater than or equal to 10 mm	Days	Precipitation
	R20mm	Annual count of days when precipitation is greater than or equal to 20 mm	Days	Precipitation
	R50mm	Annual count of days when precipitation is greater than or equal to 50 mm	Days	Precipitation
	CDD	Maximum number of consecutive days with less than 1 mm of precipitation per day	Days	Precipitation
	CWD	Maximum number of consecutive days with more than or equal to 1 mm of precipitation per day	Days	Precipitation
	R95p	Annual total precipitation when the daily precipitation exceeds the 95th percentile of the wet-day (>1 mm) precipitation	mm	Precipitation
	R99p	Annual precipitation amount when the daily precipitation exceeds the 99th percentile of the wet-day precipitation	mm	Precipitation
	SDII	Simple precipitation intensity index	mm day <sup>-1</sup>	Precipitation
	20Rx1day	One-in-20 year return value of maximum one-day precipitation	mm day <sup>-1</sup>	Precipitation
	20Rx5day	One-in-20 year return value of maximum five-day precipitation	mm day <sup>-1</sup>	Precipitation
Drought	SPI	Standardized precipitation index	Months	Precipitation
	EDDI	Potential evaporation, evaporative demand drought index	Months	Evaporation
	SMA	Soil moisture anomalies	Months	Soil moisture
	SSEMI	Standardized soil moisture index	Months	Soil moisture
	SRI	Standardized runoff index	Months	Streamflow
	SSI	Standardized streamflow index	Months	Streamflow
	PDSI	Palmer drought severity index	Months	Precipitation, evaporation
	SPEI	Standardized precipitation evapotranspiration index	Months	Precipitation, evaporation, temperature



IPCC, 2021: Annex VI: Climatic Impact-driver and Extreme Indices [Gutiérrez J.M., R. Ranasinghe, A.C. Ruane, R. Vautard (eds.)]. In Climate Change 2021: The Physical Science Basis. Contribution of Working Group I to the Sixth Assessment Report of the Intergovernmental Panel on Climate Change [Masson-Delmotte, V., P. Zhai, A. Pirani, S.L. Connors, C. Péan, S. Berger, N. Caud, Y. Chen, L. Goldfarb, M.I. Gomis, M. Huang, K. Leitzell, E. Lonnoy, J.B.R. Matthews, T.K. Maycock, T. Waterfield, Q. Yelekçi, R. Yu, and B. Zhou (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, pp. 2205–2214, doi:[10.1017/9781009157896.020](https://doi.org/10.1017/9781009157896.020).

Table AVI.2 | Regional CID indices table and relevant references.

CID Category	Climatic Impact-driver (from Table 12.1) and Potential Affected Sectors	Index	Required ECVs	Way to Calculate	Bias Adjustment	References
Heat	Change in cooling demand for energy demand and building consumption	Cooling degree days above 22°C	Tas, tasmin, tasmax	From projections	Yes	Spinoni et al. (2015, 2018)
	Heat, with thresholds important for agriculture	Number of days with Tmax >35°C or 40°C (TX35, TX40)	Tasmax	From projections	Yes	Hatfield and Prueger (2015); Hatfield et al., (2015); Grotjahn (2021)
	Heat stress index combining humidity used in occupational and industrial health	NOAA heat index (HI): number of days above 41°C threshold	Tasmax, huss, ps	From projections	Yes	Burkart et al. (2011); Lin et al. (2012); Kent et al. (2014)
Cold	Heating degree day for energy consumption	Heating degree days below 15.5°C	Tas, tasmin, tasmax	From projections	Yes	Spinoni et al. (2015, 2018)
	Frost	Number of frost days below 0°C (FD)	Tasmin	From projections	Yes	Barlow et al. (2015); Rawlins et al. (2016)
Wet	River flooding	Flood index (FI)	srrdff/mrro	From projections and simplified routing model	No	Forzieri et al. (2016); Alfieri et al. (2017)
Drought	Aridity	Soil moisture (SM)	mrso	From projections	No	Cook et al. (2020)
	Droughts	Standardized Precipitation Index accumulated over 6 months (SPI-6)	Pr	From projections	No	Naumann et al. (2018)
Wind & storm	Mean wind speed	Annual mean wind speed	sfcWind	From projections	No	Karnauskas et al. (2018); Li et al. (2018)
Snow/ice	Snow season length	Number of days with snow water equivalent >100 mm (SWE100) over the snow season (Nov–Mar for NH)	Snw	From projections	No	Damm et al. (2017); Wobus et al. (2017)
Coastal	Extreme sea level (ETWL) inducing storm surges	1-in-100-year return period level (ETWL)		Data from authors	No	Vousdoukas et al. (2018)
	Coastal erosion	Shoreline retreat by mid- and end of century		Data from authors	No	Vousdoukas et al. (2020)

CID categories are identified on the basis of relevance for risks and impacts and available literature. They are classified into seven types: heat and cold, wet and dry, wind, snow and ice, coastal, open ocean, and other





# WP4 – Hazard indices for Europe



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Remove white bloc if not  
needed

# WP4 Objectives:



The aim is to define a stakeholder-relevant set of extreme and hazard indicators for Europe, assessing locally-relevant tipping points and thresholds under different scenarios.

Instead of the classic Time of Emergence, a new approach based on Global Warming Levels (GWLs) has been applied, in order to overcome the dependency from the emissions pathways and the single models.

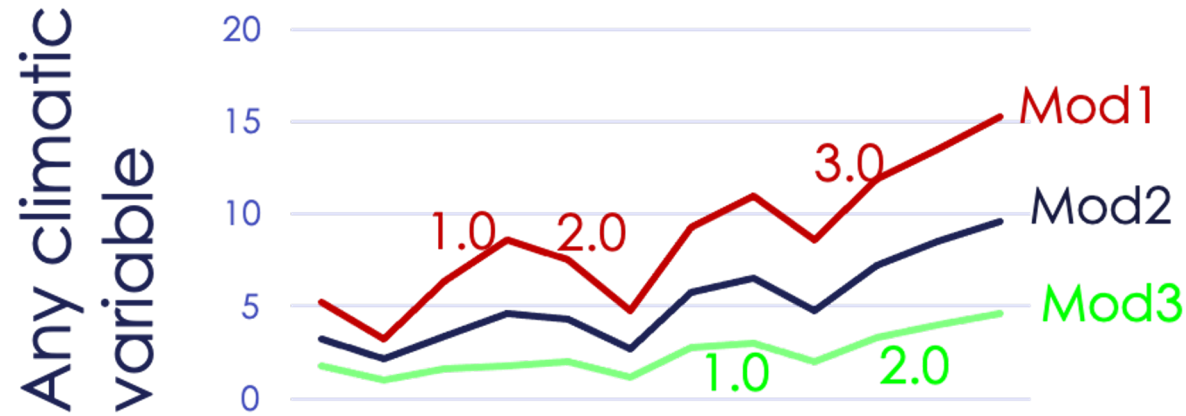
# The method: from *ToE* to *GToE*

- *Time of Emergence (ToE)* is an indicator of the magnitude of the climate change signal relative to the background variability.
- There is no single metric for *ToE*. It depends on user-driven choices of variables, space and time scales, the baseline relative to which changes are measured, and the threshold at which emergence is defined (Kirtman et al. 2013).
- For OBSERVATIONS, the *noise* is the interannual variability;
- when dealing with an ENSEMBLE OF MODELS, the inter-models' spread should be considered too;
- moreover, GCMs can refer to the pre-industrial period as the reference; RCMs should consider a more recent period since there is no availability of data before 1970.
- >> TOE assessment is affected by methodological choices!



# The method: from *ToE* to *GToE*

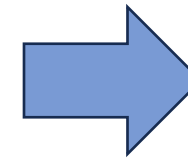
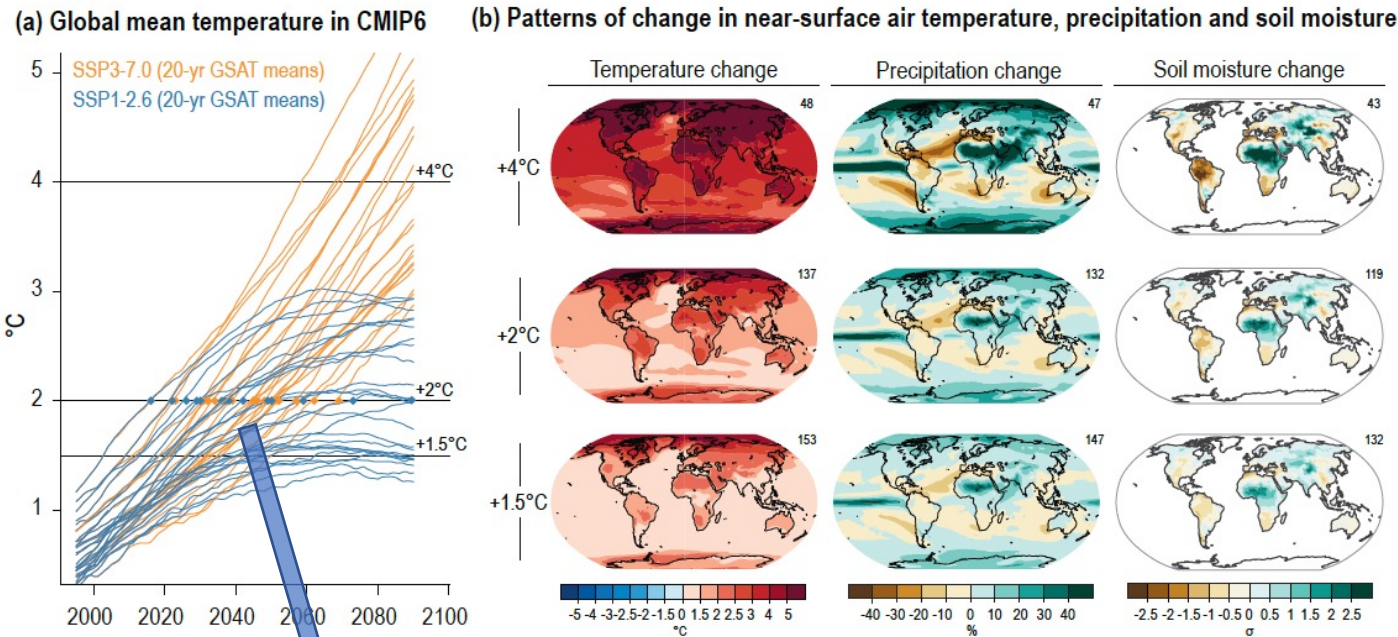
- The **Global Temperature of Emergence (GToE)** is an alternative to the *ToE* concept: by replacing time with global mean temperature, it makes no difference among models and emission pathways and it is more directly aligned with climate policy targets!



The GToE is thus defined on the basis of thresholds of temperature, the **Global Warming Levels (GWLs)**, expressed as changes in surface global temperature relative to the period 1850-1900.



The approach based on **Global Warming Levels (GWLs)** is applied, in order to overcome the dependency from the emissions pathways and the single models.



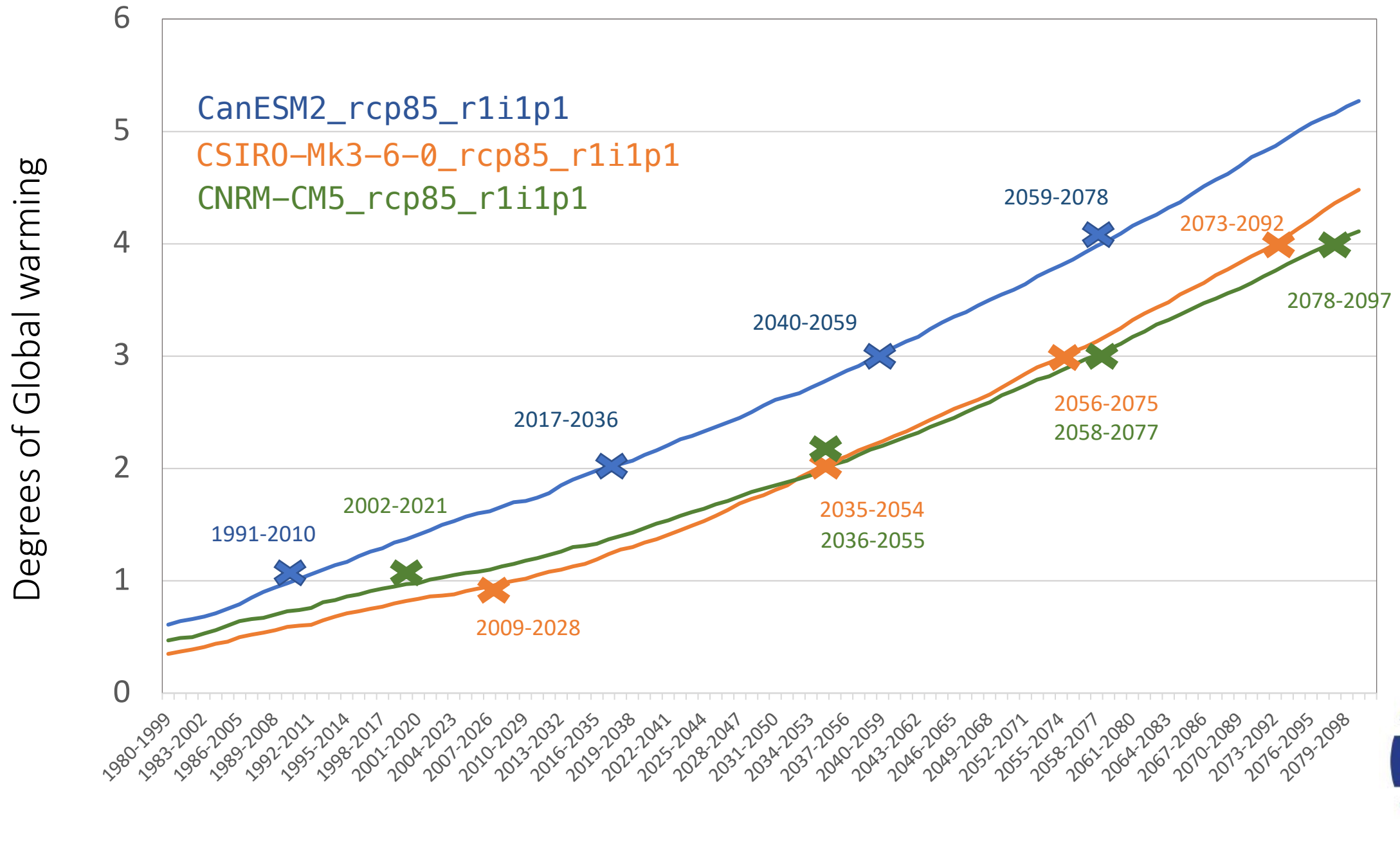
The climate response pattern for the 20-year period around when individual simulations reach a given GWL are averaged across all models and scenarios that reach that GWL.

**Figure TS.5 | Scenarios, global warming levels, and patterns of change.** The intent of this figure is to show how scenarios are linked to global warming levels (GWLs) and to provide examples of the evolution of patterns of change with global warming levels. (a) Illustrative example of GWLs defined as global surface temperature response to anthropogenic emissions in unconstrained Coupled Model Intercomparison Project Phase 6 (CMIP6) simulations, for two illustrative scenarios (SSP1-2.6 and SSP3-7.0). The time when a given simulation reaches a GWL, for example, +2°C, relative to 1850–1900 is taken as the time when the central year of a 20-year running mean first reaches that level of warming. See the dots for +2°C, and how not all simulations reach all levels of warming. The assessment of the timing when a GWL is reached takes into account additional lines of evidence and is discussed in Cross-Section Box TS.1. (b) Multi-model, multi-simulation average response patterns of change in near-surface air temperature, precipitation (expressed as percentage change) and soil moisture (expressed in standard deviations of interannual variability) for three GWLs. The number to the top right of the panels shows the number of model simulations averaged across including all models that reach the corresponding GWL in any of the five Shared Socio-economic Pathways (SSPs). See Section TS.2 for discussion. {Cross-Chapter Box 11.1}

The time when a given simulation reaches a GWL, for example, +2C, relative to 1850–1900 is taken as the time when the central year of a 20-year running mean first reaches that level of warming. See the dots for +2C, and how not all simulations reach all levels of warming.



## An example of GWLs calculation:



Hazard indices	Description/Calculation	Critical sector of exposure	References	Demonstrator(s)
TNnTrop Annual	Number of tropical nights. Minimum night-time temperature > 20 °C Annual frequency, and the maximum and average period length for periods of 3 or more days in length	Heatwaves, Human health,		Barcelona, Bergen, Prague
TNnEqua Annual	Number of equatorial nights. Minimum night-time temperature > 25 °C Annual frequency, and the maximum and average period length for periods of 3 or more days in length	Heatwaves, Human health,		Barcelona, Bergen, Prague
Tx25 Annual	Number of Summer Days with maximum daily temperature above 25 degrees Annual frequency, and the maximum and average	Human health, infrastructure, ecosystems, agriculture	Deryng et al. 2014 Petitti et al. 2016	Paris, Prague, Barcelona

	period length for periods of 3 or more days in length			
HW Annual	Heat wave as 3 consecutive days with TX above the daily threshold (p90 of TXd over ± 15 days) for the reference period 1981-2010. Annual frequency, and the maximum and average period length.	Human health, ecosystems	Russo et al., 2014; 2015	Barcelona, Prague, Paris
HEvent Annual	Heat event as 1-2 consecutive days with TX above the daily threshold (p90 of TXd over ± 15 days) for the reference period 1981-2010.	Human health, ecosystems	Russo et al., 2014; 2015	Barcelona, Prague
EHF (Excess Heat Factor)	Quantifies heat wave severity/intensity based on three-day-averaged daily mean temperature (DMT). $EHI_{avg} = (T_1 + T_{i+1} + T_{i+2})/3 - EHI_{med} = (T_1 + T_{i+1} + T_{i+2})/3 - (T_{i-1} + \dots + T_{i+3})/5$ $EHF = EHI_{avg} \times \max(1, EHI_{med})$	Heatwaves, Human Health	Nairn & Fawcett (2014)	Barcelona, Prague, Paris

NOAA Heat index (HI)	The 'Heat Index' is a measure of how the hot weather "feels" to the body. The index uses relative humidity and air temperature to produce the "apparent temperature" or the temperature the body feels <sup>4</sup> $HI = \frac{0.5 \times H_{rel} + 0.5 \times T_a}{0.5 \times H_{rel} + 0.5 \times T_a} \times 1.8 \times (T_a - 58) + 58$ with: $H_{rel} = \frac{1}{1 + \frac{0.01546717}{T_a} + \frac{0.00007474}{T_a^2} + \frac{0.00000074}{T_a^3}}$ $T_a = (33 - 85)/4 + \sqrt{(33 - 85)^2 - 35 \times 95/17}$ $H_{rel} = (0.81 - 85/10) \times (0.7 \times T_a - 75)/5$ $c_1 = 42.379 \times T_a - 2.0002125, c_2 = 32.4533237 \times T_a$ $c_3 = 0.00003763 \times T_a^2, c_4 = 0.0481717 \times T_a - 0.0012$ $c_5 = 0.00000199 \times T_a^3$ $T_a = 80^\circ F$ , the following equation is used: $HI = 0.5 \times (T_a + 45 \times T_a + 1.2 \times (T_a - 58)^2) + 0.294 \times T_a$ The calculated HI is converted into °C.	Human health, ecosystems	Annex VI: Climatic Impact-driver and Extreme Indices (IPCC Sixth Assessment Report – WG1); Burkart et al. (2011); Lin et al. (2012); Kent et al. (2014), Lu and Romps (2022)	Barcelona, Paris, Prague
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CDD Annual	Given a threshold Tb=22°C: $CDD_i = \begin{cases} 0 & \text{if } T_i - T_b \leq 0 \\ \frac{T_i - T_b}{T_i - T_b} - \frac{T_i - T_b}{T_i - T_b} & \text{if } T_i - T_b > 0 \end{cases}$ Then: $CDD = \sum_{i=1}^{365} CDD_i$	Energy consumption for cooling	Spinoni et al. 2015 <a href="https://ec.europa.eu/eurostat/cache/metadata/en/nrg_chdd_esms.htm">https://ec.europa.eu/eurostat/cache/metadata/en/nrg_chdd_esms.htm</a>	Barcelona, Paris, Prague, Bergen, Newcastle, Hamburg
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HDD Annual	Given a threshold Tb=15°C: $HDD_i = \begin{cases} \frac{T_i - T_b}{T_i - T_b} - \frac{T_i - T_b}{T_i - T_b} & \text{if } T_i - T_b > 0 \\ 0 & \text{if } T_i - T_b \leq 0 \end{cases}$ Then: $HDD = \sum_{i=1}^{365} HDD_i$	Energy demand for heating	Spinoni et al. 2015	Paris, Prague, Barcelona, Bergen, Newcastle, Hamburg
PrRnn	nn year return value of daily precipitation, e.g., nn= 2, 5, 10, 20, 50, 100, 200 years	Pluvial flooding, ecosystems and crop growth		Bergen, Prague
RX1day Annual	Maximum 1 day precipitation	Pluvial flooding, ecosystems and crop growth		Bergen, Newcastle, Barcelona, Prague
CWD Annual	CWD: consecutive wet day /maximum length of wet spell (RR ≥ 1 mm)	Water resources management, Tourists, Flooding, Ecosystems	Climate Indices and Analysis for Sectoral Application, WMO: <a href="https://indico.ictp.it/event/a10167/session/16/contribution/12/material/0/0.pdf">https://indico.ictp.it/event/a10167/session/16/contribution/12/material/0/0.pdf</a>	Bergen, Prague

Rnnmm Annual	Rnnmm: count of days where RR ≥ user-defined threshold in mm. For example, precipitation based on percentile value.	Water resources management, Tourists, Flooding, Ecosystems	Climate Indices and Analysis for Sectoral Application, WMO: <a href="https://indico.ictp.it/event/a10167/session/16/contribution/12/material/0/0.pdf">https://indico.ictp.it/event/a10167/session/16/contribution/12/material/0/0.pdf</a>	Bergen, Prague
SPI 6 months	Standardised Precipitation Index		WMO (2012)	Barcelona
NDD	Number, average length, and maximum length of dry spells. A dry spell is a period during which the daily precipitation is below a threshold of 1 mm.	Heatwaves, Droughts, Eco systems, Crop growth	Cindric et al. (2010), Manning et al., (2023)	Barcelona, Bergen, Prague, Paris
DF Decadal	Drought frequency based on a 6-month SPI	Ecosystems and agriculture	Spinoni et al. 2014	Barcelona,Prague
Fire Weather Index (FWI)	Fire Weather Index/ a meteorologically based index used worldwide to estimate fire danger. It is calculated using 24-h accumulated precipitation, instantaneous wind speed, relative humidity and daily max temperature.	Human health, ecosystems, infrastructure	Bedia et al., (2013), Bedia et al., (2018) Alternative is Fire Occurrence Probability Index (FOPi) which also considers fuel. <a href="https://www.giuseppe.it/2023">Giuseppe et al 2023</a>	Barcelona, Bergen, Prague

RxHhr	Simple monthly maxima of H-hour (1-hour and 3-hour) precipitation series	Flooding	Pritchard, D.M.W., Lewis, E., Blenkinsop, S., Patino Velasquez, L., Whitford,	Newcastle, Bergen
	(using a sliding window to identify maxima when H > 1)		A., Fowler, H.J. GSDR-I: An Observation-Based Dataset of Global Sub-Daily Precipitation Indices. Scientific Data, in press.	
RHhrTm m	Count of hours with greater than 30 or 50 mm thresholds (T)	Flooding	Pritchard, D.M.W., Lewis, E., Blenkinsop, S., Patino Velasquez, L., Whitford, A., Fowler, H.J. GSDR-I: An Observation-Based Dataset of Global Sub-Daily Precipitation Indices. Scientific Data, in press.	Newcastle, Bergen

Indices for specific Demonstrators only:

Hazard indices	Description/Calculation	Critical sector of exposure	References	Demonstrator(s)
Q100 Annual	(1) Daily discharge for each climate experiment are produced for a 130 year period (e.g. 1970-2100) (2) Annual maximum river discharge were selected and a Gumbel distribution was fitted on time slices of 30 years From the distribution, the peak corresponding to the 100 year return period is calculated	Flooding, infrastructure	Alfieri et al. 2015a, b Forzieri et al. 2016a, b  Already calculated by ICTP – Extract for individual demonstrator cities	Bergen, Prague, Barcelona
Météo France Index	A heat wave occurs when the mean temperature is greater than 25.3 °C and at least > 23,4 °C for 3 consecutive days and always above 22,4 °C	Human health, ecosystems	<a href="https://meteofrance.com/">https://meteofrance.com/</a>	Paris ONLY
H-ASI	Stagnant days defined as days on which three conditions are met simultaneously: Daily precipitation accumulation < 1mm/day, Daily mean (10m) near-surface wind speed < 3.2m/s, Daily mean 500hPa wind speed < 13m/s.	Air quality	Horton, D. E., Skinner, C. B., Singh, D., & Di enbaugh, N. S. (2014). Occurrence and persistence of future atmospheric stagnation events. Nature climate change, 4(8), 698-703.	Prague ONLY
Meteo Cat Number of heat events and	MeteoCat definition of temperature threshold is 98th percentile compared to last decade of June-August average max temperatures (calculated for June-August	Heatwaves, Human health		Barcelona ONLY

heat waves	2012-2021 period). A heatwave is a period of 3 or more consecutive days above this threshold. A heat event is 1 - 2 consecutive days above this level.			
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# Selection of hazard indicators

# Hazard Indicators: Definitions

- ❖ **TX25**: N. Summer days
- ❖ **TN20**: N. Tropical nights
- ❖ **RX1DAY**: highest 1-day precip
- ❖ **NDD**: N. dry days
- ❖ **CDDmax/mean**: consecutive dry days
- ❖ **CWDmax**: consecutive wet days
- ❖ **HI41**: Heat Index
- ❖ **CoolDD & HeatDD**: Cooling and Heating Degree days
- ❖ **DF**: Drought Frequency
- ❖ **FWI** : Fire weather Index



# Hazard Indicators: Definitions

**CDD: Cooling degree days**, a measure of the energy consumption for cooling in hot environments. It is based on the daily mean, maximum and minimum temperature and it is computed as in Spinoni et al. (2015), except that here the sum is cumulated over the whole year (instead of 6 months) so that it applies to both Hemispheres.

Input: daily min.  
max and mean  
temperature!



# Hazard Indicators: Definitions

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and  
here the sum is cumulated over the whole year (instead of 6 months) so

$$CDD_i = \begin{cases} 0 \\ \frac{T_X - T_b}{4} \\ \frac{T_X - T_b}{2} - \frac{T_b - T_N}{4} \\ T_M - T_b \end{cases} \quad \text{if} \quad \begin{cases} T_X \leq T_b \\ T_M \leq T_b < T_X \\ T_N \leq T_b < T_M \\ T_N \geq T_b \end{cases}$$

Given a threshold  $T_b = 22^\circ\text{C}$

heres.

Then:

$$CDD = \sum_{i=1}^{365} CDD_i$$

Input: daily min.  
max and mean  
temperature!



# Hazard Indicators: Definitions

**HDD: Heating Degree Days:** similarly to the CDD, it is the energy demand for heating and it is computed as in Spinoni et al. (2015), but for the whole year.

Input: daily min.  
max and mean  
temperature!



# Hazard Indicators: Definitions

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$$HDD_i = \begin{cases} \frac{T_b - T_M}{2} - \frac{T_X - T_b}{4} & \text{if } \begin{cases} T_X \leq T_b \\ T_M \leq T_b < T_X \\ T_N \leq T_b < T_M \\ T_N \geq T_b \end{cases} \\ \frac{T_b - T_N}{4} \\ 0 \end{cases}$$

Given a threshold  $T_b = 15.5^\circ\text{C}$

Input: daily min.  
max and mean  
temperature!

Then:

$$HDD = \sum_{i=1}^{365} HDD_i$$



# Hazard Indicators: Definitions

**SPI-6:** the Standardized Precipitation Index (for a time window of 6 months): a drought starts in the month when SPI-6 falls below  $-1$  and it ends when SPI-6 returns to positive values for at least two consecutive months, as in Spinoni et al. (2014).

Input: monthly precipitation!



**Drought Frequency (DF):** the total number of drought events in a decade





# Hazard Indicators: Definitions

**Heat Index (HI41- NOAA):** it is a widely used measure of apparent temperature that accounts for the effects of humidity using Steadman's model of human thermoregulation.

Input:  
DAILY Hurs,  
tasmax;  
or:  
HOURLY hurs  
and tas!

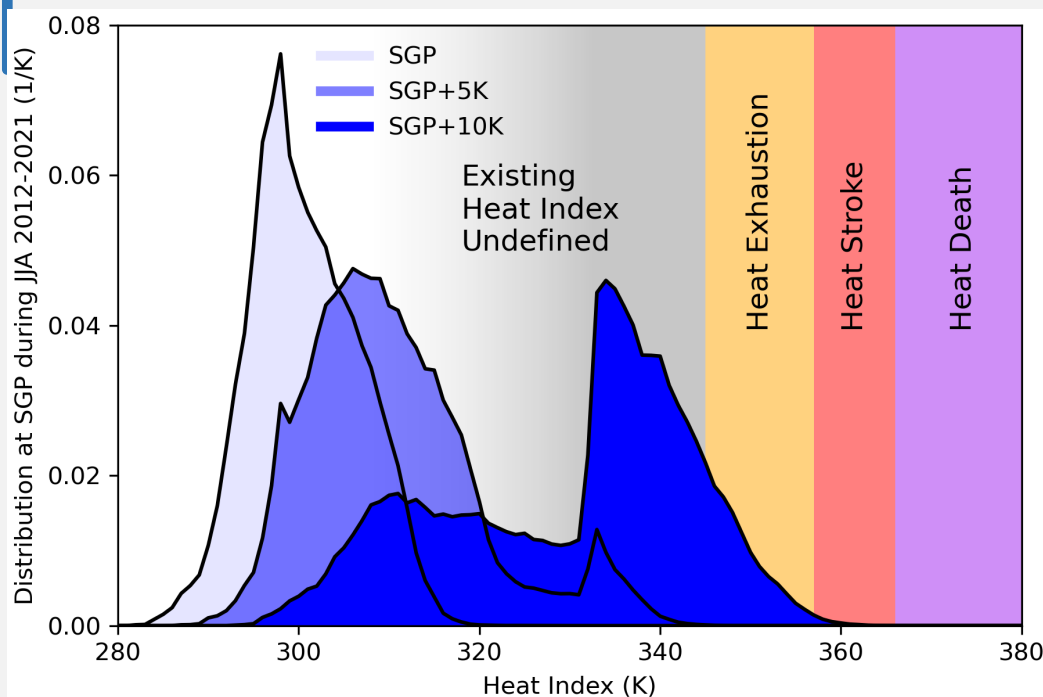
➔ It is well-defined for most combinations of high temperature and humidity experienced on Earth in the preindustrial climate, **but global warming** is increasingly generating conditions for which the heat index is undefined!



# Hazard Indicators: Definitions

**Extended Heat Index (HI41):** here, Steadman's thermoregulation model is extended to define the heat index for all combinations of temperature and humidity, allowing for an assessment of Earth's future habitability. (Lu and Romps, 2022)

Input:  
DAILY Hurs,  
tasmax;  
or:  
HOURLY hurs  
and tas!



To illustrate the change in the heat index under different warming scenarios, two other distributions are plotted by adding either 5 (blue) or 10 (dark blue) K to the SGP summer-time temperatures while keeping the relative humidities fixed. The change in the heat index is not a simple translation as the distribution in temperature–humidity space but has a nonlinear response to warming.

# Hazard Indicators: Definitions

**Extended Heat Index (HI41):** here, Steadman's thermoregulation model is extended to define the heat index for all combinations of temperature and humidity, allowing for an assessment of Earth's future habitability. (Lu and Romps, 2022)

Input:  
DAILY Hurs,  
tasmax;  
or:  
HOURLY hurs  
and tas!

Here you can read on the comparison between the “standard” Heat Index and the “extended” one:

❖ <https://iopscience.iop.org/article/10.1088/1748-9326/ac8945/pdf>

Here there are fortran and python scripts to compute the HI:

❖ <https://romps.berkeley.edu/papers/pubs-2020-heatindex.html>



GWL1.0

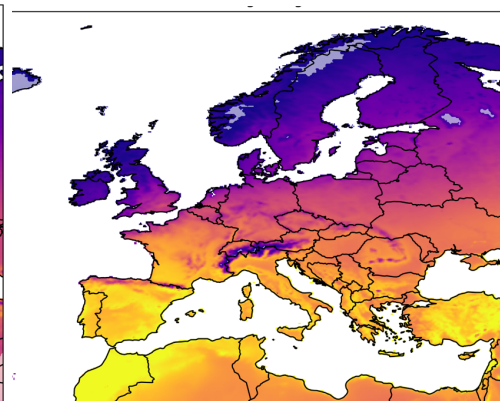
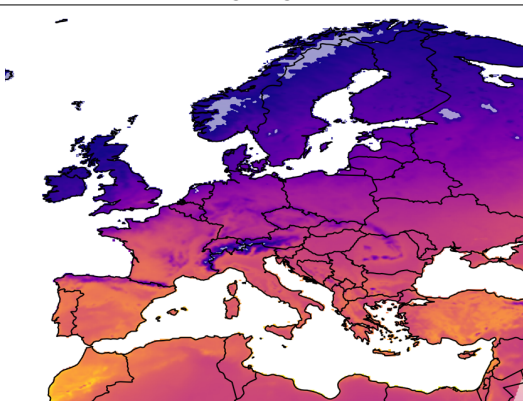
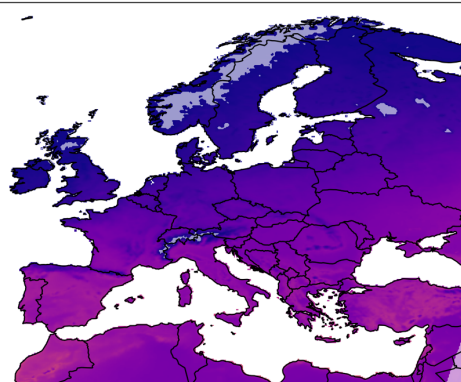
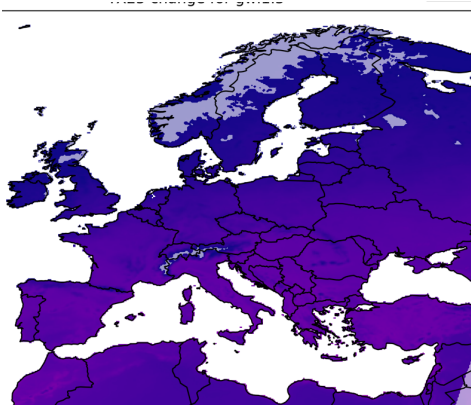
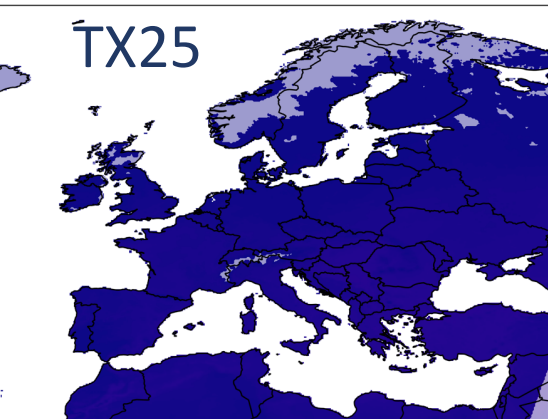
GWL1.5

GWL2.0

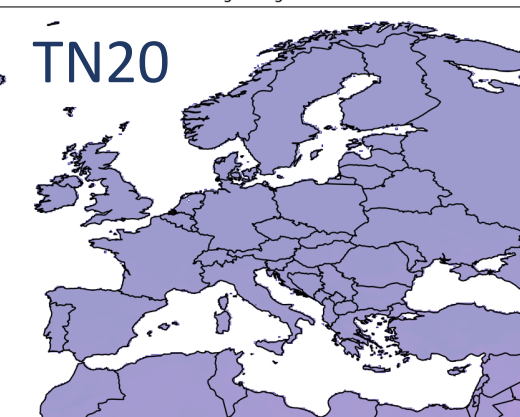
GWL3.0

GWL4.0

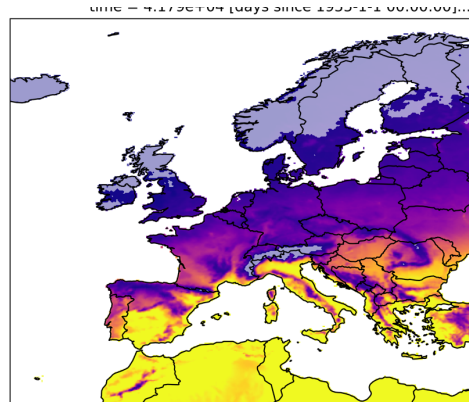
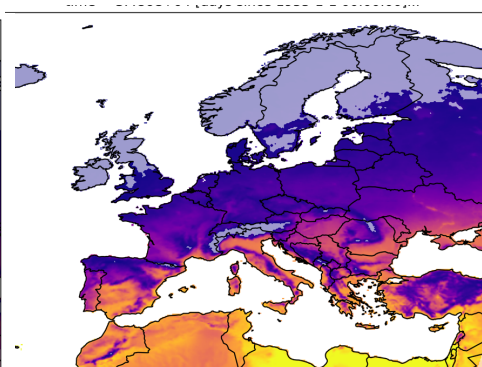
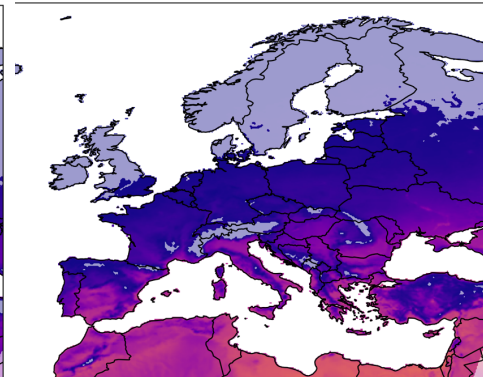
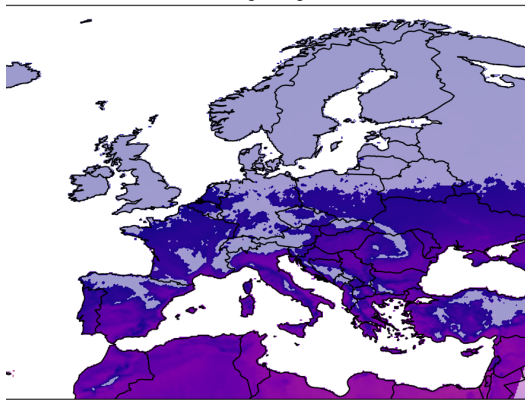
TX25



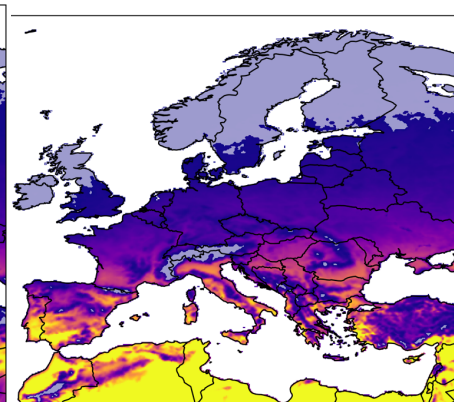
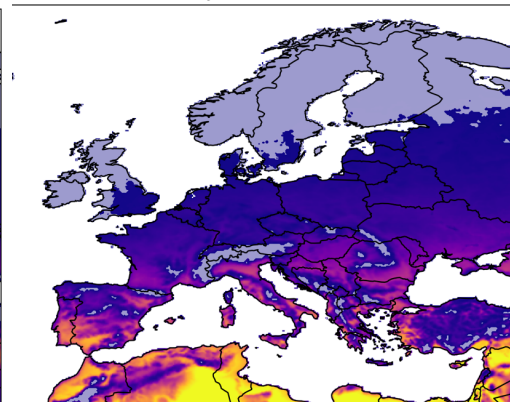
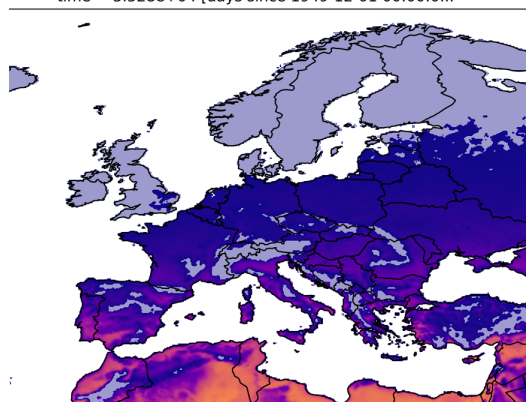
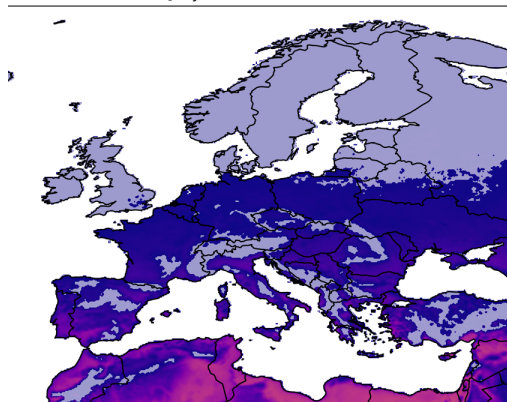
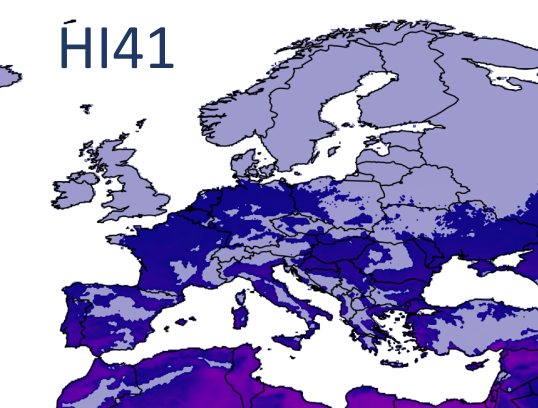
TN20 change for gw1.0



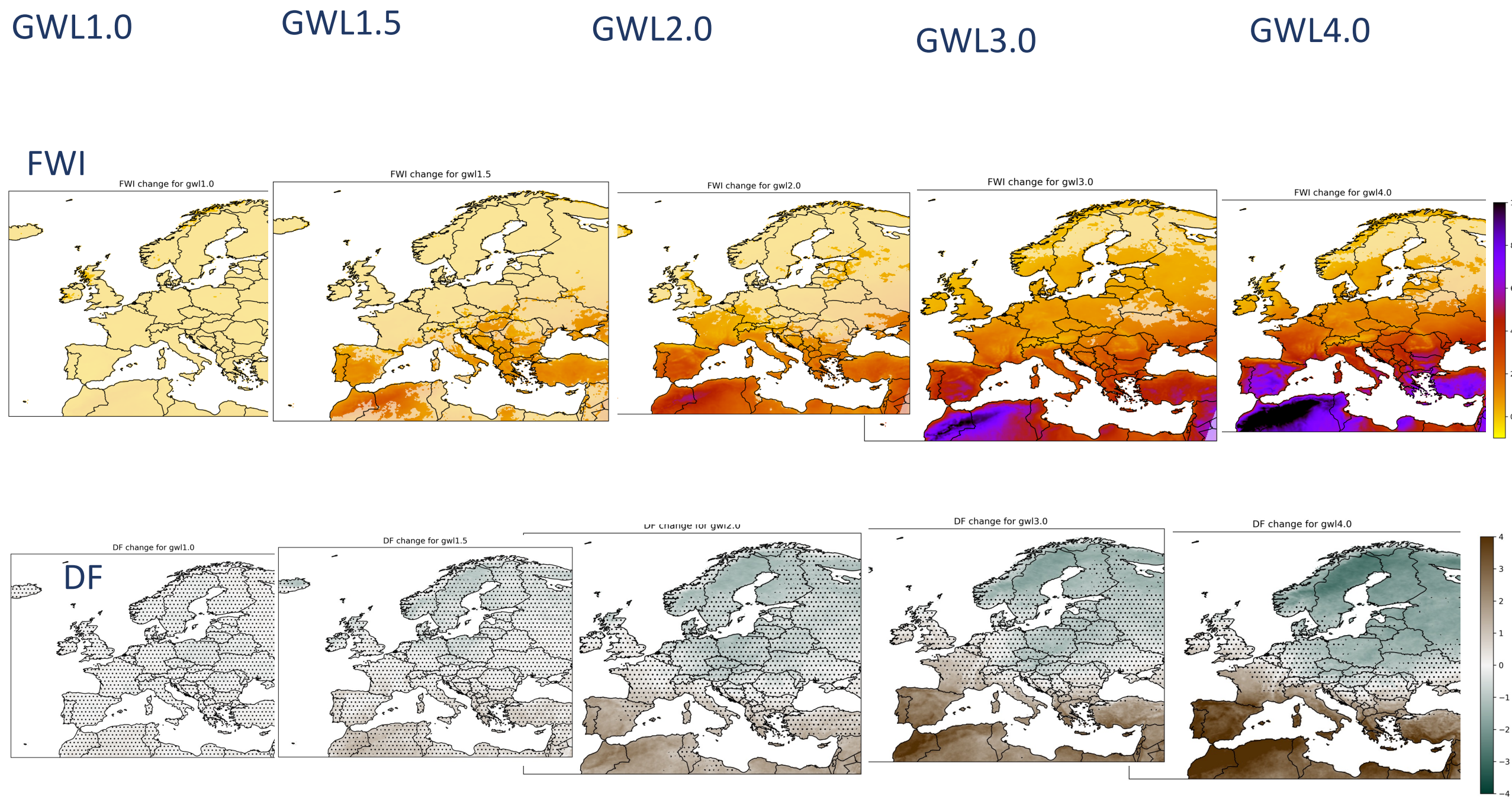
TN20 change for gw1.5



HI41









GWL1.0

GWL1.5

GWL2.0

GWL3.0

GWL4.0

CWDmax change for gwl1.0

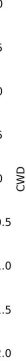
CWD

CWDmax change for gwl1.5

CWDmax change for gwl2.0

CWDmax change for gwl3.0

CWDmax change for gwl4.0



RX1DAY change for gwl1.0

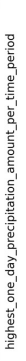
RX1DAY

RX1DAY change for gwl1.5

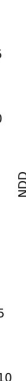
RX1DAY change for gwl2.0

RX1DAY change for gwl3.0

RX1DAY change for gwl4.0



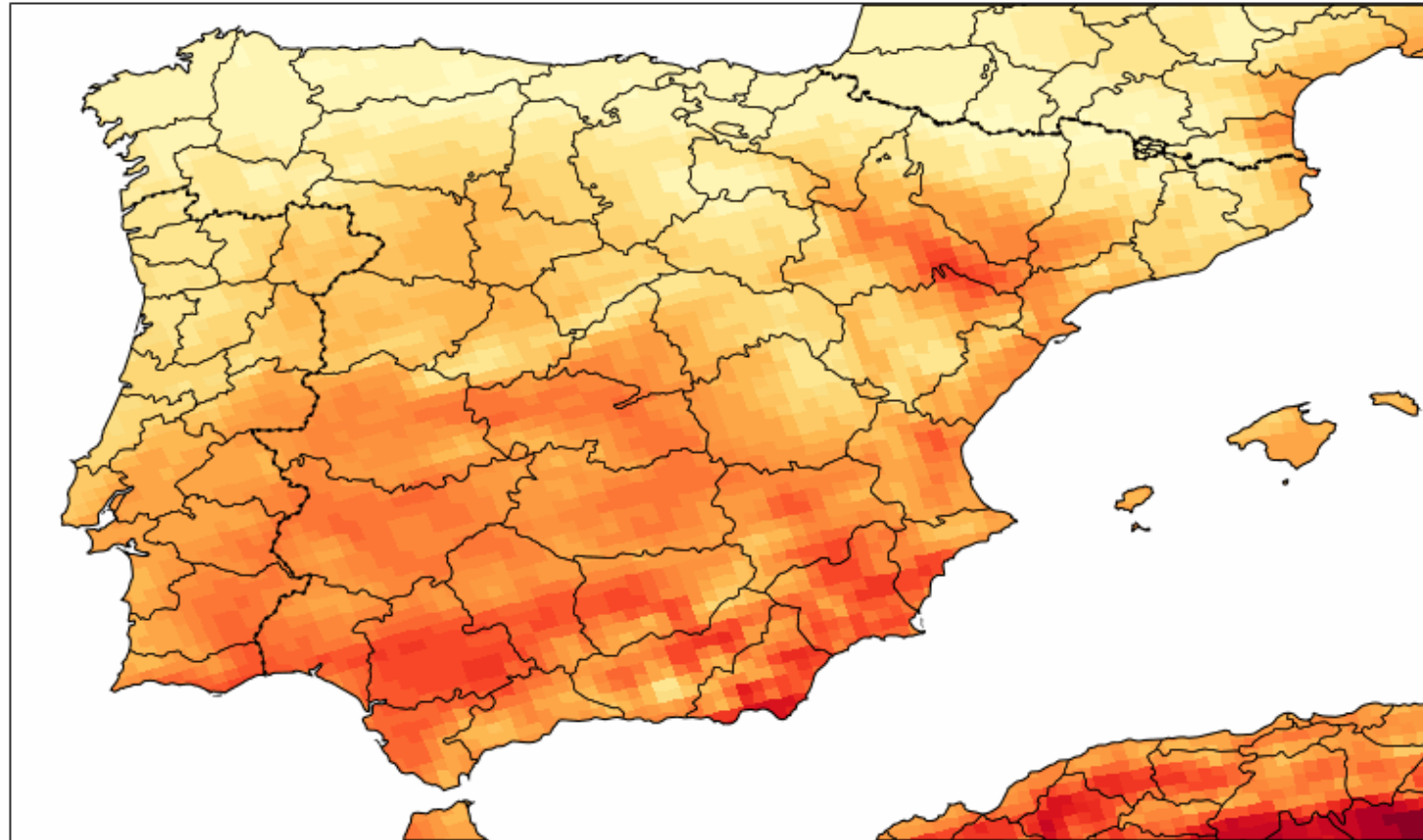
NDD



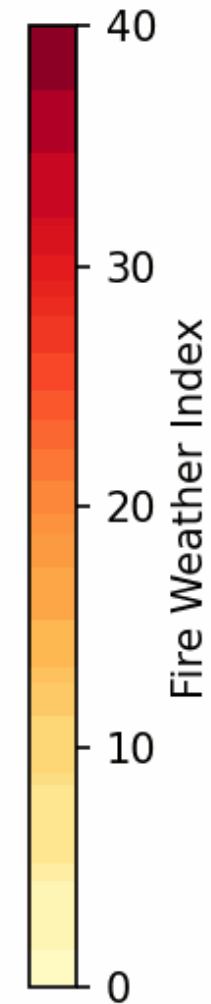


# Fire Weather Index - 2020

IMPETUS  
4CHANGE  
TECHNICAL



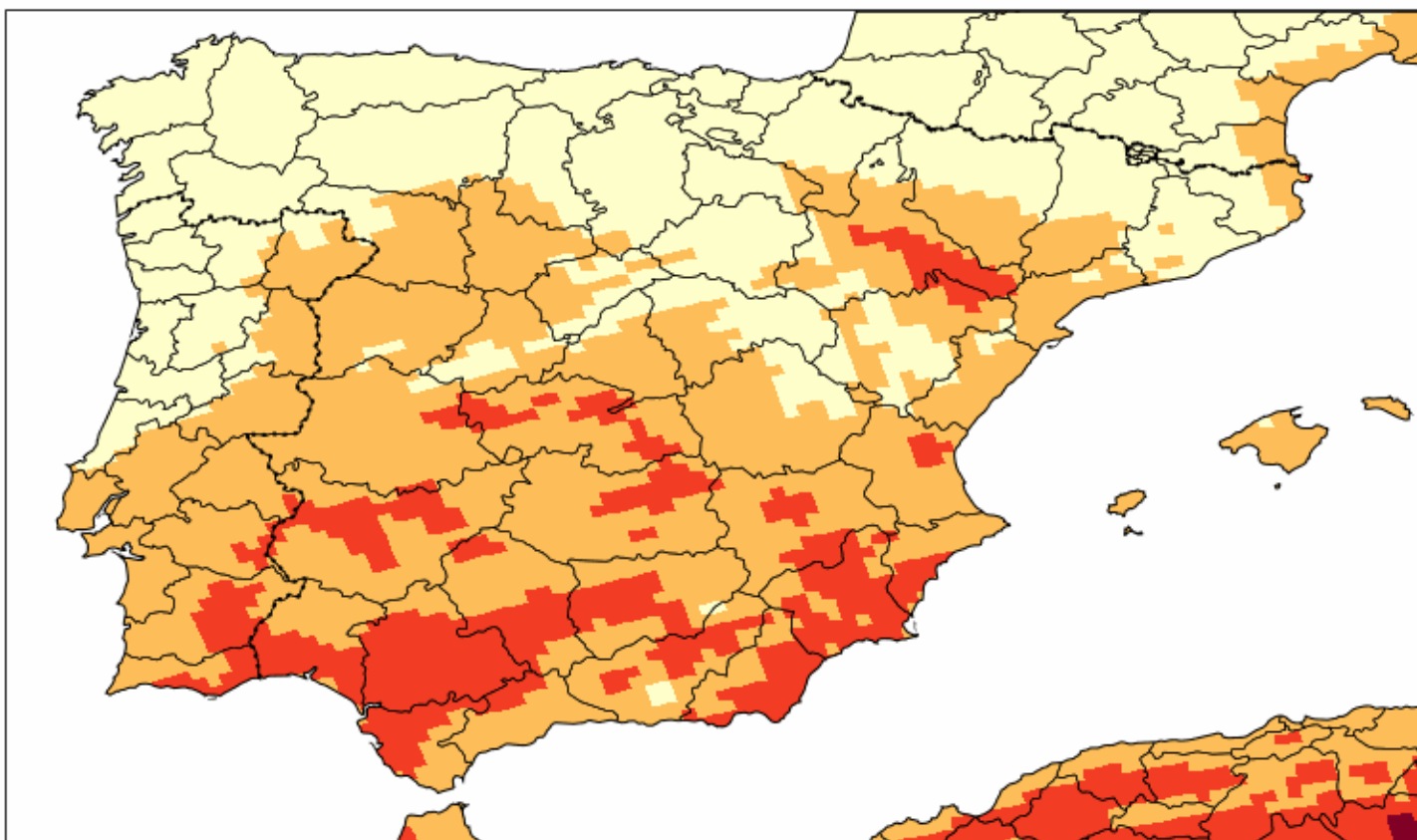
I4C EURO-CORDEX Hazard Indices - 26 models



Input:  
DAILY hurs,  
tasmax, sfwind,  
precipitation!

# Annual Fire Weather Index - 2020

IMPETUS  
4 CHANGE



I4C EURO-CORDEX Hazard Indices - 26 models

FWI category	FWI values
LOW	0 - 11.2
MODERATE	11.2 - 21.3
HIGH	21.3 - 38
VERY HIGH	38 - 50
EXTREME	50 - 70
VERY EXTREME	$\geq 70$

Table 4. The Fire danger classes according to EFFIS

**Very high**

**High**

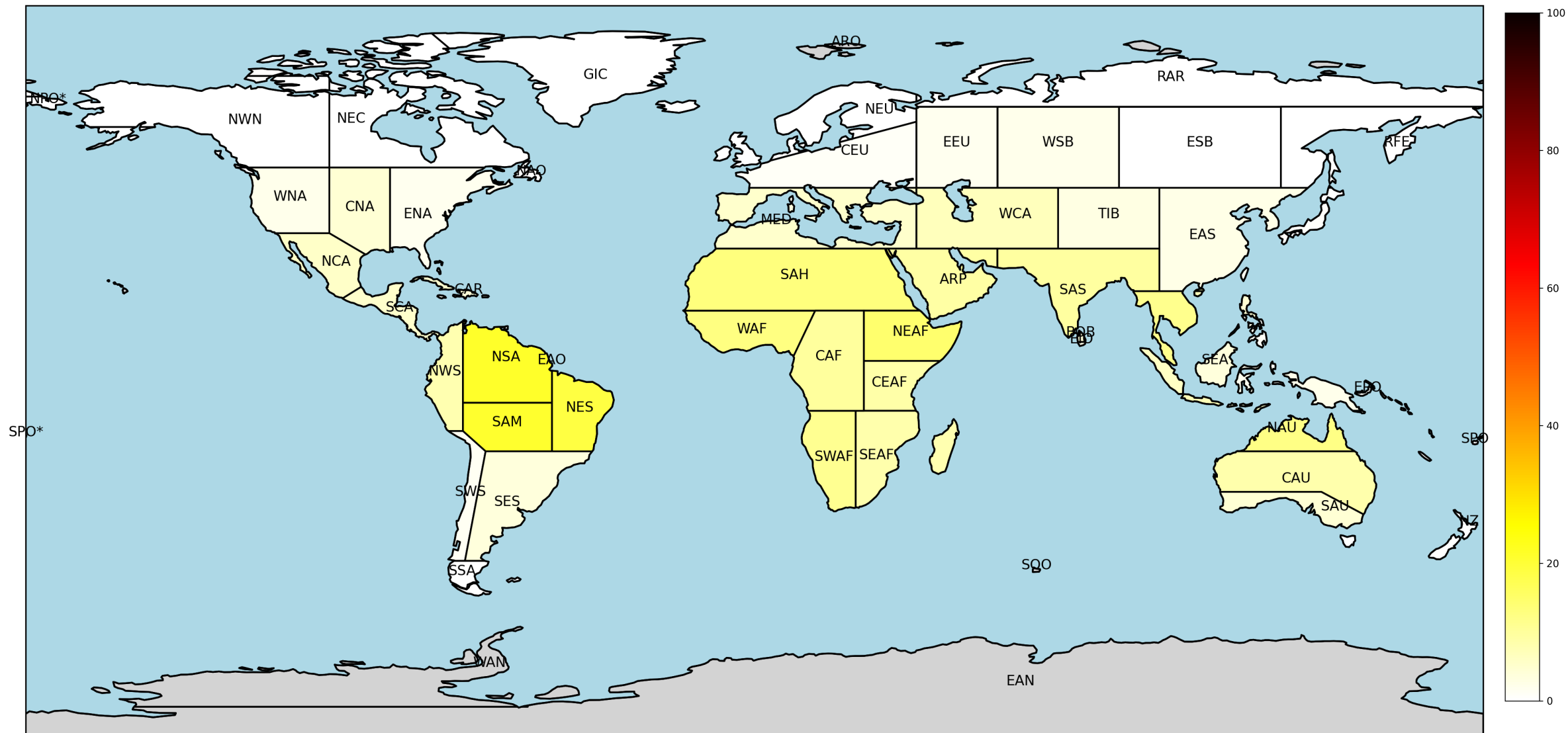
**Moderate**

**Low**

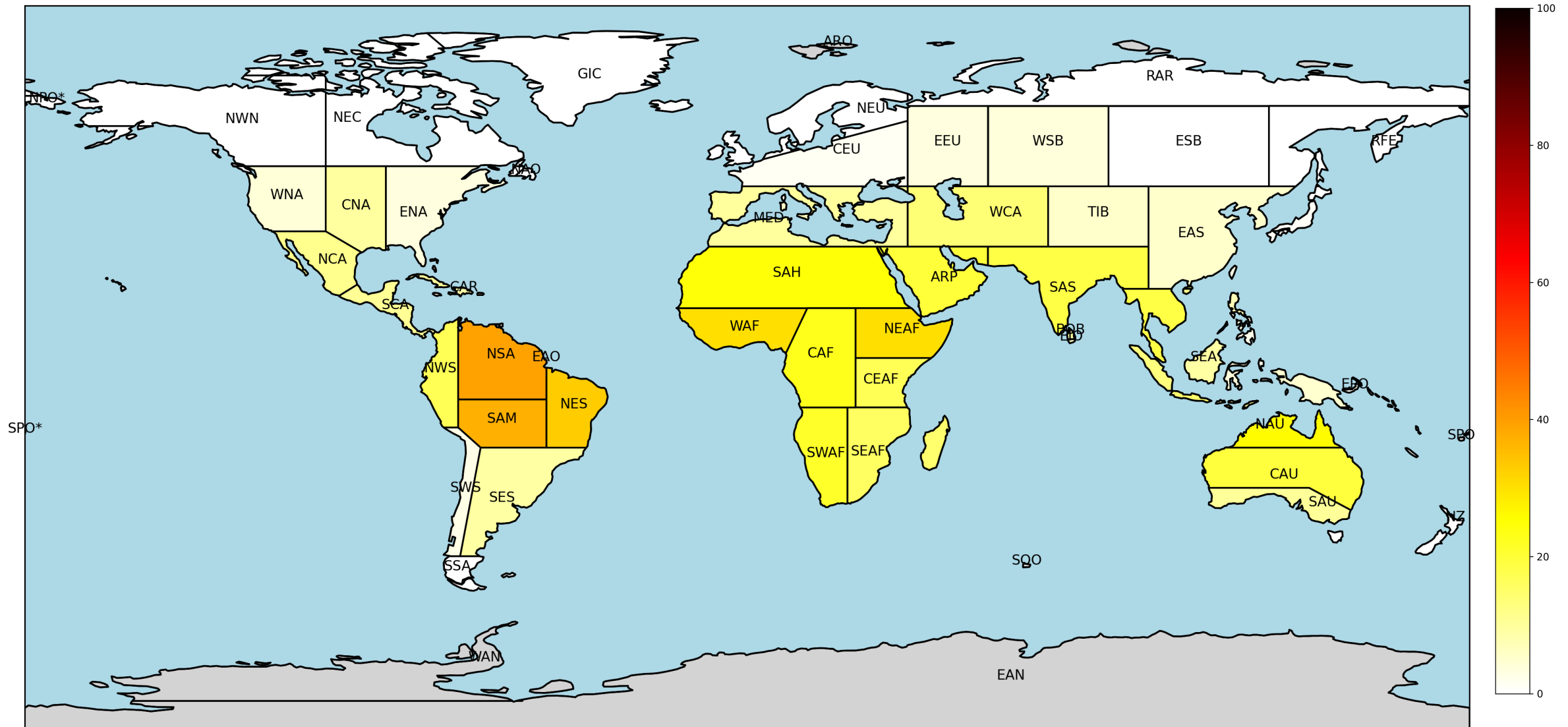
**Record wildfires in  
Spain and Portugal  
this August!!**



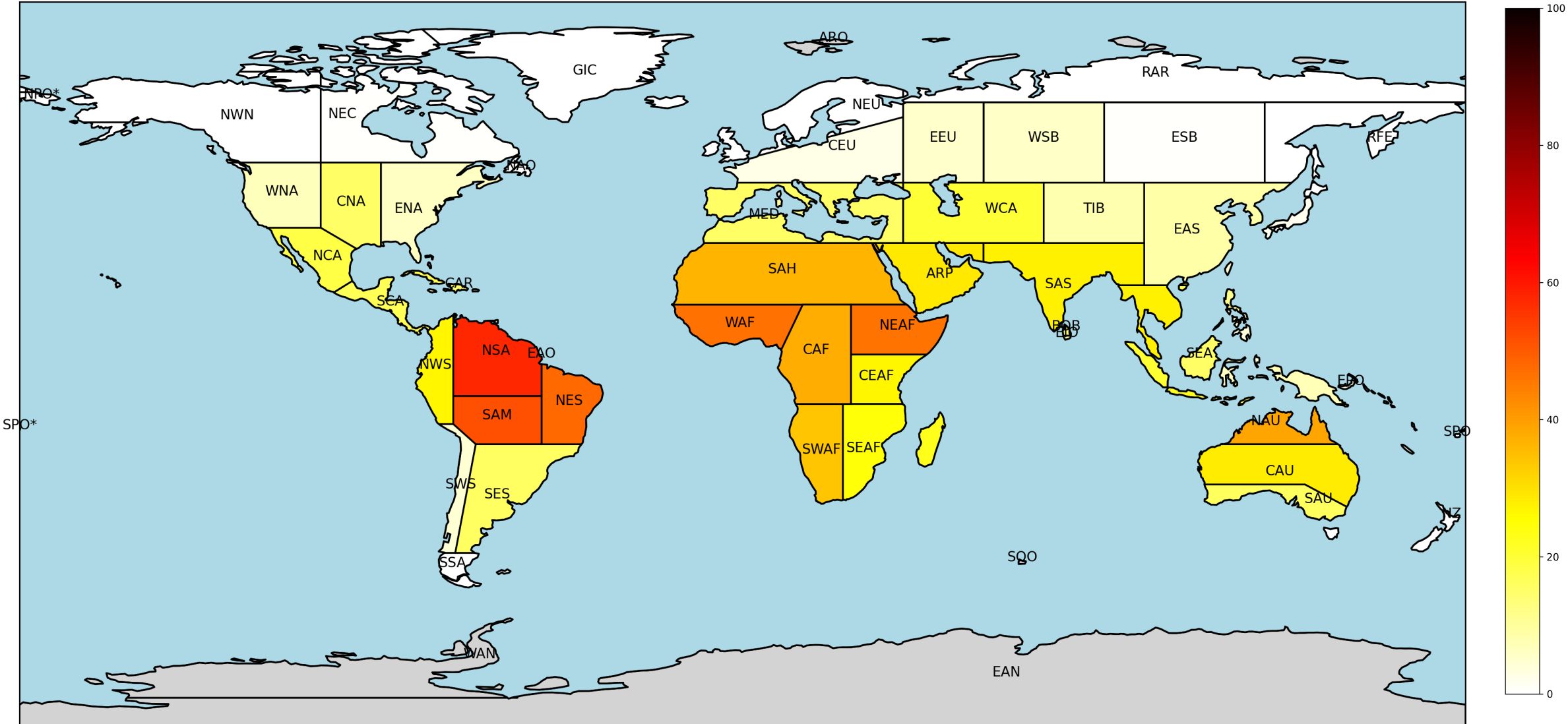
Media TX35 su regioni IPCC AR6 (solo terraferma) - gw11.0



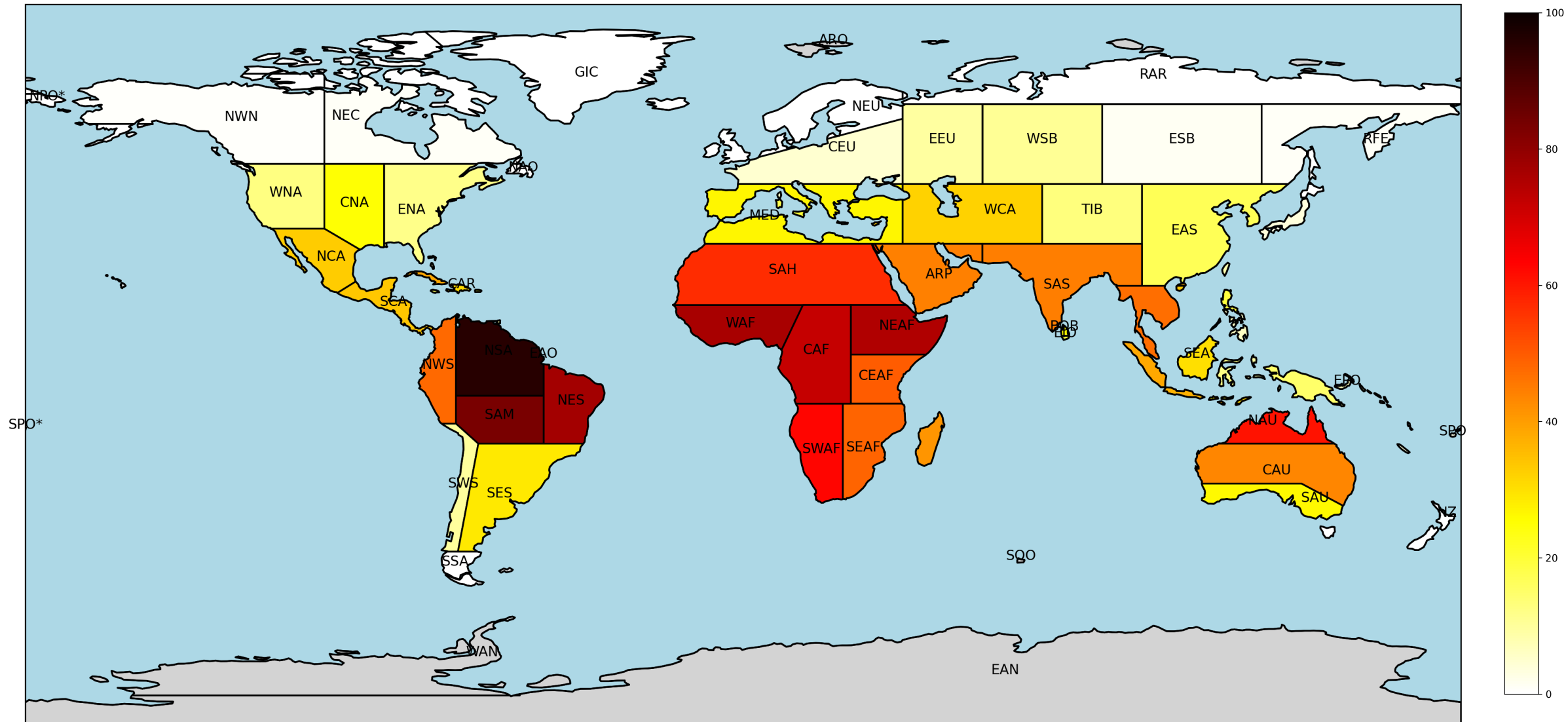
Media TX35 su regioni IPCC AR6 (solo terraferma) - gw1.5



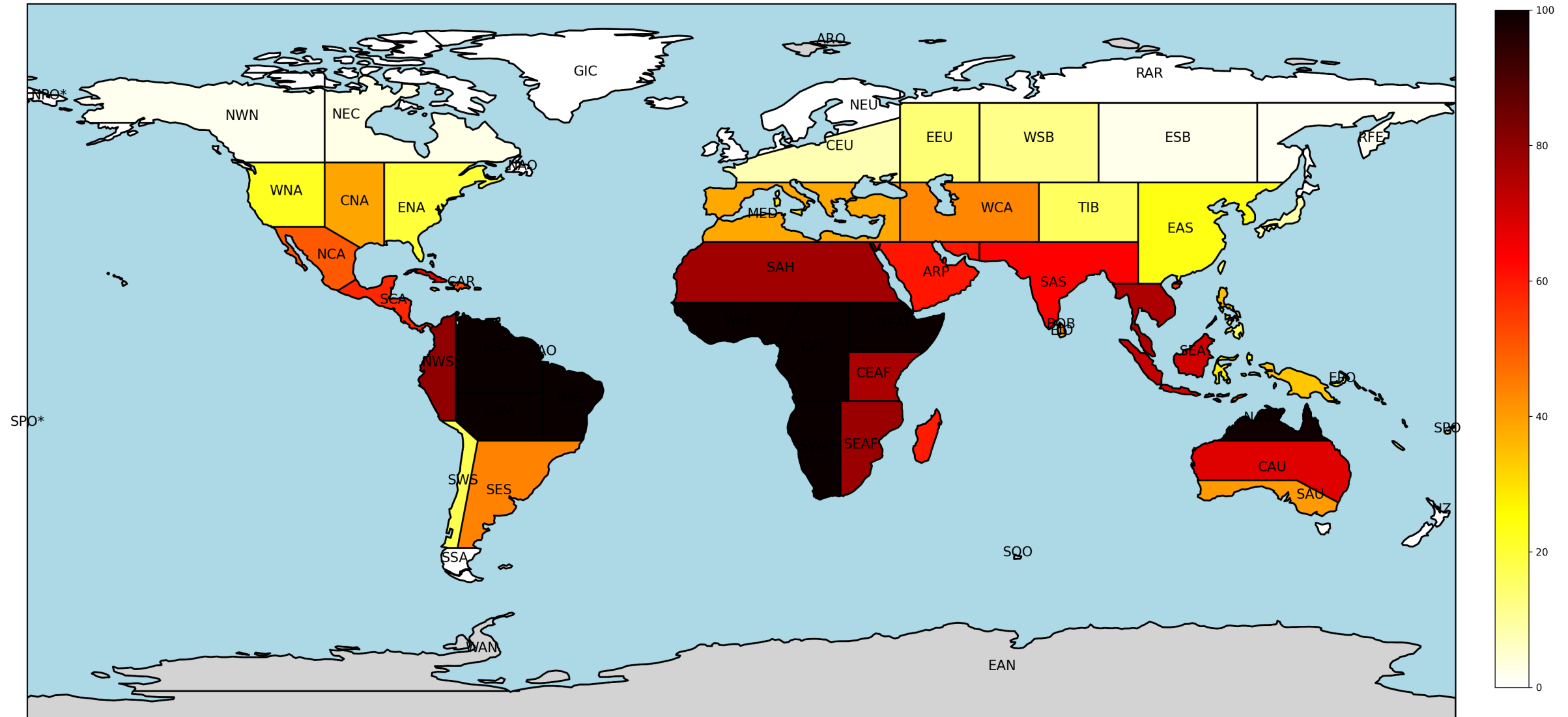
Media TX35 su regioni IPCC AR6 (solo terraferma) - gwl2.0



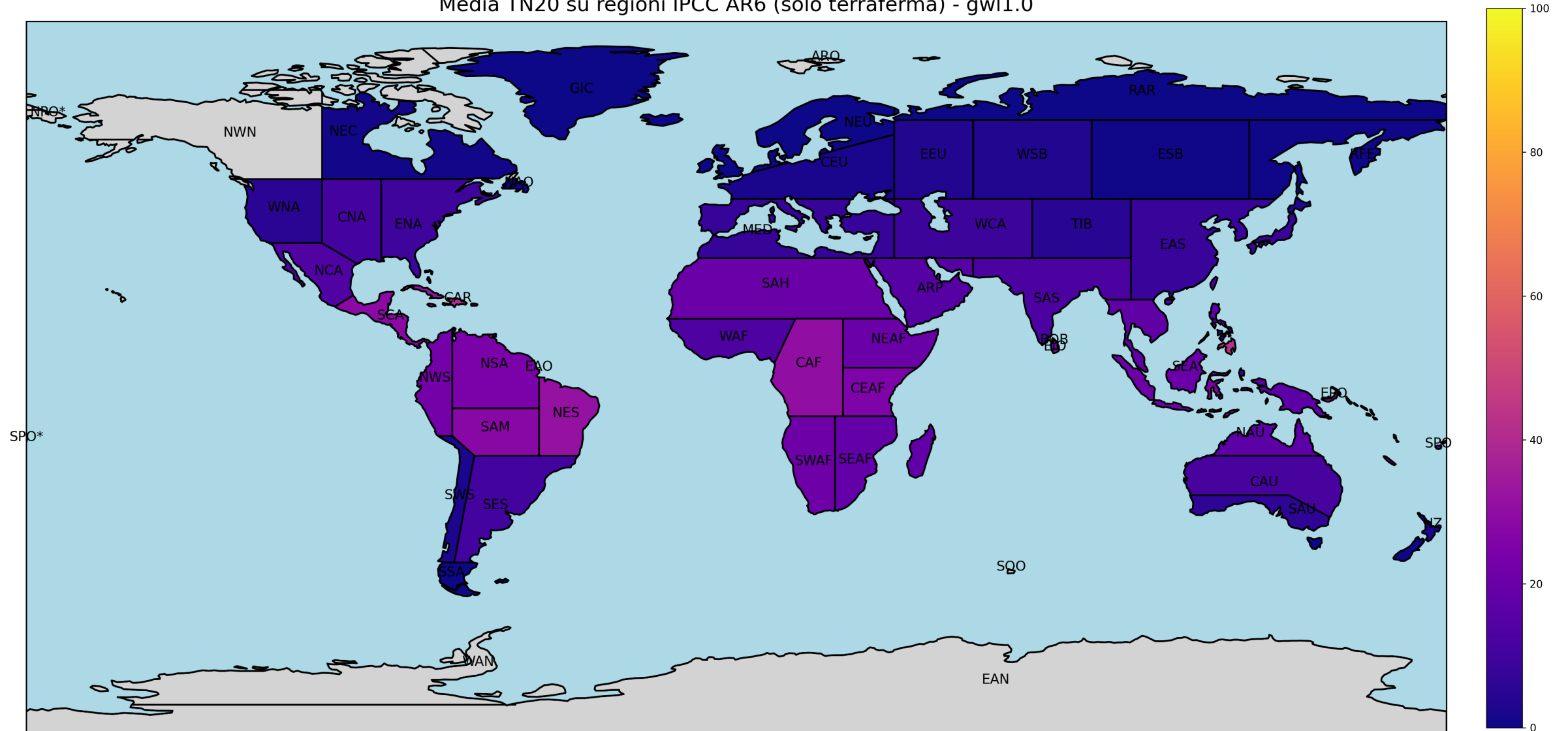
Media TX35 su regioni IPCC AR6 (solo terraferma) - gwI3.0



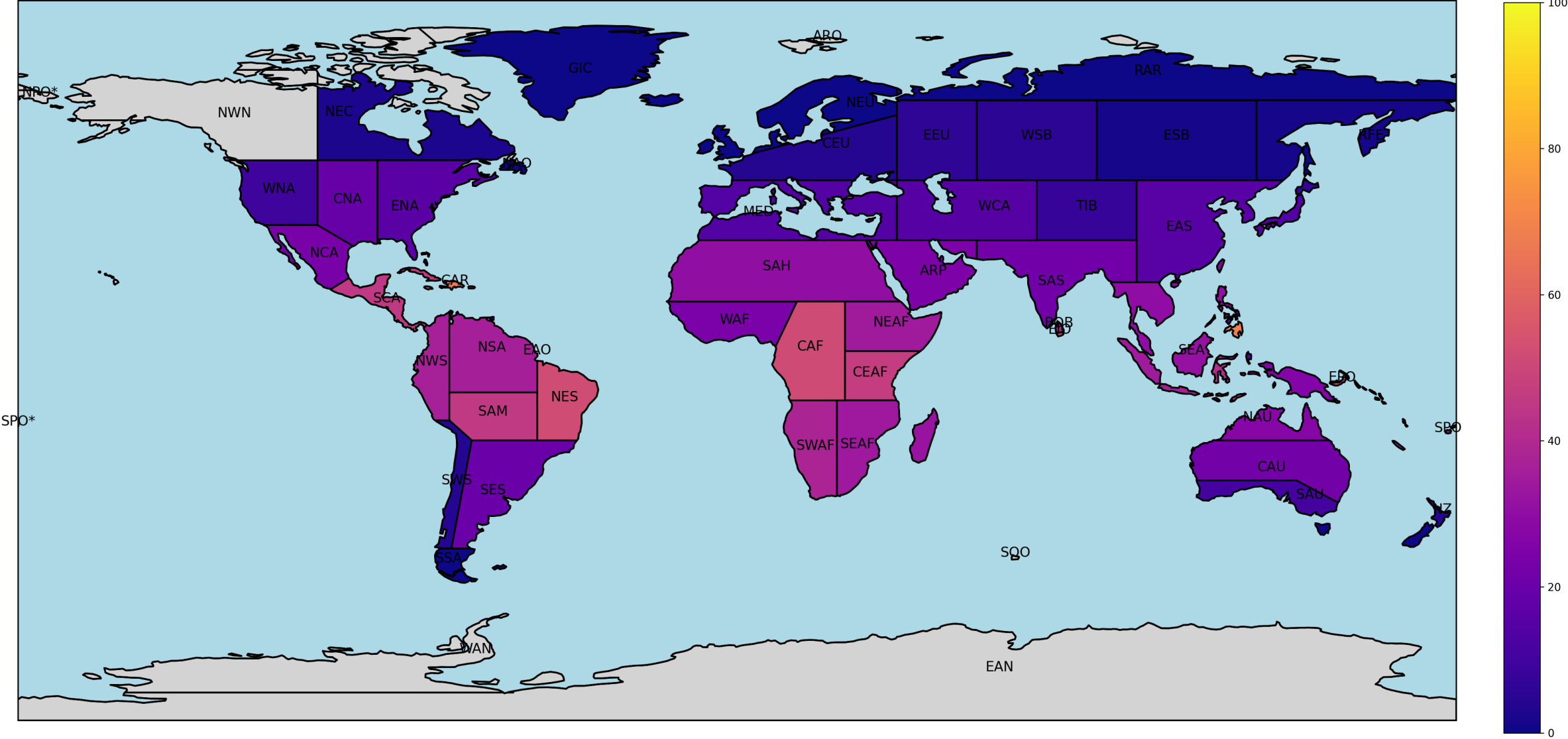
Media TX35 su regioni IPCC AR6 (solo terraferma) - gwl4.0

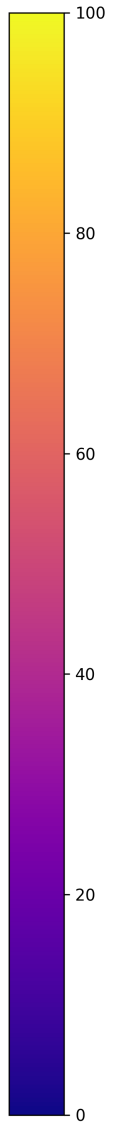


Media TN20 su regioni IPCC AR6 (solo terraferma) - gw11.0



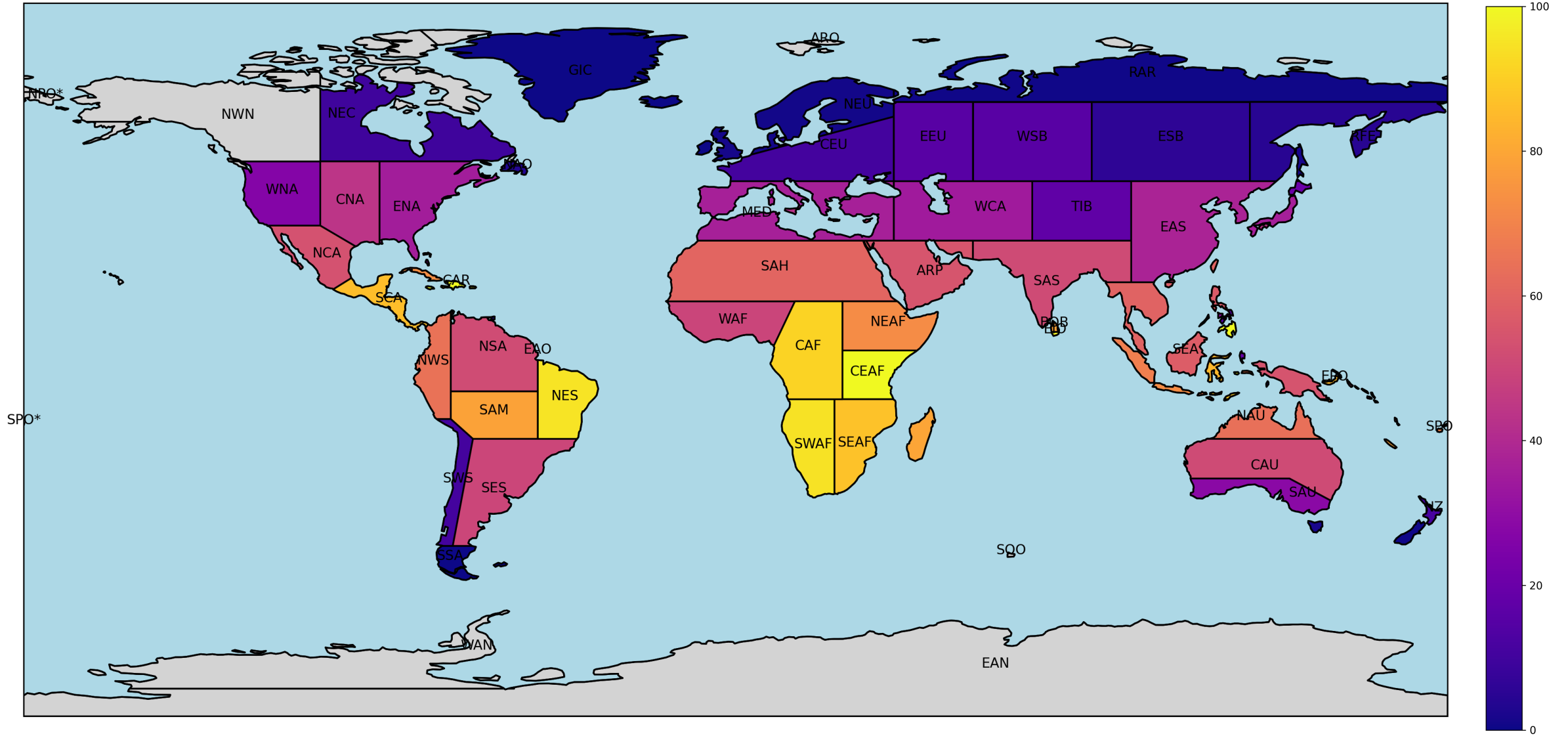
Media TN20 su regioni IPCC AR6 (solo terraferma) - gwl1.5



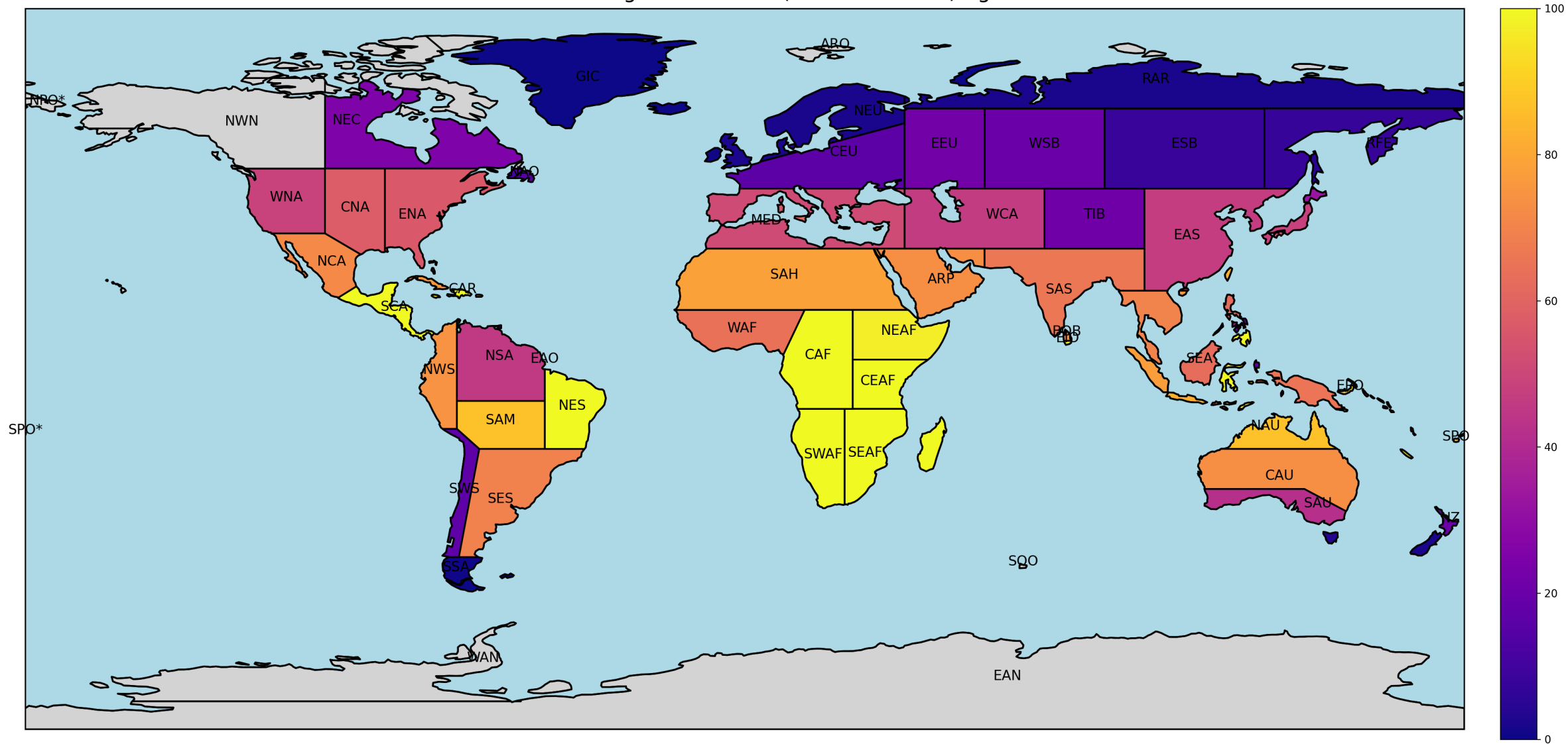




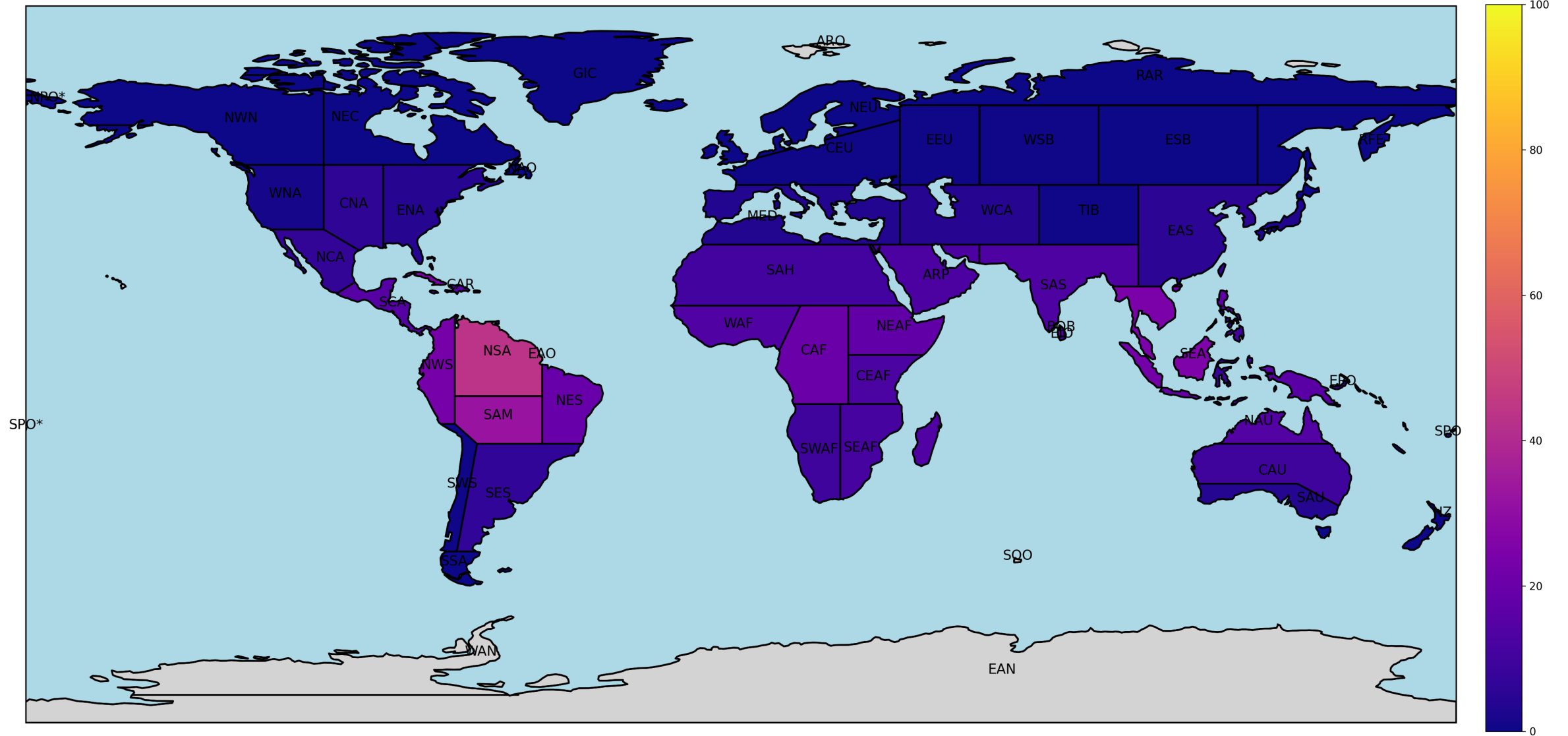
Media TN20 su regioni IPCC AR6 (solo terraferma) - gwI3.0



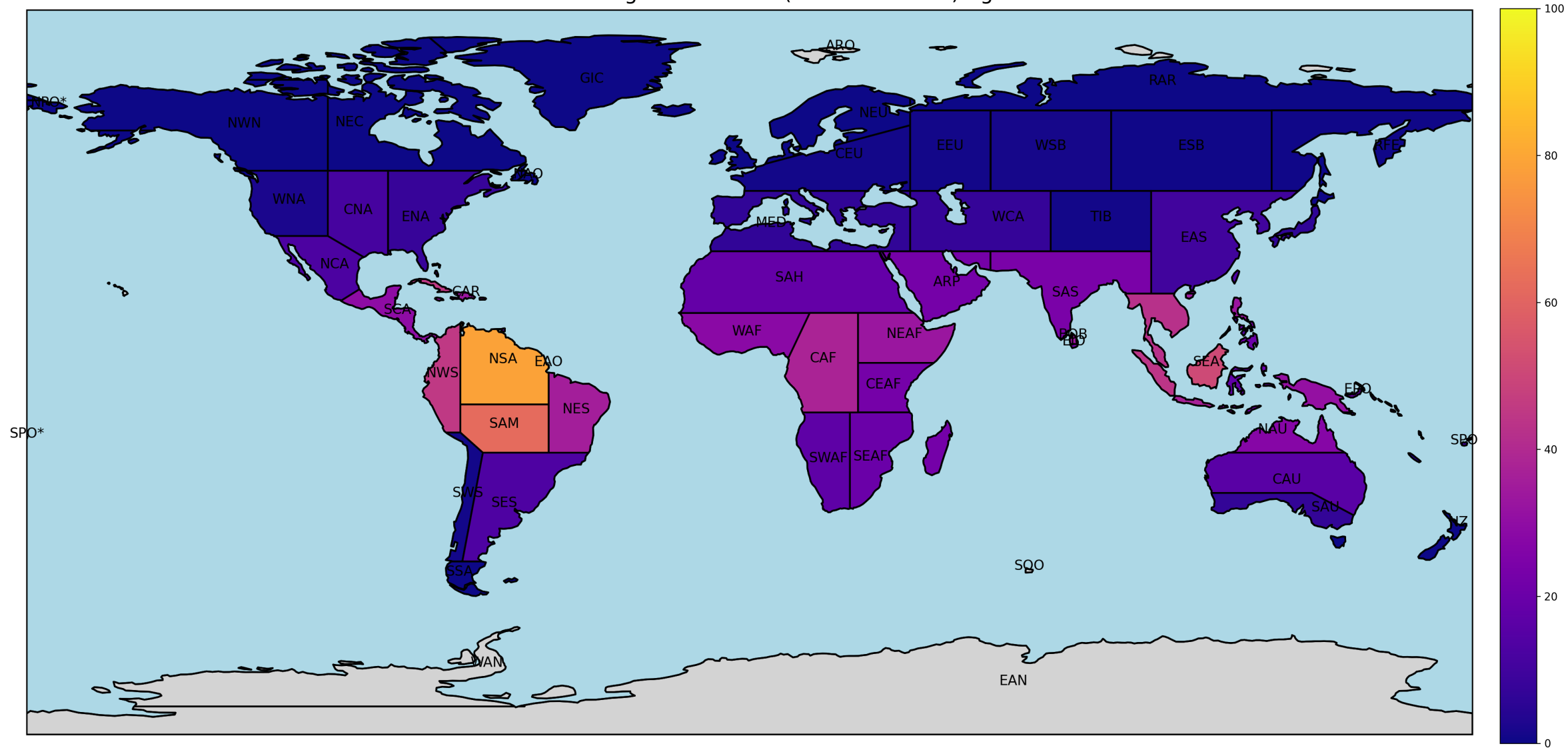
Media TN20 su regioni IPCC AR6 (solo terraferma) - gwI4.0



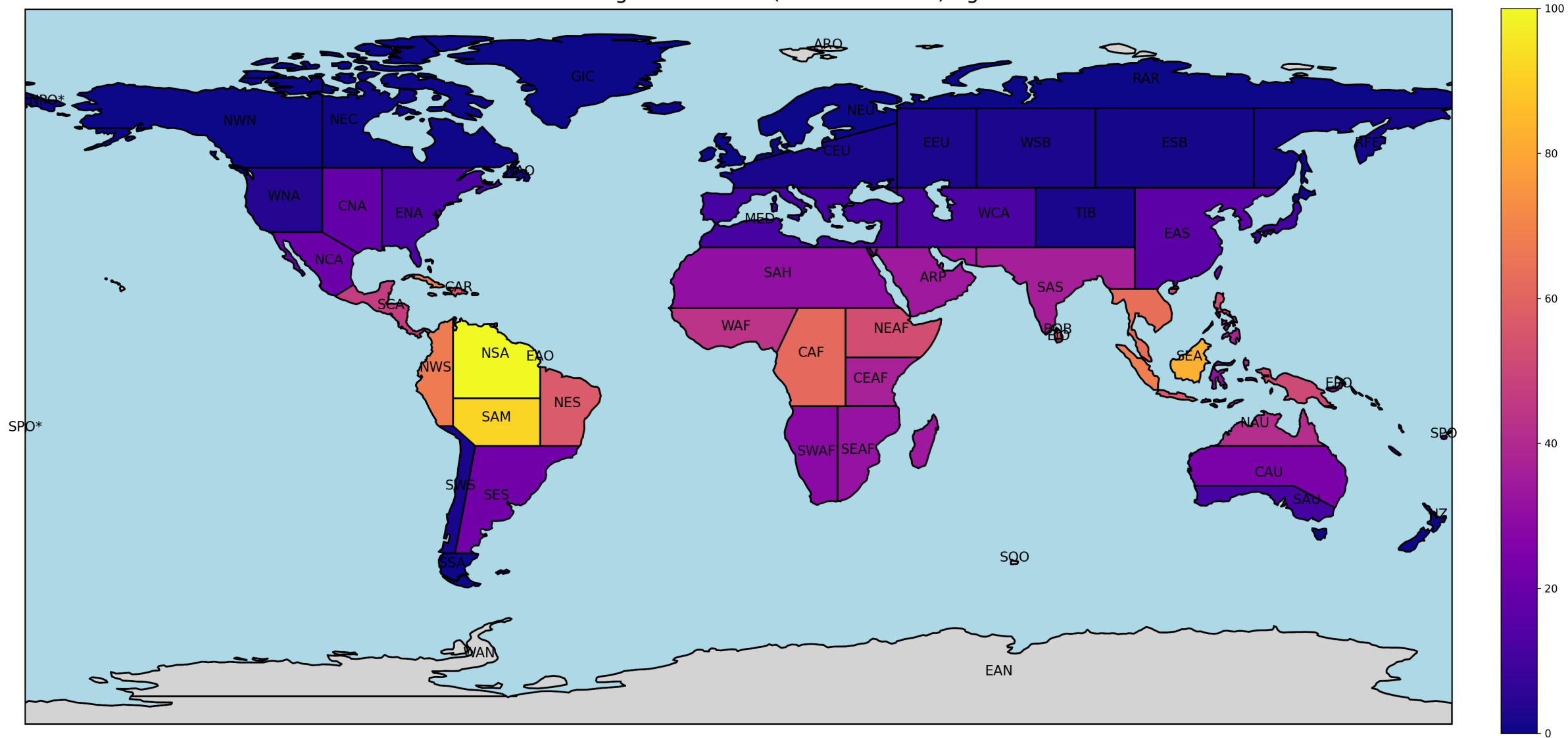
Media HI41 su regioni IPCC AR6 (solo terraferma) - gw11.0

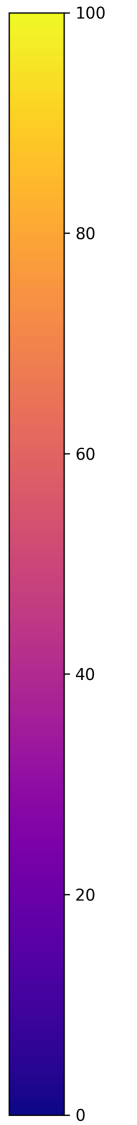


Media HI41 su regioni IPCC AR6 (solo terraferma) - gw1.5

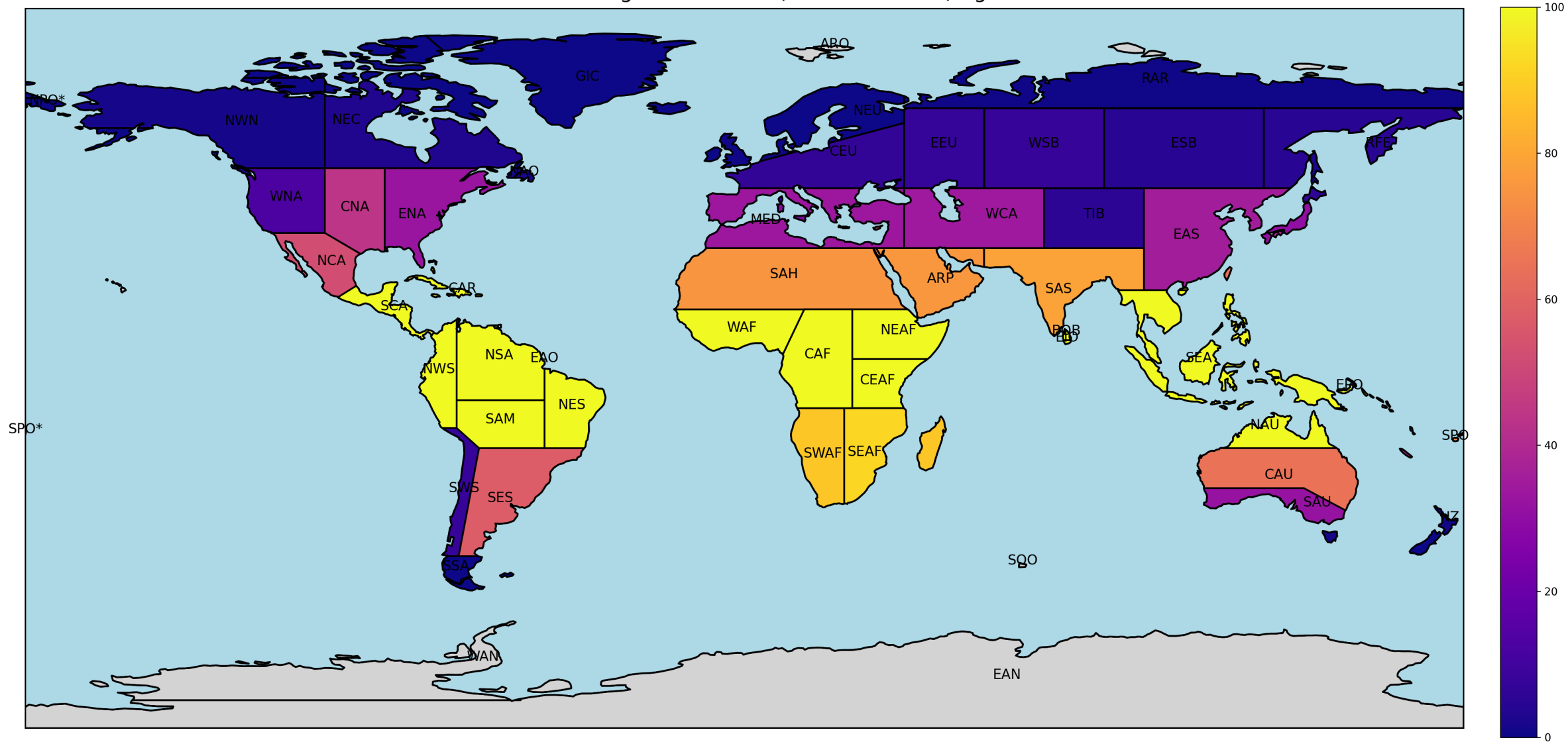


Media HI41 su regioni IPCC AR6 (solo terraferma) - gwI2.0



[illegible]

Media HI41 su regioni IPCC AR6 (solo terraferma) - gwI4.0



# The method: the probability to cross the *GToE*

The probability of reaching a specific GWL threshold is then estimated for each Climatic Impact Driver (CID) and each region of interest.

We will use GWL 1.0, 1.5, 2.0, 3.0 and 4.0 as thresholds and we will evaluate the probability of crossing them for the CIDs chosen as representative of the climate stress in a particular region.

We applied a standard method based on the signal-to-noise ratio exceedance of a specified threshold (taken as one) (Hawkins and Sutton, 2012), but enriched with the condition for the “**change robustness**”:

**a grid-point emergence occurs when the forced change is considered robust, that is when at least 66% of the models have a signal-to-noise ratio greater than one and at least 80% of them agree on the sign of change.**

The signal-to-noise ratio is estimated for each model from the ratio between the change and the standard deviation of non-overlapping 20-year means of the corresponding pre-industrial simulation; for Euro-CORDEX simulations, the reference period for the standard deviation will be 1970-1999.





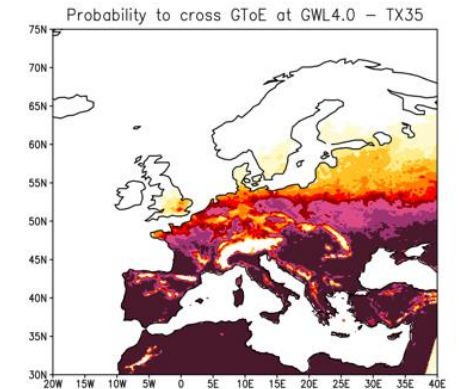
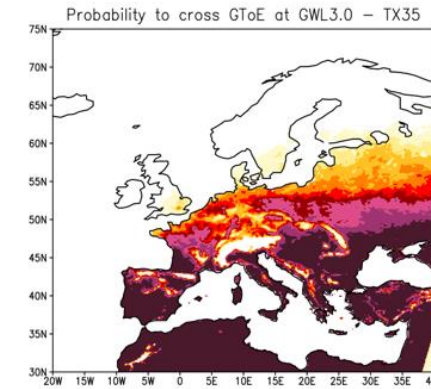
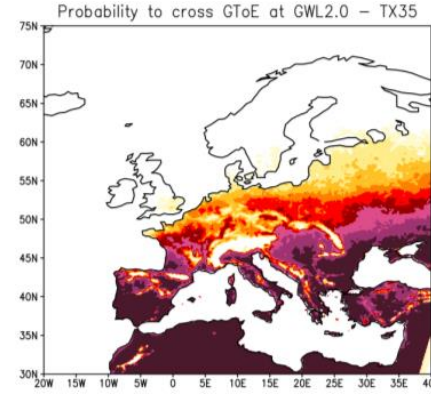
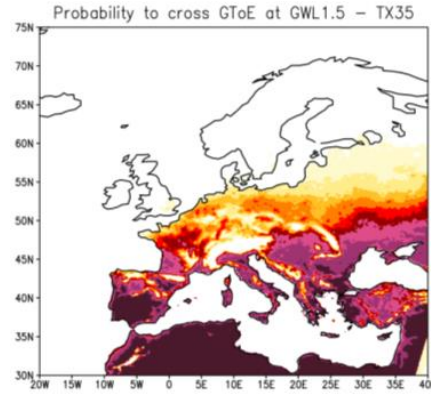
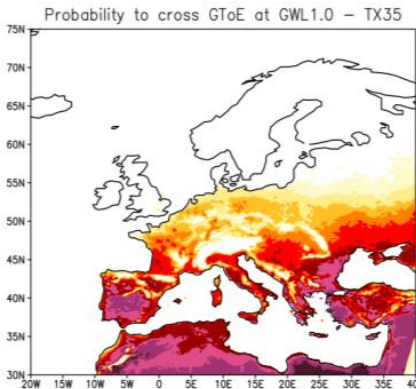
# GWL1.0

# GWL1.5

# GWL2.0

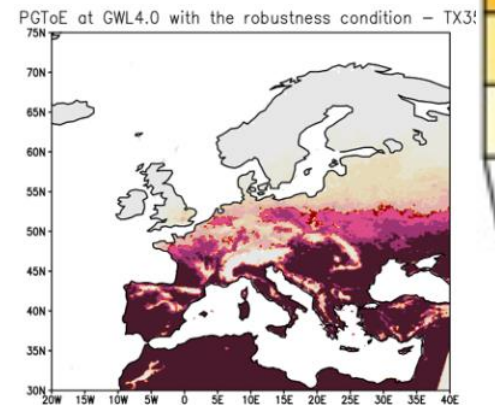
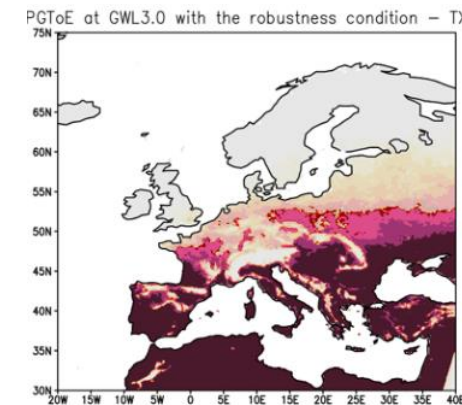
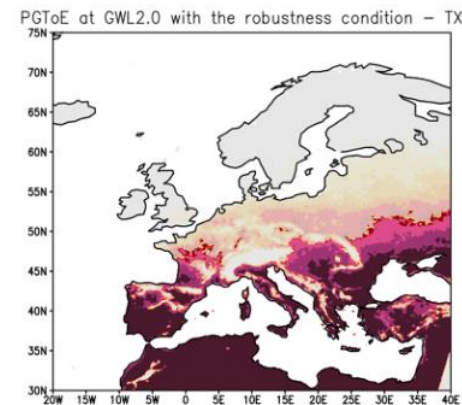
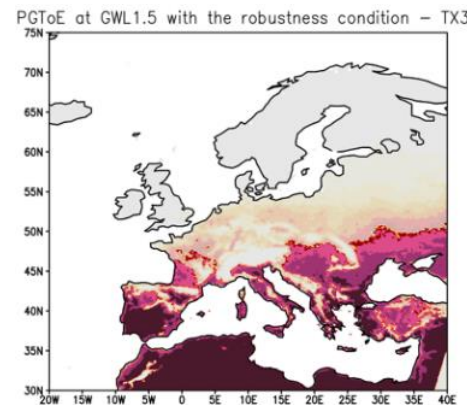
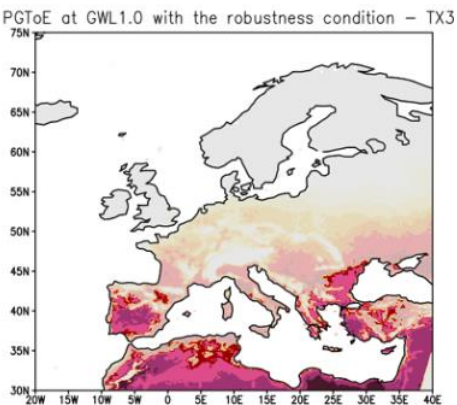
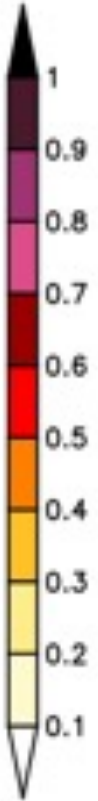
# GWL3.0

# GWL4.0



**the combination of the two conditions: at least 66% of the models have a signal-to-noise ratio greater than one and at least 80% of them agree on the sign of change**

TX35





GWL1.0

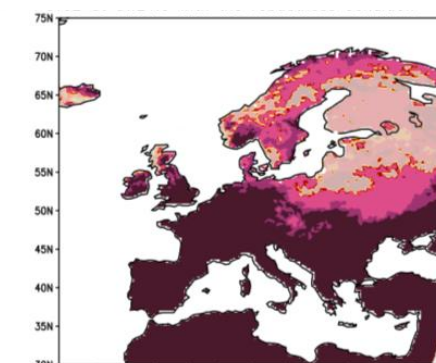
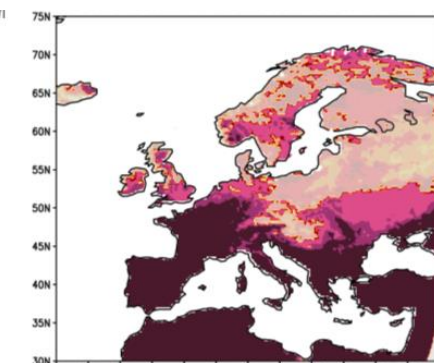
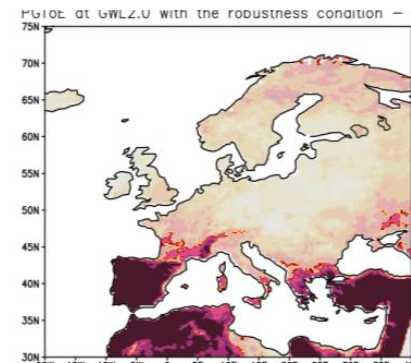
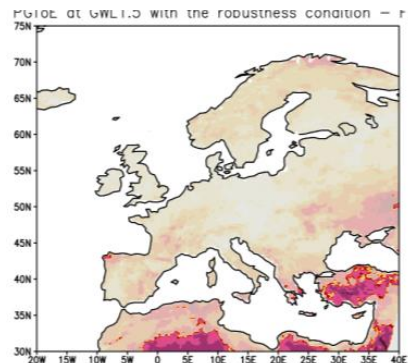
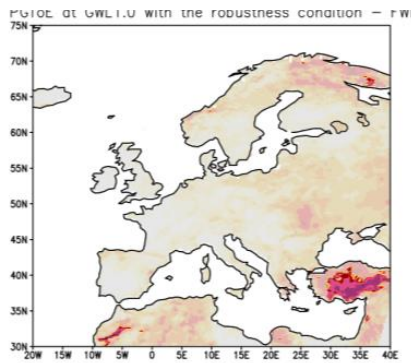
GWL1.5

GWL2.0

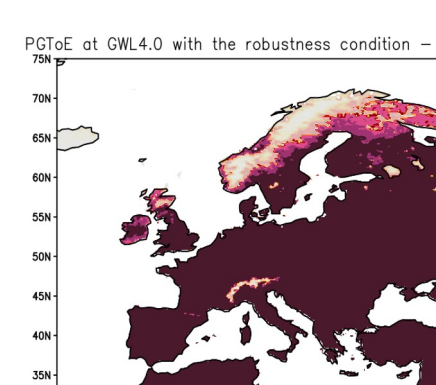
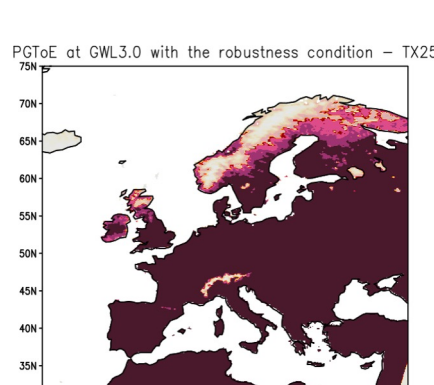
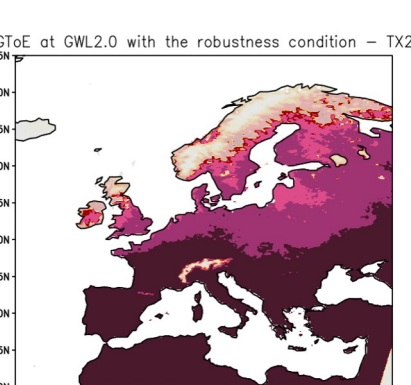
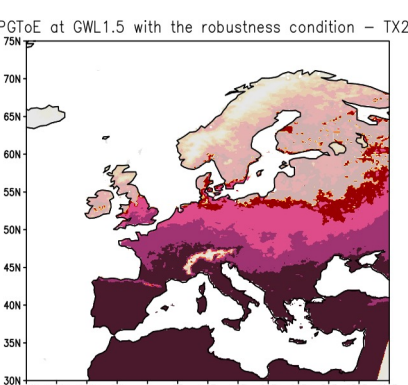
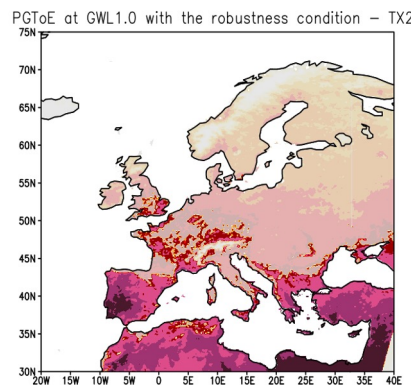
GWL3.0

GWL4.0

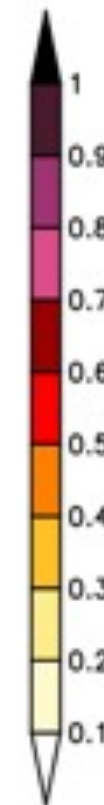
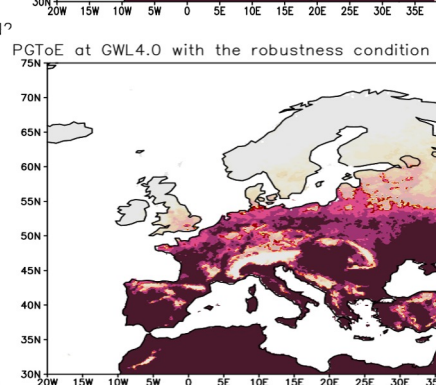
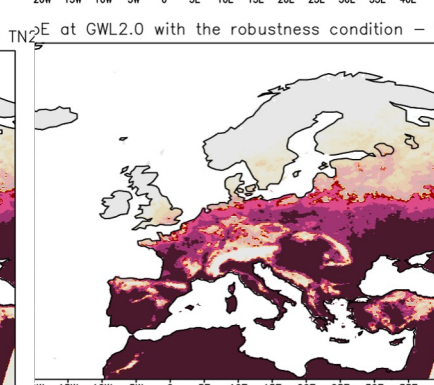
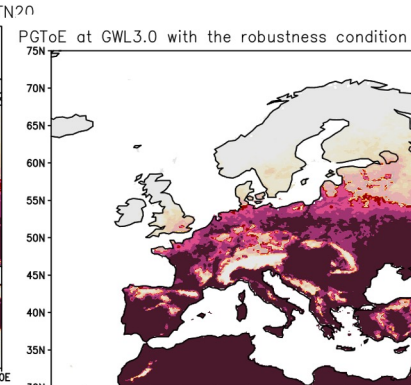
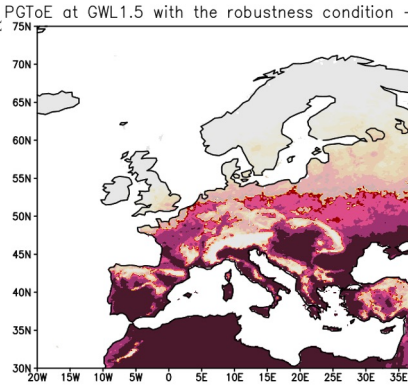
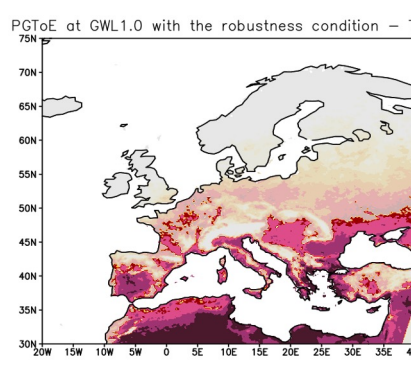
FWI



TX25



TN20





GWL1.0

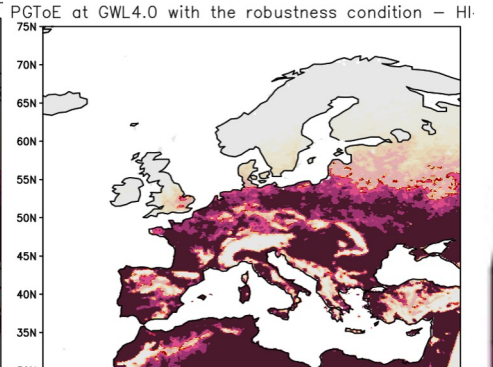
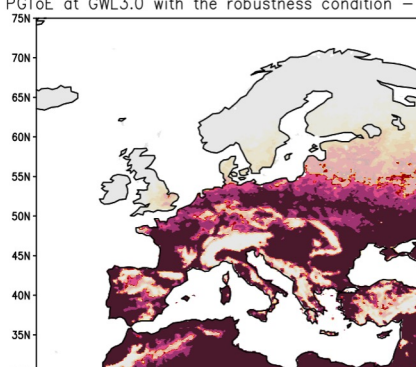
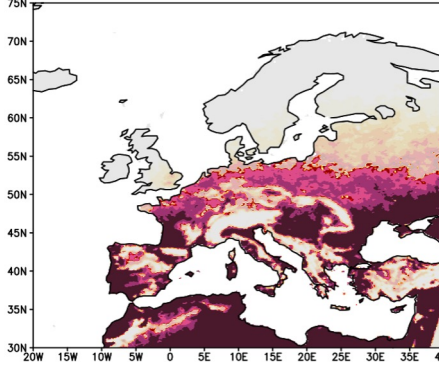
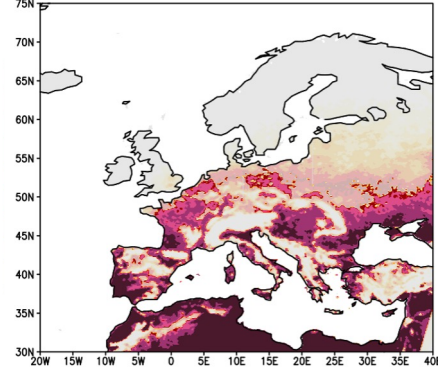
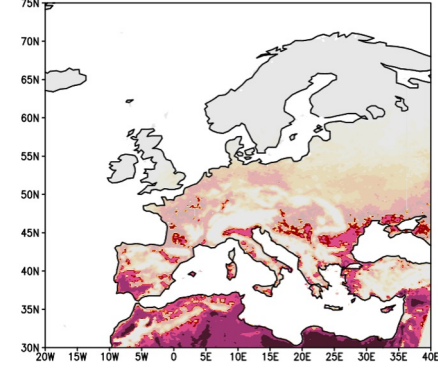
GWL1.5

GWL2.0

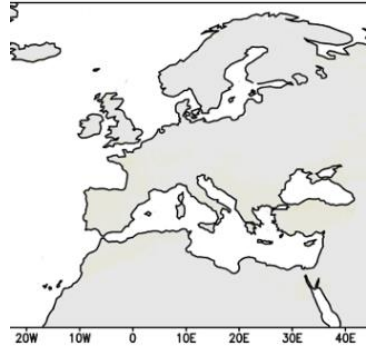
GWL3.0

GWL4.0

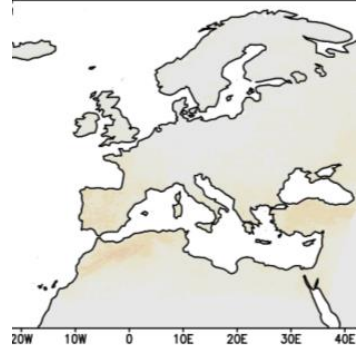
PGToE at GWL1.0 with the robustness condition - HI41



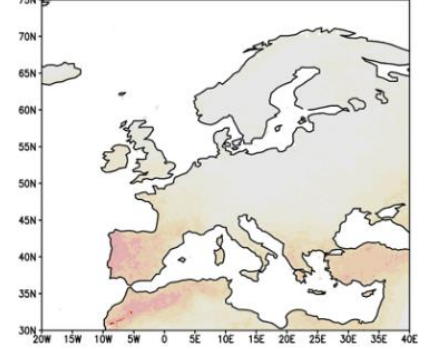
GWL1.0 with the robustness condition



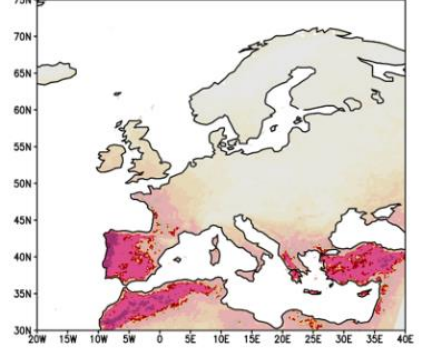
GWL1.0 with the robustness condition



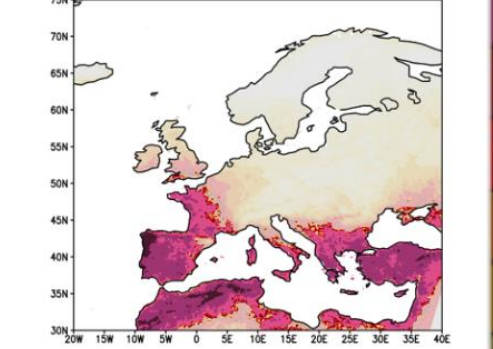
PGToE at GWL2.0 with the robustness condition - CDD



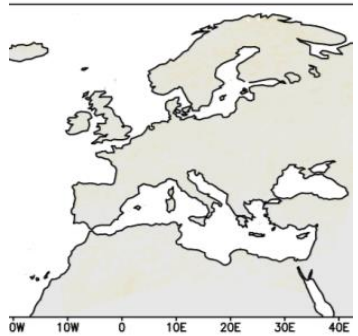
PGToE at GWL3.0 with the robustness condition - CDD



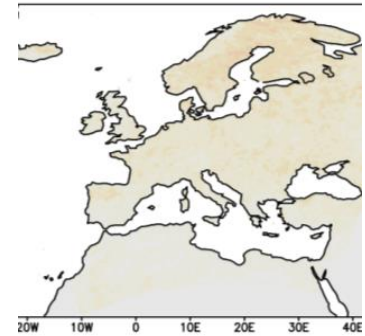
PGToE at GWL4.0 with the robustness condition - CDD



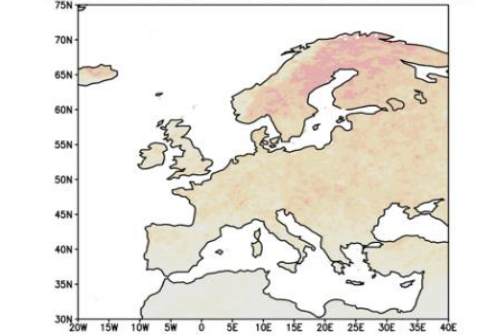
GWL1.0 with the robustness condition



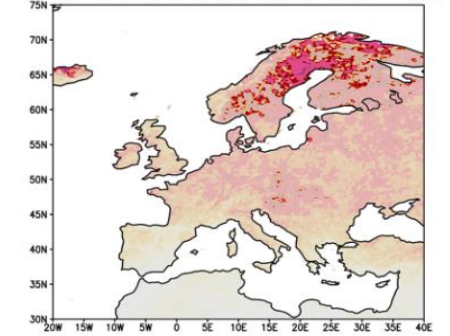
GWL1.0 with the robustness condition



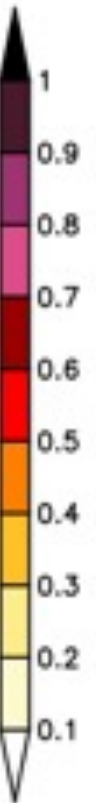
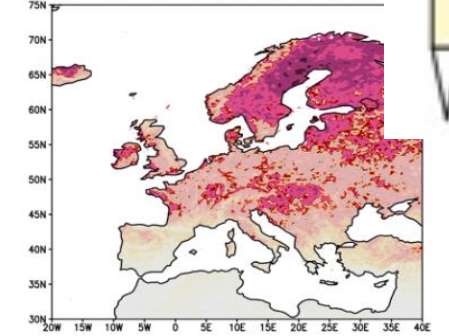
PGToE at GWL2.0 with the robustness condition - RX1DAY



PGToE at GWL3.0 with the robustness condition - RX1DAY



PGToE at GWL4.0 with the robustness condition - RX1DAY



# Let's start with the indices calculation: what do we need?

- ❖ daily time series of climatic variables (i.e.: maximum temperature and precipitation) for the historical period and the future scenario;
- ❖ GWs timeslices of each model of my ensemble;
- ❖ CDO/NCOL libraries to compute indices;





# CDO-ECA

We are going to use a set of CDO operators to compute climate indices of daily temperature and precipitation extreme developed in the frame of European Climate Assessment (ECA) project.

<b>2</b>	<b>Climate indices reference manual</b>	<b>4</b>
2.0.1	ECACDD - Consecutive dry days index per time period . . . . .	6
2.0.2	ECACFD - Consecutive frost days index per time period . . . . .	6
2.0.3	ECACSU - Consecutive summer days index per time period . . . . .	7
2.0.4	ECACWD - Consecutive wet days index per time period . . . . .	7
2.0.5	ECACWDI - Cold wave duration index w.r.t. mean of reference period . . . . .	8
2.0.6	ECACWFI - Cold-spell days index w.r.t. 10th percentile of reference period . . . . .	8
2.0.7	ECAETR - Intra-period extreme temperature range . . . . .	10
2.0.8	ECAFD - Frost days index per time period . . . . .	10
2.0.9	ECAGSL - Thermal Growing season length index . . . . .	11
2.0.10	ECAHD - Heating degree days per time period . . . . .	12
2.0.11	ECAHWDI - Heat wave duration index w.r.t. mean of reference period . . . . .	12
2.0.12	ECAHWFI - Warm spell days index w.r.t. 90th percentile of reference period . . . . .	13
2.0.13	ECAID - Ice days index per time period . . . . .	13
2.0.14	ECAR75P - Moderate wet days w.r.t. 75th percentile of reference period . . . . .	14
2.0.15	ECAR75PTOT - Precipitation percent due to R75p days . . . . .	14
2.0.16	ECAR90P - Wet days w.r.t. 90th percentile of reference period . . . . .	15
2.0.17	ECAR90PTOT - Precipitation percent due to R90p days . . . . .	15
2.0.18	ECAR95P - Very wet days w.r.t. 95th percentile of reference period . . . . .	16
2.0.19	ECAR95PTOT - Precipitation percent due to R95p days . . . . .	16
2.0.20	ECAR99P - Extremely wet days w.r.t. 99th percentile of reference period . . . . .	17
2.0.21	ECAR99PTOT - Precipitation percent due to R99p days . . . . .	17
2.0.22	ECAPD - Precipitation days index per time period . . . . .	18
2.0.23	ECARR1 - Wet days index per time period . . . . .	19
2.0.24	ECARX1DAY - Highest one day precipitation amount per time period . . . . .	19
2.0.25	ECARX5DAY - Highest five-day precipitation amount per time period . . . . .	21
2.0.26	ECASDII - Simple daily intensity index per time period . . . . .	21

[https://earth.bsc.es/gitlab/ces/cdo/raw/b4f0edf2d5c87630ed4c5dde5a4992e3e08b06a/doc/cdo\\_eca.pdf](https://earth.bsc.es/gitlab/ces/cdo/raw/b4f0edf2d5c87630ed4c5dde5a4992e3e08b06a/doc/cdo_eca.pdf)

# Where are the data you can use?

- ❖ /home/esp-shared-a/GlobalModels/**CMIP5/daily/historical**
  - ❖ /home/esp-shared-a/GlobalModels/**CMIP5/daily/rcp26**
  - ❖ /home/esp-shared-a/GlobalModels/**CMIP5/daily/rcp85**
- ❖ /home/clima-archive5/**CMIP6/\${mod}/historical**
  - ❖ /home/clima-archive5/**CMIP6/\${mod}/ssp585**
  - ❖ /home/clima-archive5/**CMIP6/\${mod}/ssp126**
  - ❖ /home/clima-archive5/**CMIP6/\${mod}/ssp370**
- ❖ /home/esp-shared-a/RegionalModels/CORDEX/**EUR-11/historical/**
  - ❖ /home/esp-shared-a/RegionalModels/CORDEX/**EUR-11/rcp26/**
  - ❖ /home/esp-shared-a/RegionalModels/CORDEX/**EUR-11/rcp85/**

# PREPARE the data:

1. Daily time series of the variables (pr, tas, tasmax, tasmin... )
2. Convert the units in the correct ones if needed!!

e.g.: `cdo mulc,86400 infile.nc outfile.nc` (to change units from "kg m-2 s-1" to "mm/day")

`cdo subc,273.15 infile.nc outfile.nc` (to change units from "Kelvin" to "degrees Celsius")

3. Select the Reference period (e.g.: 1995-2014):  
`cdo -selyear,1995/2014 infile.nc outfile.nc`

4. Select the timeslices in the future according to the GWL for a given scenario:

- search for the CMIP5/CMIP6 driver of your data in:

```
/home/netapp-clima-  
scratch/fraffael/WORKSHOP/cmip5_warming_lev  
els_all_ens_1850_1900_MIX-EUR11.csv
```



# Searching for GWLs:

If you open “cmip5\_warming\_levels\_all\_ens\_1850\_1900\_MIX-EUR11.csv”:

model, ensemble, exp, warming\_level, start\_year, end\_year

HadGEM2-ES r1i1p1 rcp85 4.0 2063 2082

HadGEM2-ES r2i1p1 rcp85 4.0 2063 2082

HadGEM2-ES r3i1p1 rcp85 4.0 2063 2082

HadGEM2-ES r4i1p1 rcp85 4.0 2060 2079

IPSL-CM5A-LR r1i1p1 rcp85 4.0 2056 2075

IPSL-CM5A-LR r2i1p1 rcp85 4.0 2057 2076

IPSL-CM5A-LR r3i1p1 rcp85 4.0 2056 2075

IPSL-CM5A-LR r4i1p1 rcp85 4.0 2054 2073

IPSL-CM5A-MR r1i1p1 rcp85 4.0 2057 2076

IPSL-CM5B-LR r1i1p1 rcp85 4.0 2075 2094

MPI-ESM-LR r1i1p1 rcp85 4.0 2072 2091

MPI-ESM-LR r2i1p1 rcp85 4.0 2071 2090

MPI-ESM-LR r3i1p1 rcp85 4.0 2071 2090

MPI-ESM-MR r1i1p1 rcp85 4.0 2073 2092



# PREPARE the data:

5. Compute the index using the daily values for the reference and future timeslices
6. Compute the mean of each period:

```
cdo timmean tx35_ref.nc tx35_ref-mean.nc  
cdo timmean tx35_gwl4.0.nc tx35_gwl4.0-mean.nc
```

7. Compute the change:

```
cdo sub tx35_gwl4.0-mean.nc tx35_ref-mean.nc tx35_gwl4.0ch.nc
```





# How can I compute some of the indices seen so far?

❖ Here: </home/netapp-clima-scratch/fraffael/WORKSHOP/recipes/>

CDDmax.readme CoolDD.readme CWDmax.readme  
HeatDD.readme NDD.readme RX1DAY.readme  
TN20.readme TX25.readme

**PAY ATTENTION to the description files:**  
**they tell you the units you need to use to**  
**compute a specific index.**