

MLD Visualization

In [8]:

```
import pandas as pd
import numpy as np

df_mld = pd.read_csv('mld_means_new.csv')
df_mld.sample(5)
```

Out[8]:

	Lat	Long	Year	mld_09	mld_10	mld_11	mld_12	mld_01	mld_02	mld_04	mld_05	mld_06	mld_07
27851	85.130000	-150.100000	2004	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN
24324	-53.583000	149.298000	2002	131.247990	117.021190	77.375790	61.464188	40.685440	69.604710	85.205890	111.676570	108.085530	117.880330
23880	-49.393000	150.389000	1999	154.046110	138.263580	110.081390	60.175407	56.818960	74.987236	129.418820	203.046880	195.944060	182.051100
25032	-49.393000	150.389000	2007	130.276960	128.302760	94.973440	62.438824	49.911537	70.163260	100.360860	96.325485	216.694170	113.814644
9789	-21.383333	-114.283333	2015	62.993843	42.302597	37.278122	31.571268	18.048435	34.572000	33.426777	64.821540	72.289024	85.209114

In [9]:

```
month_name = ["01", "02", "03", "04", "05", "06", "07", "08", "09", "10", "11", "12"]
for month in month_name:
    unique_values = df_mld['mld_' + month].unique()
    print(f"Unique values in mld_{month}: {unique_values}")
df_mld['yearly_mean_mld'] = df_mld[['mld_' + month for month in month_name]].mean(axis=1)
```

Unique values in mld_01:	[nan	59.190742	74.78616	...	52.03397	56.460392	29.661045]
Unique values in mld_02:	[nan	49.773827	35.400013	...	60.	61.673435	37.826736]
Unique values in mld_03:	[nan	46.68332	49.884098	...	38.184013	31.833996	30.288898
17.02741	12.002083	26.825031	26.181917	26.030293	...	93.04824		
80.38718	72.32047	24.185448	56.322937	33.22253	...	83.10309		
100.92002	103.811066	52.168526	37.515858	20.897945	...	14.418519		
32.599373	16.160042	11.540199	12.323704	20.374668	...	11.457884		
13.94981	11.977362	11.862844	11.447563	12.123527	...	18.625072		
16.815994	19.163023	13.573595	63.92895	51.077114	...	52.32264		
34.726547	48.35926	34.37816	29.334148	43.83156	...	39.03835		
15.920845	12.362115	14.100921	13.29261	17.283278	...	16.540775		
17.33564	23.95749	50.37726	55.224663	51.95207	...	30.566118		
29.68642	28.577381	24.702831	11.061262	33.36479	...	34.150272		
33.60605	78.41152	65.080635	70.14343	15.915667	...	76.70798		
29.557009	71.788635	94.443146	92.75221	63.558002	...	67.923904		
21.33516	16.390392	32.970474	12.609028	11.081683	...	13.654188		
17.988289	11.004198	13.96397	12.18118	11.306923	...	10.869328		
13.252601	15.720172	14.152401	17.822493	16.453873	...	43.94004		
42.63427	43.3705	35.318478	41.962383	31.668152	...	36.476177		
38.186897	38.741646	15.316448	11.180578	13.126181	...	12.145503		
14.598671	15.014437	17.958092	28.207222	49.13414	...	67.69017		
42.858658	28.96141	30.56711	26.866621	19.410261	...	10.239864		
35.637238	32.960636	15.638529	88.882034	95.396225	...	71.03752		
21.985825	56.84028	30.344213	90.03328	100.69904	...	86.32275		
54.135635	34.84503	18.477564	13.850554	29.828133	...	15.801043		
13.606934	15.467265	12.774769	11.489546	12.02545	...	11.834646		
11.277831	11.516845	13.68969	13.730432	15.3076	...	20.335733		
12.587433	55.250996	54.74045	39.98867	28.922535	...	46.981903		
32.873085	37.729446	32.6529	31.333	16.281578	...	12.6219635		
13.710141	13.2013645	16.41327	14.735512	15.505895	...	23.784622		
43.81012	48.82713	41.51069	27.043816	37.630703	...	29.035809		
27.613533	10.402879	29.260057	28.899536	21.146149	...	102.14203		
78.137505	74.12517	14.690193	60.032913	29.630009	...	97.76229		
97.5387	89.08707	57.93315	64.59768	18.16858	...	12.289931		
22.260088	14.563419	11.238113	19.088863	16.786629	...	10.205648		
11.93749	11.251547	13.726524	12.347916	15.555095	...	20.920021		
13.000104	18.083693	16.77168	42.679737	42.609463	...	39.83705		
36.38923	52.081474	52.580505	46.553185	41.89772	...	34.442917		
16.320782	12.736565	15.635392	14.954952	13.815558	...	14.481466		
15.963743	24.496094	44.53227	45.11705	34.52865	...	30.691652		
22.272566	22.128416	35.703983	10.649602	31.760994	...	28.27356		
46.86721	73.34161	70.33372	56.44673	27.567007	...	54.730415		
27.67414	67.73166	83.859436	167.87724	74.78989	...	35.402718		
29.983028	18.12189	25.165836	23.643148	11.953906	...	14.867546		
18.996878	10.943403	13.692637	11.767998	15.448111	...	10.838652		
17.012245	12.640356	13.944568	25.64437	17.931345	...	38.060295		
33.528175	29.420294	29.276869	40.706104	25.527157	...	56.612286		
30.62378	32.744873	18.297026	13.375544	15.865741	...	14.247347		
16.983227	16.815523	21.060104	25.317524	37.602104	...	57.685925		
35.157295	33.497375	32.266804	31.323578	41.20026	...	10.491228		
31.728638	31.547215	18.852026	62.70542	74.43291	...	67.56932		
19.303944	100.21544	33.86123	65.88552	62.87404	...	43.159447		
44.024418	46.09613	18.621643	11.894947	21.826162	...	16.580418		
11.410844	11.522807	12.971571	10.299462	17.705149	...	12.459601		
14.177959	12.421661	16.426285	15.777992	11.424084	...	15.923822		
13.126342	40.86251	40.976494	31.486725	27.653378	...	36.968613		
41.6904	29.291164	28.940727	26.417686	17.993	...	10.610989		
12.860018	11.331513	14.838426	14.467826	18.568813	...	33.401157		
39.888824	40.157524	75.47687	58.77908	52.529484	...	48.52508		
42.71496	82.903984	87.62325	270.28723	94.88211	...	18.12834		
17.17894	14.587991	45.272404	31.2422	38.889427	...	33.770226		
39.6777	35.415676	24.587427	150.50912	55.69331	...	49.086338		
50.840847	33.280098	40.80266	28.447828	73.777626	...	46.527752		
45.849346	82.65818	77.10479	88.227516	83.214035	...	71.715744		
82.07949	218.10837	87.61901	17.806362	17.815285	...	15.567294		
36.63142	34.7304	24.626076	34.588917	37.78976	...	33.560616		
26.743141	200.22025	102.63633	69.577835	84.774345	...	54.354362		
33.814102	33.249718	81.1691	50.61573	76.52467	...	85.551		
63.48659	61.80553	42.860188	91.41139	62.987137	...	399.68558		
62.470448	18.224997	18.11895	14.450679	35.77406	...	27.899052		
31.359735	31.182379	33.477085	29.662216	20.656092	...	193.2743		
96.95215	67.34224	74.10622	38.490967	29.26565	...	24.33581		
74.742744	68.36346	63.29127	72.20084	70.65854	...	69.868225		
67.33255	50.802128	67.63814	261.43732	67.82506	...	18.805313		
18.15773	15.122427	26.658173	35.236053	32.53841	...	34.828125		
35.342407	29.89152	24.46597	111.57527	54.266018	...	51.912678		
55.98564	42.376614	30.250658	22.92851	61.965134	...	61.711803		
80.11076	70.2779	80.88376	82.15735	78.86055	...	105.509636		
68.67624	299.93964	62.281635	23.847729	25.286201	...	20.086746		
18.403952	32.761044	23.708704	30.19468	34.500267	...	33.2527		
22.804293	264.65195	130.68063	79.34828	93.09299	...	46.69384		

```
31.881697 33.09513 78.85542 50.42646 70.213 53.689407
56.087395 53.84992 70.863785 76.56824 78.426765 317.97742
70.46883 20.525375 21.0591 19.562641 26.424696 29.636038
30.617859 29.99942 34.268593 33.993256 14.442732 346.375
66.634964 42.913986 77.80798 53.968105 25.029003 29.026928
43.091484 47.465626 30.07672 44.41519 41.193947 59.002975
57.120667 33.92782 43.698204 44.176247 35.404312 31.688866
94.84898 50.073357 64.70474 57.753666 28.5886 46.341026
40.034523 78.96427 76.30058 88.286026 77.65579 92.46178
69.41588 52.70084 62.588463 44.81572 52.935997 75.45704
59.710846 77.557755 79.250015 84.91603 65.39437 77.11854
48.824257 53.957317 53.90949 50.39672 47.069332 74.49314
55.22364 89.01294 68.65846 85.122574 83.064026 78.511215
64.83432 38.74573 47.02197 51.849564 56.633743 71.945526
51.301643 78.42591 59.949364 63.60362 63.11573 76.36619
67.50949 45.796974 50.970108 47.325333 61.23971 63.442802
50.42344 79.69373 74.15845 71.76153 68.22099 71.54631
66.193275 40.53823 99.915924 46.70443 69.43073 58.646866
38.666176 71.890335 81.905174 53.731487 70.28738 61.848656
56.031654 27.037727 45.366203 39.681816 49.538773 41.79024
36.120728 18.483051 34.527348 34.563736 36.919933 39.997803
27.995497 44.87967 53.982346 19.453651 21.424091 26.136286
22.720448 19.548307 44.905098 35.305817 24.130838 13.26565
14.722732 20.85156 18.432291 44.930073 46.24465 43.317158
25.71245 16.706652 20.909798 24.77006 20.677132 32.723568
32.37201 15.095542 26.139425 12.046389 13.192694 15.13933
15.4079895 41.649345 61.730183 63.94348 46.216827 11.494131
14.018851 20.386786 17.64575 50.736443 46.847565 70.08035
68.37741 26.166395 ]
Unique values in mld_04: [ nan 44.65589 42.84558 ... 60.065083 61.28432 29.450932]
Unique values in mld_05: [ nan 34.855434 27.71458 ... 60.870667 60.682606 21.377972]
Unique values in mld_06: [ nan 31.309702 31.371567 ... 20.30401 24.228731 12.53417 ]
Unique values in mld_07: [ nan 28.587702 20.42932 ... 10.161933 22.982924 10.586878]
Unique values in mld_08: [ nan 28.293474 21.498158 ... 11.352995 30.780376 10.42367 ]
Unique values in mld_09: [34.264526 28.403645 15.140156 ... 14.382504 37.82218 19.463236]
Unique values in mld_10: [32.723934 32.41229 19.495127 ... 26.024498 43.935303 38.688454]
Unique values in mld_11: [32.89129 40.468742 18.960192 ... 37.603374 47.111786 32.160187]
Unique values in mld_12: [35.58445 48.187347 35.324955 ... 46.18387 50.637802 27.500666]
```

```
In [10]: df_mld.sample(5)
```

	Lat	Long	Year	mld_09	mld_10	mld_11	mld_12	mld_01	mld_02	mld_04	mld_05	mld_06	mld_07	mld_03
2896	15.000000	-107.0000	2002	16.903954	16.984198	19.306920	21.439137	16.497860	23.557991	22.905160	18.071108	15.481211	20.696335	22.267599
7189	15.000000	-107.0000	2010	13.071715	10.950776	15.594679	16.348997	26.245897	18.353033	17.849016	17.438065	22.289875	13.407757	14.78164
22076	42.241817	-19.1961	2021	24.715183	34.133858	38.718925	53.666298	101.153350	84.525930	50.515840	25.594824	15.047014	13.807727	16.08730
14992	42.241817	-19.1961	1998	25.190685	32.255085	47.185516	69.096250	164.372300	102.731964	98.633650	24.717926	17.227640	13.716379	14.01481
24199	-67.189000	145.7210	2001	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN

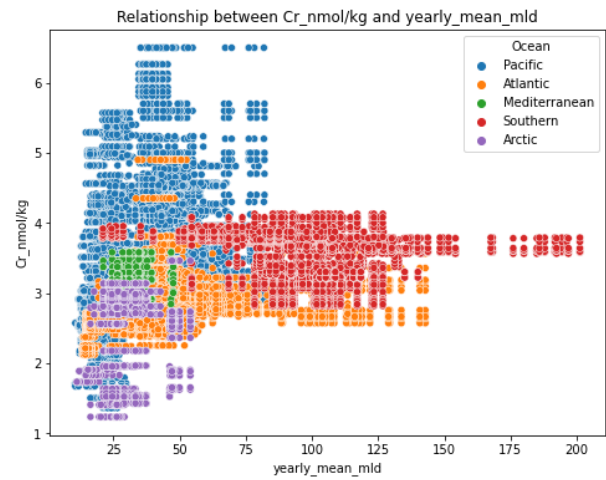
```
In [11]: for col in df_mld.columns:
if 'mld' in col:
print(f"Correlation between 'Cr_nmol/kg' and '{col}': {df_mld['Cr_nmol/kg'].corr(df_mld[col])}")
```

```
Correlation between 'Cr_nmol/kg' and 'mld_09': 0.18894732169041442
Correlation between 'Cr_nmol/kg' and 'mld_10': 0.1714894678008805
Correlation between 'Cr_nmol/kg' and 'mld_11': 0.10920656108655881
Correlation between 'Cr_nmol/kg' and 'mld_12': -0.00878505665673678
Correlation between 'Cr_nmol/kg' and 'mld_01': -0.06255206450950722
Correlation between 'Cr_nmol/kg' and 'mld_02': -0.036047074439683836
Correlation between 'Cr_nmol/kg' and 'mld_04': 0.06599786236657355
Correlation between 'Cr_nmol/kg' and 'mld_05': 0.17803507014595613
Correlation between 'Cr_nmol/kg' and 'mld_06': 0.20501011864004795
Correlation between 'Cr_nmol/kg' and 'mld_07': 0.19838237152518817
Correlation between 'Cr_nmol/kg' and 'mld_08': 0.192596327685962
Correlation between 'Cr_nmol/kg' and 'mld_03': -0.012898459732326674
Correlation between 'Cr_nmol/kg' and 'yearly_mean_mld': 0.15284549984126186
```

```
In [12]: import seaborn as sns
import matplotlib.pyplot as plt

mld_cols = ['yearly_mean_mld']

for col in mld_cols:
plt.figure(figsize=(8, 6))
sns.scatterplot(data=df_mld, x=col, y='Cr_nmol/kg', hue='Ocean')
plt.title(f'Relationship between Cr_nmol/kg and {col}')
plt.show()
```



In []: