AI/ML for Climate Workshop

International Livestock Research Institute (ILRI)

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ML-based Subseasonal & Seasonal Prediction

Interactive Learning



Click the Binder button above to launch an interactive Jupyter notebook for NumPy and Pandas climate data analysis!

Rainfall Data Acquisition and Processing

- Download daily CHIRPS Rainfall data from 1981–2024
- · Clip by Uganda
- Compute areal-weighted monthly mean (use pixel area or cos(lat) weights).
- Build seasonal totals:
 - OND = Oct+Nov+Dec
 - MAM = Mar+Apr+May
- · Climatology & anomalies
- Define a base pevarianceriod (WMO 1991–2020)
- Compute seasonal climatology and anomalies: anom = total clim

• Standardize: z = (anom / std) to stabilize

```
import os
import os, glob, math, re, io, warnings, requests, yaml
from typing import Dict, Tuple, List
import argparse
from pathlib import Path
import sys, os
import requests
import numpy as np
import pandas as pd
import xarray as xr
import geopandas as gpd
import regionmask
from shapely.geometry import mapping
from tqdm import tqdm
import matplotlib as mpl
import matplotlib.pyplot as plt
import cartopy.crs as ccrs
import cartopy.feature as cfeature
import matplotlib.ticker as mticker
from cartopy.mpl.ticker import LongitudeFormatter, LatitudeFormatter
xr.set_options(keep_attrs=True)
# Set working directory
os.chdir("C:\\Users\\yonas\\Documents\\ICPAC\\ea seasonal pred\\seasonal-ml-pred")
# Paths (matches your repo)
RAW DIR = "data/raw"
PROCESSED DIR = "data/processed"
EXTERNAL_DIR = "data/external/indices"
LOGS DIR = "reports/logs"
os.makedirs(RAW DIR, exist ok=True)
os.makedirs(PROCESSED DIR, exist ok=True)
os.makedirs(EXTERNAL DIR, exist ok=True)
os.makedirs(LOGS DIR, exist ok=True)
```

Download CHIRPS Daily Rainfall Data (1981–2024)

1) Download CHIRPS daily NetCDF (v2.0) by year range 2) Clip to a region 3) Merge all (clipped or raw) files into a single NetCDF

Source: "https://data.chc.ucsb.edu/products/CHIRPS-2.0/global_daily/netcdf/"

Spatial resolution: [p25=0.25°, p05=0.05°]

```
.....
Download CHIRPS daily NetCDF (v2.0) by year range, optionally clip to a region,
and merge all (clipped or raw) files into a single NetCDF saved in.
def download_file(url: str, dest: Path, chunk=2**20):
    dest.parent.mkdir(parents=True, exist ok=True)
    tmp = dest.with suffix(dest.suffix + ".part")
    with requests.get(url, stream=True, timeout=180) as r:
       r.raise_for_status()
       with open(tmp, "wb") as f:
            for blk in r.iter content(chunk size=chunk):
                if blk: f.write(blk)
    tmp.replace(dest)
def build url(year: int, res: str) -> str:
    base = f"https://data.chc.ucsb.edu/products/CHIRPS-2.0/global daily/netcdf/{res}"
    return f"{base}/chirps-v2.0.{year}.days {res}.nc"
def standardize for merge(ds):
   ren = {}
    if "latitude" in ds.dims: ren["latitude"] = "lat"
    if "longitude" in ds.dims: ren["longitude"] = "lon"
        ds = ds.rename(ren)
    try:
       lat = ds["lat"]
       if lat[0] > lat[-1]:
           ds = ds.reindex(lat=list(reversed(lat.values)))
    except Exception:
       pass
    return ds
def clip box(ds, N, S, W, E):
    import numpy as np
    # Validate latitude bounds
    if S \gg N:
        raise ValueError(f"Invalid latitude bounds: South ({S}) must be less than North
    # Work with original coordinate names first
    lat name = "latitude" if "latitude" in ds.dims else "lat"
    lon name = "longitude" if "longitude" in ds.dims else "lon"
    lat = ds[lat name].values
    lon = ds[lon name].values
    # For xarray.sel(), we need to provide bounds in the correct order
```

```
# CHIRPS latitude goes from south to north (-49.875 to 49.875)
    \# So we select from S (southern bound) to N (northern bound)
    lat slice = slice(S, N)
   lon min, lon max = float(lon.min()), float(lon.max())
   W2, E2 = W, E
    # Handle longitude wrapping if needed (CHIRPS uses -180 to 180)
    if lon_min \ge 0 and W < 0: # convert input -180..180 to 0..360, if needed
       W2 = (W + 360) \% 360
       E2 = (E + 360) \% 360
    # Create selection dictionary
    sel dict = {lat name: lat slice}
    if W2 <= E2:
       sel dict[lon name] = slice(W2, E2)
       ds_sub = ds.sel(sel_dict)
    else:
       # Handle longitude wrapping case
       left dict = {lat name: lat slice, lon name: slice(W2, lon max)}
       right_dict = {lat_name: lat_slice, lon_name: slice(lon_min, E2)}
       left = ds.sel(left dict)
        right = ds.sel(right dict)
       ds sub = type(ds).concat([left, right], dim=lon name)
    # Now standardize for merge
   ds sub = standardize for merge(ds sub)
    return ds sub
def merge_to_netcdf(nc_paths, out_path: Path):
   import xarray as xr
   if not nc paths:
        raise ValueError("No input files found to merge.")
   print(f"[merge] {len(nc paths)} files -> {out path.name}")
    ds = xr.open mfdataset(
        [str(p) for p in nc paths],
       combine="by coords",
       preprocess=standardize_for_merge,
       parallel=False,
   data vars = list(ds.data vars)
   if not data vars:
       raise ValueError("No data variables in opened datasets.")
    enc = {v: {"zlib": True, "complevel": 3} for v in data vars}
   out path.parent.mkdir(parents=True, exist ok=True)
   ds.to netcdf(out path, encoding=enc)
   print("[ok] merged saved:", out path)
def main():
   ap = argparse.ArgumentParser(description="Download CHIRPS daily v2.0 by year range;
   ap.add argument("--start", type=int, required=True, help="Start year (e.g., 2018)")
    ap.add argument("--end", type=int, required=True, help="End year (inclusive, e.g.,
    ap.add_argument("--outdir", default="chirps_downloads", help="Directory to save year
```

```
ap.add argument("--res", choices=["p25", "p05"], default="p25", help="Spatial resolu
ap.add argument("--clip", nargs=4, type=float, metavar=("N", "S", "W", "E"),
                help="Optional clip box (degrees): North South West East")
ap.add argument("--merge-name", type=str, default=None,
                help="Merged filename (no path). If omitted, an automatic name is \tau
ap.add argument("--overwrite", action="store true", help="Overwrite existing yearly
args = ap.parse args()
years = list(range(args.start, args.end + 1))
outdir = Path(args.outdir)
outdir.mkdir(parents=True, exist ok=True)
downloaded = []
clipped = []
for y in years:
   url = build url(y, args.res)
    raw nc = outdir / f"chirps-v2.0.{y}.days_{args.res}.nc"
    if not raw nc.exists() or args.overwrite:
       print(f"[GET] {url}")
        try:
            download file(url, raw nc)
            print(f"[ok ] saved {raw nc}")
        except Exception as e:
            print(f"[ERR] download failed for {y}: {e}")
            continue
    else:
        print(f"[skip] {raw nc} exists")
    downloaded.append(raw nc)
    if args.clip:
        N, S, W, E = args.clip
        out clip = raw nc.with name(raw nc.stem + " clip.nc")
        if not out clip.exists() or args.overwrite:
            try:
               import xarray as xr
                ds = xr.open dataset(raw nc)
               ds sub = clip box(ds, N, S, W, E)
                enc = {v: {"zlib": True, "complevel": 3} for v in ds sub.data vars}
                ds sub.to netcdf(out clip, encoding=enc)
                print(f"[ok ] clipped → {out clip}")
            except Exception as e:
                print(f"[warn] clip failed for {y} ({e}); skipping clip")
        else:
            print(f"[skip] {out clip} exists")
        if out clip.exists():
            clipped.append(out clip)
# Make merged filename inside outdir
if args.merge name:
   merge name = Path(args.merge name).name # drop any directory parts
else:
    suffix = " clip" if args.clip else ""
    merge_name = f"chirps_{args.res}_{years[0]}-{years[-1]}{suffix}.nc"
```

```
target = outdir / merge_name

# Merge if we have files
to_merge = clipped if args.clip else downloaded
to_merge = [p for p in to_merge if p.exists()]

if to_merge:
    try:
        merge_to_netcdf(to_merge, target)
    except Exception as e:
        print(f"[ERR] merge failed: {e}")
        sys.exit(2)

else:
    print("[warn] nothing to merge (no downloaded or clipped files).")

# if __name__ == "__main__":
# main()
```

```
ug_chirps81_25 = xr.open_dataset(RAW_DIR + "/chirps_p25_uganda/chirps_p25_1981-2024_clug_chirps81_25
```

```
<xarray.Dataset> Size: 58MB
Dimensions: (time: 16071, lat: 29, lon: 31)
Coordinates:
  * time
           (time) datetime64[ns] 129kB 1981-01-01 1981-01-02 ... 2024-12-31
  * lat
           (lat) float32 116B -2.375 -2.125 -1.875 ... 4.125 4.375 4.625
         (lon) float32 124B 28.62 28.88 29.12 29.38 ... 35.62 35.88 36.12
Data variables:
   precip (time, lat, lon) float32 58MB ...
Attributes: (12/15)
   Conventions:
                    CF-1.6
   title:
                     CHIRPS Version 2.0
   history:
                    created by Climate Hazards Group
   version:
                    Version 2.0
   date created:
                    2015-10-07
   creator name:
                    Pete Peterson
                    Funk, C.C., Peterson, P.J., Landsfeld, M.F., Pedreros,...
   reference:
   comments:
                     time variable denotes the first day of the given day.
   acknowledgements: The Climate Hazards Group InfraRed Precipitation with ...
                    ftp://chg-ftpout.geog.ucsb.edu/pub/org/chg/products/CH...
   ftp url:
   website:
                    http://chg.geog.ucsb.edu/data/chirps/index.html
    faq:
                     http://chg-wiki.geog.ucsb.edu/wiki/CHIRPS FAQ
```

Mean Annual Total Rainfall Map

```
# annual total
pr_ann_sum = ug_chirps81_25['precip'].resample(time="YS").sum()

# Then climatological mean (mm/year)
mean_annual = pr_ann_sum.mean("time") # (lat, lon)
```

```
# Plot with Cartopy
fig = plt.figure(figsize=(10, 6))
ax = fig.add subplot(1, 1, 1, projection=ccrs.PlateCarree())
# Add data
im = mean_annual.plot.imshow(
   transform=ccrs.PlateCarree(),
   cmap="YlGnBu",
   add colorbar=True,
    cbar kwargs={"label": "mm/year"}
# Add map features
ax.add feature(cfeature.COASTLINE)
ax.add_feature(cfeature.BORDERS, linestyle="-")
ax.add feature(cfeature.LAKES, alpha=0.9)
#ax.add feature(cfeature.RIVERS)
# Set title and labels
ax.set title("Mean annual total precipitation (mm/year), CHIRPS 1981-2024")
ax.set xlabel("Longitude")
ax.set_ylabel("Latitude")
# Set extent (optional, adjust as needed)
ax.set extent([mean annual.lon.min(), mean annual.lon.max(), mean annual.lat.min(), mea
plt.tight layout()
plt.show()
```

Output:

```
<Figure size 1000x600 with 2 Axes>
```

Monthly Totals

```
# MONTHLY total
```

```
pr_mon_sum = ug_chirps81_25['precip'].resample(time="MS").sum() # monthly total (mm)
```

```
mon clim = pr mon sum.groupby("time.month").mean("time") # (month, lat, lon)
# --- plotting config
proj = ccrs.PlateCarree()
extent = [float(mon clim.lon.min()), float(mon clim.lon.max()),
          float(mon_clim.lat.min()), float(mon_clim.lat.max())]
lon2d, lat2d = np.meshgrid(mon clim.lon, mon clim.lat)
# shared color scale
vmin, vmax = 0.0, float(mon clim.quantile(0.98).values) # or set vmax=400
norm = mpl.colors.Normalize(vmin=vmin, vmax=vmax)
cmap = "YlGnBu"
# layout: 3x4 panels + 1 slim colorbar column
fig = plt.figure(figsize=(13, 8), constrained layout=True)
gs = fig.add gridspec(nrows=3, ncols=5, width ratios=[1,1,1,1,0.045])
axes = [fig.add subplot(gs[r, c], projection=proj) for r in range(3) for c in range(4)]
cax = fig.add subplot(gs[:, 4])
month labels = ["Jan", "Feb", "Mar", "Apr", "May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov", "Dec"
def style axis(ax, left=False, bottom=False):
    ax.add feature(cfeature.OCEAN, zorder=100, edgecolor="0.2", facecolor="navy")
    ax.add feature(cfeature.BORDERS, edgecolor="black", linewidth=0.5)
    ax.add feature(cfeature.LAKES, edgecolor="black", facecolor="lightsteelblue", linew
    ax.coastlines(resolution="10m", linewidth=0.6)
    ax.set extent(extent, crs=proj)
    # --- THIS is the important change ---
    gl = ax.gridlines(crs=proj, draw labels=True,
                      linewidth=0.4, color="white", linestyle="--",
                      x inline=False, y inline=False)
    gl.top labels = False
    gl.right labels = False
    gl.left labels = left
    gl.bottom labels = bottom
    gl.xformatter = LongitudeFormatter()
    gl.yformatter = LatitudeFormatter()
    # set locators AFTER creating gridliner
    q1.xlocator = mticker.FixedLocator(np.arange(np.floor(extent[0]), np.ceil(extent[1])
    gl.ylocator = mticker.FixedLocator(np.arange(np.floor(extent[2]), np.ceil(extent[3])
mappables = []
for i, ax in enumerate(axes):
   month = i + 1
   da = mon clim.sel(month=month)
    im = ax.pcolormesh(lon2d, lat2d, da.values, transform=proj,
                       cmap=cmap, norm=norm, shading="auto")
    ax.set title(month labels[i], fontsize=11, pad=6)
```

```
style_axis(ax, left=(i % 4 == 0), bottom=(i // 4 == 2))
mappables.append(im)

# single colorbar with units
cb = fig.colorbar(mappables[0], cax=cax)
cb.set_label("mm / month", fontsize=11)

fig.suptitle("CHIRPS Monthly Precipitation Climatology (mm/month) 1981 - 2024", fontsiz
plt.show()
```

```
<Figure size 1300x800 with 13 Axes>
```

```
def _std_dims(ds: xr.Dataset) -> xr.Dataset:
    rename = {}
    if "latitude" in ds.dims: rename["latitude"] = "lat"
    if "longitude" in ds.dims: rename["longitude"] = "lon"
    if rename:
        ds = ds.rename(rename)
    # shift lon to [-180,180] if needed
    if float(ds.lon.max()) > 180:
        ds = ds.assign_coords(lon=(((ds.lon + 180) % 360) - 180)).sortby("lon")
    return ds

def _get_varname(ds: xr.Dataset) -> str:
    for cand in ["precip", "precipitation", "pr", "rain"]:
        if cand in ds.data_vars:
            return cand
    return list(ds.data_vars)[0]
```

```
ds = _std_dims(ug_chirps81_25)
ds
```

```
<xarray.Dataset> Size: 58MB
Dimensions: (time: 16071, lat: 29, lon: 31)
Coordinates:
           (time) datetime64[ns] 129kB 1981-01-01 1981-01-02 ... 2024-12-31
  * time
            (lat) float32 116B -2.375 -2.125 -1.875 ... 4.125 4.375 4.625
  * lat
            (lon) float32 124B 28.62 28.88 29.12 29.38 ... 35.62 35.88 36.12
Data variables:
   precip (time, lat, lon) float32 58MB ...
Attributes: (12/15)
   Conventions: CF-1.6
   title:
                     CHIRPS Version 2.0
   history:
                    created by Climate Hazards Group
```

```
version:
          Version 2.0
                2015-10-07
date_created:
creator name:
                Pete Peterson
                 Funk, C.C., Peterson, P.J., Landsfeld, M.F., Pedreros,...
reference:
                  time variable denotes the first day of the given day.
comments:
acknowledgements: The Climate Hazards Group InfraRed Precipitation with ...
ftp url:
                 ftp://chg-ftpout.geog.ucsb.edu/pub/org/chg/products/CH...
                 http://chg.geog.ucsb.edu/data/chirps/index.html
website:
faq:
                  http://chg-wiki.geog.ucsb.edu/wiki/CHIRPS FAQ
```

```
var = _get_varname(ds)
var
```

```
'precip'
```

```
def uganda polygon(crs: str = "EPSG:4326") -> gpd.GeoDataFrame:
   Return Uganda admin-0 polygon in requested CRS.
   Tries Cartopy Natural Earth; falls back to NACIS CDN zip.
   world = None
   try:
       import cartopy.io.shapereader as shpreader
       shpfile = shpreader.natural earth(resolution="110m", category="cultural", name=
       world = gpd.read file(shpfile).to crs(crs)
   except Exception:
       url = "https://naciscdn.org/naturalearth/110m/cultural/ne 110m admin 0 countrie
       world = gpd.read file(url).to crs(crs)
    # match by any of the typical name fields
   name_fields = [c for c in ["ADMIN", "SOVEREIGNT", "NAME", "NAME LONG"] if c in worl
   if not name fields:
       raise RuntimeError("No country name fields in Natural Earth schema.")
   mask = None
   for col in name fields:
       m = world[col].str.casefold() == "uganda"
       mask = m if mask is None else (mask | m)
   ug = world.loc[mask].copy()
   if ug.empty:
       raise RuntimeError("Uganda polygon not found.")
    return ug
```

```
def ensure_lon_minus180_180(ds: xr.Dataset, lon_name: str = "lon") -> xr.Dataset:
    """Shift lon from [0,360) to [-180,180) if needed."""
    if float(ds[lon_name].max()) > 180:
```

```
ds = ds.assign coords({lon name: (((ds[lon name] + 180) % 360) - 180)}).sortby(
    return ds
def uganda_mask(ds: xr.Dataset, uganda_gdf: gpd.GeoDataFrame,
               lat name: str = "lat", lon name: str = "lon") -> xr.DataArray:
   Build a boolean mask (lat, lon) marking cells inside Uganda.
   Uses regionmask.from geopandas to avoid outline-shape errors.
    # 1) make sure lon ranges are compatible with Natural Earth (-180..180)
   ds = ensure lon minus180 180 (ds, lon name=lon name)
    # 2) dissolve to a single (Multi)Polygon and fix minor topology issues
   gdf = uganda gdf.to crs("EPSG:4326").copy()
   gdf = gdf.dissolve() # merge multipart shapes into one row
    gdf = gdf.reset index(drop=True)
    # buffer(0) fixes self-intersections if present
    gdf["geometry"] = gdf.buffer(0)
    # 3) regionmask directly from GeoDataFrame
    gdf["name"] = "Uganda"
    regs = regionmask.from_geopandas(gdf[["name", "geometry"]], names="name", name="adm
    # 4) mask on ds grid: inside=0, outside=NaN -> convert to boolean
    rm = regs.mask(ds, lon name=lon name, lat name=lat name) # DataArray (lat, lon)
   inside = rm.notnull()
    inside.name = "uganda_mask"
   return inside
```

```
uga = uganda_polygon()
uga
```

```
featurecla scalerank LABELRANK SOVEREIGNT SOV_A3 ADMO_DIF LEVEL \
168 Admin-0 country 1 3 Uganda UGA 0 2

TYPE TLC ADMIN ... FCLASS_TR FCLASS_ID FCLASS_PL \
168 Sovereign country 1 Uganda ... None None None

FCLASS_GR FCLASS_IT FCLASS_NL FCLASS_SE FCLASS_BD FCLASS_UA \
168 None None None None None None

geometry
168 POLYGON ((33.90371 -0.95, 31.86617 -1.02736, 3...)

[1 rows x 169 columns]
```

```
def ensure_lon_minus180_180(ds: xr.Dataset, lon_name: str = "lon") \rightarrow xr.Dataset: """Shift longitudes from [0,360) \rightarrow [-180,180) if needed."""
```

```
if float(ds[lon_name].max()) > 180:
        ds = ds.assign\_coords(\{lon\_name: (((ds[lon\_name] + 180) % 360) - 180)\}).sortby(
    return ds
def uganda mask (
   ds: xr.Dataset,
    uganda_gdf: gpd.GeoDataFrame,
    lat name: str = "lat",
    lon_name: str = "lon",
) -> xr.DataArray:
    11 11 11
    Boolean mask (lat, lon) for cells inside Uganda.
    Handles regionmask versions with/without lon_name/lat_name arguments.
    # 1) Ensure lon in [-180, 180]
    ds = ensure lon minus180 180 (ds, lon name=lon name)
    # 2) Single clean (Multi) Polygon
    gdf = uganda gdf.to crs("EPSG:4326").copy()
    gdf = gdf.dissolve().reset index(drop=True)
    gdf["geometry"] = gdf.buffer(0) # fix tiny topology issues
    gdf["name"] = "Uganda"
    # 3) Build regions and mask
    regs = regionmask.from geopandas(gdf[["name", "geometry"]], names="name", name="adm
    try:
        # Newer regionmask: supports lon name/lat name with Dataset
        rm = regs.mask(ds, lon name=lon name, lat name=lat name)
    except TypeError:
       # Older regionmask: pass arrays/coords directly
        lon = ds[lon name]
       lat = ds[lat name]
        rm = regs.mask(lon, lat)  # returns a NumPy array or DataArray depending on ver
        # If it's NumPy, wrap into DataArray
        if not isinstance(rm, xr.DataArray):
           # dims from coords
           lat dim = lat.dims[0]
            lon dim = lon.dims[0]
            rm = xr.DataArray(rm, coords={lat dim: lat, lon dim: lon}, dims=(lat dim, l
    \# regionmask returns region numbers inside, NaN outside \rightarrow convert to boolean
    inside = rm.notnull()
    inside.name = "uganda mask"
    return inside
```

```
mask = uganda_mask(ds, uga)
print(mask.shape, mask.dtype, mask.name)
```

```
(29, 31) bool uganda_mask
```

```
# plot uganda_mask
plt.figure(figsize=(8, 6))
mask.plot(cmap="Set1")
plt.title("Uganda Mask (True=inside, False=outside)")
plt.show()
```

```
<Figure size 800x600 with 2 Axes>
```

```
# Plot the data within the mask
# Create subplots for comparison - 1 row, 3 columns
fig, axes = plt.subplots(1, 3, figsize=(18, 6), subplot_kw={'projection': ccrs.PlateCar
# 1. Original mean annual precipitation (Full Domain)
ax1 = axes[0]
mean annual.plot.imshow(
    ax=ax1, transform=ccrs.PlateCarree(), cmap="YlGnBu",
    add colorbar=True, cbar kwargs={"label": "mm/year", "shrink": 0.8}
ax1.add feature(cfeature.COASTLINE)
ax1.add_feature(cfeature.BORDERS, linestyle="-")
ax1.set title("Mean Annual Precipitation\n(Full Domain)", fontsize=12)
ax1.set extent([ds.lon.min(), ds.lon.max(), ds.lat.min(), ds.lat.max()], crs=ccrs.Plate
# 2. Uganda mask
ax2 = axes[1]
mask.plot.imshow(ax=ax2, transform=ccrs.PlateCarree(), cmap="RdYlBu",
                add colorbar=True, cbar kwargs={"label": "Uganda Mask", "shrink": 0.8}
ax2.add feature(cfeature.COASTLINE)
ax2.add feature(cfeature.BORDERS, linestyle="-")
ax2.set title("Uganda Mask\n(True=inside, False=outside)", fontsize=12)
ax2.set extent([ds.lon.min(), ds.lon.max(), ds.lat.min(), ds.lat.max()], crs=ccrs.Plate
# 3. Masked precipitation (Uganda only)
masked precip = mean annual.where(mask)
ax3 = axes[2]
masked precip.plot.imshow(
    ax=ax3, transform=ccrs.PlateCarree(), cmap="YlGnBu",
    add colorbar=True, cbar kwargs={"label": "mm/year", "shrink": 0.8}
ax3.add feature(cfeature.COASTLINE)
ax3.add feature(cfeature.BORDERS, linestyle="-")
ax3.set title("Mean Annual Precipitation\n(Uganda Only)", fontsize=12)
ax3.set extent([ds.lon.min(), ds.lon.max(), ds.lat.min(), ds.lat.max()], crs=ccrs.Plate
```

```
plt.tight_layout()
plt.show()

# Print some statistics
print(f"Grid cells in Uganda: {mask.sum().values}")
```

```
<Figure size 1800x600 with 6 Axes>
Grid cells in Uganda: 325
```

```
def area_weights(lat: xr.DataArray) -> xr.DataArray:
    return xr.DataArray(np.cos(np.deg2rad(lat)), coords={"lat": lat}, dims=["lat"])

def uganda_area_mean_monthly(ds: xr.Dataset, var: str, mask: xr.DataArray) -> pd.Series
    w_lat = area_weights(ds["lat"])
    w2d = w_lat.broadcast_like(mask)
    eff_w = (w2d.where(mask, 0.0))
    eff_w = eff_w / eff_w.sum(("lat", "lon"))
    monthly = (ds[var] * eff_w).sum(("lat", "lon"))
    s = monthly.to_series().sort_index()
    s.name = "rf_mm"
    s.index.name = "time"
    return s
```

```
monthly_mean = uganda_area_mean_monthly(ds, var, mask)
monthly_mean.head()
```

Output:

```
monthly_path = os.path.join(PROCESSED_DIR, "uganda_monthly_chirps_1981_2024.csv")
monthly_mean.to_csv(monthly_path)
```

Build Seasonal Totals:

- OND = Oct+Nov+Dec
- MAM = Mar+Apr+May

```
def monthly to seasonal total(monthly: pd.Series, season: str) -> pd.Series:
    if season == "OND":
       months = [10, 11, 12]
    elif season == "MAM":
        months = [3, 4, 5]
    else:
       raise ValueError("season must be 'OND' or 'MAM'")
    df = monthly.to frame("rf").dropna()
   df["year"] = df.index.year
    df["mon"] = df.index.month
    sel = df[df["mon"].isin(months)]
    tot = sel.groupby("year")["rf"].sum()
    tot.index.name = "season year"
    return tot
ond = monthly_to_seasonal_total(monthly_mean, "OND")
mam = monthly_to_seasonal_total(monthly_mean, "MAM")
ond df = ond.to frame("OND total mm")
mam_df = mam.to_frame("MAM_total_mm")
ond path csv = os.path.join(PROCESSED DIR, "kenya OND 1981 2024.csv")
mam path csv = os.path.join(PROCESSED DIR, "kenya MAM 1981 2024.csv")
ond_df.to_csv(ond_path_csv)
mam_df.to_csv(mam_path_csv)
ond_df.head(), mam_df.head()
```

```
OND_total_mm
season year
1981 251.316055
1982
           373.333588
1983
           269.184723
           294.980530
1984
1985
           262.542297,
         MAM total mm
season_year
         459.408356
1981
1982
           435.969727
1983
           354.696686
1984
           312.646851
1985
           479.866699)
```

Calculate Climatology (1991–2020), Anomaly (mm), & **Standardized Anomally**

- · Climatology & anomalies
- Define a base period (WMO 1991–2020)
- Compute seasonal climatology and anomalies: anom = total clim

```
• Standardize: z = (anom / std) to stabilize variance
def climatology and anoms(seasonal: pd.Series, base: Tuple[int,int]):
   base mask = (seasonal.index >= base[0]) & (seasonal.index <= base[1])</pre>
   clim mean = seasonal[base mask].mean()
   clim std = seasonal[base_mask].std(ddof=1)
   anom_mm = seasonal - clim_mean
   anom std = (seasonal - clim mean) / clim std if clim std and not math.isclose(clim
    # broadcast clim to all years (useful for adding back later)
    clim_series = pd.Series(clim_mean, index=seasonal.index)
    return clim_series, anom_mm, anom_std
START YEAR, END YEAR = 1981, 2024
CLIM START, CLIM END = 1991, 2020
ond clim, ond anom mm, ond anom std = climatology and anoms(ond, (CLIM START, CLIM END)
mam clim, mam anom mm, mam anom std = climatology and anoms(mam, (CLIM START, CLIM END)
OND out = pd.DataFrame({
   "season year": ond.index,
    "OND total mm": ond.values,
    "OND clim1991 2020 mm": ond clim.values,
```

```
"OND anom mm": ond_anom_mm.values,
    "OND anom std": ond anom std.values
}).set index("season year")
MAM out = pd.DataFrame({
    "season year": mam.index,
    "MAM total mm": mam.values,
    "MAM clim1991 2020 mm": mam clim.values,
    "MAM anom mm": mam anom mm.values,
    "MAM anom std": mam anom std.values
}).set index("season year")
```

```
OND_out.head()
```

```
OND_total_mm OND_clim1991_2020_mm OND_anom_mm OND_anom_std
season year
1981
           251.316055
                              328.404297 -77.088242 -1.053309
1982
           373.333588
                              328.404297 44.929291
                                                     0.613899
1983
           269.184723
                              328.404297 -59.219574
                                                    -0.809157
                             328.404297 -33.423767
1984
           294.980530
                                                    -0.456691
          262.542297
                              328.404297 -65.862000
                                                    -0.899917
1985
```

```
MAM_out.head()
```

Output:

```
MAM total mm MAM clim1991 2020 mm MAM anom mm MAM anom std
season_year
1981 459.408356
                            416.463226 42.945129 0.838471
1982
          435.969727
                            416.463226 19.506500
                                                   0.380850
          354.696686
1983
                            416.463226 -61.766541
                                                  -1.205945
1984
                            416.463226 -103.816376
          312.646851
                                                  -2.026936
1985
          479.866699
                            416.463226 63.403473 1.237905
```

```
OND_out.to_csv(os.path.join(PROCESSED_DIR, "uganda_OND_1981_2024.csv"))
MAM_out.to_csv(os.path.join(PROCESSED_DIR, "uganda_MAM_1981_2024.csv"))
```

```
def missing_report(df: pd.DataFrame, title: str = ""):
    col_na = df.isna().sum().sort_values(ascending=False)
    col_rate = (df.isna().mean()*100).round(1).sort_values(ascending=False)
    row_na = df.isna().sum(axis=1)
    row_rate = (df.isna().mean(axis=1)*100).round(1)
    print(f"=== Missingness report: {title} ===")
    print(f"Rows: {len(df)} | Cols: {df.shape[1]}")
    print(f"Rows with ≥1 missing: {int((row_na>0).sum())} ({(row_na>0).mean()*100:.1f}%
    print("\nTop 10 columns by % missing:")
    display(col_rate.head(10).to_frame("% missing"))
    print("\nYears with any missing (first 10):")
    display(row_rate[row_rate>0].head(10).to_frame("% missing in row"))
```

```
missing_report(OND_out, "OND 1981-2024")
```

```
missing_report(MAM_out, "MAM 1981-2024")
```

```
=== Missingness report: MAM 1981-2024 ===
Rows: 44 | Cols: 4
Rows with \geq 1 missing: 0 (0.0%)
Top 10 columns by % missing:
                   % missing
MAM_total_mm
                     0.0
MAM_clim1991_2020_mm
                         0.0
MAM_anom_mm
                         0.0
MAM anom std
                         0.0
Years with any missing (first 10):
Empty DataFrame
Columns: [% missing in row]
Index: []
```

```
def plot_missingness(df: pd.DataFrame, title: str):
    m = df.isna().astype(int)
    fig, ax = plt.subplots(figsize=(10, 4))
    ax.imshow(m.values, aspect="auto", interpolation="nearest")
    ax.set_title(title)
    ax.set_xlabel("features")
    ax.set_ylabel("season_year (row)")
```

```
ax.set_yticks([]); ax.set_xticks([])
plt.show()

plot_missingness(OND_out, "Missingness Heatmap: OND 1981-2024")

Output:

<Figure size 1000x400 with 1 Axes>

plot_missingness(MAM_out, "Missingness Heatmap: MAM 1981-2024")

Output:

<Figure size 1000x400 with 1 Axes>
```

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