

chl_visual

June 22, 2023

```
[ ]: import pandas as pd
import numpy as np

df_chl = pd.read_csv("chl_means_new.csv")
df_chl.head(5)
```

```
[ ]:   Lat  Long   Ocean  Cr_nmol/kg   Year  chl_01  chl_02  chl_03  \
0   21  -132  Pacific         NaN  1983.0  0.06898  0.062931  0.052969
1   21  -132  Pacific    3.190219  1983.0  0.06898  0.062931  0.052969
2   21  -132  Pacific    3.132063  1983.0  0.06898  0.062931  0.052969
3   21  -132  Pacific         NaN  1983.0  0.06898  0.062931  0.052969
4   21  -132  Pacific    3.424319  1983.0  0.06898  0.062931  0.052969

      chl_04  chl_05  chl_06  chl_07  chl_08  chl_09  chl_10  \
0  0.047427  0.049164  0.048803  0.045213  0.042394  0.044499  0.049289
1  0.047427  0.049164  0.048803  0.045213  0.042394  0.044499  0.049289
2  0.047427  0.049164  0.048803  0.045213  0.042394  0.044499  0.049289
3  0.047427  0.049164  0.048803  0.045213  0.042394  0.044499  0.049289
4  0.047427  0.049164  0.048803  0.045213  0.042394  0.044499  0.049289

      chl_11  chl_12
0  0.060621  0.066887
1  0.060621  0.066887
2  0.060621  0.066887
3  0.060621  0.066887
4  0.060621  0.066887
```

```
[ ]: month_name = ["01", "02", "03", "04", "05", "06", "07", "08", "09", "10", "11", "12"]
df_chl['yearly_mean_chl'] = df_chl[['chl_' + month for month in month_name]].
    ↪mean(axis=1)
df_chl.head(5)
```

```
[ ]:   Lat  Long   Ocean  Cr_nmol/kg   Year  chl_01  chl_02  chl_03  \
0   21  -132  Pacific         NaN  1983.0  0.06898  0.062931  0.052969
1   21  -132  Pacific    3.190219  1983.0  0.06898  0.062931  0.052969
2   21  -132  Pacific    3.132063  1983.0  0.06898  0.062931  0.052969
3   21  -132  Pacific         NaN  1983.0  0.06898  0.062931  0.052969
4   21  -132  Pacific    3.424319  1983.0  0.06898  0.062931  0.052969
```

	chl_04	chl_05	chl_06	chl_07	chl_08	chl_09	chl_10	\
0	0.047427	0.049164	0.048803	0.045213	0.042394	0.044499	0.049289	
1	0.047427	0.049164	0.048803	0.045213	0.042394	0.044499	0.049289	
2	0.047427	0.049164	0.048803	0.045213	0.042394	0.044499	0.049289	
3	0.047427	0.049164	0.048803	0.045213	0.042394	0.044499	0.049289	
4	0.047427	0.049164	0.048803	0.045213	0.042394	0.044499	0.049289	

	chl_11	chl_12	yearly_mean_chl
0	0.060621	0.066887	0.053265
1	0.060621	0.066887	0.053265
2	0.060621	0.066887	0.053265
3	0.060621	0.066887	0.053265
4	0.060621	0.066887	0.053265

```
[ ]: for col in df_chl.columns:
      if 'chl' in col:
          print(f"Correlation between 'Cr_nmol/kg' and '{col}': {df_chl['Cr_nmol/
          ↳kg'].corr(df_chl[col])}")
```

```
Correlation between 'Cr_nmol/kg' and 'chl_01': 0.2222380030471767
Correlation between 'Cr_nmol/kg' and 'chl_02': 0.239509176886822
Correlation between 'Cr_nmol/kg' and 'chl_03': 0.1575600091263584
Correlation between 'Cr_nmol/kg' and 'chl_04': 0.07302255062946351
Correlation between 'Cr_nmol/kg' and 'chl_05': -0.06234617009495116
Correlation between 'Cr_nmol/kg' and 'chl_06': -0.08492823354130775
Correlation between 'Cr_nmol/kg' and 'chl_07': -0.06148940897283865
Correlation between 'Cr_nmol/kg' and 'chl_08': 0.017685228049793285
Correlation between 'Cr_nmol/kg' and 'chl_09': 0.03633780981839884
Correlation between 'Cr_nmol/kg' and 'chl_10': 0.20825374588132703
Correlation between 'Cr_nmol/kg' and 'chl_11': 0.20822047938850002
Correlation between 'Cr_nmol/kg' and 'chl_12': 0.17539221611321804
Correlation between 'Cr_nmol/kg' and 'yearly_mean_chl': 0.11206869836067768
```

```
[ ]: #From the code above, we find that the correlation between CHL and Chromium
      ↳concentration is very little
```

```
[ ]: import seaborn as sns
      import matplotlib.pyplot as plt

      df_chl[df_chl['yearly_mean_chl'] < 0] = np.nan #Remove points with missing data
      ↳since missing data are filled with -9999

      plt.figure(figsize=(8, 6))
      sns.scatterplot(data=df_chl, x='yearly_mean_chl', y='Cr_nmol/kg', hue='Ocean')
      plt.title('Relationship between Cr_nmol/kg and yearly_mean_chl')
      plt.show()
```

