Africa Supraregional

Adaptation to Climate Change in the Lake Chad Basin

CLIMATE CHANGE STUDY

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Climate change therefore has an acute impact on agriculture, cattle farming and fisheries. The high level of rainfall variability poses a risk to production on land that is traditionally farmed through rainfed agriculture, irrigation and flood recession farming, as well as from pastureland and fishing. A lack of rainfall leads to failed harvests and damages natural vegetation. Heavy rainfall causes flooding and erodes farmland. These changes in climate conditions result in increased food insecurity, social tensions and poverty, and cause a rise in the number of refugees among the affected population.

To address these challenges, the German cooperation project “Adaptation to Climate Change in the Lake Chad Basin’ (ACC) was initiated. The project is part of the programme “Sustainable Water Management Lake Chad Basin” of the German Federal Ministry for Economic Cooperation and Development (BMZ) and the Lake Chad Basin Commission, implemented by the Deutsche Gesellschaft für Internationale Zusammenarbeit (GIZ) GmbH. Basis for this project is the Climate Change Study, executed by the AHT GROUP AG, with several modelling, addressing various assumptions in climate change. This allows the assessment of possible future climate trends, possible impacts on ecosystems within the Lake Chad Basin and socio-economic consequences. The climate change study provides the knowledge base as input and contribution for the LCBC Climate Change Adaptation Strategy Development.

Climate change under various change scenarios is mostly driven by a temperature increase. Precipitation slightly increased until mid-20th century, however
started decreasing thereafter, reaching a minimum during the droughts of the 70ies and 80ies. No significant difference in temperature development is forecasted until the 2040ies. After 2049, temperatures under an A2 scenario will rise more quickly while B1 temperatures gradually taper off.

The modeling results forecast the highest projected increase of mean annual temperature for the central and the eastern Lake Chad Basin reaching in some areas up to 3 °C (B1) or almost 6 °C (A2) by the end of this century. Increases will be highest during spring and fall.

Rising temperatures will increase evaporation/evapotranspiration, thus reducing the available water resources by up to 4 % (B1) and nearly 10 % (A2) until 2099. While at the beginning of this century, clear differences exist in reduced availability of water depending on climatic region, reductions will be almost equal to all climate zones by the end of the 21st century. Projected climate change will be the cause for a gradual southward shift of ecosystems within the Lake Chad Basin. Displacements will be smallest in dry areas (north) and increase to the south causing larger shifts and fragmentation in ecosystems of the more humid climates. A crucial transition is the line marking the 120-day growing period. It roughly separates the more intensely cultivated areas from areas with only scattered farming activities. According to FAO, this line is a threshold between drought resistant breeds of a short vegetative period and more valuable and water demanding crops. The line is calculated from precipitation and potential evapotranspiration and ignores the special situation of wetland areas.

Over the 21st century the east-west stretching line will gradually shift south with largest losses of cropping areas occurring in the central part of the Lake Chad Basin. Shifts are smallest towards the eastern (upper Komadugu-Yobe basin) and western basin periphery (area south of Am Timan). Total area losses amount to 70,960 km² (B1 scenario) and 135,150 km² (A2 scenario) until the end of the century. Because of large displacements in the central basin (east of Bongor), these areas could well be targeted for introducing alternative cropping practices. Parts of this area fall into the intervention zone of the GIZ program module.

The future situation of wetland areas that do not fall under the 120-day line rule, because these areas receive additional water from river discharge, is more difficult to predict. Upstream surface run-off, a major source of water for downstream wetlands, cannot be modeled reliably from monthly GCM data. A comparison of the wetland situation during a dry and a wet year suggests a general shrinking of e.g. the Yaere-Naga wetlands with some fragmentation of wetted areas. Human activities seem to have adjusted already to this situation and are concentrated on the wetland core that receives substantial wetting even during dry years. Due to increased evapotranspiration rates, the duration of the growing period of wetland vegetation will likely be shorter, but without falling below typical cropping periods.

Evaporation from open water surfaces such as Lake Chad will increase by about 5 % (B1) to 11 % (A2). Assuming a continuous discharge into the lake at the current level will certainly allow preserving Lake Chad within its current shorelines and some spillovers into the northern pond to sustain agricultural activities. Where such water resources (northern pond) are not available, agricultural activities should focus on the basin’s hinterland, where evaporation rates are lower.
This study analyzes possible impacts of projected future climate change on ecosystems of the Lake Chad Basin. Results are intended to serve as an orientation to realize the severity of changes and their spatial distribution and serve as a basis for developing adaptation strategies. For the agricultural areas, it is of particular interest to understand how the agricultural potentials may change under various climate change scenarios. Further, this study will provide the knowledge base (as input and contribution) for the LCBC Climate Change Adaptation Strategy Development.

In a first step, the study (i) calculates the impact on the boundaries of the Sahel for different time horizons and (ii) assesses the impact on monthly flood volumes of Logone and Chari rivers. In step 2, the study assesses the impact of changing discharge patterns on agricultural production models, considering (i) change in precipitation, (ii) availability of floodwater from the spillover of the two principal rivers, and (iii) last not least increased evapotranspiration.

Due to size of the Lake Chad Basin, a detailed mapping/analysis of e.g. individual agricultural areas and their situation under different climate conditions is not feasible. The reliable distinction of agricultural areas in Central Africa is particularly complex due to the spectral and textural appearance, the seasonal cycles and the similarity of these features with natural vegetation cover. A reliable distinction requires the analysis of a considerable number of high-resolution satellite data, a time-consuming process that was impossible to accomplish within the allotted time for this study.

This study focuses on the analysis of agricultural potentials primarily controlled by rainfall and temperature. Other parameters affecting agricultural potentials are various soil parameters, topography and sometimes groundwater. These parameters are considered as temporally stable, and are therefore not included in the analyses. Significant agricultural changes however may be experienced due to changes in the run-off situation. In areas where agriculture not only depends on local rainfall but also on stream water (irrigated agriculture, wetland areas), changes in total discharge and in peak flow (affecting inundation) may lead to a deterioration of agricultural potentials. The situation of drainage systems is addressed in this study as far as data are available.

Different hydro-meteorological (e.g. potential evapotranspiration) or climatic parameters (e.g. aridity index) are calculated for improved comprehension of the combined impact of rainfall and temperature on vegetation. Other variables used include outputs from hydrologic modeling i.e. soil moisture and surface run-off.

Climate parameters describing future climate change scenarios are taken from GCMs. While GCMs are rather accurate in forecasting global climate variation and change these are significantly less detailed in the prediction of local climate change. In parts, this is a matter of the coarse resolution of GCMs smoothing or neglecting physical processes at a local scale. To lessen the influence of a coarse spatial resolution, several statistical improvements (downscaling) are applied. Chapter 7 describes the details on the models and algorithms applied in this study. All figures included in this report except for Figure 2 are prepared using the calculation and modelling results as described within the respective chapters.
To put past and today climate situations, moisture distributions and their driving forces into proper relationship a brief review of the pre-historic development of the climate and of Lake Chad is presented in the following chapters.

Processes that still define the present climate in this region (COOK AND VIZY, 2006) triggered conditions responsible for humid periods in Africa about 7 million years ago and younger. These include the two major drivers of climate in this region, (i) the seasonal migration of the Intertropical Convergence Zone (ITCZ) and (ii) the contrast between sea surface temperature and continental temperature in the Gulf of Guinea. Scientists assume that differences in the path of the ITCZ are the major cause for the formation of pre-historic humid climates in an ecosystem, which today is arid. Forces responsible for long-term variations in sun orbital parameters that induce such differences in the ITCZ are, however unrelated to drivers responsible for climate variation/change that we observe today.

3.1 Messinian (Neogene) Lake Chad

Stratigraphic evidence in the “Zeit Formation” (Messinian, 7.2 – 5.3 million years) in North Africa reveals that sediments of this period were deposited at a higher rate as other units of Miocene – Pliocene times (23 – 2 million years) (Table 1). Thus, it is concluded that the Messinian was a time marked by high sediment yields and high rainfall (GRIFIT, 2006). Investigations in the sedimentary rocks of the Sahabi Formation in the area of the Gulf of Surt further proved the existence of rivers originating from a lake in the Chad Basin flowing via the East Tibesti Valley into the Mediterranean (Figure 1). The Lake was termed Messinian Lake Chad (sometimes also called Neogene Lake Chad).
3.2 Lake Mega Chad

While evidence for the spatial extent of the Messinian Lake Chad exclusively comes from lithological analyses, the extent of the younger Lake Mega Chad is also documented in geomorphological features. Recent interpretations of data from the Shuttle Radar Topographic Mission (SRTM), supported by sedimentological evidence, and geomorphic analyses clearly support the interpretation of extended sand-ridges as lake deposits (GHIENNE et al., 2002; LEBLANC et al., 2006a; LEBLANC et al., 2006b; SCHUSTER et al., 2005) rather than being the result of neotectonic faulting as argued by some authors. These investigations produced convincing results about the existence and the spatial extent of a Lake Mega Chad as well as its temporal episodes.

During times of maximum extension, Lake Mega Chad covered an area of approximately 350,000 km² with its deepest point in the Bodélé Depression (~210m) and linked with modern Lake Chad through the Bab Al Ghazal. Its maximum level at about 325m was controlled by the Mayo Kebbi spill point, from where water would discharge through the Benue River into the Atlantic (Figure 1).

Geomorphological investigations suggest several Lake Mega Chad episodes (since at least 3 to 3.5 million years) with the latest lasting until approximately 5,300 to 4,400 years BP (before presence) (GRIFFIN, 2006), and coinciding with the mid-Holocene wet period of northern and central Africa.

Explaining the existence of a lake in what is today one of the world’s driest areas, was the focus of many research studies. Differences in vegetation, surface albedo, or the differences in temperature between sea surface and continent are considered a major or at least a contributing factor. Modern General Circulation Models (GCM) used in climate modelling now provide convincing evidence, attributing the wetter conditions during this period to differences in sun orbital parameters (SEPULCHRE et al., 2008).

During Lake Mega Chad times, the sun reached further to the north (today Tropic of Cancer at Latitude 23.5N), causing a northward shift of the Inter-tropical Convergence Zone (ITCZ). The migration of the ITCZ affects seasonal precipitation patterns, bringing in moist air (Monsoon) from the Gulf of Guinea during summer, but hot and dry air from the Sahara (Harmattan) as the sun moves to the south (winter) (Figure 2).

With a northward shift of the ITCZ, moist air, dragged in from the Gulf of Guinea could move further north and provide the necessary moisture. This modification in the ITCZ was responsible for the existence of the Messinian Lake Chad and of Lake Mega Chad.

It is one theory that the trigger for shifts in ITCZ at such time-scales is due to variations in the earth axis’ rotational parameters including eccentricity and axial tilt and variations in the Earth’s orbit (Milankovic theory). Each of these parameters completes a full cycle at a different time scale (axis eccentricity ~26,000 years, axial tilt ~41,000 years, orbit eccentricity ~100,000 years) with their combined effect determining Earth-Sun position (insolation) and its timing and thus temporal variations in ITCZ position. A more recent study (ZHANG et al., 2014) also employing climate modeling techniques, argues that the closure of the Tethys Sea in the north weakened the African monsoon, making it solely dependent on sun orbital parameters and thus initiating the aridification of the Sahara.
As described above, climate in the LCB region is strongly controlled by the seasonally varying position of the Inter-Tropical Convergence Zone (ITCZ). The inter-tropical front crosses the region from south to north supplying the south of Chad with typically more than 900 mm of rain per year which becomes continuously less as one moves north (Figure 3). Seasonal rains occur between July and September, a period that may be considerably shorter (longer) towards northern (southern) latitudes. With the arrival of the rains, temperatures typically drop by five and more degrees, caused by dense cloud coverage, blocking solar radiation and by evaporative cooling. The months of August to June are mostly dry with lowest temperatures reached during wintertime and highest temperatures during spring.

Strong inter-annual climate fluctuations are common in this region, with rainfall variations often exceeding +/- 200%. Inter-annual climate variations are believed to be controlled by differences between sea surface temperature and continental temperatures. Drivers of sea surface temperature in the Gulf of Guinea - the dominant moisture resource for rain in the Lake Chad Basin - are linked to a global network of ocean currents of which the mechanism is not yet fully understood.

During the past decades, the region was affected by several dry spells, marked by consecutive years of low rainfall, with impacts on water resources, vegetation covers and on agricultural production. The last severe dry spells in the region occurred at the beginning of the seventies and at the beginning of the eighties.

Since then, the region gradually recovered, primarily driven by increasing seasonal rainfall (Figure 9). Even though a general recovery from the droughts of the 70ies and 80ies of the 20th century can be observed, inter-annual rainfall variation is still high. Poor rainfall years take turns with good rainfall years, thus preventing environments (water resources, vegetation) to reach a pre-drought level. Extended dry spells, however, did not occur since then. Other than rainfall records, this trend is similarly reflected in the recovery of natural vegetation covers, and in the rising of lake levels such as those of Lake Fitri.

Whether such inter-annual rainfall fluctuations can be attributed to global climate change is hard to say. However, it is a fact that similar high rainfall fluctuations did occur at the beginning of the 19th century, though at higher overall rainfall amounts (Figure 9).

4.1 Current Climate Situation in the Lake Chad Basin

Annual migration of the ITCZ causes a strong North-South precipitation gradient (Figure 3). In the south, the length of the rainfall period is longest and rainfall amounts are highest. In the utmost south year-round precipitation leads to tropical conditions. Moving North, precipitation periods quickly shorten and amounts drop to zero creating hyper-arid conditions (see Figure 4).

Annual temperature cycles also show a characteristic North-South gradient controlled by the position of the sun and by cloud cover. An increase in cloud
Following HARGREAVES, the annual total potential evapotranspiration (PET) is calculated using minimum, maximum and mean temperature as one input. According to UNEP’s classification of aridity, the majority of areas in the Lake Chad Basin used for some kind of agricultural activities, fall into the arid and the semi-arid class. Arid and semi-arid climates are generally considered as little productive. However, to assess the agricultural potential, aridity levels based on total annual P and PET are less important. More important is the length of the growing period where climate conditions exceed certain thresholds. FAO (2004) gives some orientation on required climate conditions for growing different crop types (Table 2).

The temperature and precipitation distribution predominantly define the region’s agricultural potential. The Aridity Index (AI) is a parameter that allows expressing temperature and precipitation in a single parameter. The Aridity Index calculates as:

\[ AI = \frac{\text{Precipitation}}{\text{Potential Evapotranspiration}} \]  

Equation 1

Table 2: Some typical crops under rainfed conditions (FAO, 2004)

<table>
<thead>
<tr>
<th>Class</th>
<th>Length of growing period</th>
<th>Typical crops</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hyper-arid</td>
<td>0</td>
<td>No crop, no pasture</td>
</tr>
<tr>
<td>Arid</td>
<td>1 – 59</td>
<td>No crops, marginal pasture</td>
</tr>
<tr>
<td>Semi-arid</td>
<td>60 – 119</td>
<td>Bush millet, sorghum, sesame</td>
</tr>
<tr>
<td>Dry sub-humid</td>
<td>110 – 179</td>
<td>Maize, bean, groundnut, peas, barley, wheat, tall (suitable for rainfed agriculture)</td>
</tr>
</tbody>
</table>

* Modeled precipitation (right) from downscaled CGCM3T47 model. Observed and modeled data represent 2000-2009 averages. Modeled data show higher rainfall particularly in the south. To correct this anomaly a relative increase between years, added to the observed rainfall was used in modeling future scenarios instead of absolute modeled precipitation values.

5 Temporal intervals of 10-days for different climate zones as shown in Figure 5.
When calculating the aridity index (AI) as an annual average (An in the diagram), the AI does not even reach semi-arid conditions. This is different for AI calculations at 10-day intervals where both cycles considerably exceed the annual average for a longer period of time offering a window for agricultural activities.

To calculate the length of the period with conditions that exceed dry sub-humid conditions for at least 120 days, the aridity index is calculated for every 10 days throughout a year \( (P_{10}/PET_{10}) \). Calculations were done on a 2000-2009 mean for precipitation and PET. The resulting 120-day line matches well with more intensely cultivated areas (Figures 7 and 8). The identification of areas that are more intensely cultivated is based on the interpretation of seasonal vegetation cycles \(^6\) (Figure 7) and on the frequency of bush fires (Figure 8). The first distinguishes between vegetation classes (vegetation phenologies) based on differences in the shape of their seasonal cycle considering parameters including density, vigor, and length of the green period (see diagram in Figure 7). The classification of vegetation phenologies shows vegetation covers with extended green periods in purple colors, followed to the north by dark green colors describing dense vegetation covers with strong seasonal NDVI peaks. Further to the north, vegetation covers become increasingly scattered, showing shorter green periods indicated by light green colors that gradually turn into brown (bare soil, Figure 7). The second, the fire interpretation, assumes a strong linkage between some kind of agriculture and the frequency of fires. Fires however may also be linked to hunting and fishing activities or may occur naturally. Nevertheless, the two interpretations – fires and vegetation classification – show good agreement in their spatial distribution confirming each other and thus verifying the assumption to be indicative for agricultural activities.

Highest fire activities are found in the wetlands of the Yaere-Naga and within the seasonally flooded, former shorelines of Lake Chad. Fire activity is also high in

\(^6\) A brief description of the technique of classifying the shapes of vegetation cycles is presented in the Annex.
4.2 Past Climate Variation and Change

Studies of recent climate variation and change are not only a strong indicator for future climate change developments but also useful for understanding possible climate change impacts and for judging an area’s climate sensitivity. In this respect the analyses of patterns e.g. of past temperature change may be a more reliable indicator for future temperature developments than results from General Circulation Models or from Regional Climate Models.

For assessing climate variation and change over the past 40 to 100 years, different precipitation and temperature data sets were used, provided by the Climate Research Unit Temperature (CRUTEM) of the British Hadley Climate Center, by the U.S. National Climate Data Center (NCDC) and by the Global Precipitation Climate Center (GPCC) of the Deutsche Wetterdienst (DWD).

Named centers continuously collect, store and distribute data for different climate variables in their original form (NCDC) or pre-processed. Originally, all data are station measurements that the climate centers, for some products, interpolated to grids of different resolution. These data sets come in resolutions of daily or monthly intervals and at spatial resolutions of 0.5 to 5 degrees. Data from the NCDC are provided as individual, daily station records that can be interpolated to grids as needed.

4.2.1 Long-term precipitation development 1901 – 2010

The analyses of precipitation fluctuations and trends during the last century is based on pre-processed station data provided by the GPCC (RUDOLF et al., 2010), spatially interpolated to monthly raster data of 0.5 degree resolution. Though spatially not very detailed these data give an excellent insight into long-term regional precipitation development.

Rainfall in the LCB is submitted to pronounced inter-annual rainfall variations throughout all climate zones, which – during the observed period – only became temporarily lower (more stable) during the fifties. During this decade, the area received consistently high rainfall amounts. Since the nineties inter-annual variations are back to a level comparable to the beginning of 20th century, however at much lower overall rainfall amounts. The high inter-annual variability makes this area and its agriculture extremely vulnerable.

4.2.2 Long-term temperature development between 1953/73 and 2013

In contrast to precipitation, no temperature data exists for this area that cover a similarly long period (1901 – present). The longest temperature data set available is gridded, coarse resolution (5 degrees) data provided by CRUTEM (Hadley Center). This data set contains temperature data averaged over a 5 x 5 degree area (about 600 km x 600 km) as indicated in Figure 10 and covers different periods depending on data availability (absolute earliest records start in 1850, earliest in the LCB in 1901). There are no CRUTEM data for the northeastern LCB.

The most complete records, starting at the beginning of the 20th century (western LCB), show positive temperature anomalies at the beginning of the century that are at a level similar or even higher than those today. The most southern area (pixels 1407 to 1409) takes a contrasting development compared to all other areas located to the north. A mid-century temperature low visible in most northern pixels seems to correlate with a period of high rainfall (compare Figure 9). Higher cloud cover during rainy years may be one explanation for the temperature drop but there may be more.

\* Deviation from mean between 1901 and 2010 for the arid (top) and the semi-arid climate zone of the LCB as delineated in Figure 5. Anomalies are calculated relative to a 1981-2001 mean. Mean precipitation for the arid zone and semi-arid zone is 340 mm and 860 mm resp.

\* Temperatures from available climate stations (red dots) are averaged across the pixel area. Figure 11 presents temperature diagrams (anomalies) for each pixel area. Numbers stand for the pixel Id in the CRUTEM grid.
A strong impact on water resources (soil moisture) as changes in annual temperatures suggest. In contrast, agricultural activities extending into months of October and November are likely to struggle more with the effects of a temperature increase.

Along the southern basin periphery, representing the area with maximum precipitation and hence most river discharge, temperature increases are among the lowest during all seasons. As the rainy season is least affected by temperature increases, past climate change does not have such a detailed impression on spatio-temporal variations of temperature change. Over the past 40 years, the eastern Chad experiences the highest temperature increase in the entire basin. Increases across the entire basin are strongest during spring and fall (up to 6 °C). During summer and during winter months we see the "least" temperature increases (3-4 °C). Along the southern basin periphery, representing the area with maximum precipitation and hence most river discharge, temperature increases are among the lowest during all seasons. As the rainy season is least affected by temperature increases, past climate change does not have such a strong impact on water resources (soil moisture) as changes in annual temperatures suggest. In contrast, agricultural activities extending into months of October and November are likely to struggle more with the effects of a temperature increase.

In the above map, diagrams approximately cover the area they represent (each pixel is about 600 km x 600 km, compare Figure 10).
Outputs from a range of different general circulation models (GCM) are available for the analysis of possible impacts of future climate change on natural resources of the Lake Chad Basin. This climate study considers model outputs from ocean-atmosphere coupled models only as required for a model’s consideration in the IPCC report (IPCC, 2013). Typical GCM outputs have spatial resolutions between 1.4 and 5 degrees (150 km² to 600 km²). While detailed in their global climate forecasts, GCMs are much less accurate in the prediction of regional climate change. Only in part this is due to their coarse spatial resolution. A further reason is the complexity of processes that are not yet fully understood.

5.1 General Circulation Models (GCM)

Different GCMs perform differently well in different areas. While temperatures are usually well captured by most models they often perform much poorer in the projection of future precipitation. The model outputs for a reference period (observed period), usually the period between 1990 and 2000, provide guidance for the performance of a distinct model in an area of interest. The outputs of three different GCMs were examined that included those from the American CCSM model, the Canadian CGSM model and the British HadGEM1. The CGSM data from the Canadian Climate Change Scenarios Network (CCCSN) best describe the two parameters, temperature and precipitation. An ensemble mean of the CGSM model is applied for the analysis.

5.2 Climate Change Scenarios

The international Panel on Climate Change (IPCC) has developed a set of emission scenarios that describe possible developments of different parameters with an impact on greenhouse gas concentrations. Parameters include among others population growth, economic development, and the introduction of new and more efficient technology (IPPC, 2007a; IPPC, 2007b). From these so-called story lines, this study analyzed the B1, the A1b and the A2 scenarios. Key elements of either story line are the following:

- The A1 story line is marked by a very rapid economic growth, a global population that will peak in mid-century and decline thereafter and a rapid introduction of new and more efficient technologies.
- The A2 story line describes a regionally fragmented economic growth, where new technologies are introduced much slower and only locally. Population growth is continuously increasing.
- The B1 story line has a population growth similar as in A1, with a quicker change in economic structures toward a service and information economy, with reductions in material intensity and the introduction of clean and resource-efficient technologies.

As starting climate conditions are unknown, model runs are typically initiated using different starting conditions (usually seven runs). An ensemble is the total number of model runs.
5.3 Data Preparation

Because of the coarse spatial (3.7 degrees) and temporal (monthly) resolution of the CGSM products, they have only little meaning at a local scale. Therefore both climate parameters are downscaled to 10-day intervals and 5 km (temperature) and 12 km (precipitation) respectively (WILBY et al., 1998). The process of downscaling relies on statistical relationships that are in case of temperature linkages between topographic elevation and temperature (lapse rate), in the case of precipitation observed precipitation distribution patterns that are applied to future change scenario data. This assumes that observed statistical relationships do not change over time.

For the downscaling of temperature, the 1 km resolving GMTED2010 data and a lapse rate of 5 °C are used. For precipitation, statistics of spatial and temporal patterns were derived from 12 km resolving RFE data (LAWS et al., 2004). For both parameters, data covering the period from 2000 to 2099 were processed.

5.4 Climate Change under Different Scenarios

Where data processing times and data evaluation were within a manageable amount, results are presented for all three scenarios (B1, A1b and A2, e.g. parameter temperature). However, where time consuming processing steps are required the focus of the study was on the B1 and the A2 scenario (rainfall, various modeled, hydro-meteorological parameters). Time intervals at which future change situations are presented may vary, depending on the magnitude of change that occurs between two points in time.

5.4.1 Temperature

The B1 story line represents the mildest temperature increase of the discussed scenarios. In a B1 scenario, annual average temperatures will increase until the end of the century by an approximate 2 °C, plus/minus 0.5 °C depending on the climate zone. A significant difference in temperature increase between different scenarios does not occur before the end of the 2030ies. Latest at the beginning of 2040 in an A1b and an A2 scenario, temperatures will start increasing faster resulting in a total temperature increase of 3 °C (A1b) and 4 °C (A2) by 2099.

It is worthwhile to mention that future temperature increases over the next 100 years are expected to be considerably lower than the increases observed between 1974 and 2013.

Temperature increase is highest in the arid (Figure 13, top) and lowest in the dry sub-humid zone (Figure 13, bottom). The high inter-annual temperature variation in the arid zone is related to inter-annual variations in cloud cover that are less pronounced in the semi-arid and dry sub-humid zone.

Increasing temperature will lift water losses through evapotranspiration. A rough approximation for assessing the water loss to evapotranspiration is as follows: Each 1 °C increase in temperature will increase evapotranspiration by about 0.4 mm per day. Assuming sufficient soil moisture will be available for about 120 days, the additional water loss amounts to 48 mm. However, the exact amount will further depend on variables such as vegetation cover, wind speed, air humidity and soils.

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According to CGSM data (CGCM3T47) from the Canadian Climate Change Scenarios Network (CCCSN). Note, color scale is not comparable with Figure 14 showing temperature changes between 1973 and 2013, due to large scale differences (~4°C/40years compared to ~2.4°C/100years).

Figure 15: Seasonal temperature change under an A1b scenario

Figure 16: Seasonal temperature change under an A2 scenario

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According to CGSM data (CGCM3T47) from the Canadian Climate Change Scenarios Network (CCCSN). Note, color scale is not comparable with Figure 14 showing temperature changes between 1973 and 2013, due to large scale differences (~4°C/40years compared to ~2.4°C/100years).
5.4.2 Precipitation

While temperature predictions under various change scenarios are relatively accurate, future changes in precipitation are not. The Canadian (CCGM) model describes average annual precipitation for the reference period in the region reasonably well; other models are either too high or too low. Still modeled rainfall from CCGM is somewhat high. Therefore, instead of using the absolute modeled rainfall amounts, relative rainfall increases were calculated and added to the observed rainfall average. Precipitation calculated this way was used for the calculation of various parameters such as aridity indices or in hydrological modeling of soil moisture and surface run-off.

All GCMs perform rather poorly in describing inter-annual precipitation variability. Observed data suggest a considerably higher inter-annual variability in comparison to modelled data (Figure 17). In contradiction to inter-annually very homogeneous model outputs, there is a general agreement among scientists that variability is very likely to increase under future climate change scenarios (IPCC, 2013). This view is primarily based on recent observations less so on model outputs. As mentioned above, the sea surface temperature (SST) in the Gulf of Guinea controls the rainfall in the region (Afiesimama et al., 2006; Cook and Vizy, 2006). The poor performance of GCMs in describing inter-annual variations is a result of the models’ weakness in modeling SST variation (Abraham et al., 2013; Cook, 1999, Dai, 2010; Hagos and Cook, 2007).

This weakness has effects not only on the modeling of inter-annual variations but also on the description of realistic seasonal rainfall variations such as the modeling of ‘extreme’ events. Different to observed rainfall, showing eventual outliers even based on monthly data, modeled climate shows normally distributed rainfall across the wet, without any remarkable peaks. However, the outliers are important for triggering floods that allow rivers to spill over into areas such as the Yaere-Naga.

In hydrological models, normally distributed rainfall in the headwaters may produce no run-off at all, because soils never saturate, which in response dramatically reduces river discharge. This may occur even if the annual total rainfall is higher than observed rainfall creating substantial run-off. Thus, temporal rainfall distribution and rainfall intensity play an important role in the partitioning of water between soil moisture and run-off (GERBAUX et al., 2009). Due to the models’ weakness in the description of variability, caution is necessary in the interpretation of climate impacts on the quantification of hydrological parameters including soil moisture, run-off, deeper infiltration and river discharge. Still, the total available water will be less.

Figure 17: Annual total precipitation in the arid (top) and semi-arid (bottom) zone

5.5 Impact of Climate Change on the Length of the Growing Period

For monitoring the impact of projected climate change on agricultural potentials (IMPACT ASSESSMENT INC., 2006), the parameter Aridity Index (AI) is used (Equation 1). The AI combines precipitation and temperature in one variable. Due to no rainfall and high temperatures throughout most of the year, an annual AI is little meaningful. Rainfall is concentrated on a few months during which it is crucial for how long the AI exceeds a threshold that is considered favorable for growing distinct crops (Table 2). Accordingly, the AI was calculated for every 10-day interval and the duration the AI exceeded dry sub-humid conditions was derived thereof. A minimum duration of 120 days under dry sub-humid conditions is required to grow crops as listed in Table 2. A mean 120-day line, constructed from AI values averaged over years 2000-2013, is used as a baseline against which future changes are measured.

The average 120-day line, well describes the beginning of intense, rainfed cultivation in the Basin. The line is solely based on temperature – or rather PET derived thereof - and the distribution of local precipitation. Where areas receive additional water from other sources, intense cultivation may be possible beyond this line, particularly in wetlands like the Yaere-Naga.
Massenya, Salamat, or around and in the former extents of Lake Chad.

Inter-annual shifts in the 120-day line do not occur as a North-South movement along the entire front but rather as local extensions (withdrawals), reflecting the spotty rainfall distribution. However, the 2001-2013 average is located close to the northern boundary of the ‘area of variation’, the entire hatched area in Figure 18 must be considered as vulnerable, where harvest losses may occur in case of droughts (PULAWARTY AND SIVAKUMAR, 2014).

The impact of climate change on the 120-day line under a B1 and an A2 scenario was calculated as a 10-year average and at intervals of ten years from 2010/2019 to 2090/2099. Until around 2030, there are no significant climate differences between the various scenarios. After 2030 temperatures under an A2 scenario will rise more quickly, while temperatures start tapering off at the beginning of 2060 under a B1 scenario.

By the end of the century, the positioning of the 120-day line under a B1 scenario shows a significant southward shift indicating deteriorating conditions particularly in the central basin. The Eastern and the Western areas remain less affected. Considering surface effects only (temperature and precipitation), the displacement of the 120-day line from its current position means that an area of approximately 70,960 km² will no longer be usable for growing more water demanding crops, but only for crops such as sorghum and millet (see Table 2). The situation for tree crops may be different as they are further controlled by groundwater and do not instantaneously respond to inter-annual rainfall variation. Generally, tree vegetation gradually increases to the south, e.g., as shown by continuously lower correlations between precipitation and biomass. The present situation of aquifers, including ground water levels/variations, flow and replenishment are still under investigation. A completion of this study will be a pre-condition for drawing meaningful conclusions on the impact of future climate change scenarios on groundwater.

Under an A2 scenario, an even larger area must be reclassified to a lower agricultural potential. Because of a retreat of the 120-day line, an area of about 135,150 km² will no longer fulfill the conditions for growing crops that are more valuable. A reduction of agricultural potentials of course does not only occur along the limits of the 120-day line but progressively causes an entire southward shift of potentials by a similar amount. An exact shift of the entire zone of variability (within which the 120-day line fluctuates) cannot be calculated from the GCM data, as models only poorly describe inter-annual rainfall variability. Although, current fluctuations of the 120-day line exceed the magnitude of retreat under both, a B1 and an A2 scenario, future fluctuations may even be larger and increase the risk for droughts.

A time-series of more detailed maps for a B1 and an A2 scenario, with further sub-divisions for the length of the growing period is presented in Chapter 7. The time-series visualizes spatially distributed effects of climate change on the Aridity Index. E.g., the transition from arid to hyper-arid conditions remains almost unaffected under either change scenario and has little impact on the location of the arid to semi-arid boundary.

Impacts of climate change tend to become continuously larger – causing larger displacements of boundaries – as one moves south, towards longer growing periods. Classes 201-203 (growing days) and 231-260, approximately covering areas with tropical conditions, become increasingly fragmented or disappear (231-260 class). Climate conditions supporting the growth of tropical forests may no longer exist in these areas at the end of the century.
Some boundaries float back and forth (e.g. 2070-2079, B1 scenario) what is caused by a slight increase in precipitation until mid-century.

Strongest negative offsets are visible in the displacement of the 120-day contour line between Maroua and Bongor including some additional areas east and west of these towns. Since the 120-day line, according to FAO definition, marks an important transition in crop suitability, these areas are likely to experience drastic changes in crop potential. They are best suited for interventions aiming at an introduction of alternative crops, more drought resistant seeds or an adapted crop management. Parts of this area fall into the projected intervention zone of the GIZ project.

As discussed above, agricultural potentials in other areas will likewise be reduced with effects on productivity but less on crop potential/suitability.

To better illustrate possible climate change impacts on vegetation covers, additional growing periods were identified that best describe boundaries separating distinct vegetation phenologies as classified from biomass time-series (NDVI) and shown in Figure 7, Figure 20 and Figure 21. A perfect match between the distribution of vegetation phenologies and growing period contours was not found, since the parameter ‘growing period’ only depends on precipitation and PET neglecting other relevant parameters including soils, topography, additional water resources, etc. Growing periods that best describe the transition between the interpreted phenology classes are the 20-, 60-, 120- and 170-day periods. These contours were applied to visualize the projected displacement of vegetation phenologies or ‘ecosystems’ under a B1 (Figure 20) and an A2 scenario (Figure 21).

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The Naga-Yaere Wetlands are a seasonally flooded area of roughly 7,500 km² and located within the semi-arid climate zone. Annual rainfall amount in the Yaere-Naga wetlands is between 600 and 800 mm. The vegetation of the wetlands shows growing periods that are significantly longer than in its surrounding areas.

With evapotranspiration rates of around 430 mm between July and September and more than 1,000 mm until the end of the year (annual total about 2,400 mm), local rainfall is far from being sufficient to support such growth periods. During summer time, spillovers of the Logone River into the wetlands, contribute an estimated equivalent of about 900 mm of rainfall to the Yaere-Naga wetlands (400 mm during dry years). These numbers result from evaluations of river discharge measurements, showing downstream discharge losses as the Logone passes the Yaere-Naga wetlands (Figure 22). River water discharges into the Yaere-Naga through natural pathways and through man-made canals constructed for fishing (canaux de pêches). The floodwater supports an extended growing period of (natural) vegetation cover that may be 3 to 4 months longer than in areas outside the Yaere-Naga.

While this potential is rarely exploited for extended farming activities, it represents a fodder resource for animals, for fishing and for replenishing groundwater. Larger irrigation schemes inside the Yaere-Naga such as those at Lake Maga and near Logone-gana do not depend on inundations. To sustain multiple crops these schemes draw water directly from the Maga reservoir or from the Logone River.

The Yaere-Naga owes its particular situation of seasonal flooding to its smooth morphology and discharge characteristics of the Logone River. Located at the former estuary of the Palaeo Logone into Lake Mega Chad (Figure 22), the flat morphology causes a slowdown of flow velocity, triggering the accumulation of river sediments and a spillover of river waters into adjacent areas.

Figure 22: Geomorphological setting of the Yaere-Naga wetlands

At the former estuary of the Logone River into Lake Mega Chad (former shore line of Lake Mega Chad shown as brown lines).
Depending on river discharge at Bongor the Logone river is losing water as far downstream as Logone-gana. The outflow into the Yaere wetlands (greyish area in Figure 23) between Katoa and Logone-gana starts, once discharge rates at Bongor exceed about 356 m³/sec, between Bongor and Katoa outflow starts when discharge rates exceed 1,055 m³/sec (Bongor station). As the diagram in Figure 23 shows, there is some return flow between Katoa and Logone-gana during November and December (crossover of discharge curves).

Figure 24 summarizes water losses of the Logone River between Bongor and Katoa with discharge rates at Bongor plotted against water losses to the Yaere between Bongor-Katoa and between Katoa-Logone

As discussed under 5.4.2, future rainfall characteristics (intensity and timing and spatial distribution), therefore are critical in determining future surface run-off and possible flooding scenarios of the Yaere-Naga wetlands. Rainfall intensity and timing are however, poorly described in general circulation models due to physical model limitations and their spatial (several 100 km²) and temporal resolution (monthly). The above can be approximated with some confidence for the total water losses to evaporation, at the expense of soil moisture, deeper infiltration and surface run-off, while long lasting or repeated events of low intensity may never reach a point of saturation, creating no run-off at all.

The model results however do not allow distinguishing between losses to soil moisture, deeper infiltration water and run-off. Though hydrologic modeling suggests dramatic run-off losses (see Chapter 7.5), these losses bear little credibility due to the weaknesses in GCM data. One has to assume, however that flooding of the Yaere-Naga wetlands is likely to happen only at a lesser extent, and even if flooding continues at the current level, increasing evaporation rates will shorten the period with abundant soil moisture. In order to understand the impact of these influences on the situation of the Yaere-Naga wetlands, a comparison of two vegetation classifications during a wet and a dry year was undertaken. The results of this analysis are presented in Figure 27.
larger river courses. Otherwise, it is only man-made canals extending into these areas. From the eastern banks of the Logone River a dense network of fishing canals, several kilometers in length, diverts water into the hinterland. No farming activities were detected in these areas. In contrast, along a branch of the Logone, intense farming activities occur but only few traces of fishing canals. Apart from accessibility, morphology and flooding intensity are the major controls for farming, fishing and settlements.

6.1 Lake Chad

Lake Chad receives most of its water from the Chari and Logone rivers with only minor contributions from the Komadugu-Yobe River. Figure 29 presents the annual total inflow from Chari and Logone rivers into Lake Chad. Ungauged inflow to the lake coming from e.g. the El Beid River and other intermittent flow paths fed by the Yaere-Naga flood plains, contribute an additional, unknown volume of water. A rising PET in the headwaters will similarly reduce the total water available, as described for the Yaere-Naga wetlands, with effects on river discharge. Total upstream water losses range between 4% and 10% of the currently available water. Again, a quantification of future discharge reduction is not possible, since GCM data do not allow reliable modeling of water partitioning among soil moisture, deeper infiltration and surface run-off.

People in the Yaere-Naga wetlands seem to be aware of this situation and concentrate their activities on these core areas of inundation (Figure 28). However, it needs to be mentioned that only few areas have been interpreted for the distribution of settlements and activities due to the limited availability of high-resolution satellite data.

Settlements and associated small-scale farming activities are only found near the course of main rivers (Logone and its branches). As satellite analyses suggest farmers grow one crop per year only. Moving away from the main river, settlements and farming activities disappear, unless they are accessible through larger river courses. Otherwise, it is only man-made canals extending into these areas. From the eastern banks of the Logone River a dense network of fishing canals, several kilometers in length, diverts water into the hinterland. No farming activities were detected in these areas. In contrast, along a branch of the Logone, intense farming activities occur but only few traces of fishing canals. Apart from accessibility, morphology and flooding intensity are the major controls for farming, fishing and settlements.

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Evaporation over the open water surface of the lake will cause further water losses that vary with lake surface area. Figure 30 shows the situation for a B1 and an A2 scenario, assuming evaporation from the current open water surface (left) and from a hypothetical water surface equaling the lake’s extent in 1963 (right).

The above quantities add to the current evaporation of about 5.2 km³ for the current lake extent and about 39 km³ for the 1963 extent. This is an evaporation increase of 5.2 % (B1) and 11.6 %, (A2) by 2099. Lake surfaces for this calculation are 2,142 km² and 15,753 km², respectively.

Considering current river discharge (~27 km³) and 2099 evaporative losses of up to 6 km³ will still allow to maintain the lake within its current boundaries, even under the assumption of additional losses e.g. through groundwater infiltration or spillovers into the northern pond. Evaporation from the lake surface as in 1963 however totals 42 to 45 km³ in 2099. These values thus leave the possibility of a restoration of the lake within its 1963 shorelines doubtful, even if modeled rainfall data are not entirely correct.

Since the shrinking of Lake Chad, people settled in elevated areas of the former lake, making a living on fishing and farming (Figure 31). Particularly people in the northern pond are at risk to be affected by the impacts of climate change. Spillovers into the northern pond are expected to be less frequent and intervals of abundant soil moisture will become shorter, however will still be long enough to sustain the growth of crops. Current green periods in the northern pond are of similar length as can be seen in the Yaere-Naga wetlands (Figure 27). It is expected that the flooded area in the northern pond will shrink, but there will still be locations receiving abundant water. Whether the spillover can be controlled and water be allocated to farmed areas is a matter for technical, political and social consideration.

Background is a mosaic of Landsat images from 2003, with vegetation displayed in green colors and water in dark blue to black. Brownish colors are sparsely or non-vegetated areas. Spillovers into the northern pond occur east of the east-west oriented barrier and through passages in the barrier itself. Interpretation of settlements is based on very high resolution satellite data.
7.1 Seasonal Temperature Change

Seasonal Temperature Change B1 Scenario (2040-2049)

Figure 32: Absolute seasonal temperature change between 2000-2009 - B1 scenario

(10-year average) and 2040-2049 (10-year average), under the assumption of a B1 scenario.
Figure 33: Absolute seasonal temperature change between 2000-2009 and 2040-2049 - A2 scenario

Figure 34: Absolute seasonal temperature change between 2000-2009 and 2070-2079 - B1 scenario

10-year average), under the assumption of an A2 scenario

10-year average), under the assumption of a B1 scenario
Seasonal Temperature Change A2 Scenario (2070-2079)

Figure 35: Absolute seasonal temperature change between 2000-2009 and 2070-2079 - A2 scenario

Seasonal Temperature Change B1 Scenario (2090-2099)

Figure 36: Absolute seasonal temperature change between 2000-2009 and 2090-2099 - B1 scenario

---

34 10-year average), under the assumption of an A2 scenario

35 10-year average), under the assumption of a B1 scenario
7.2 Spatial Downscaling of GCM Data

Because of the coarse spatial (3.7 deg.) and temporal (monthly) resolution of the CGSM products, they have only little meaning at a local scale. Therefore downscaled both climate parameters to 10-day intervals and 5 km (temperature) and 12 km (precipitation) respectively. The process of downscaling relies on statistical relationships that are in case of temperature linkages between topographic elevation and temperature (lapse rate), in the case of precipitation observed precipitation distribution patterns that are applied to future change scenario data. This assumes that observed statistical relationships do not change over time.

7.3 Spatial Downscaling, Temperatures

Because of the strong linkage between temperature and topographic elevation, a high-resolution digital elevation model is needed and a lapse rate that best describes the temperature – elevation relationship. For the downscaling of temperature, the 1-km resolving GMTED2010 data and a lapse rate of 5°C is used. In a first step, the topographic data (1 km GMTED2010) are degraded to the resolution of GCM temperatures where the resulting elevation is an average calculated from the high resolution pixels. Pixels of the GCM temperature file and of the degraded topography file need to align. Using elevations from the degraded topography, temperatures are projected to sea level using a lapse rate of 5 °C/1000 m. Sea level temperatures are (i) resampled to the desired resolution (e.g. the resolution of the topographic file) and (ii) projected back to their original elevation using the high-resolution topographic data.

7.4 Spatial Downscaling, Precipitation

The downscaling of precipitation is more complicated and the results less reliable as compared to downscaled temperatures. For redistribution of precipitation to a higher spatial resolution, statistics of spatial and temporal patterns were derived from 12 km resolving RFE data (Figure 39).
This procedure involved the following two steps:
1. Spatial downscaling from about 450 km² (1.7 deg) to 12 km².
2. Temporal downscaling from monthly to 10-day intervals.

For spatial downscaling the high-resolution “% precipitation distribution” pattern within the domain relative to the domain “mean precipitation” is determined. This is calculated from observed high-resolution precipitation data (RFI data), where the temporal resolution of the observed data is in accordance with resolution intended for the GCM data (here 10-day intervals):

\[ \frac{100 \text{ OBS annual precipitation}}{10 \text{ days} \times 10 \text{ days}} \times \text{GCM annual precipitation} = \% \text{ Distribution Pattern} \]

The % distribution pattern may be calculated individually for each observed year or as an average for all observed years. The % Distribution file can inversely be used to spatially downscale GCM data:

\[ \frac{\% \text{ Distribution Pattern}}{100} \times \text{GCM annual precipitation} = \text{GCM high spatial resolution} \]

The same approach was applied for temporal downscaling:

\[ \frac{10 \text{ days} \times 10 \text{ days} \times \text{OBS annual precipitation}}{100} = \% \text{ Distribution Pattern} \]

For both parameters, temperature and precipitation data covering the period from 2000 to 2099 were processed. For temperature data of the B1, A1b and A2 scenario were downscaled for precipitation only the B1 and the A2 scenario.

### 7.5 Hydrologic Modelling

#### 7.5.1 Potential Evapotranspiration

Gridded PET was calculated using a modified HARGREAVES method. In a first step Top of Atmosphere Radiation (TOA) was calculated considering topography (slope and aspect) from GMTED2010 data. TOA is calculated using the common equations that describe the earth’s orbit around the sun. The output forms an input into the HARGREAVES equation.

\[
\text{PET}_{\text{HARG}} = 0.0023 \times L_s \left[ (T_{\text{mean}} + 17.8) \right] \left[ (T_{\text{max}} - T_{\text{min}}) \right] \left[ \frac{238.8}{935.5 - 0.55 T_s} \right] \text{[mm/day]}
\]

Equation 2

with:
- \( L_s \): Top of Atmosphere Radiation (Extraterrestrial Radiation) [mm/day]
- \( T_{\text{mean}} \): Mean Temperature
- \( T_{\text{max}} \): Maximum Temperature
- \( T_{\text{min}} \): Minimum Temperature

Prior to insertion of Top of Atmosphere radiation (TOA) into equation 2, TOA must be converted into its water equivalent in mm:

\[
\text{Water Depth} = L_s \times \left[ \frac{238.8}{935.5 - 0.55 T_s} \right] \text{[mm]}
\]

Equation 3

Using minimum and maximum temperature accounts for variations in cloud coverage PETHARGREAVES values are linearly correlated to PETPENMAN, but with lower values for PETHARGREAVES. PETHARGREAVES was corrected to fit the more accurate PETPENMAN using a look-up curve, created from measured data recorded at a station, located in an arid climate.

#### 7.5.2 Equations used in hydrologic modelling

The general equation for calculating soil moisture (SM) is:

\[
\text{SM}(t) = P(t) + M(t)/0.4 - AET(t) - R(t)
\]

Equation 4

where precipitation (P), snow melt (M), actual evapotranspiration (AET) and run-off (R) are measured in mm. AET is assumed to be proportional to PET and the available soil moisture (SM) according to

\[
AET(t) = \frac{\text{PET}(t) \left( \frac{\text{SM}(t)}{\text{MSM}} \right)}{\text{MIN}}
\]

Equation 5

Equation 5 indicates that with constant PET, the AET will decrease as the soil becomes drier. The 10-day precipitation, snow melt, and potential evapotranspiration are assumed uniform during a 10-day interval. The time development of soil moisture can then be written as:

\[
\frac{d\text{SM}}{dt} = P - M - R - F + \text{SM}_{\text{EXP}}
\]

Equation 6

where
- \( F(T) = \frac{\text{PET}}{\text{MSM}} \)

Equation 7

is the factor for evaporation. The term “excess” removes water in excess of the maximum soil moisture (MSM) and becomes runoff. The soil moisture at the end of a 10-day period (SM{10}) is given by

\[
\text{SM}_{10} = \text{SM}_0 - (\text{SM}_0 - \text{SM}_{\text{EXP}}) \left( \text{F}^{-1} \right)
\]

Equation 8

where \( \text{SM}_0 \) is a steady soil moisture state,

\[
\text{SM}_{\text{EXP}} = \left( P + M / F \right)
\]

Equation 9

SM_{10} is the soil moisture at the beginning of the 10-day period, \( t \) marks the end of the time interval e.g. 10 for a period of 10 days. Values for \( P \) and \( \text{PET} \) in equation 6 and 7 depend on the time steps chosen: days, hours, seconds or other.

The amount of subsurface flow is calculated as a user defined percentage from the total soil moisture. If temperature drops below Tcrit, precipitation is assumed to accumulate as snow. If temperature is above Tcrit, accumulated snow is reduced by melting (M) using the equation of EGGLESTON and RILEY (1971):

\[
M = C_m \times (T_{\text{mean}} - T_{\text{crit}})
\]

Equation 10

Without considering vegetation cover as done in the original equation, the melt rate factor \( C_m \) [mm/°C per day] is determined by

\[
C_m = k_s \times R_i (1 - A)
\]

Equation 11

where \( k_s \) is a proportionally constant, \( R_i \) the solar radiation index, and \( A \) the snow albedo. Albedo changes with time \( t \) (in days) described by

\[
A = 0.4 \times \left( 1 + 0.2^t \right)
\]

Equation 12

where the time constant \( k_i \) is 0.2. \( R_i \) is the ratio between TOA radiation on a surface to TOA radiation on this surface if it was horizontal. For periods of rain \( C_m \) is adjusted according to

\[
C_m(\text{rain}) = C_m \times 0.00126 \times P
\]

Equation 13
where \( P \) is precipitation (mm). In addition to the excess water (Equation 3), the amount of precipitation that becomes runoff (R) is triggered by slope according to:

\[
R = C_{m} \times P
\]

Equation 14

Values for CRO range between 0.08 and 0.35 and depend on slope ranges as defined in Caltran’s Storm Water Quality Handbook (2003).

To calculate the time it takes surface run-off to reach a defined pour point we used the equations from Manning for sheet flow:

\[
T = \frac{0.007 \times n \times L^{1/3}}{(P)^{1/2}}
\]

Equation 15

where

- \( T \) = travel time [hr]
- \( n \) = Manning’s roughness coefficient
- \( L \) = flow length [ft]
- \( P \) = 2-year, 24-hour rainfall [in]
- \( s \) = slope of hydraulic grade line [ft/ft]

and for open channel flow

\[
V = \frac{1}{2} \times \frac{n \times L^{1/2}}{P^{1/2}}
\]

Equation 16

where

- \( V \) = Velocity
- \( K \) = 1 when measured in meters and seconds, 1.49 for feet and seconds
- \( S \) = channel slope
- \( n \) = Manning coefficient; differs for different materials
- \( A \) = Area of channel [ft²]
- \( P \) = Perimeter of channel [ft] (wetted perimeter)
- \( A/P \) = also called hydraulic radius \( R_h \)

Knowing pixel size and flow direction, the calculated velocities can be converted to travel times. The criterion for open channel is a minimum number of pixels that flow into a pixel, which is defined by the user.

Travel time is calculated for each pixel. Total travel time from each pixel to the pour point are calculated as described for the surface flow.

### 7.6 Classification of Vegetation Cycles (Ecosystems)

Biomas classifications of the Chad Basin and of the Yaere-Naga wetlands were implemented using an algorithm developed by authors of this study. Some of its advantages include (i) a subtle distinction of vegetation cycles of different shapes, (ii) the interpretation of vegetation covers based on their seasonal biomass variation (not just a single point in time), and (iii) the classifications’ spatial and temporal comparability, required for change detection.

A major element of the classification is the conversion of NDVI time series from temporal space to Fourier space using a Discrete Fourier Transform (DFT). Using a DFT, the shape of a complex, and periodic cycle like an NDVI cycle can be decomposed into its individual frequencies, where each frequency is described by its individual magnitude and phase. The DFT is written as:

\[
F(u) = \sum_{x=0}^{N-1} f(x) e^{iu2\pi x/T}
\]

Equation 17

where \( h \) is the head gradient given by terrain slope and \( l \) is flow distance given by pixel size and flow direction. Total travel times from each pixel to the pour point are calculated as described for the surface flow.

The calculated magnitudes and phases are the inputs for measuring shape similarity. By excluding the higher magnitudes and phases from the similarity calculation, the influence of high-frequency noise is eliminated from the shape identification process. The objective is to eliminate most of the noise, however, to preserve the diagnostic phenological features (Figure 42).

Through a Fourier transform and the elimination of the higher harmonics, a reduction of parameters is achieved which facilitates similarity measures. After the conversion from temporal space to Fourier space (magnitude and phase), the number of variables is reduced from 46 (for 16-day interval MODIS NDVI) to a total number of ten, if using the first five harmonics (Figure 43) or even six if just using the first three harmonics.

Following selection of a set of reference cycles, the algorithm searches for pixels with matching Fourier variables (magnitude and phase values). The challenge in shape classification is not so much to identify matching shapes but rather to allow for reasonable deviations from a perfect match due to natural shape variability and data noise. For this purpose, a set of controls is available to define individual accuracy thresholds.

\[
F_{\text{phasemag}} = \tan \left( \frac{F_{\text{phasemag}}}{F_{\text{phasemag}}} \right)
\]

Equation 22

The calculated magnitudes and phases are the inputs for measuring shape similarity. By excluding the higher magnitudes and phases from the similarity calculation, the influence of high-frequency noise is eliminated from the shape identification process. The objective is to eliminate most of the noise, however, to preserve the diagnostic phenological features (Figure 42).

Through a Fourier transform and the elimination of the higher harmonics, a reduction of parameters is achieved which facilitates similarity measures. After the conversion from temporal space to Fourier space (magnitude and phase), the number of variables is reduced from 46 (for 16-day interval MODIS NDVI) to a total number of ten, if using the first five harmonics (Figure 43) or even six if just using the first three harmonics.

Following selection of a set of reference cycles, the algorithm searches for pixels with matching Fourier variables (magnitude and phase values). The challenge in shape classification is not so much to identify matching shapes but rather to allow for reasonable deviations from a perfect match due to natural shape variability and data noise. For this purpose, a set of controls is available to define individual accuracy thresholds.

**Figure 42:** Fourier filtered biomass cycle (red curve; created through inverse Fourier transform) after removing noise (blue curve) from the original biomass cycle (green curve) using a discrete Fourier transform function.

**Figure 43:** Magnitudes (left) and phases (right) of a biomass cycle as shown in Figure 42.

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From the original 11 harmonics in the original cycle only harmonics 1 to 5 are preserved in the Fourier filtered cycle. The 23 intervals along the x-axis represent one year: The biomass cycle is double peaked, representing a double-cropped area.
7.7 Length of Growing Period

Length of Growing Period (2010-2019)

Figure 44: Average number of days an area exceeds dry sub-humid conditions in 2010-2019\textsuperscript{b}

\textsuperscript{a} Under the assumption of a B1 scenario (left) and an A2 scenario (right)

Length of Growing Period (2020-2029)

Figure 45: Average number of days an area exceeds dry sub-humid conditions in 2020-2029\textsuperscript{b}

Length of Growing Period (2030-2039)

Figure 46: Average number of days an area exceeds dry sub-humid conditions in 2030-2039\textsuperscript{b}

\textsuperscript{a} Under the assumption of a B1 scenario (left) and an A2 scenario (right)
Length of Growing Period (2040-2049)

Figure 47: Average number of days an area exceeds dry sub-humid conditions in 2040-2049.

Length of Growing Period (2050-2059)

Figure 48: Average number of days an area exceeds dry sub-humid conditions in 2050-2059.

Length of Growing Period (2060-2069)

Figure 49: Average number of days an area exceeds dry sub-humid conditions in 2060-2069.

Length of Growing Period (2070-2079)

Figure 50: Average number of days an area exceeds dry sub-humid conditions in 2070-2079.

Under the assumption of a B1 scenario (left) and an A2 scenario (right).

Under the assumption of a B1 scenario (left) and an A2 scenario (right).

Under the assumption of a B1 scenario (left) and an A2 scenario (right).

Under the assumption of a B1 scenario (left) and an A2 scenario (right).
Length of Growing Period (2080-2089)

Figure 51: Average number of days an area exceeds dry sub-humid conditions in 2080-2089.

Length of Growing Period (2090-2099)

Figure 52: Average number of days an area exceeds dry sub-humid conditions in 2090-2099.

Annex

7.8 PET Increase

Figure 53: PET Increase by 2010-2019 for a B1 (left) and an A2 Scenario (right).

Figure 54: PET Increase by 2020-2029 for a B1 (left) and an A2 Scenario (right).
Figure 55: PET Increase by 2030-2039 for a B1 (left) and an A2 Scenario (right)

Figure 56: PET Increase by 2040-2049 for a B1 (left) and an A2 Scenario (right)

Figure 57: PET Increase by 2050-2059 for a B1 (left) and an A2 Scenario (right)

Figure 58: PET Increase by 2060-2069 for a B1 (left) and an A2 Scenario (right)
Figure 59: PET Increase by 2070-2079 for a B1 (left) and an A2 Scenario (right)

Figure 60: PET Increase by 2080-2089 for a B1 (left) and an A2 Scenario (right)

Figure 61: PET Increase by 2090-2099 for a B1 (left) and an A2 Scenario (right)
7.9 Impacts of Rainfall Characteristics on Surface Run-off

Modelled rainfall is a normally distributed characteristic that repeats itself over the modeling period from 2000 to 2099. Even on a basis of 10-day interval or monthly basis observed data show outliers (between DOY 213 and DOY 243) that are crucial for the creation of run-off water. Despite a lower annual total rainfall (upper diagram), the outliers lead to substantially higher run-off amounts compared to the normally distributed modeled rainfall (lower diagram).


