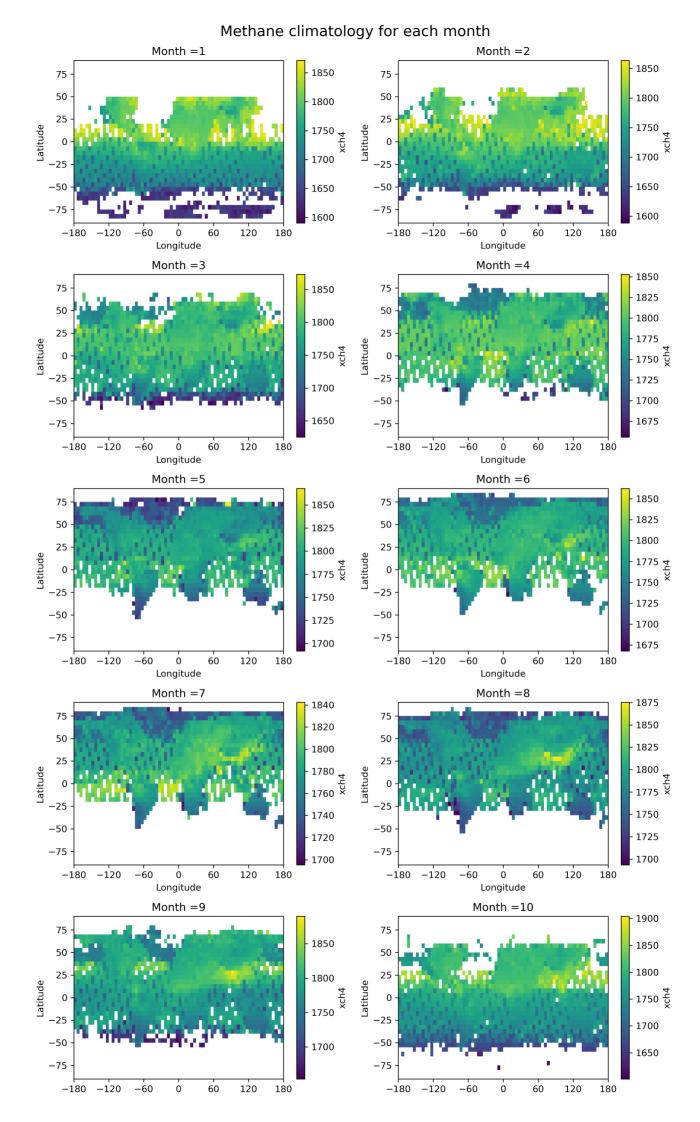
陈禹凡 12232261

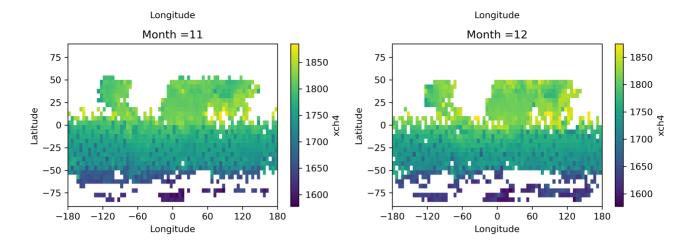
```
In [1]: import numpy as np
import pandas as pd
import datetime
import netCDF4
import xarray as xr
import matplotlib as mpl
import matplotlib.pyplot as plt
import matplotlib.gridspec as gridspec
```

1. Global methane levels from 2002

1.1 Methane climatology for each month

```
In [2]: #读取文件命名为Methane
        Methane = xr. open dataset ("200301 202006-C3S-L3 GHG-PRODUCTS-OBS4MIPS-MERGED-v4.3.nc", engine="netcdf4")
        #将xch4数据改为ppb浓度
        Methane['xch4']. data = Methane['xch4']. data*1e9
        #将xch4数据按月平均分辨率提取出来命名为Methane_month
        Methane_month = Methane.xch4.groupby('time.month').mean()
        plt. figure(figsize=(10,20), dpi=300)
        #绘制1-12月xch4含量图
        for i in range(1,13):
           #子图显示位置六行两列(6,2,i)
            plt. subplot (6, 2, i)
            Methane_month.sel(month = i).plot()
            #美化图片
            plt.title('Month = ' + str(i))
            plt. xlabel('Longitude'); plt. ylabel('Latitude')
            plt. xticks (range (-180, 181, 60))
        plt.suptitle('Methane climatology for each month', verticalalignment='bottom', fontsize=16)
        plt. tight_layout()
        plt. show()
```

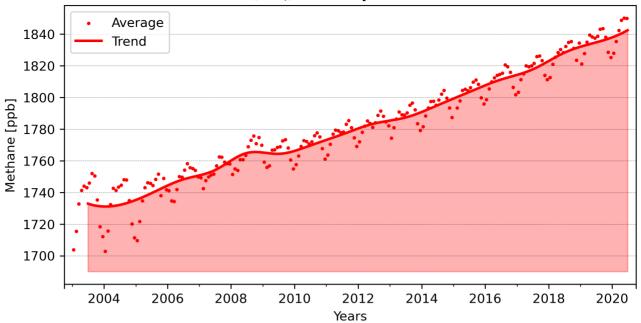




1.2 Globally-averaged methane from 2003-01 to 2020-06

```
#将全球数据按时间(time)统计起来,绘制成散点图
In [3]:
        fig, axs = plt. subplots (1, 1, figsize = (8, 4), dpi = 300)
        #散点图
        x = Methane['time'].data
        y=Methane.xch4.mean(dim=['lat', 'lon'])
        plt. scatter(x, y, s=3, c='r', marker='o')
        plt.\ legend (labels = \hbox{\tt ["Average"]}, loc = \hbox{\tt Methane}.\ xch4.\ mean (dim = \hbox{\tt ['lat', 'lon']}), fontsize = 6)
        plt. title ('Global Methane (CH$ {4}$) - Monthly Means from 2003 to 2020')
        plt. xlabel('Years'); plt. ylabel('Methane [ppb]')
        #绘制趋势曲线,在这里我使用年均值数据,因此我先将年均值提取出来
        #Methane_month为全球月均值数据(Methane为每月数据)
        Methane_month = Methane.mean(dim=['lat', 'lon'])
        #将其转化为二维列表数据(time+xch4)
        all_Methane_month = pd. DataFrame(index=Methane_month['time'], columns=['xch4'])
        for i in range(len(all_Methane_month)):
            all_Methane_month['xch4'][i] = Methane_month.xch4.values[i]
        #通过resample函数实现对每年(Y)的重采样
        Methane Year = all Methane month.resample('Y').mean()['xch4']
        #绘制趋势曲线(该方法为在百度搜索得到)
        #数据起始,从年中开始
        startDate = '2003-07-01'
endDate = '2020-07-01'
        #将年均数据新建为df trend列表
        df_trend = pd. DataFrame (Methane_Year. values, index=pd. date_range(startDate, endDate, freq='365D'), columns=['xch4'])
        #运用pandas自带的插值函数interpolate,以频率为1天,方法为立方(cubic),对数据进行平滑
        df_smooth = df_trend.reindex(index= pd.date_range(startDate, endDate, freq='1D')).interpolate('cubic')
        #将xch4列重命名为Trend,图例
        df_smooth = df_smooth.rename(columns={'xch4': 'Trend'})
        #绘制曲线图
        df_smooth.plot(ax=axs, c = 'r', linewidth=2)
        #趋势线下填充红色
        axs. fill_between(x = df_smooth. index, y1 = 1690, y2 = df_smooth. values. reshape(-1,), color = 'r', alpha = 0.3)
        axs. set_xlim('2002-10-01', '2020-10-01')
        axs.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k', axis='y')
        #同时,还可以使用课堂上讲的rolling means方法进行重采样,在此不再重复~
```

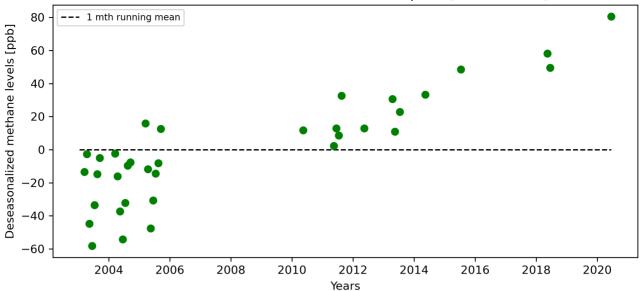
Global Methane (CH₄) - Monthly Means from 2003 to 2020



1.3 Deseasonalized methane levels at point [15°S, 150°W] from 2003-01 to 2020-06

```
In [4]: #挑选point [15°S, 150°W]区域
        Methane_region = Methane. sel(lat=15, lon=-150, method='nearest')
        #将time数据以月份分组统计 (group)
        Methane_month = Methane_region.xch4.groupby('time.month')
        #每月数据减去气候学数据即为该区域的异常值
        #Methane month.mean(dim = 'time')为2003-2020年每月某地方的sst平均值
        Methane anom = Methane month - Methane month. mean(dim = 'time')
        fig, axm = plt. subplots(1, 1, figsize = [8, 4], dpi=300)
        #将数据导入到图
        axm. hlines(y = 0, xmin=Methane_anom['time'][0], xmax=Methane_anom['time'][-1], color='k', ls='--',
                   lw=1.2, label='1 mth running mean')
        axm. plot(Methane_anom['time'], Methane_anom. values, 'go')
        axm. set_title('The deseasonalized methane levels at point [15° S, 150° W]')
        axm. set_xlabel('Years')
        axm. set_ylabel('Deseasonalized methane levels [ppb]')
        axm. legend(loc='best', fontsize=8)
        #使用紧凑的布局
        plt. tight_layout()
        plt. show()
        #下方warning有点不懂
        C:\Users\Administrator\anaconda3\lib\site-packages\xarray\core\indexes.py:234: FutureWarning: Passing method to Flo
        at64Index.get_loc is deprecated and will raise in a future version. Use index.get_indexer([item], method=...) inste
         indexer = self.index.get_loc(
        C:\Users\Administrator\anaconda3\lib\site-packages\xarray\core\indexes.py:234: FutureWarning: Passing method to Flo
        at64Index.get_loc is deprecated and will raise in a future version. Use index.get_indexer([item], method=...) inste
        indexer = self.index.get_loc(
```

The deseasonalized methane levels at point [15°S, 150°W]



从数据(去季节性)散点分布我们可以发现:①数据在2006-2010中有缺失,可能与地点以及数据采集方式有关;②在大约2006年之前,甲烷水平是相对较低的,而2010年之后,甲烷水平相对较高;③从2003-2021年甲烷总体水平是呈逐年增高的趋势(2006年之前增高相对缓慢),说明甲烷排放量逐年增大,这可能与人类活动有关,或与整个生态环境的改变有关。

2. Niño 3.4 index

2.1 Monthly climatology for SST from Niño 3.4 region

```
In [5]: #读取文件,命名为SST_data SST_data SST_data = xr.open_dataset("NOAA_NCDC_ERSST_v3b_SST.nc",engine="netcdf4")

#挑选the South American coast(5N-5S, 170W-120W)区域
SST_region = SST_data.sel(lat = slice(-5,5), lon = slice(190,240))

#将time数据以月份分组统计(group)
SST_month = SST_region.sst.groupby('time.month')

#每月数据减去气候学数据即为该区域的异常值(SST_anom还是个三维数据,有时间,纬度,经度三个方向)
#SST_month.mean(dim = 'time')为1960-2016年每月某地方的sst平均值
SST_anom = SST_month - SST_month.mean(dim = 'time')
```

2.2 Visualize the computed Niño 3.4

```
In [6]: #np. nanmean为沿指定轴计算算术平均值,忽略NaN(在百度中查找到)
#因为SST_anom为三维,方向分别为0: 时间,1: 纬度,2: 经度
#因此我们沿着axis=(1,2)将异常值进行平均计算,得到整个区域在不同时间段的异常数据(SST_anom_average)
SST_anom_average = np. nanmean(SST_anom. values, axis=(1,2))

#转化为二维列表,方便后续用pandas统计画图
#新建一个列表all_data_month,index为时间, columns为异常值
all_data_month = pd. DataFrame(index=SST_anom['time'], columns=['Anomaly'])

#将区域异常数据SST_anom_average赋值到all_data_month
for i in range(len(all_data_month)):
    all_data_month['Anomaly'][i] = SST_anom_average[i]

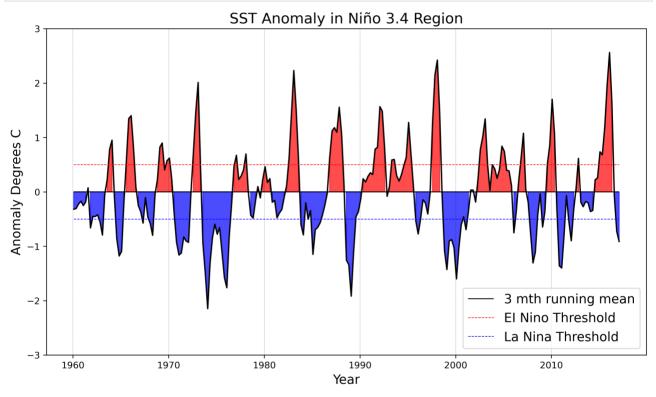
#通过resample函数实现对每3个月(3M)的重采样
SST_3M = all_data_month. resample('3M'). mean()['Anomaly']
```

```
In [7]: #创建图, 大小, 分辨率
fig, axs = plt. subplots(1,1,figsize = [10,6],dpi=300)

#将数据导入到图
axs. plot(SST_3M. index, SST_3M. values, color='k')

# 绘制网格线
axs. grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k', axis='x')
```

```
# 在图中指定位置画线 (+/- 0.5°C)
axs. hlines(y = 0, xmin=SST_3M. index[0], xmax=SST_3M. index[-1], color='k', ls='solid', lw=1, label='3 mth running mean')
axs. hlines (y = 0.5, xmin=SST_3M. index[0], xmax=SST_3M. index[-1], color='r', ls='--', lw=0.7, label='EI Nino Threshold') axs. hlines (y = -0.5, xmin=SST_3M. index[0], xmax=SST_3M. index[-1], color='b', ls='--', lw=0.7, label='La Nina Threshold')
# El Niño 和 La Niña地方填充颜色
axs. fill_between (SST_3M. index, 0, SST_3M. values, where=(SST_3M. values>0), color='r', alpha=0.7)
axs. fill between (SST 3M. index, 0, SST 3M. values, where=(SST 3M. values<0), color='b', alpha=0.7)
# x, y轴范围
axs. set ylim(-3, 3)
#添加图例
axs. legend(loc='best', fontsize=14)
# 设置x,y轴和tile格式
axs.set_xlabel('Year', color='k', fontsize=14)
axs.set_ylabel('Anomaly Degrees C', color='k', fontsize=14)
axs. set_title("SST Anomaly in Niño 3.4 Region", color='k', fontsize=16)
#使用紧凑的布局
plt. tight layout()
plt. show()
```



3. Explore a netCDF dataset

3.1

```
wind_data = xr.open_dataset("wind_CMFD_V0106_B-01_01dy_010deg_201801-201812.nc",engine="netcdf4")

#将time数据按月份分组统计(group)
wind_month = wind_data.wind.groupby('time.month')

#每日数据减去气候学数据即为该区域的日异常值(每天值-月平均值)
wind_anom = wind_month - wind_month.mean(dim = 'time')

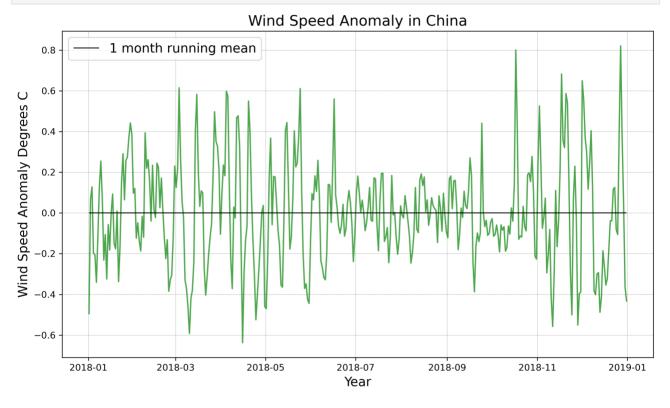
#因为wind_anom为三维,方向分别为0,时间,1:纬度,2:经度

#因此我们沿着axis=(1,2)将异常值进行平均计算,得到整个区域在不同时间段的异常数据(wind_anom_average)
wind_anom_average = np. nanmean(wind_anom.values,axis=(1,2))

#转化为二维列表,方便后续用pandas统计画图
#新建一个列表allwind_data_month, index为时间,columns为异常值
allwind_data_month = pd. DataFrame(index=wind_anom['time'],columns=['Anomaly'])

#将区域异常数据wind_anom_average赋值到allwind_data_month
for i in range(len(allwind_data_month)):
    allwind_data_month['Anomaly'][i] = wind_anom_average[i]
```

```
In [9]: #创建图,大小,分辨率
        fig, axw = plt. subplots (1, 1, figsize = [10, 6], dpi=300)
        #将数据导入到图
        axw. plot(allwind_data_month. index, allwind_data_month. values, color='g', alpha=0.7)
        # 绘制网格线
        axw.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
        #在图中y=0画线(每y月均值)
        axw. hlines(y = 0, xmin=allwind data month. index[0], xmax=allwind data month. index[-1],
                   color='k', ls='solid', lw=1, label='1 month running mean')
        #添加图例
        axw. legend(loc='best', fontsize=14)
        # 设置x,y轴和tile格式
        axw. set xlabel ('Year', color='k', fontsize=14)
        axw. set_ylabel('Wind Speed Anomaly Degrees C', color='k', fontsize=14)
        axw.set_title("Wind Speed Anomaly in China", color='k',fontsize=16)
        #使用紧凑的布局
        plt.tight_layout()
        plt. show()
```



3.2

```
In [10]: plt. figure(figsize=(10, 10), dpi=300)
           #中国2018年1月平均风速
           plt. subplot (3, 2, 1)
           wind_data_Jan = wind_data.sel(time = slice('2018-01-01','2018-01-31'))
           wind_data_Jan['wind']. mean(dim='time'). plot(robust=True, cmap='gist_rainbow', cbar_kwargs={'label':'Wind Speed (m/s)}
           plt. title ('The average wind speed of China in January 2018') plt. grid (linestyle='--', linewidth=0.3, alpha=0.5, color='k')
           #中国2018年7月平均风速
           plt. subplot (3, 2, 2)
           wind_data_Jul = wind_data.sel(time = slice('2018-07-01', '2018-07-31'))
           wind_data_Jul['wind']. mean(dim='time'). plot(robust=True, cmap='gist_rainbow', cbar_kwargs={'label':'Wind Speed (m/s)}
           plt.title('The average wind speed of China in July 2018') plt.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
           #2018年中国平均风速
           plt. subplot (3, 2, 3)
           wind_data['wind'].mean(dim='time').plot(robust=True,cmap='gist_rainbow',cbar_kwargs={'label':'Wind Speed (m/s)'})
           plt. title('The average wind speed of China in 2018')
           plt.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
           #2018年中国区域风速异常
```

```
plt. subplot (3, 2, 4)
wind_anom. mean(dim='time'). plot(robust=True, cmap='gist_rainbow', cbar_kwargs={'label':'Anomalous wind Speed (m/s)'
plt. title('The anomalies of wind speed in 2018')
plt. grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
#北京市2018年风速变化曲线
plt. subplot (3, 2, 5)
wind data beijing = wind data. wind. sel(lat= '39', lon='115', method='nearest')
wind_data_beijing.plot()
plt. xlabel('Year'); plt. ylabel('Wind Speed (m/s)')
plt.title('Annual Wind Speed Change Curve in Beijing')
plt.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
#中国2018年风速变化曲线
plt. subplot (3, 2, 6)
wind_data['wind']. mean(['lat', 'lon']). plot(color='r')
plt. xlabel('Year'); plt. ylabel('Wind Speed (m/s)')
plt.title('Annual Wind Speed Change Curve in China')
plt.grid(linestyle='--', linewidth=0.3, alpha=0.5, color='k')
plt. suptitle ('6 different plots using the dataset', verticalalignment='bottom', fontsize=20)
plt. tight_layout()
plt. show()
#下方warning不太懂
C:\Users\Administrator\anaconda3\lib\site-packages\xarray\core\indexes.py:234: FutureWarning: Passing method to Flo
at64Index.get_loc is deprecated and will raise in a future version. Use index.get_indexer([item], method=...) inste
ad.
 indexer = self.index.get_loc(
C:\Users\Administrator\anaconda3\lib\site-packages\xarray\core\indexes.py:234: FutureWarning: Passing method to Flo
at64Index.get_loc is deprecated and will raise in a future version. Use index.get_indexer([item], method=...) inste
indexer = self.index.get_loc(
```

6 different plots using the dataset

