

# HWRF

March 12, 2022

## 1 Local testing

I copied a nc file and track file to this directory.

```
[1]: import numpy as np
import xarray as xr

import pandas as pd

from matplotlib import pyplot as plt
%config InlineBackend.figure_format = 'retina'
%matplotlib inline
```

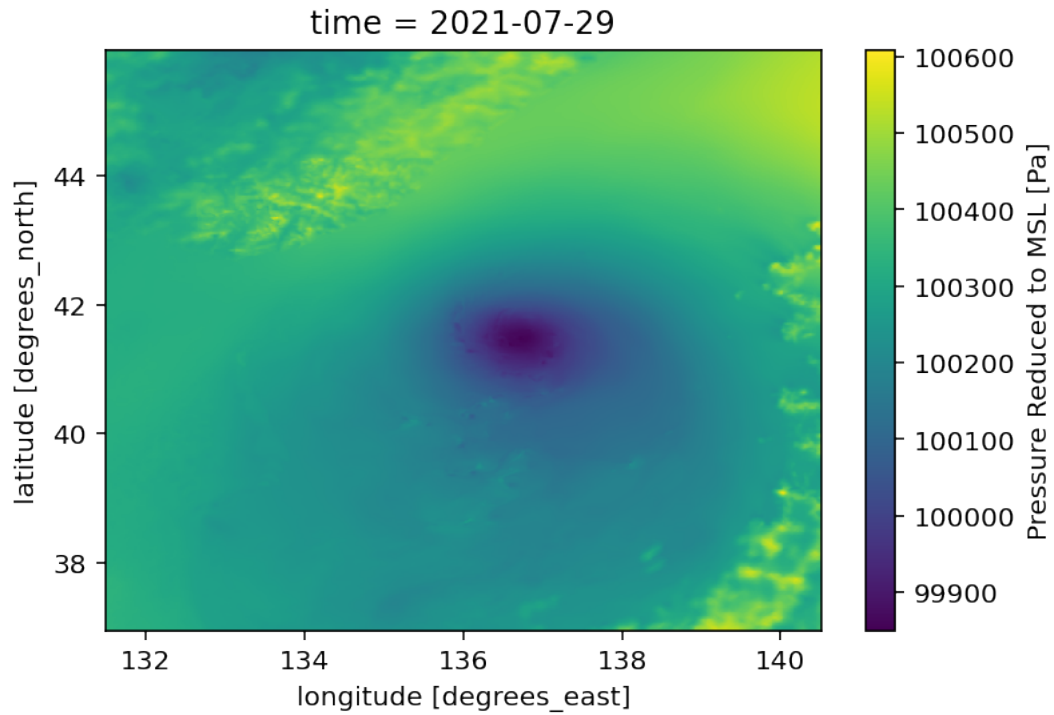
## 2 Read one forecast file

```
[2]: #ds = xr.open_dataset('/data2/qy2216/HWRF_wind/hwrf.2021072718/nepartak11w/nc/
→nepartak11w.2021072718.f030.nc')
ds = xr.open_dataset('nepartak11w.2021072718.f030.nc')
ds
```

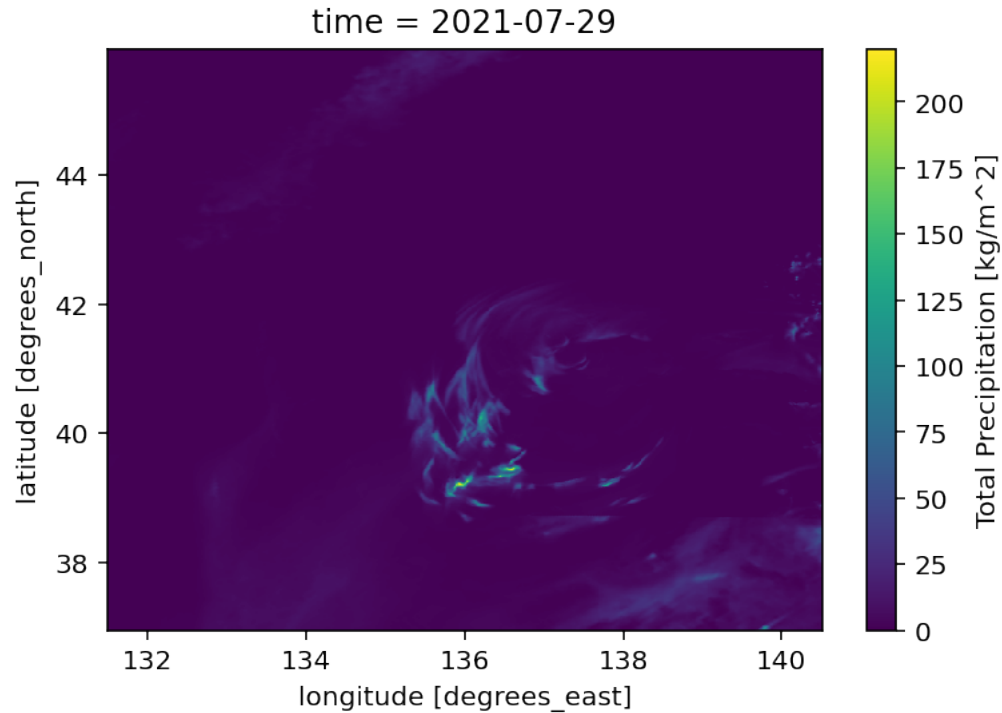
```
[2]: <xarray.Dataset>
Dimensions:                (latitude: 601, longitude: 601, time: 1)
Coordinates:
  * latitude                (latitude) float64 36.95 36.97 36.98 ... 45.94 45.95
  * longitude               (longitude) float64 131.5 131.5 131.5 ... 140.5 140.5
  * time                    (time) datetime64[ns] 2021-07-29
Data variables:
  PRMSL_meansealevel        (time, latitude, longitude) float32 ...
  PRES_surface              (time, latitude, longitude) float32 ...
  UGRD_10maboveground        (time, latitude, longitude) float32 ...
  VGRD_10maboveground        (time, latitude, longitude) float32 ...
  PRATE_surface              (time, latitude, longitude) float32 ...
  APCP_surface              (time, latitude, longitude) float32 ...
  ACPCP_surface              (time, latitude, longitude) float32 ...
  NCPCP_surface              (time, latitude, longitude) float32 ...
Attributes:
  Conventions:              COARDS
```

History: created by wgrib2  
GRIB2\_grid\_template: 0

```
[3]: ds.PRMSL_meansealevel.plot();
```



```
[4]: ds.APCP_surface.plot();
```



### 3 Read tpc file

With center etc

0	1	2	3	4	5	6	7	8
9	0							

012345678901234567890123456789012345678901234567890123456789012345678901234567890123

HOUR: 0.0 LONG: 142.00 LAT: 37.80 MIN PRES (hPa): 985.00 MAX SURF WIND  
(KNOTS): 41.00

```
[5]: import pandas as pd
colspecs = [(5, 10), (18, 25), (32, 38), (56, 63), (89, 95)]
# df = pd.read_fwf('/data2/qy2216/HWRF_wind/hwrf.2021072718/nepartak11w/track/
↳ NEPARTAK11W.2021072718.stats.tpc',
df = pd.read_fwf('NEPARTAK11W.2021072718.stats.tpc',
                 colspecs=colspecs, header=None, skiprows=12, names=['hour',
↳ 'lon', 'lat', 'P', 'V'])
df = df[:41]
df.head()
```

```
[5]:    hour    lon    lat    P    V
     0    0.  142.0  37.8  985.0  41.0
```

```

1   3.  141.8  38.4  992.0  43.0
2   6.  141.6  38.9  994.0  39.0
3   9.  141.2  39.7  997.0  34.0
4  12.  140.7  40.4  997.0  29.0

```

```

[6]: # they are strings, convert to float
df['hour'] = df['hour'].astype(float)
df['lat'] = df['lat'].astype(float)
df['lon'] = df['lon'].astype(float)

xcenter = df[df.hour == 30].lon.values[0]
ycenter = df[df.hour == 30].lat.values[0]

print(xcenter, ycenter)

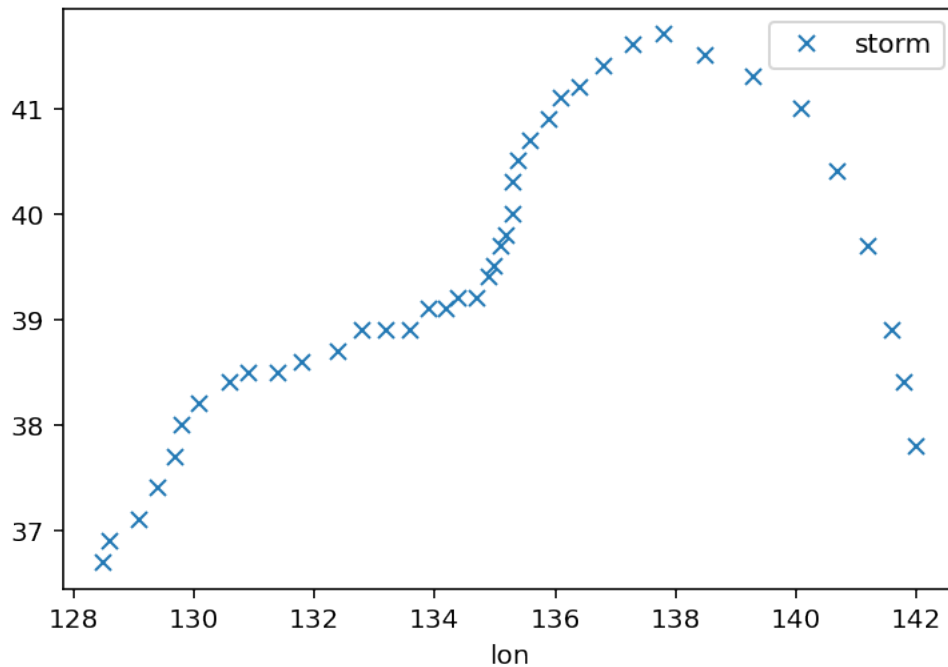
```

```
136.8 41.4
```

```

[7]: df.plot(x='lon', y='lat', style='x', label='storm')
plt.show()

```



## 4 Start plotting

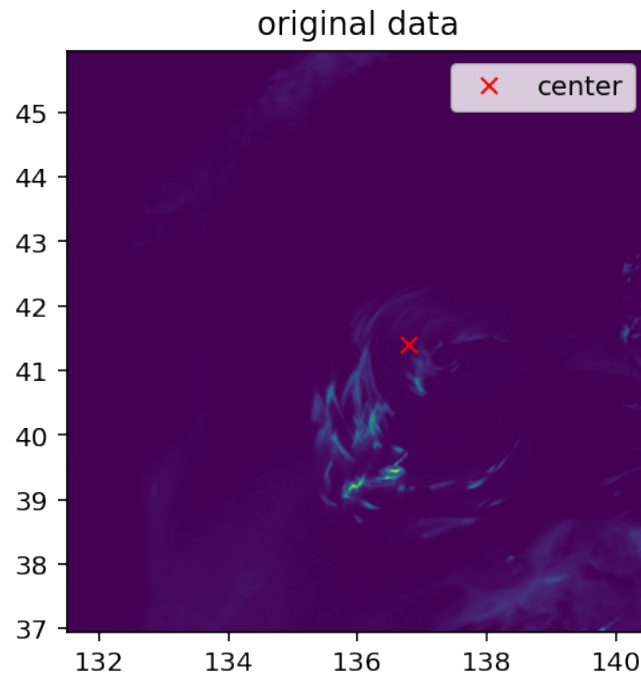
Strategy 1. Define storm-centered polar coordinate grid 2. Convert lat-lon to km - Use cosine of center lat value - Use cosine of lat values <- sensible - Use catopy

## 4.1 Original (lat-lon)

```
[8]: #z = ds.PRMSL_meansealevel.isel(time=0).values
z = ds.APCP_surface.isel(time=0).values

lon = ds.longitude.values
lat = ds.latitude.values

plt.pcolormesh(lon, lat, z, shading='gouraud')
plt.title('original data')
plt.gca().set_aspect('equal')
plt.plot(xcenter, ycenter, 'rx', label='center')
plt.gca().set_aspect('equal')
plt.legend()
plt.show()
```



## 4.2 Storm-centered polar coordinate grid

```
[9]: # new polar coordinates that are uniform in r and theta
# units of r are km

ntheta_pts = 201
nr_pts = 200
rmax = 300
```

```

#FIX rmax name
r1 = np.linspace(0, rmax, nr_pts)
theta1 = np.linspace(-np.pi, np.pi, ntheta_pts)

theta2, r2 = np.meshgrid(theta1, r1)

# cartesian coordinates (in km) of the polar coordinates
polar_x2 = r2 * np.cos(theta2)
polar_y2 = r2 * np.sin(theta2)
polar_x2.shape

```

[9]: (200, 201)

### 4.3 Cosine lat weighting (center value)

```
[10]: np.cos(np.deg2rad(ycenter))
```

[10]: 0.7501110696304596

```

[11]: # uses one cos lat value for scaling
# units are km
x1 = (ds.longitude.values - xcenter) * np.cos(np.deg2rad(ycenter)) * 110.567
y1 = (ds.latitude.values - ycenter) * 110.567
x2, y2 = np.meshgrid(x1, y1)

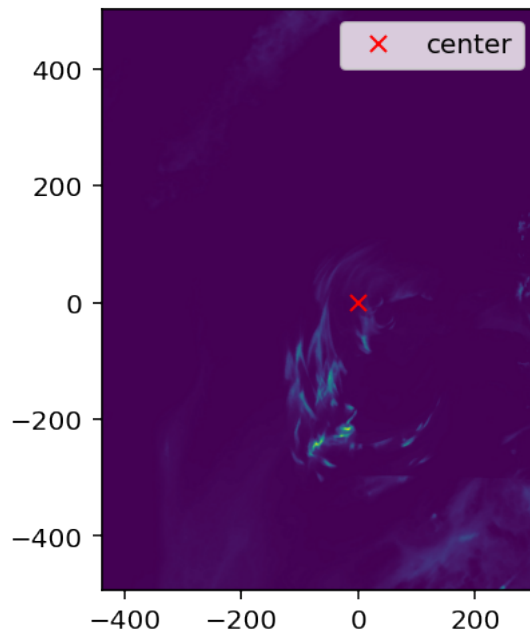
# number of points X 2 matrix of x y
xy = np.stack((x2.flatten(), y2.flatten())).T
print('number of grid points', xy.shape, 601 * 601)

plt.pcolormesh(x1, y1, z, shading='gouraud')
plt.plot(0, 0, 'rx', label='center')
plt.gca().set_aspect('equal')
plt.legend()
plt.title('Note the width of the domain in km is constant from top to bottom')
plt.show()

```

number of grid points (361201, 2) 361201

Note the width of the domain in km is constant from top to bottom



```
[12]: from scipy.interpolate import NearestNDInterpolator
      #from scipy.interpolate import LinearNDInterpolator
      #interp = LinearNDInterpolator(xy, z.flatten())
      interp = NearestNDInterpolator(xy, z.flatten())
```

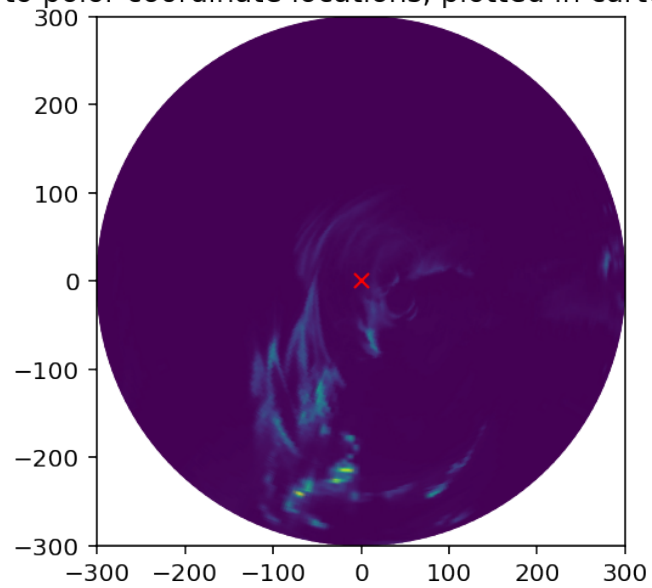
```
[13]: # z interpolated to the uniform polar coordinate locations
      Z = interp(polar_x2, polar_y2)
```

```
[14]: Z.shape
```

```
[14]: (200, 201)
```

```
[15]: plt.pcolormesh(polar_x2, polar_y2, Z, shading='gouraud')
      plt.plot(0, 0, 'rx', label='center')
      plt.gca().set_aspect('equal')
      plt.title('interpolated to polar coordinate locations, plotted in cartesian_
      ↪coordinates');
```

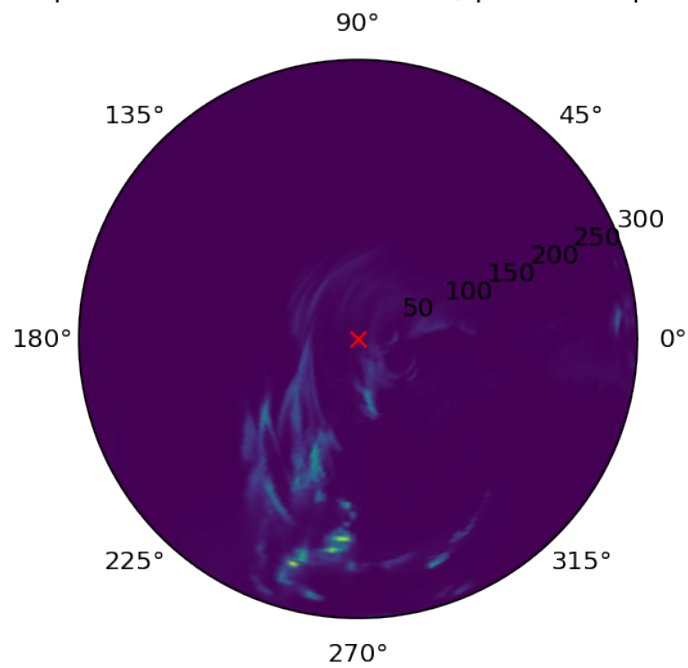
interpolated to polar coordinate locations, plotted in cartesian coordinates



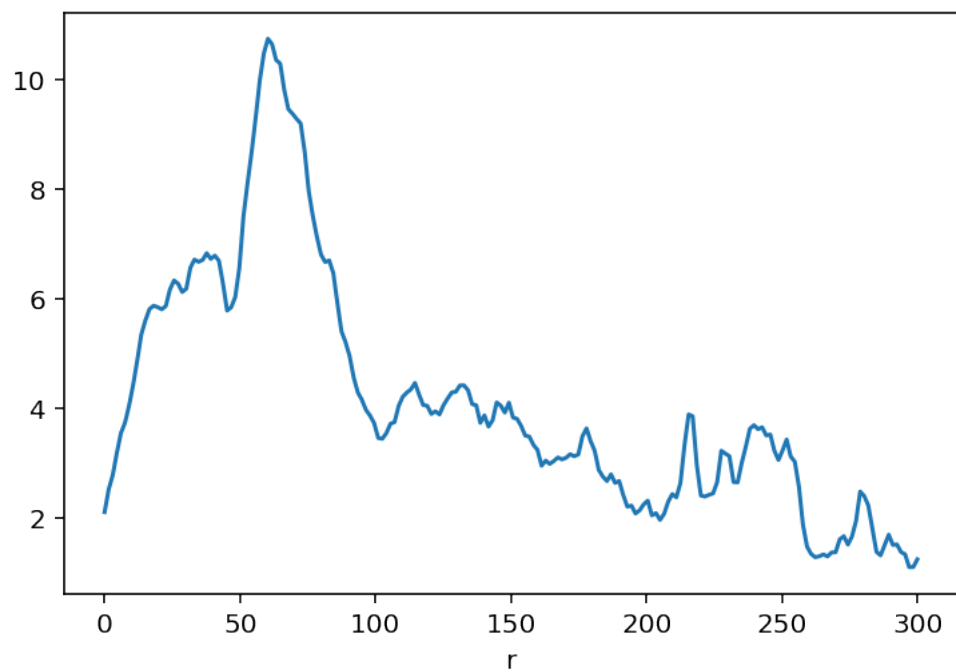
```
[16]: plt.subplot(projection="polar")
plt.pcolormesh(theta1, r1, Z, shading='gouraud')
plt.plot(0, 0, 'rx', label='center')
plt.title('interpolated to polar coordinate locations, plotted in polar_
↪coordinates')
plt.show()
```



interpolated to polar coordinate locations, plotted in polar coordinates



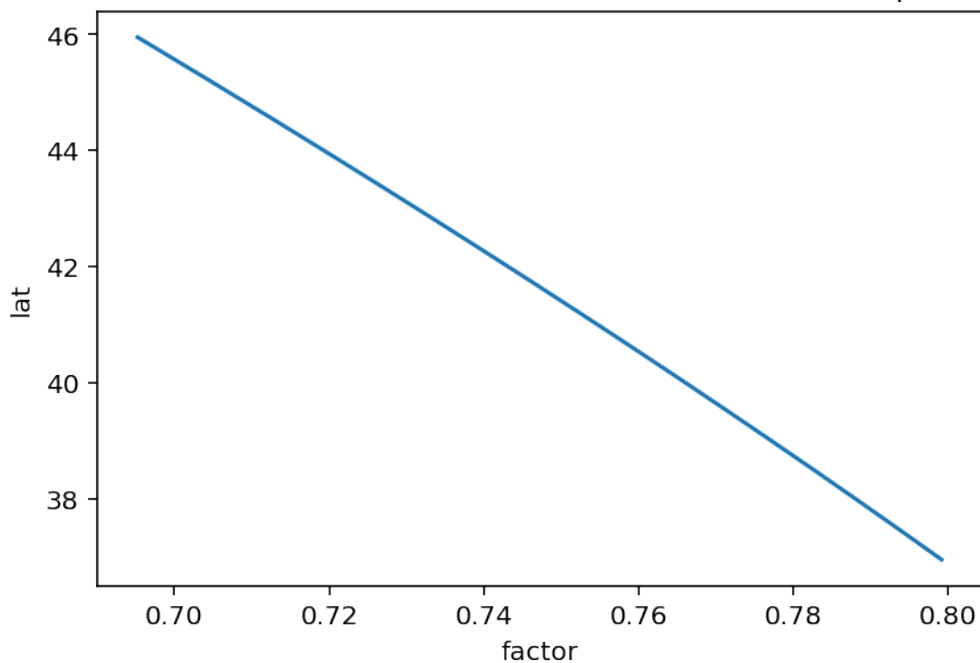
```
[17]: plt.plot(r1, np.mean(Z, axis=1))  
      plt.xlabel('r')  
      plt.show()
```



```
[18]: # rename result to compare latter
Z_center = Z
```

```
[19]: plt.plot(np.cos(np.deg2rad(ds.latitude.values)), lat)
plt.title('A fair amount of variation in the cosine lat factor from top to_
↳bottom')
plt.xlabel('factor')
plt.ylabel('lat')
plt.show()
```

A fair amount of variation in the cosine lat factor from top to bottom



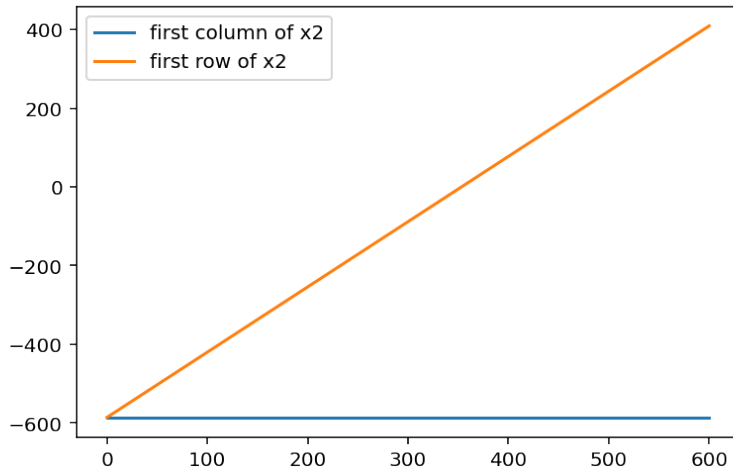
#### 4.4 Cosine lat weighting (varying)

```
[20]: # uses all the cos lat values km/deg = 110.567
x1 = (ds.longitude.values - xcenter) * 110.567
y1 = (ds.latitude.values - ycenter) * 110.567
x2, y2 = np.meshgrid(x1, y1)
```

```
[21]: plt.plot(x2[:, 0], label='first column of x2')
plt.plot(x2[0, :], label='first row of x2')
plt.title('Each column of x2 contains one value of x1 and each row of x2_
↳contains all the values of x1')
```

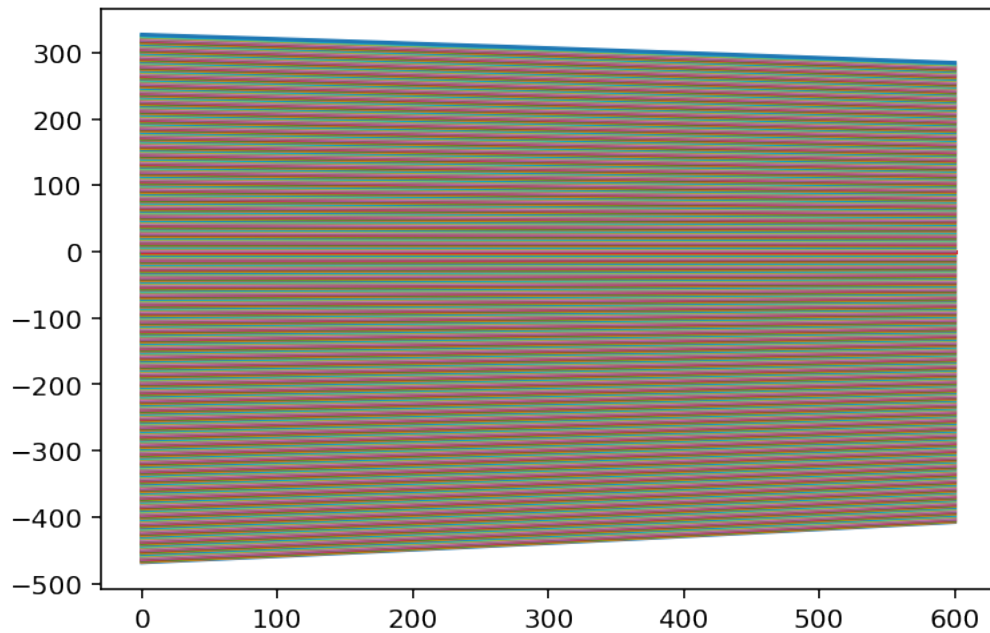
```
plt.legend()
plt.show()
```

Each column of x2 contains one value of x1 and each row of x2 contains all the values of x1

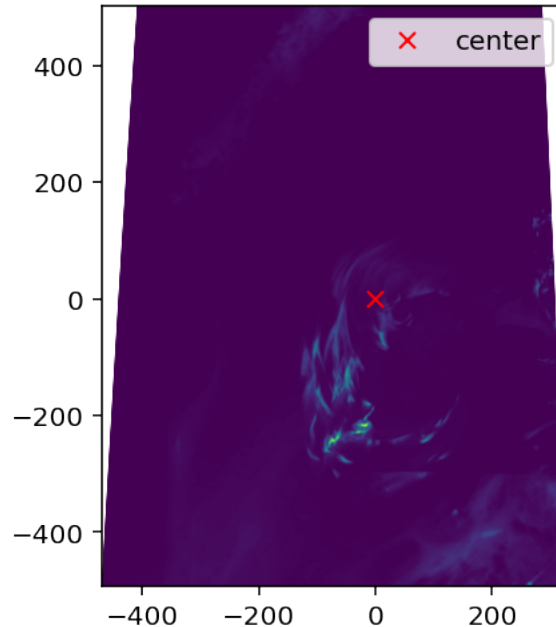


```
[22]: # choose the shape of the factor to broadcast along the 2nd
cosine_factor = np.cos(np.deg2rad(ds.latitude.values)).reshape(-1, 1)
x2 = cosine_factor * x2
```

```
[23]: plt.plot(x2);
```



```
[24]: plt.pcolormesh(x2, y2, z, shading='gouraud')
plt.plot(0, 0, 'rx', label='center')
plt.gca().set_aspect('equal')
plt.legend()
plt.show()
```



```
[25]: # number of points X 2 matrix of x y
xy = np.stack((x2.flatten(), y2.flatten())).T
xy.shape
```

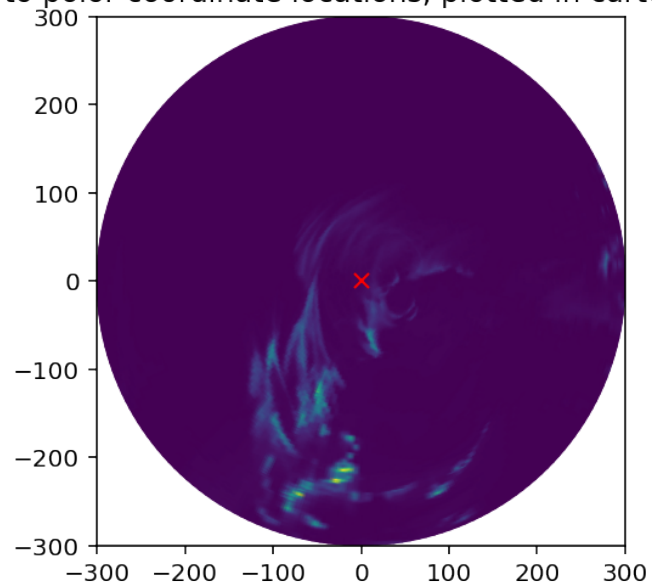
```
[25]: (361201, 2)
```

```
[26]: from scipy.interpolate import NearestNDInterpolator
#from scipy.interpolate import LinearNDInterpolator
#interp = LinearNDInterpolator(xy, z.flatten())
interp = NearestNDInterpolator(xy, z.flatten())
```

```
[27]: # z interpolated to the uniform polar coordinate locations
Z = interp(polar_x2, polar_y2)
```

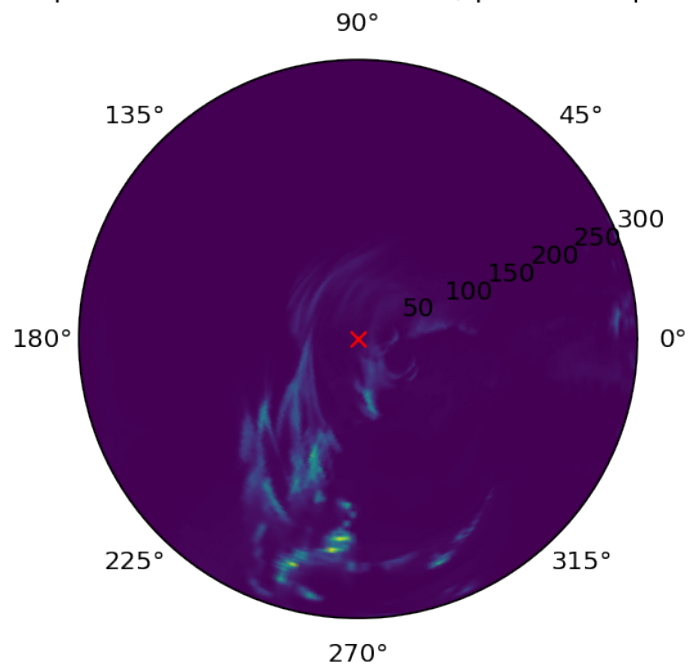
```
[28]: plt.pcolormesh(polar_x2, polar_y2, Z, shading='gouraud')
plt.plot(0, 0, 'rx', label='center')
plt.gca().set_aspect('equal')
plt.title('interpolated to polar coordinate locations, plotted in cartesian_
↳coordinates');
```

interpolated to polar coordinate locations, plotted in cartesian coordinates

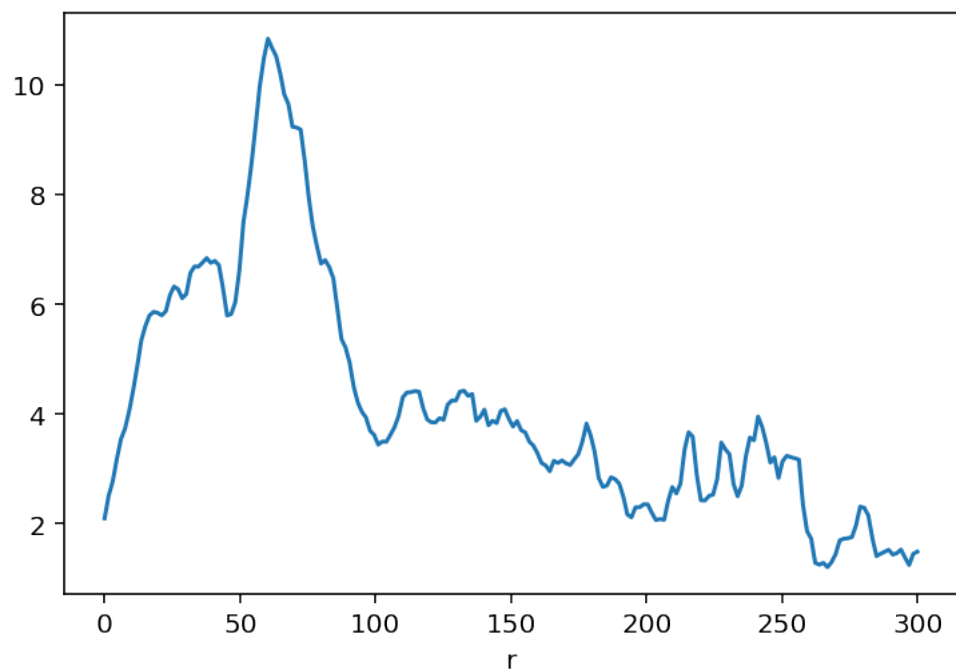


```
[29]: plt.subplot(projection="polar")
plt.pcolormesh(theta1, r1, Z, shading='gouraud')
plt.plot(0, 0, 'rx', label='center')
plt.title('interpolated to polar coordinate locations, plotted in polar_
↪coordinates')
plt.show()
```

interpolated to polar coordinate locations, plotted in polar coordinates



```
[30]: plt.plot(r1, np.mean(Z, axis=1))  
plt.xlabel('r')  
plt.show()
```



```
[31]: # rename result to compare latter
      Z_varying = Z
```

## 4.5 Cartopy

```
[32]: import cartopy.crs as ccrs

proj_local = ccrs.TransverseMercator(central_longitude=xcenter,
    ↪central_latitude=ycenter, approx=False)
#proj_local = ccrs.AzimuthalEquidistant(central_longitude=xcenter,
    ↪central_latitude=ycenter)

proj_data = ccrs.PlateCarree()

# define point
p_data = (xcenter, ycenter)

# convert from data to cartesian coordinates
p_local = proj_local.transform_point(*p_data, src_crs=proj_data)

print(p_local)
```

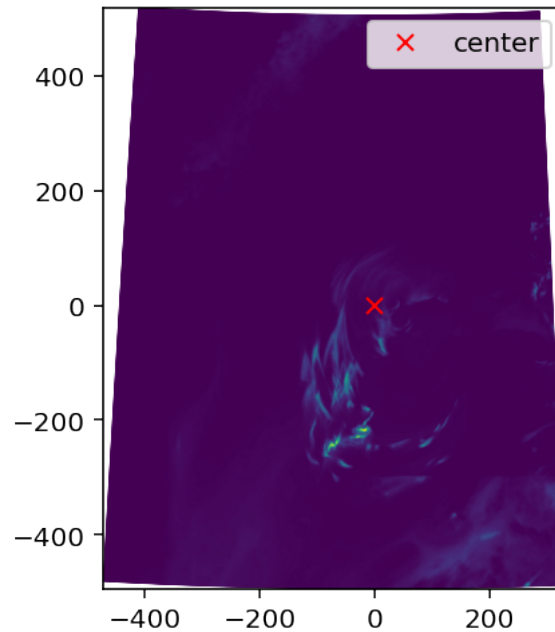
(0.0, 0.0)

```
[33]: # number of points X 2 matrix of x y
lon2, lat2 = np.meshgrid(lon, lat)
lonlat = np.stack((lon2.flatten(), lat2.flatten())).T
```

```
[34]: x2 = np.empty_like(lon2)
      y2 = np.empty_like(lat2)

      for ii in range(0, len(lon)):
          for jj in range(0, len(lat)):
              x2[ii, jj], y2[ii, jj] = proj_local.transform_point(lon2[ii, jj],
    ↪lat2[ii, jj], src_crs=proj_data)
      # m -> km
      x2, y2 = x2/1000, y2/1000
```

```
[35]: plt.pcolormesh(x2, y2, z, shading='gouraud')
      plt.plot(0, 0, 'rx', label='center')
      plt.gca().set_aspect('equal')
      plt.legend()
      plt.show()
```



```
[36]: # number of points X 2 matrix of x y
xy = np.stack((x2.flatten(), y2.flatten())).T
xy.shape
```

```
[36]: (361201, 2)
```

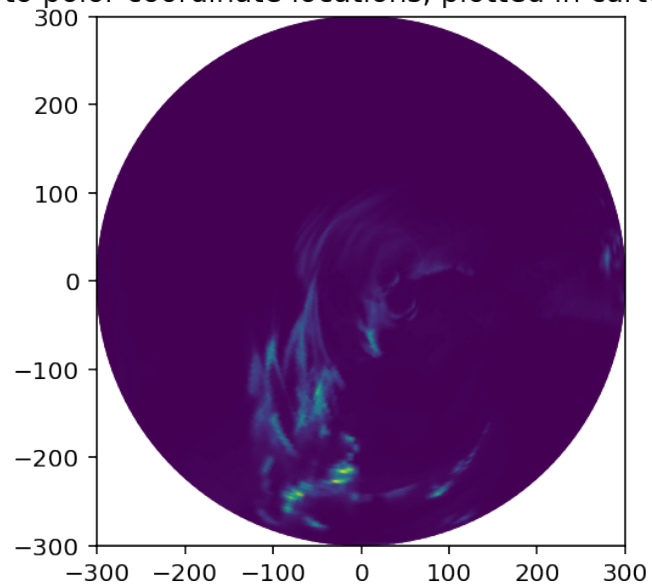
```
[37]: #from scipy.interpolate import LinearNDInterpolator
#interp = LinearNDInterpolator(xy, z.flatten())
from scipy.interpolate import NearestNDInterpolator
interp = NearestNDInterpolator(xy, z.flatten())
```

```
[38]: # z interpolated to the uniform polar coordinate locations
Z = interp(polar_x2, polar_y2)
```

```
[39]: plt.pcolormesh(polar_x2, polar_y2, Z, shading='gouraud')
plt.gca().set_aspect('equal')
plt.title('interpolated to polar coordinate locations, plotted in cartesian_
↪coordinates');
```

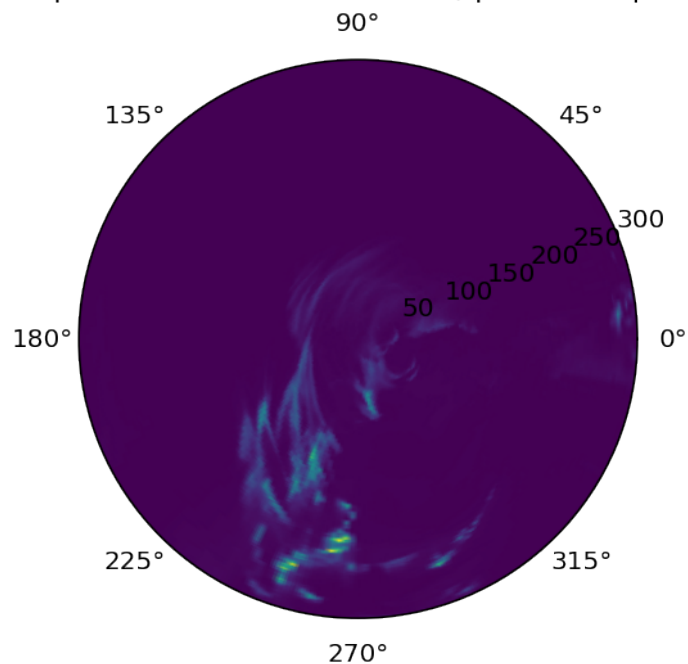


interpolated to polar coordinate locations, plotted in cartesian coordinates



```
[40]: plt.subplot(projection="polar")
plt.pcolormesh(theta1, r1, Z, shading='gouraud')
plt.title('interpolated to polar coordinate locations, plotted in polar_
↪coordinates')
plt.show()
```

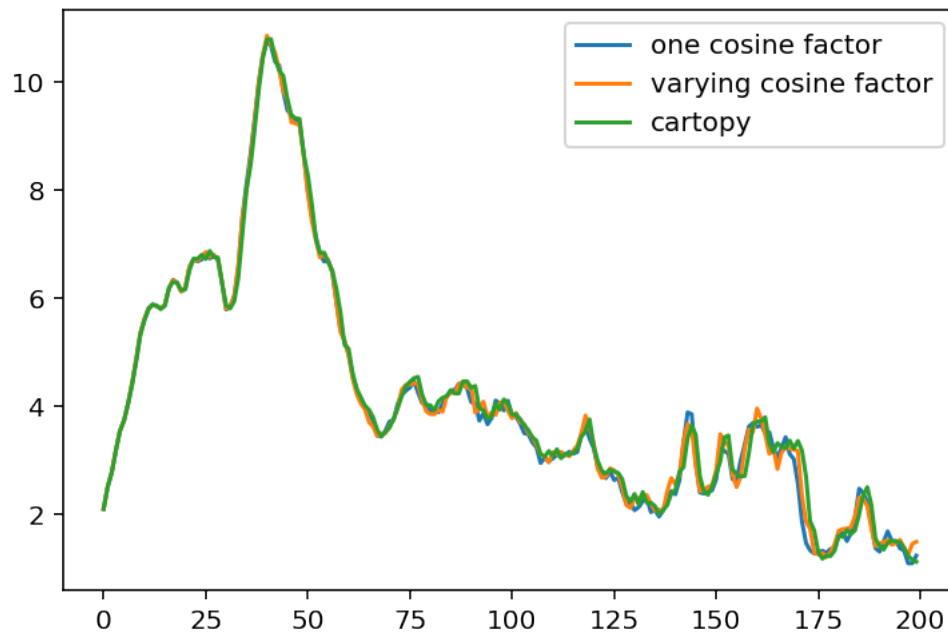
interpolated to polar coordinate locations, plotted in polar coordinates



```
[41]: # rename result to compare latter
Z_proj = Z
```

## 5 Compare

```
[42]: plt.plot(np.mean(Z_center, axis=1), label='one cosine factor')
plt.plot(np.mean(Z_varying, axis=1), label='varying cosine factor')
plt.plot(np.mean(Z_proj, axis=1), label='cartopy')
plt.legend()
plt.show()
```



```
[43]: fig, axs = plt.subplots(1, 3, figsize=(15, 5), subplot_kw=dict(polar=True))

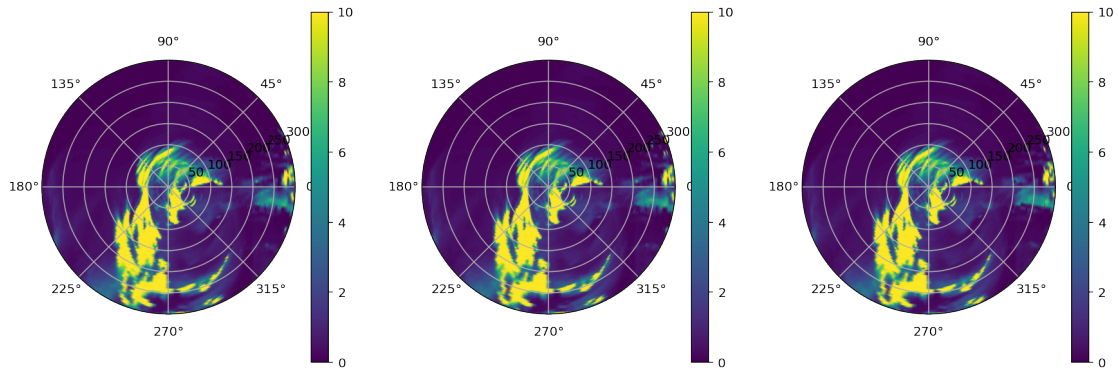
ax = axs[0]
g = ax.pcolormesh(theta1, r1, Z_center, vmax=10, shading='gouraud')
plt.colorbar(g, ax=ax)
ax.grid()

ax = axs[1]
g = ax.pcolormesh(theta1, r1, Z_varying, vmax=10, shading='gouraud')
plt.colorbar(g, ax=ax)
ax.grid()
```

```

ax = axs[2]
g = axs[2].pcolormesh(theta1, r1, Z_proj, vmax=10, shading='gouraud')
plt.colorbar(g, ax=ax)
ax.grid()

```



```

[45]: fig, axs = plt.subplots(1, 2, figsize=(15, 5), subplot_kw=dict(polar=True))

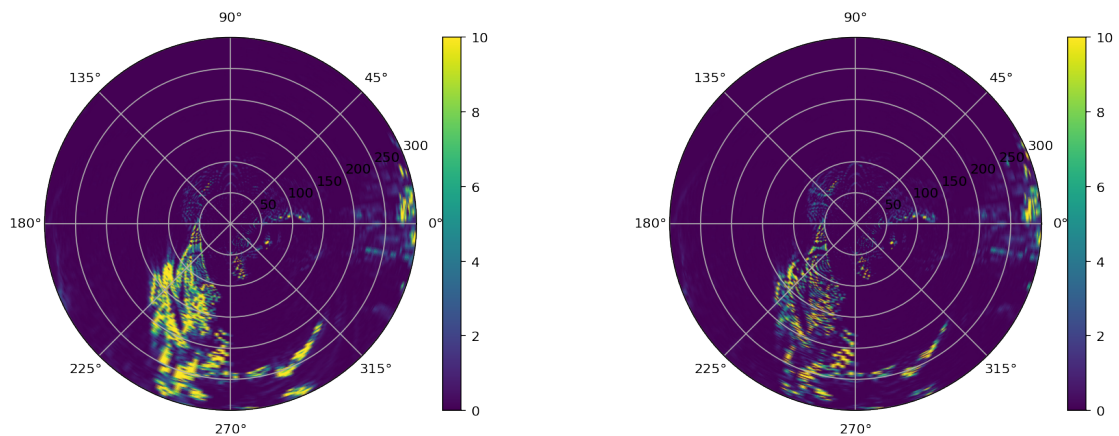
```

```

ax = axs[0]
g = ax.pcolormesh(theta1, r1, np.abs(Z_center - Z_proj), vmax=10,
    ↪shading='gouraud')
plt.colorbar(g, ax=ax)
ax.grid()

ax = axs[1]
g = ax.pcolormesh(theta1, r1, np.abs(Z_varying - Z_proj), vmax=10,
    ↪shading='gouraud')
plt.colorbar(g, ax=ax)
ax.grid()

```



[ ]:

[ ]: