

Smart Beach Canada

Smart Beach Phase II: Data-Driven Safety



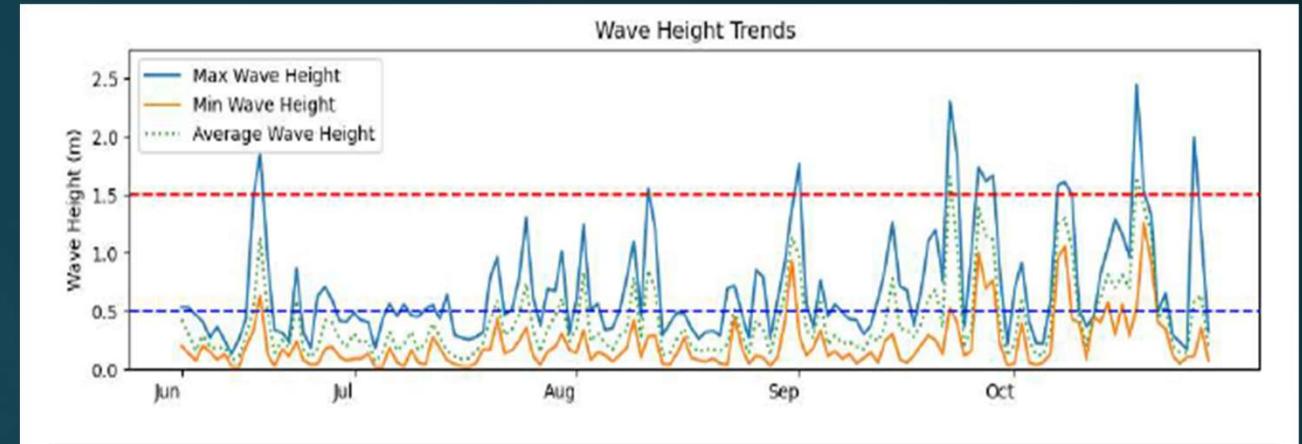
Wave Height Exploration

Wave Height Trends: June - October

```
In [43]: merged_df.groupby('conditions')[['maxwaveheight']].agg(['count','mean','median','min','max'])
```

```
Out[43]:
```

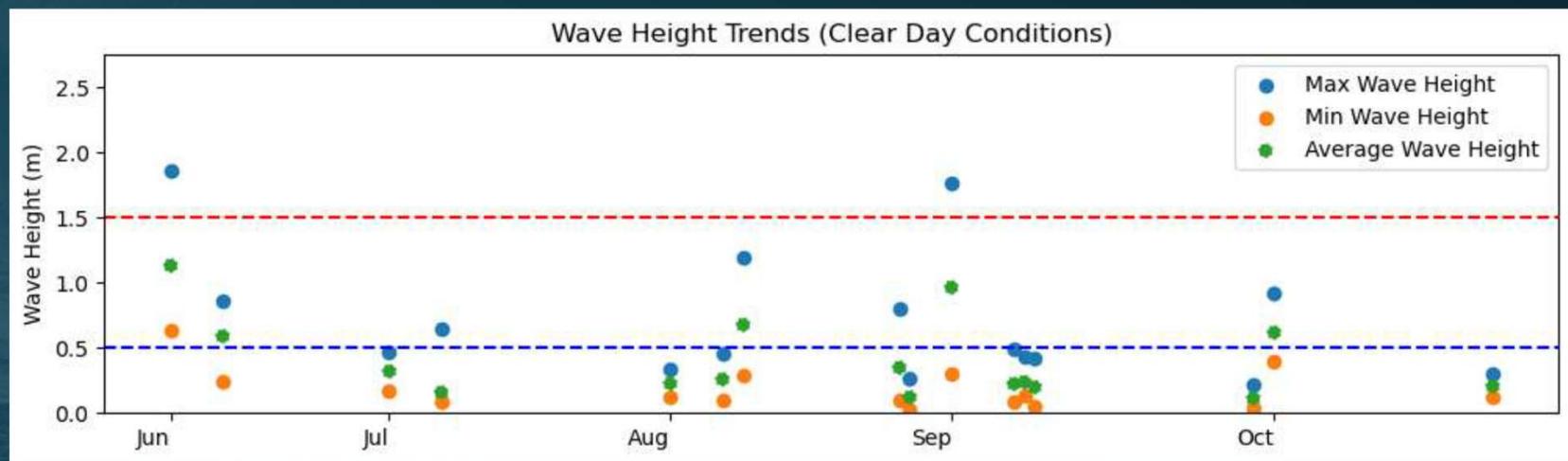
conditions	maxwaveheight				
	count	mean	median	min	max
clear-day	16	0.713188	0.479	0.215	1.855
partly-cloudy-day	58	0.574448	0.455	0.137	1.826
rain	69	0.783015	0.589	0.261	2.447
snow	1	1.538000	1.538	1.538	1.538



Wave Height Trends

Clear Day Conditions

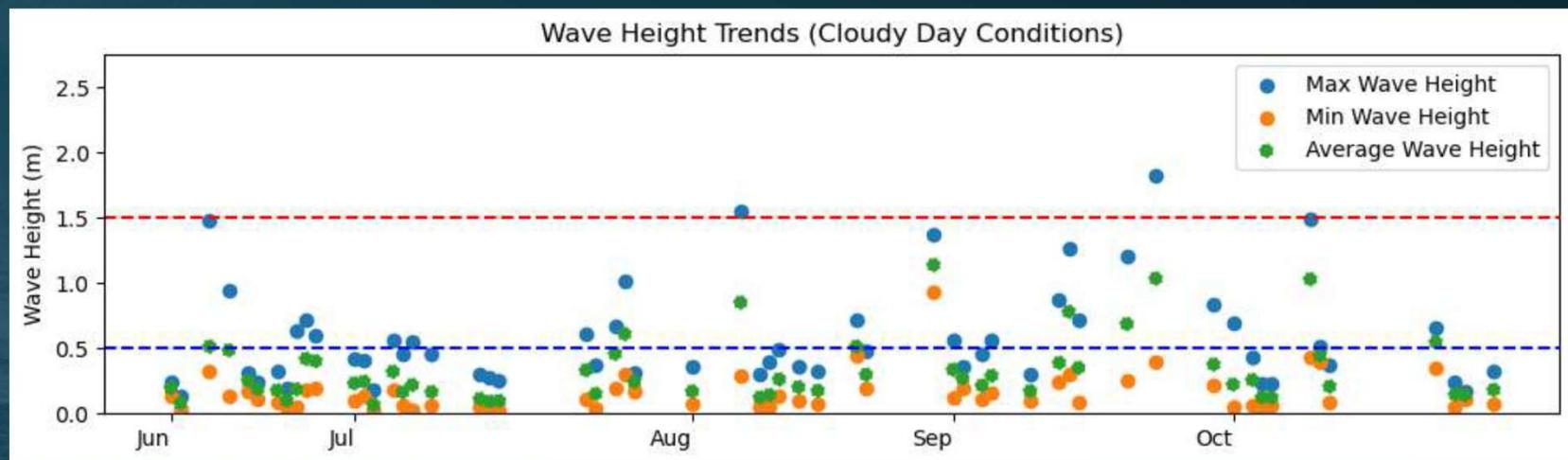
conditions	maxwaveheight				
	count	mean	median	min	max
	16	0.713188	0.479	0.215	1.855
partly-cloudy-day	58	0.574448	0.455	0.137	1.826
rain	69	0.783015	0.589	0.261	2.447
snow	1	1.538000	1.538	1.538	1.538



Wave Height Trends

Partly Cloudy Days

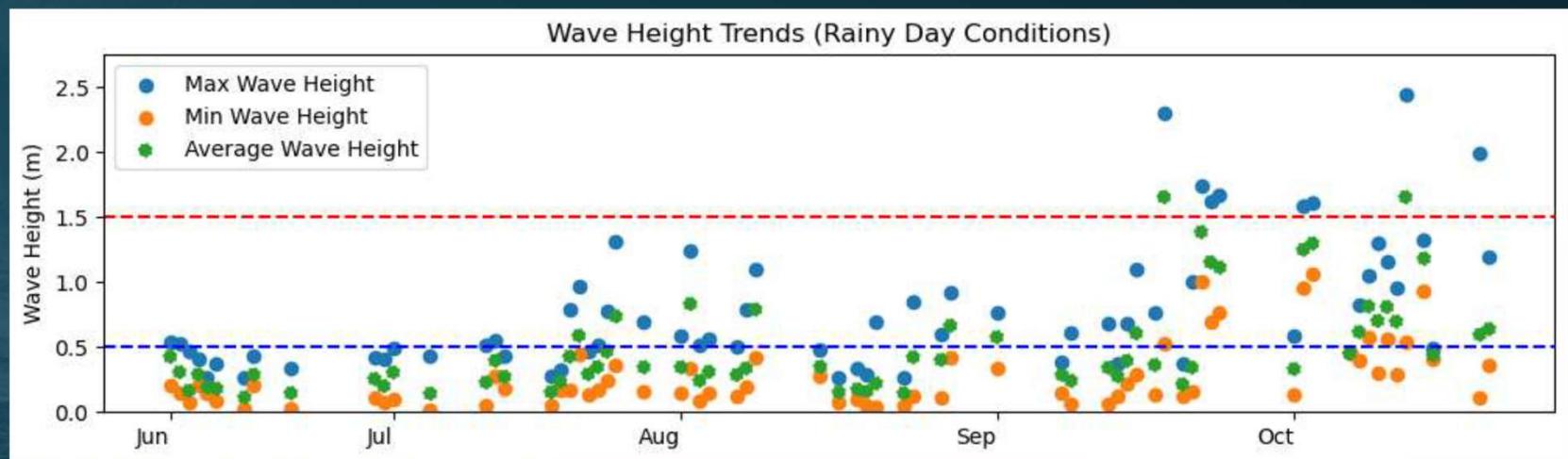
conditions	maxwaveheight				
	count	mean	median	min	max
clear-day	16	0.713188	0.479	0.215	1.855
partly-cloudy-day	58	0.574448	0.455	0.137	1.826
rain	69	0.783015	0.589	0.261	2.447
snow	1	1.538000	1.538	1.538	1.538



Wave Height Trends

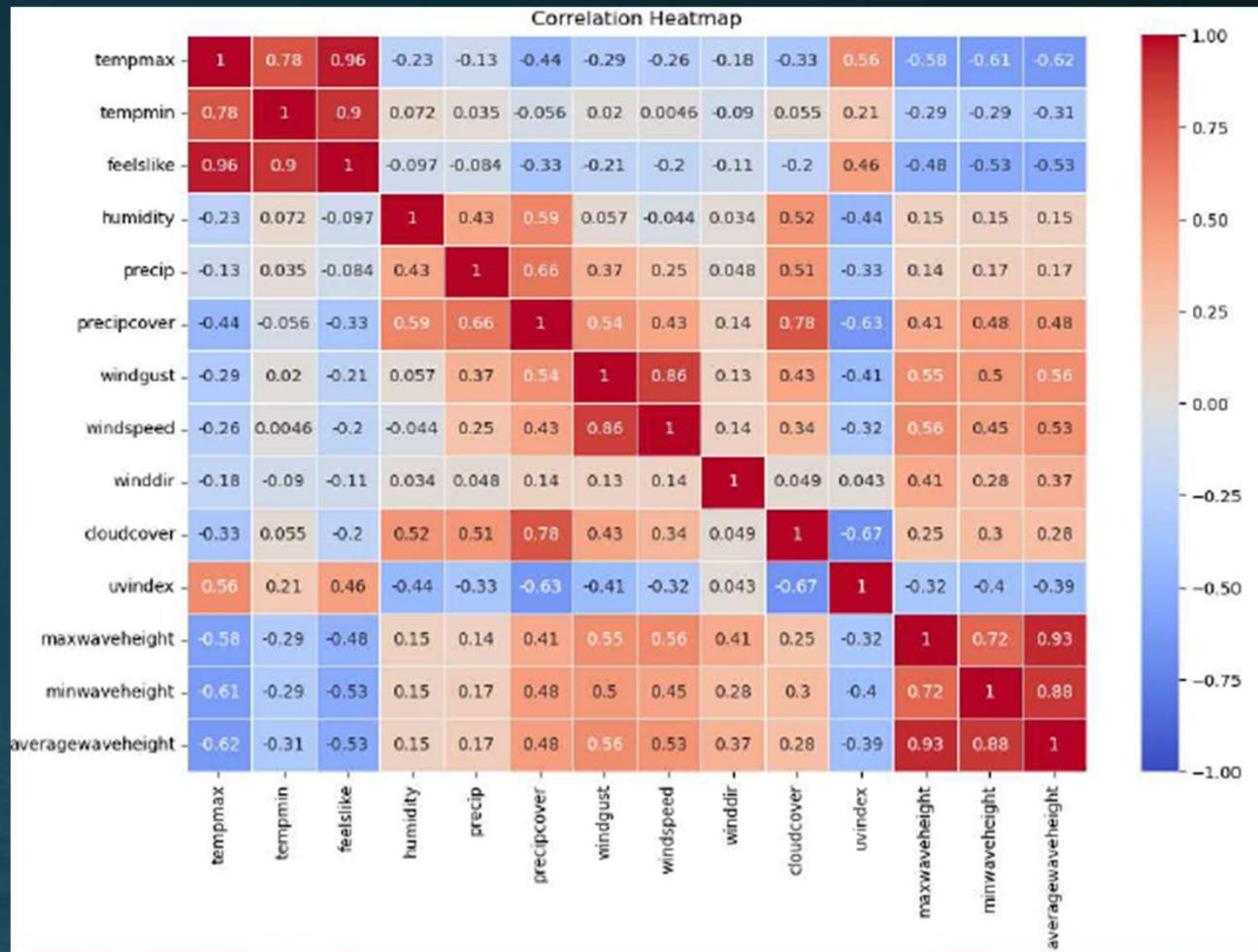
Rainy Days

	maxwaveheight				
	count	mean	median	min	max
conditions					
clear-day	16	0.713188	0.479	0.215	1.855
partly-cloudy-day	58	0.574448	0.455	0.137	1.826
rain	69	0.783015	0.589	0.261	2.447
snow	1	1.538000	1.538	1.538	1.538

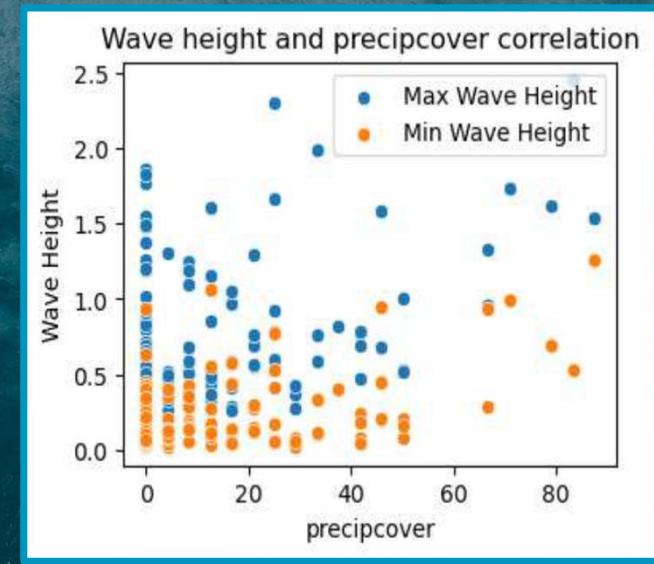
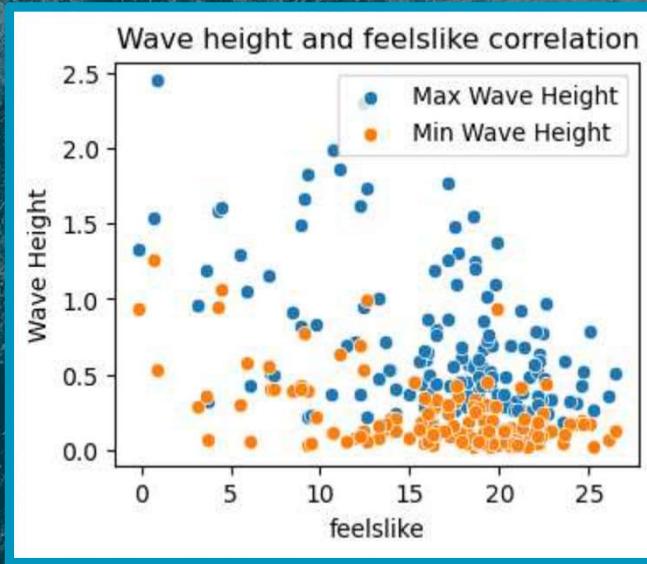
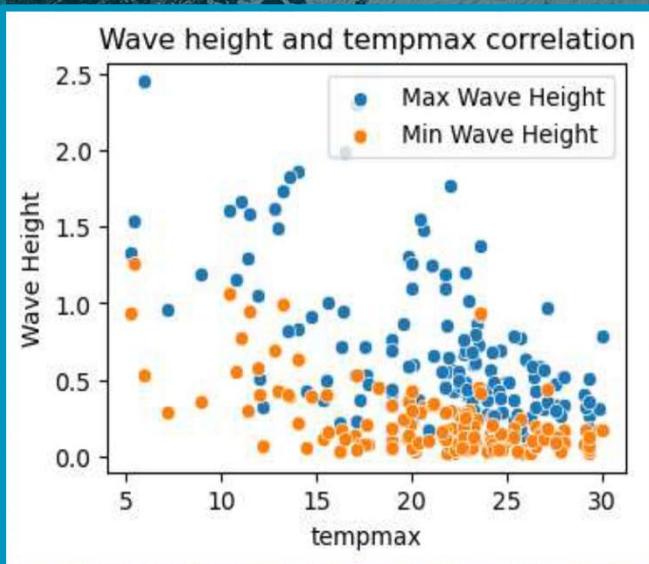


Correlation Heatmap

Wave Height vs Weather Variables



Correlation of wave height and other Variables



maxwaveheight

-0.58

maxwaveheight

-0.61

maxwaveheight

-0.48

minwaveheight

-0.53

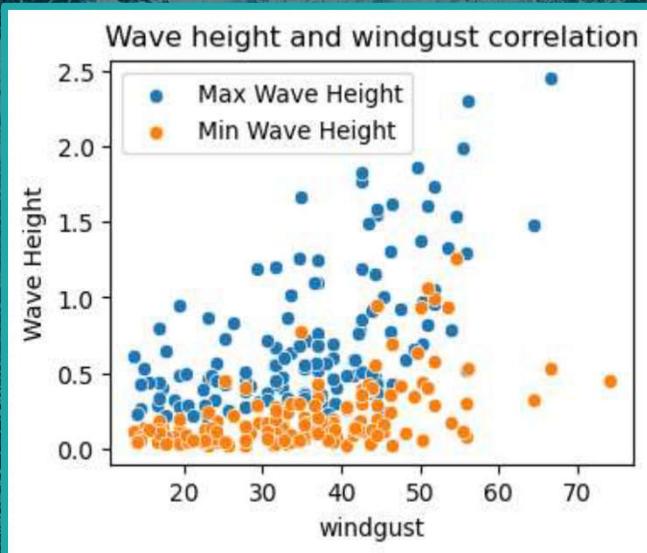
maxwaveheight

0.41

minwaveheight

0.48

Correlation of wave height and other Variables

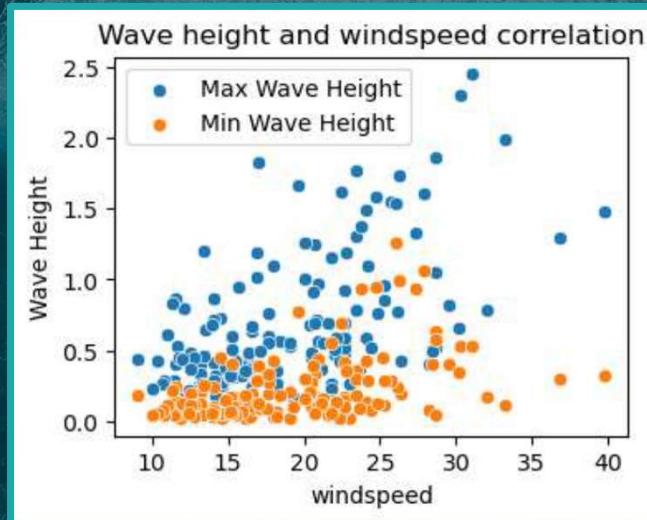


maxwaveheight

0.55

maxwaveheight

0.5



maxwaveheight

0.56

minwaveheight

0.45

Wind Direction Conversion: Degrees to Direction

```
def wind_direction(x):
    if 0 <= x < 45:
        return 'N-NE'
    elif 45 <= x < 90:
        return 'NE-E'
    elif 90 <= x < 135:
        return 'E-SE'
    elif 135 <= x < 180:
        return 'SE-S'
    elif 180 <= x < 225:
        return 'S-SW'
    elif 225 <= x < 270:
        return 'SW-W'
    elif 270 <= x < 315:
        return 'W-NW'
    elif 315 <= x < 360:
        return 'NW-N'
    else:
        return 'Unknown'

merged_df['windir_cat'] = merged_df['winddir'].apply(wind_direction)

print(merged_df)
```

Maximum Wave Height with Wind Direction

	count	mean	median	min	max
windir_cat					
W-NW	28	0.942071	0.6635	0.228	2.447
NW-N	20	1.082350	1.0240	0.276	2.305
N-NE	12	0.576667	0.4910	0.215	1.665
SW-W	21	0.752429	0.6740	0.261	1.611
S-SW	36	0.519389	0.4630	0.189	1.297
NE-E	8	0.462125	0.4125	0.137	0.915
E-SE	4	0.526500	0.5190	0.271	0.797
SE-S	15	0.336533	0.3050	0.173	0.606



Exploration of People Data

```
In [ ]: #People count
```

```
In [190]: PCdf = pd.read_csv('image_counts_all.csv')
```

```
In [191]: PCdf.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 312 entries, 0 to 311
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   File_name        312 non-null    object  
 1   Year             312 non-null    int64  
 2   Total People     312 non-null    int64  
 3   People in Water 312 non-null    int64  
 4   People on Sand   312 non-null    int64  
dtypes: int64(4), object(1)
memory usage: 12.3+ KB
```

```
In [195]: PCdf.head()
```

```
Out[195]:
```

	File_name	Year	Total People	People in Water	People on Sand
0	2022-08-25_15-35-39.jpg.rf.f29c550c1dfa9e50ded...	2022	14	1	13
1	2022-08-25_15-35-40.jpg.rf.5fbf5ece1d5e39664fd...	2022	16	2	14
2	2022-08-25_15-38-19.jpg.rf.592c8ceddf472ad8562...	2022	18	5	13
3	2022-08-25_15-38-19.jpg.rf.592c8ceddf472ad8562...	2022	18	5	13
4	2022-08-25_15-40-27.jpg.rf.ceb8e5145f8ec9240ed...	2022	15	3	12

Removing Extra Copies

```
In [199]: PCdf.File_name.str.contains('Copy')
```

```
Out[199]: 0      False
1      False
2      True
3     False
4      True
...
307    False
308    False
309    False
310    False
311    False
Name: File_name, Length: 312, dtype: bool
```

```
print(PeopleCountdf)
```

```
          File_name  Year  Total People \
0  2022-08-25_15-35-39.jpg.rf.f29c550c1dfa9e50ded...  2022       14
1  2022-08-25_15-35-40.jpg.rf.5fb5ce1d5e39664fd...  2022       16
3  2022-08-25_15-38-19.jpg.rf.592c8ceffd472ad8562...  2022       18
5  2022-08-25_15-40-27.jpg.rf.ceb8e5145f8ec9240ed...  2022       15
7  2022-08-25_15-41-33.jpg.rf.5d69e86774bd8db8810...  2022       18
...
307   ...
308   ...
309   ...
310   ...
311   ...

307  2023-08-17_06-40-17.jpg.rf.1a45cf9a03b13987afb...  2023       0
308  2023-08-21_15-55-19.jpg.rf.545cc1b11449621bb2b...  2023       19
309  2023-08-21_17-35-20.jpg.rf.3abe7a112ec79a468be...  2023       20
310  2023-08-22_06-20-17.jpg.rf.e0d9e001700c794b191...  2023       0
311  2023-08-22_08-10-18.jpg.rf.4f2622d52d4785cfa9c...  2023       2
```

```
People in Water  People on Sand
```

```
0              1            13
1              2            14
3              5            13
5              3            12
7              3            15
...
307             0            0
308             3            16
309             0            20
310             0            0
311             0            2
```

```
[261 rows x 5 columns]
```

Extracting Data and Time from File_name

```
In [215]: PeopleCountdf = PeopleCountdf.copy()  
  
PeopleCountdf['DateTime'] = PeopleCountdf['File_name'].str.extract(r'(\d{4}-\d{2}-\d{2}_\d{2}-\d{2}-\d{2})')  
PeopleCountdf['DateTime'] = pd.to_datetime(PeopleCountdf['DateTime'], format='%Y-%m-%d %H-%M-%S', errors='coerce')
```

```
In [216]: PeopleCountdf.head()
```

```
Out[216]:
```

	File_name	Year	Total People	People in Water	People on Sand	DateTime
0	2022-08-25_15-35-39.jpg.rf.f29c550c1dfa9e50ded...	2022	14	1	13	2022-08-25 15:35:39
1	2022-08-25_15-35-40.jpg.rf.5bfb5ece1d5e39664fd...	2022	16	2	14	2022-08-25 15:35:40
3	2022-08-25_15-38-19.jpg.rf.592c8ceddf472ad8562...	2022	18	5	13	2022-08-25 15:38:19
5	2022-08-25_15-40-27.jpg.rf.ceb8e5145f8ec9240ed...	2022	15	3	12	2022-08-25 15:40:27
7	2022-08-25_15-41-33.jpg.rf.5d69e86774bd8db8810...	2022	18	3	15	2022-08-25 15:41:33

```
PeopleCount2022['DateTime'].value_counts()
```

```
2022-08-27 17:20:16    2  
2022-08-27 17:00:16    2  
2022-08-25 15:35:39    1  
2022-09-25 07:10:16    1  
2022-09-17 16:10:16    1  
..  
2022-08-27 16:00:16    1  
2022-08-27 16:10:15    1  
2022-08-27 16:40:16    1  
2022-08-27 17:10:15    1  
2022-10-27 08:50:16    1  
Name: DateTime, Length: 218, dtype: int64
```

Limitations Due to No Consistency

```
In [245]: date_counts = PeopleCount2022['DateTime'].dt.date.value_counts().reset_index()
date_counts.columns = ['Date', 'Count']
print(date_counts)
```

	Date	Count
0	2022-08-27	44
1	2022-08-26	27
2	2022-08-25	15
3	2022-08-28	14
4	2022-08-29	7
5	2022-09-30	7
6	2022-10-18	6
7	2022-09-10	5
8	2022-09-13	5
9	2022-10-14	5
10	2022-10-09	4
11	2022-09-29	4
12	2022-10-01	4
13	2022-10-07	4
14	2022-09-02	4
15	2022-10-19	3
16	2022-09-14	3
17	2022-09-11	3
18	2022-10-08	3
19	2022-10-16	3
20	2022-09-19	3
21	2022-10-10	2
22	2022-08-31	2
23	2022-10-06	2
24	2022-10-20	2
25	2022-10-02	2
26	2022-10-13	2
27	2022-09-07	2
28	2022-09-04	2
29	2022-09-08	2
30	2022-09-25	2
31	2022-09-24	2
32	2022-09-22	2
33	2022-10-21	2
34	2022-09-17	2
35	2022-09-09	2
36	2022-09-27	2
37	2022-10-22	1
38	2022-10-25	1
39	2022-10-23	1
40	2022-10-24	1
41	2022-09-28	1
42	2022-10-15	1
43	2022-10-12	1
44	2022-10-11	1
45	2022-10-05	1
46	2022-09-20	1
47	2022-09-15	1
48	2022-09-12	1
49	2022-09-06	1
50	2022-09-05	1
51	2022-10-27	1

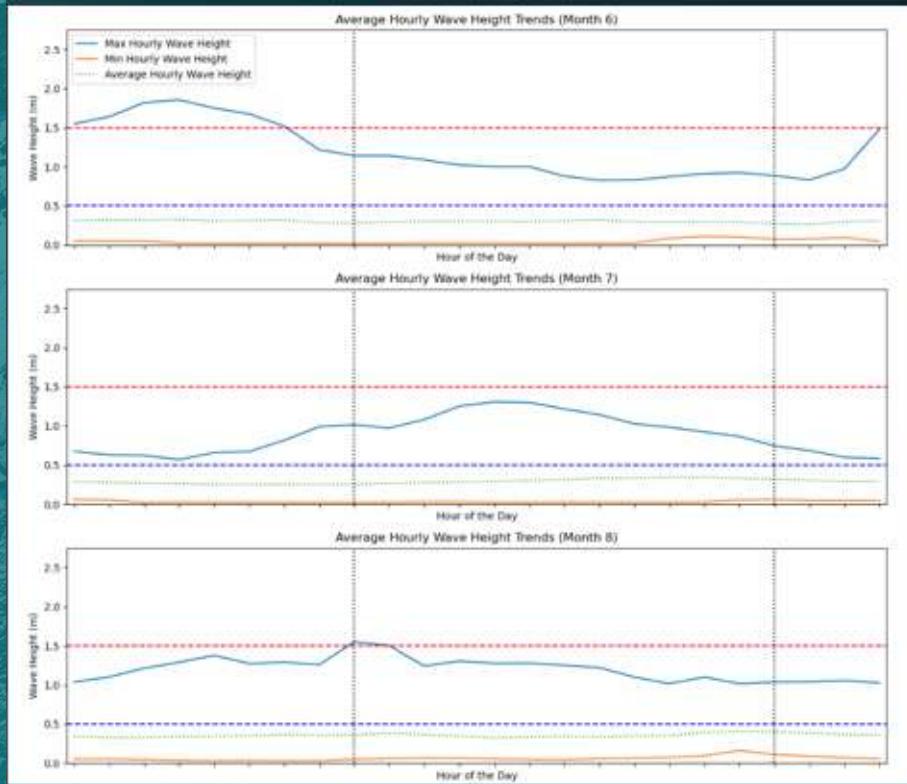
Year	Total People	People in Water	People on Sand	DateTime
2022	14	1	13	2022-08-25 15:35:39
2022	16	2	14	2022-08-25 15:35:40
2022	18	5	13	2022-08-25 15:38:19
2022	15	3	12	2022-08-25 15:40:27
2022	18	3	15	2022-08-25 15:41:33

Some days have too many readings while some days do not have enough.

Some readings are from night hours while some are from day

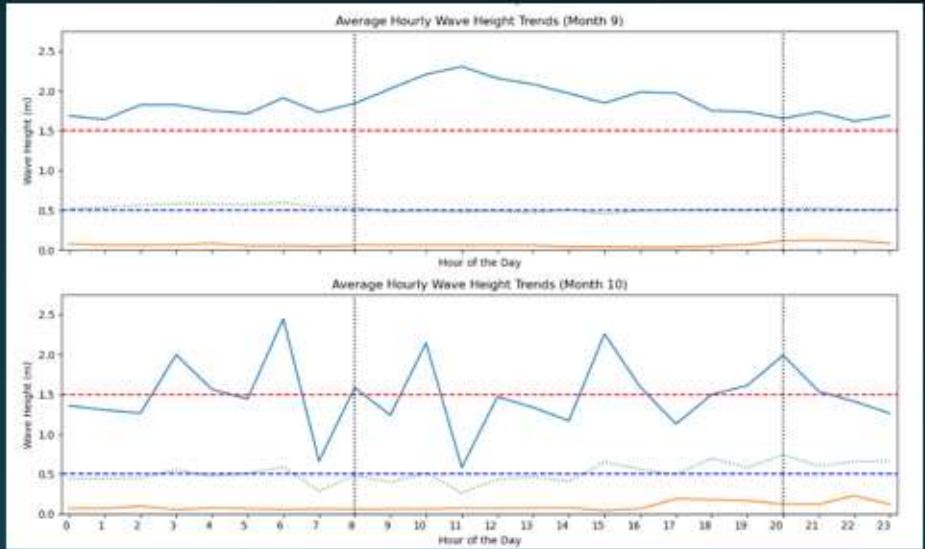
Data inconsistency and limited availability can affect the output negatively

Why Time Matters when Collecting Data



June - August

Max, Min and Average Wave height trends
Showcasing hourly variations with separators at
8AM and 8PM as a high beach usage band



September - October

Maximum Wave heights: 8:00AM - 8:00PM

```
In [141]: buoy_data['time'] = pd.to_datetime(buoy_data['time (UTC)'])

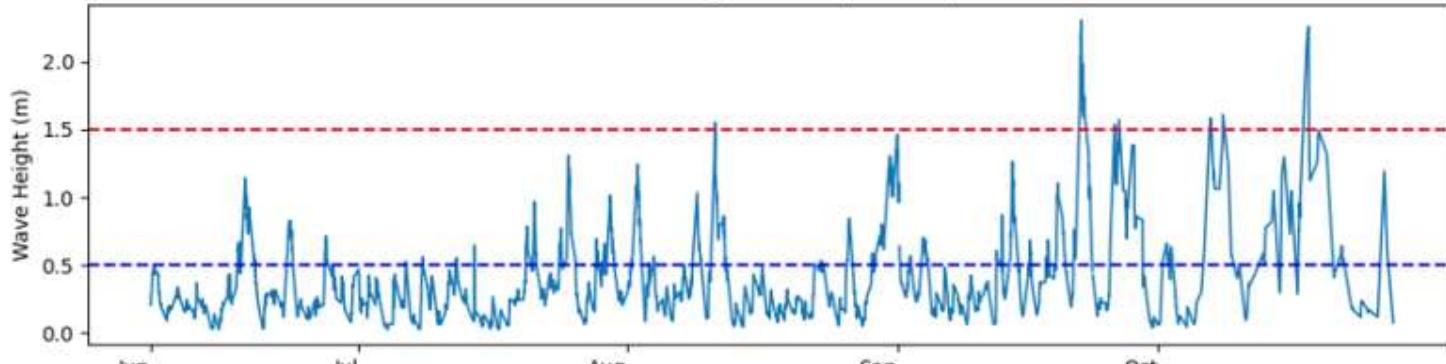
buoy_data.set_index('time', inplace=True)

filtered_data = buoy_data.between_time('08:00:00', '20:00:00')
filtered_data.reset_index(inplace=True)
```

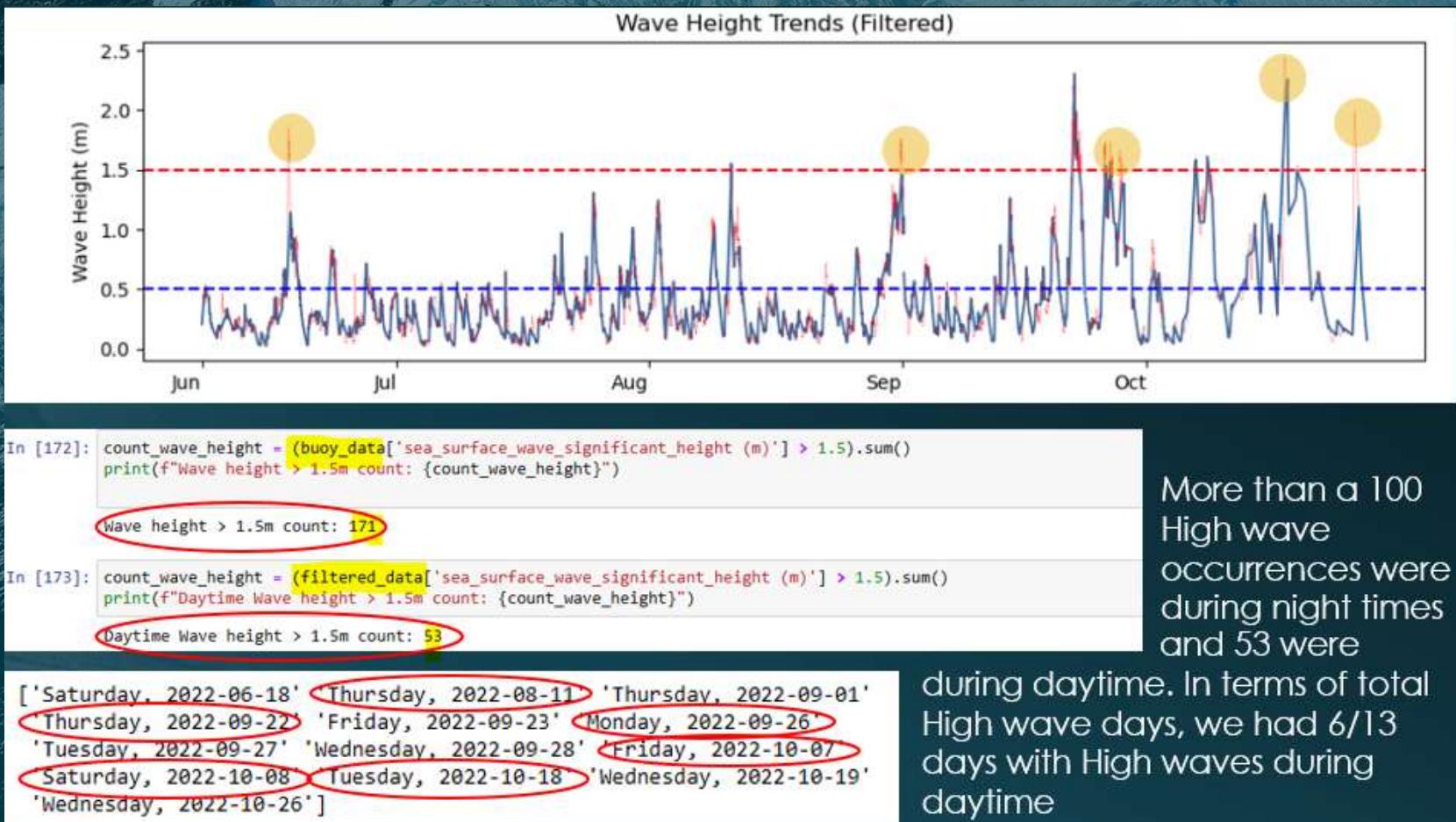
```
In [142]: filtered_data.info() #From 8000 rows we went to 4000
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4207 entries, 0 to 4206
Data columns (total 41 columns):
 #   Column           Non-Null Count  Dtype  
 ---  -- 
 0   time             4207 non-null   datetime64[ns, UTC]
 1   time (UTC)       4207 non-null   object  
 2   air_pressure_at_mean_sea_level (Pa) 4207 non-null   float64
 ...  ...
```

Wave Height Trends (Filtered)



Wave Data In terms of High Wave Days



Daytime High Wave Days

['Thursday, 2022-08-11' 'Thursday, 2022-09-22' 'Monday, 2022-09-26'
 'Friday, 2022-10-07' 'Saturday, 2022-10-08' 'Tuesday, 2022-10-18']

	time	weekday	waveheight				
2276	2022-08-11 08:20:00+00:00	Thursday	1.513	3741	2022-09-22 16:00:00+00:00	Thursday	1.807
2277	2022-08-11 08:40:00+00:00	Thursday	1.551	3742	2022-09-22 16:20:00+00:00	Thursday	1.769
2278	2022-08-11 09:00:00+00:00	Thursday	1.503	3743	2022-09-22 16:40:00+00:00	Thursday	1.986
3717	2022-09-22 08:00:00+00:00	Thursday	1.722	3744	2022-09-22 17:00:00+00:00	Thursday	1.972
3718	2022-09-22 08:20:00+00:00	Thursday	1.733	3745	2022-09-22 17:20:00+00:00	Thursday	1.630
3719	2022-09-22 08:40:00+00:00	Thursday	1.843	3746	2022-09-22 17:40:00+00:00	Thursday	1.730
3720	2022-09-22 09:00:00+00:00	Thursday	1.843	3747	2022-09-22 18:00:00+00:00	Thursday	1.717
3721	2022-09-22 09:20:00+00:00	Thursday	2.028	3748	2022-09-22 18:20:00+00:00	Thursday	1.752
3722	2022-09-22 09:40:00+00:00	Thursday	2.027	3749	2022-09-22 18:40:00+00:00	Thursday	1.592
3723	2022-09-22 10:00:00+00:00	Thursday	2.024	3750	2022-09-22 19:00:00+00:00	Thursday	1.736
3724	2022-09-22 10:20:00+00:00	Thursday	2.076	3751	2022-09-22 19:20:00+00:00	Thursday	1.716
3725	2022-09-22 10:40:00+00:00	Thursday	2.208	3752	2022-09-22 19:40:00+00:00	Thursday	1.645
3726	2022-09-22 11:00:00+00:00	Thursday	2.175	3753	2022-09-22 20:00:00+00:00	Thursday	1.651
3727	2022-09-22 11:20:00+00:00	Thursday	2.129	3865	2022-09-26 08:00:00+00:00	Monday	1.531
3728	2022-09-22 11:40:00+00:00	Thursday	2.305	3866	2022-09-26 08:20:00+00:00	Monday	1.536
3729	2022-09-22 12:00:00+00:00	Thursday	2.158	3867	2022-09-26 08:40:00+00:00	Monday	1.511
3730	2022-09-22 12:20:00+00:00	Thursday	2.080	3898	2022-09-26 19:20:00+00:00	Monday	1.540
3731	2022-09-22 12:40:00+00:00	Thursday	2.084	3899	2022-09-26 19:40:00+00:00	Monday	1.540
3732	2022-09-22 13:00:00+00:00	Thursday	2.084	3900	2022-09-26 20:00:00+00:00	Monday	1.572
3733	2022-09-22 13:20:00+00:00	Thursday	1.752	4106	2022-10-07 08:00:00+00:00	Friday	1.585
3734	2022-09-22 13:40:00+00:00	Thursday	1.793	4107	2022-10-07 10:00:00+00:00	Friday	1.543
3735	2022-09-22 14:00:00+00:00	Thursday	1.928	4115	2022-10-08 19:00:00+00:00	Saturday	1.611
3736	2022-09-22 14:20:00+00:00	Thursday	1.904	4160	2022-10-18 10:20:00+00:00	Tuesday	2.145
3737	2022-09-22 14:40:00+00:00	Thursday	1.969	4161	2022-10-18 15:20:00+00:00	Tuesday	2.260
3738	2022-09-22 15:00:00+00:00	Thursday	1.848	4162	2022-10-18 15:40:00+00:00	Tuesday	2.005
3739	2022-09-22 15:20:00+00:00	Thursday	1.752	4163	2022-10-18 16:00:00+00:00	Tuesday	1.591
3740	2022-09-22 15:40:00+00:00	Thursday	1.773				

August 8, Thursday

- 8AM – 9AM

September 22, Thursday

- 8AM – 8PM

September 26, Monday

- 8AM – 9AM
- 7PM – 8PM

October 7, Friday

- 8AM, 10AM

October 8, Saturday

- 7PM

October 18, Tuesday

- 10AM
- 3PM – 4PM

Why is Data Consistency Needed?

High Wave Days:

August 8, Thursday(3)

- 8AM – 9AM

September 22, Thursday(37)

- 8AM – 8PM

September 26, Monday(6)

- 8AM – 9AM
- 7PM – 8PM

October 7, Friday(2)

- 8AM, 10AM

October 8, Saturday(1)

- 7PM

October 18, Tuesday(4)

- 10AM
- 3PM – 4PM

People Count Data:

August 8, Thursday(0)

- No data

September 22, Thursday(2)

- 1:23PM, 6:54PM

September 26, Monday(0)

- No data

October 7, Friday(3)

- 9:10AM, 10:10AM, 10:37AM

October 8, Saturday(3)

- 12:30PM, 2:10PM, 3:10PM

October 18, Tuesday(6)

- 8:20AM, 11:20AM, 12:20PM,
- 1:30PM, 2:10PM, 3:40PM

Machine Learning Challenges:

Determining Maximum Number of People on any given day for predicting based on weather variables:

- Summation of all counts per day is not valid as the frequency of readings is not consistent and may provide misleading and skewed information.
- Average of all counts per day is not valid for the same reason because the time when readings were taken is not consistent, some days have readings taken at unusual times like early morning or late night when no one is at the beach while some days have more readings during day times.
- Max count among all readings for a given date would result in a similar problem as some dates/times which could potentially have higher traffic due to favorable weather might not have been recorded.

Proposed Solution:

We can take max count for each day; however, we will study the frequency of reading times for each day and limit the selection to a timeframe that had a significant number of readings taken every day within that frame.

Note: The predicted data may or may not provide correct results due to limitations from the input data. Although we will create a machine learning model for the scope of this project even if more data is not available, retraining the model with a better dataset is highly advised.

Selecting Timeframes with most records

```
In [23]: PeopleCount2022['date'] = PeopleCount2022['DateTime'].dt.date  
PeopleCount2022['month'] = PeopleCount2022['DateTime'].dt.month  
PeopleCount2022['hour'] = PeopleCount2022['DateTime'].dt.hour  
PeopleCount2022
```

Out[23]:

	File_name	Year	Total People	People in Water	People on Sand	Datetime	date	month	hour
0	2022-08-25_15-35-39.jpg.rf.f29c550c1dfa9e50ded...	2022	14	1	13	2022-08-25 15:35:39	2022-08-25	8	15
1	2022-08-25_15-35-40.jpg.rf.5fb5ece1d5e39664fd...	2022	16	2	14	2022-08-25 15:35:40	2022-08-25	8	15
2	2022-08-25_15-38-19.jpg.rf.592c8ceddf472ad8562...	2022	18	5	13	2022-08-25 15:38:19	2022-08-25	8	15
3	2022-08-25_15-40-27.jpg.rf.cebb5e5145f8ec9240ed...	2022	15	3	12	2022-08-25 15:40:27	2022-08-25	8	15
4	2022-08-25_15-41-33.jpg.rf.5d69e86774bd8db8810...	2022	18	3	15	2022-08-25 15:41:33	2022-08-25	8	15
...
215	2022-10-22_15-43-52.jpg.rf.f0f727fb618ad2bf751...	2022	2	0	2	2022-10-22 15:43:52	2022-10-22	10	52
216	2022-10-23_17-20-16.jpg.rf.c9fb83271b930be8bd...	2022	4	0	4	2022-10-23 17:20:16	2022-10-23	10	52
217	2022-10-24_17-30-16.jpg.rf.359aea533c623c69372...	2022	7	0	7	2022-10-24 17:30:16	2022-10-24	10	52
218	2022-10-25_18-33-52.jpg.rf.1ee485d7e774f4d046c...	2022	2	0	2	2022-10-25 18:33:52	2022-10-25	10	52
219	2022-10-27_08-50-16.jpg.rf.b579b95f7074c1aa942...	2022	2	0	2	2022-10-27 08:50:16	2022-10-27	10	52

220 rows × 9 columns

(a) Extracting hours from Datetime to check hourly records

```
In [30]: PeopleCount2022['hour'].value_counts().sort_index()  
Out[30]: 6    2  
7    12  
8    10  
9     3  
10    9  
11   16  
12   15  
13   20  
14   25  
15   34  
16   27  
17   28  
18   19  
Name: hour, dtype: int64
```

```
In [28]: hourly_entries_per_day = PeopleCount2022.groupby(['date', 'hour']).size().reset_index(name='Count')  
print(hourly_entries_per_day)
```

	date	hour	Count
0	2022-08-25	15	7
1	2022-08-25	16	6
2	2022-08-25	17	2
3	2022-08-26	11	2
4	2022-08-26	13	1
...
128	2022-10-22	15	1
129	2022-10-23	17	1
130	2022-10-24	17	1
131	2022-10-25	18	1
132	2022-10-27	8	1

133 rows × 3 columns

(b) Grouping Data based on hours and checking for duplicates

Hourly records without duplicates and filtering the timeframe with most records

```
In [34]: total_distinct_hourly_counts = PeopleCount2022.groupby('hour')['date'].nunique().reset_index(name='Count')
print(total_distinct_hourly_counts)
```

	hour	Count
0	6	2
1	7	10
2	8	10
3	9	3
4	10	8
5	11	9
6	12	10
7	13	12
8	14	14
9	15	17
10	16	12
11	17	12
12	18	14

```
In [35]: filtered_data = PeopleCount2022[(PeopleCount2022['hour'] >= 12) & (PeopleCount2022['hour'] <= 17)]
```

Out[53]:

	Datetime	File_name	Year	Total People	People in Water	People on Sand	month	hour
0	2022-08-25 15:35:39	2022-08-25_15-35-39.jpg.rf.f29c550c1dfa9e50ded...	2022	14	1	13	8	15
1	2022-08-25 15:35:40	2022-08-25_15-35-40.jpg.rf.5bfb5ece1d5e39664fd...	2022	16	2	14	8	15
2	2022-08-25 15:38:19	2022-08-25_15-38-19.jpg.rf.592c8ceffd472ad8562...	2022	18	5	13	8	15
3	2022-08-25 15:40:27	2022-08-25_15-40-27.jpg.rf.ceb8e5145f8ec9240ed...	2022	15	3	12	8	15
4	2022-08-25 15:41:33	2022-08-25_15-41-33.jpg.rf.5d69e86774bd8db8810...	2022	18	3	15	8	15
...
144	2022-10-18 15:40:15	2022-10-18_15-40-15.jpg.rf.d3ff33ba3fe8db1d542...	2022	5	2	3	10	15
145	2022-10-21 13:30:16	2022-10-21_13-30-16.jpg.rf.fc5f4afa75e1c39ffcd...	2022	2	0	2	10	13
146	2022-10-22 15:43:52	2022-10-22_15-43-52.jpg.rf.f0f727fb618ad2bf751...	2022	2	0	2	10	15
147	2022-10-23 17:20:16	2022-10-23_17-20-16.jpg.rf.c9fb83271b930be8bd...	2022	4	0	4	10	17
148	2022-10-24 17:30:16	2022-10-24_17-30-16.jpg.rf.359aea533c623c69372...	2022	7	0	7	10	17

149 rows × 8 columns

Taking Max People count for each unique day within the 12PM - 5PM timeframe

```
In [61]: MaxPeopleCount = filtered_data.groupby(filtered_data['DateTime'].dt.date)['Total People'].max().reset_index()
MaxPeopleCount.columns = ['Date', 'Daily Max People']
MaxPeopleCount
```

Out[61]:

	Date	Max Total People	Month	Day	Day_Name					
0	2022-08-25	22	8	25	Thursday					
1	2022-08-26	25	8	26	Friday					
2	2022-08-27	32	8	27	Saturday					
3	2022-08-28	29	8	28	Sunday					
4	2022-08-29	28	8	29	Monday					
5	2022-09-02	23	9	2	Friday					
6	2022-09-04	21	9	4	Sunday					
7	2022-09-05	19	9	5	Monday					
8	2022-09-06	13	9	6	Tuesday					
9	2022-09-07	16	9	7	Wednesday					
10	2022-09-09	1	9	9	Friday					
11	2022-09-10	15	9	10	Saturday					
12	2022-09-11	14	9	11	Sunday					
13	2022-09-12	2	9	12	Monday					
14	2022-09-13	3	9	13	Tuesday					
15	2022-09-14	10	9	14	Wednesday					
16	2022-09-15	3	9	15	Thursday					
17	2022-09-17	21	9	17	Saturday					
18	2022-09-19	10	9	19	Monday					
19	2022-09-22	7	9	22	Thursday					
20	2022-09-25	0	9	25	Sunday					
21	2022-09-27			0	9	27	Tuesday			
22	2022-09-28			0	9	28	Wednesday			
23	2022-09-29			2	9	29	Thursday			
24	2022-09-30			3	9	30	Friday			
25	2022-10-01			4	10	1	Saturday			
26	2022-10-08			7	10	8	Saturday			
27	2022-10-09			7	10	9	Sunday			
28	2022-10-12			0	10	12	Wednesday			
29	2022-10-13			3	10	13	Thursday			
30	2022-10-14			4	10	14	Friday			
31	2022-10-16			3	10	16	Sunday			
32	2022-10-18			5	10	18	Tuesday			
33	2022-10-21			2	10	21	Friday			
34	2022-10-22			2	10	22	Saturday			
35	2022-10-23			4	10	23	Sunday			
36	2022-10-24			7	10	24	Monday			

Selecting Timeframes with most records

Checking the hourly data within the timeframe on daily level

- First figure shows the record with the maximum people for any given hour
- Second figure shows the number of records for any given hour

Max People	Dates	2022-08-25	2022-08-26	2022-08-27	2022-08-28	2022-08-29	2022-09-02	2022-09-04	2022-09-05	2022-09-06
Hours										
12.00					24	26				
13.00				15	29	12	28	17		
14.00				22	29	21			4	
15.00		20	25	32	17		8	23	21	13
16.00		22	22	29	29			6		
17.00		16	19	28	10	22			19	
Max Value		22	25	32	29	28	23	21	19	13

Max People Count	Dates	2022-08-25	2022-08-26	2022-08-27	2022-08-28	2022-08-29	2022-09-02	2022-09-04	2022-09-05	2022-09-06
Hours										
12.00					6	1				
13.00			1	8	1	1	1			
14.00			5	8	1			1		
15.00		7	5	5	2	1	2	1		1
16.00		6	4	3	4		1			
17.00		2	6	8	1	3			1	
Grand Total		15	21	38	10	5	4	2	1	1

Selecting Timeframes with most records

Merging the selected People Count data to other weather and wave variables

	Date	Max Total People	Month	Day	Day_Name	tempmax	tempmin	temp	feelslikemax	feelslikemin	...	winddir	cloudcover	visibility	uvindex	conditions	maxwaveheight	minwaveheight	averagewaveheight	month_name	windir_cat
0	2022-08-25	22	8	25	Thursday	24.3	17.3	20.8	24.3	17.3	...	132.9	63.5	24.0	7	rain	0.271	0.052	0.145986	Aug	E-SE
1	2022-08-26	25	8	26	Friday	21.8	12.7	19.1	21.8	12.7	...	347.5	75.3	20.3	5	rain	0.854	0.121	0.423131	Aug	NW-N
2	2022-08-27	32	8	27	Saturday	23.3	9.2	16.5	23.3	8.9	...	122.3	9.6	24.9	9	clear-day	0.797	0.102	0.349016	Aug	E-SE
3	2022-08-28	29	8	28	Sunday	27.8	13.8	20.9	28.4	13.8	...	174.9	19.3	24.0	9	clear-day	0.264	0.030	0.121794	Aug	SE-S
4	2022-08-29	28	8	29	Monday	26.3	21.1	23.6	26.3	21.1	...	191.3	74.6	24.0	6	rain	0.594	0.105	0.401742	Aug	S-SW
5	2022-09-02	23	9	2	Friday	26.9	17.2	21.8	27.4	17.2	...	185.3	29.4	24.0	9	partly-cloudy-day	0.567	0.121	0.337455	Sep	S-SW
6	2022-09-04	21	9	4	Sunday	18.9	15.8	16.5	18.9	15.8	...	57.6	74.5	24.0	1	rain	0.768	0.336	0.577739	Sep	NE-E
7	2022-09-05	19	9	5	Monday	21.7	15.4	17.8	21.7	15.4	...	63.7	69.0	24.0	8	partly-cloudy-day	0.456	0.114	0.215522	Sep	NE-E
8	2022-09-06	13	9	6	Tuesday	21.6	12.7	17.4	21.6	12.7	...	34.8	21.8	24.0	9	partly-cloudy-day	0.558	0.158	0.291101	Sep	N-NE
9	2022-09-07	16	9	7	Wednesday	22.4	9.9	16.1	22.4	9.5	...	9.6	15.8	24.0	9	clear-day	0.492	0.092	0.224268	Sep	N-NE
10	2022-09-09	1	9	9	Friday	24.5	12.6	19.0	24.5	12.6	...	183.2	13.0	24.0	9	clear-day	0.423	0.052	0.199431	Sep	S-SW
30	2022-10-14	4	10	14	Friday	11.9	5.9	8.5	11.9	2.3	...	194.7	44.6	23.0	7	rain	1.046	0.577	0.814500	Oct	S-SW
31	2022-10-16	3	10	16	Sunday	10.8	6.7	8.7	10.8	4.0	...	230.8	40.5	22.5	6	rain	1.160	0.558	0.809615	Oct	SW-W
32	2022-10-18	5	10	18	Tuesday	6.0	2.1	4.6	2.3	-0.8	...	286.6	83.1	22.6	3	rain	2.447	0.534	1.655364	Oct	W-NW
33	2022-10-21	2	10	21	Friday	15.5	2.0	9.1	15.5	-1.6	...	180.8	30.5	23.0	6	rain	0.497	0.405	0.436667	Oct	S-SW
34	2022-10-22	2	10	22	Saturday	21.1	11.9	15.8	21.1	11.9	...	184.6	28.8	24.0	7	partly-cloudy-day	0.656	0.350	0.548857	Oct	S-SW
35	2022-10-23	4	10	23	Sunday	23.2	12.2	16.9	23.2	12.2	...	153.8	18.2	24.0	6	clear-day	0.296	0.122	0.206455	Oct	SE-S
36	2022-10-24	7	10	24	Monday	22.6	10.8	16.0	22.6	10.8	...	147.8	36.1	24.0	6	partly-cloudy-day	0.237	0.053	0.146818	Oct	SE-S

37 rows × 28 columns

Check for correlations and Patterns

Checking records available based on weather conditions and monthly outlook.

```
In [72]: allVals.groupby('conditions')[['Max Total People']].agg(['count','mean','median','min','max'])
```

```
Out[72]:
```

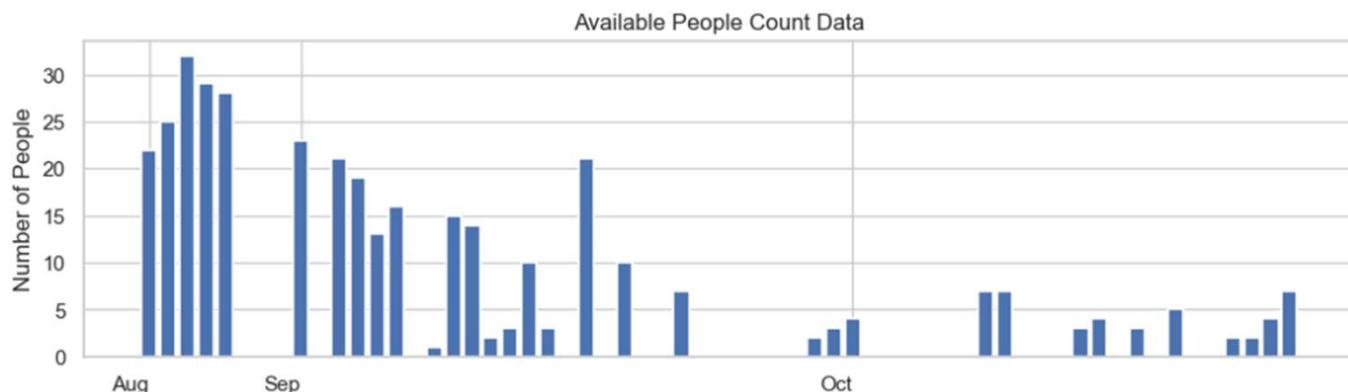
conditions	Max Total People				
	count	mean	median	min	max
clear-day	6	14.166667	10.0	1	32
partly-cloudy-day	12	9.000000	7.0	2	23
rain	19	9.157895	5.0	0	28

```
In [91]: plt.figure(figsize=(12, 3))
```

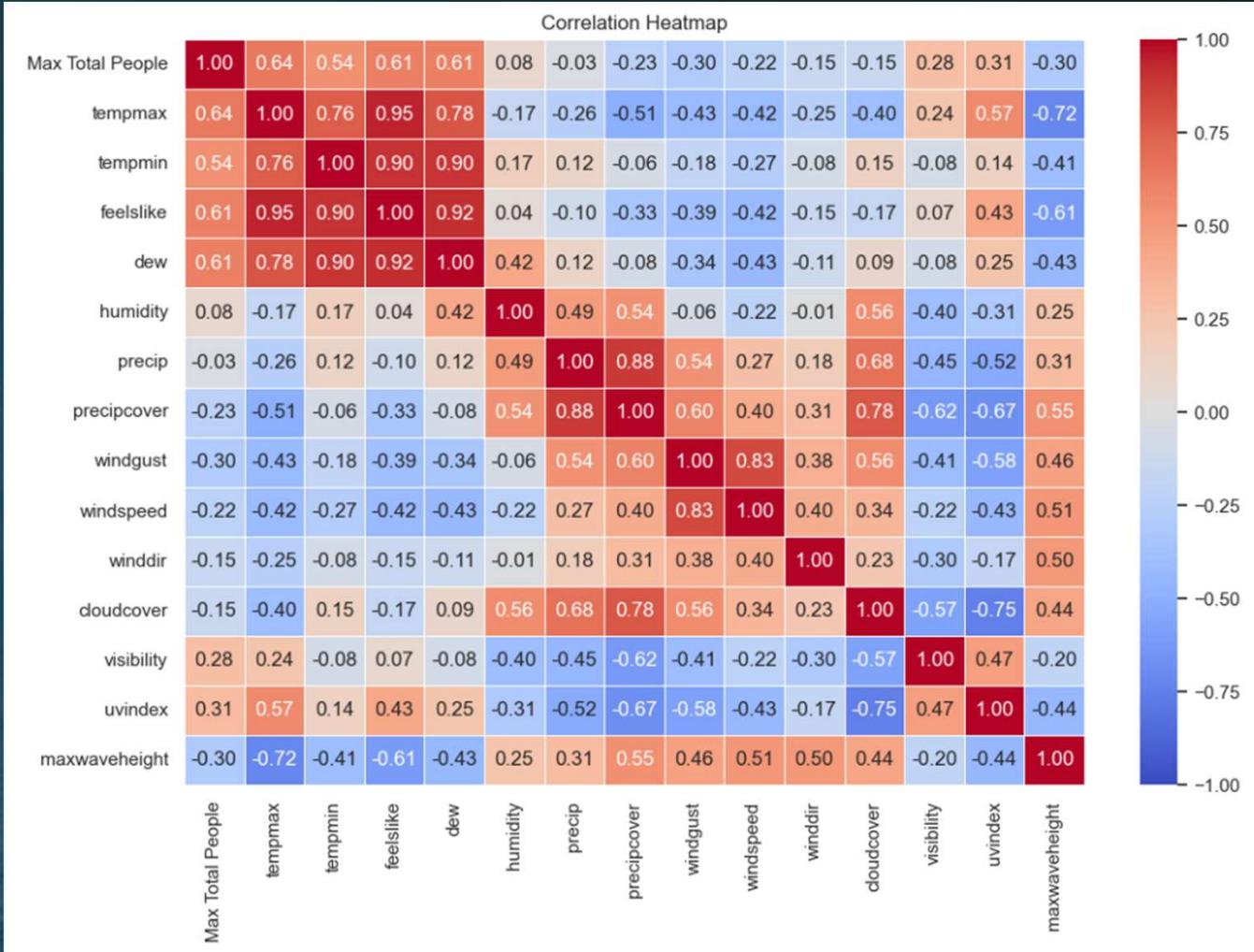
```
plt.bar(allVals['Date'], allVals['Max Total People'], linewidth=1.2)

plt.title('Available People Count Data')
plt.ylabel('Number of People')
month_index = allVals.groupby('Month').head(1).index
plt.xticks(allVals.loc[month_index, 'Date'], allVals.loc[month_index, 'month_name'], rotation=0, ha='right')

plt.show()
```



Check for correlations and Patterns



Check for correlations and Patterns

Scatter Plot for Max People vs Max temperature.

```
In [100]: plt.figure(figsize=(6, 3))
sns.scatterplot(x='tempmax', y='Max Total People', data=allVals)
plt.title('Number of People vs Max temp')
plt.show()
```

