Project 1

Performing Analysis of Meteorological Data

Objective: Perform Data Cleaning, Perform Analysis for Testing the (given) Hypothesisand finally put-forth your conclusion by writing a blog article.

Data Source: https://www.kaggle.com/muthuj7/weather-dataset (<a href="https://www.kaggle.com/muthuj7/weather-dataset (<a href="https://www.kaggle.com/muthuj7/weather-datase

The Meteorological Dataset Consists of:

- Total Number of Data Point: 96453
- Total Number of Attribute/Columns in data set is 11.

Attribute Information:

- 1. Formatted Date
- 2. Summary
- 3. Precip Type
- 4. Temperature (C)
- 5. Apparent Temperature (C)
- 6. Humidity
- 7. Wind Speed (km/h)
- 8. Wind Bearing (degrees)
- 9. Visibility (km)
- 10. Pressure (millibars)
- 11. Daily Summary

Loading the Data

In [1]:

```
#Python Library
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
%matplotlib inline
```

In [2]:

```
#Reading weatherHistory.csv file
df = pd.read_csv("weatherHistory.csv")
df.head(10)
```

Out[2]:

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Wind Bearing (degrees)	Vis
0	2006-04-01 00:00:00.000 +0200	Partly Cloudy	rain	9.472222	7.388889	0.89	14.1197	251	15
1	2006-04-01 01:00:00.000 +0200	Partly Cloudy	rain	9.355556	7.227778	0.86	14.2646	259	15
2	2006-04-01 02:00:00.000 +0200	Mostly Cloudy	rain	9.377778	9.377778	0.89	3.9284	204	14
3	2006-04-01 03:00:00.000 +0200	Partly Cloudy	rain	8.288889	5.944444	0.83	14.1036	269	15
4	2006-04-01 04:00:00.000 +0200	Mostly Cloudy	rain	8.755556	6.977778	0.83	11.0446	259	15
5	2006-04-01 05:00:00.000 +0200	Partly Cloudy	rain	9.222222	7.111111	0.85	13.9587	258	14
6	2006-04-01 06:00:00.000 +0200	Partly Cloudy	rain	7.733333	5.522222	0.95	12.3648	259	9
7	2006-04-01 07:00:00.000 +0200	Partly Cloudy	rain	8.772222	6.527778	0.89	14.1519	260	9
8	2006-04-01 08:00:00.000 +0200	Partly Cloudy	rain	10.822222	10.822222	0.82	11.3183	259	9
9	2006-04-01 09:00:00.000 +0200	Partly Cloudy	rain	13.772222	13.772222	0.72	12.5258	279	9

In [3]:

#shape of Dataset df.shape

Out[3]:

(96453, 11)

In [4]:

```
#Statistical Summary of DataFrame
df.describe()
```

Out[4]:

	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Wind Bearing (degrees)	Visibility (km)	
count	96453.000000	96453.000000	96453.000000	96453.000000	96453.000000	96453.000000	96
mean	11.932678	10.855029	0.734899	10.810640	187.509232	10.347325	1
std	9.551546	10.696847	0.195473	6.913571	107.383428	4.192123	
min	- 21.822222	- 27.716667	0.000000	0.000000	0.000000	0.000000	
25%	4.688889	2.311111	0.600000	5.828200	116.000000	8.339800	1
50%	12.000000	12.000000	0.780000	9.965900	180.000000	10.046400	1
75%	18.838889	18.838889	0.890000	14.135800	290.000000	14.812000	1
max	39.905556	39.344444	1.000000	63.852600	359.000000	16.100000	1
4							•

In [5]:

#Concise Summary of the DataFrame
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 96453 entries, 0 to 96452
Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	Formatted Date	96453 non-null	object
1	Summary	96453 non-null	object
2	Precip Type	95936 non-null	object
3	Temperature (C)	96453 non-null	float64
4	Apparent Temperature (C)	96453 non-null	float64
5	Humidity	96453 non-null	float64
6	Wind Speed (km/h)	96453 non-null	float64
7	Wind Bearing (degrees)	96453 non-null	int64
8	Visibility (km)	96453 non-null	float64
9	Pressure (millibars)	96453 non-null	float64
10	Daily Summary	96453 non-null	object
	67		

dtypes: float64(6), int64(1), object(4)

memory usage: 8.1+ MB

In [6]:

```
#Missing Values on Dataset from String to Date Time
df.isnull().sum()
```

Out[6]:

Formatted Date	0
Summary	0
Precip Type	517
Temperature (C)	0
Apparent Temperature (C)	0
Humidity	0
Wind Speed (km/h)	0
Wind Bearing (degrees)	0
Visibility (km)	0
Pressure (millibars)	0
Daily Summary	0
dtype: int64	

Observation:

- In 'Precip Type' Column there are 517 missing values.
- 'Wind Bearing (degrees)' has only integer values and Formatted Date is in String.
- Minimum values of <u>Humidity (https://www.chicagotribune.com/news/ct-xpm-2011-12-16-ct-wea-1216-asktom-20111216-story.html)</u>, <u>Wind Speed (km/h) (https://www.researchgate.net/post/Can-wind-speed-be-zero)</u>, <u>Wind Bearing (degrees)</u>

(https://en.wikipedia.org/wiki/Wind_direction#:~:text=Wind%20direction%20is%20usually%20reported,to%20 Visibility (km)

(https://en.wikipedia.org/wiki/Visibility#:~:text=Visibility%20of%20less%20than%20100,be%20activated%20to are Zero and they can be Zero.

In [7]:

```
#Changing Formatted Date from String to Datetime
df['Formatted Date'] = pd.to_datetime(df['Formatted Date'], utc=True)
```

In [8]:

#Now Formatted Date is in Date Time Format df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 96453 entries, 0 to 96452
Data columns (total 11 columns):
#
    Column
                             Non-Null Count Dtype
    -----
                              -----
0
    Formatted Date
                             96453 non-null datetime64[ns, UTC]
1
    Summary
                             96453 non-null object
2
    Precip Type
                             95936 non-null object
                             96453 non-null float64
3
    Temperature (C)
4
    Apparent Temperature (C) 96453 non-null float64
5
    Humidity
                             96453 non-null float64
6
    Wind Speed (km/h)
                             96453 non-null float64
                             96453 non-null int64
7
    Wind Bearing (degrees)
8
    Visibility (km)
                             96453 non-null float64
9
    Pressure (millibars)
                             96453 non-null float64
10 Daily Summary
                             96453 non-null object
dtypes: datetime64[ns, UTC](1), float64(6), int64(1), object(3)
memory usage: 8.1+ MB
```

In [9]:

df.sample(20)

Out[9]:

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Win Bearin (degree
59633	2012-11-25 17:00:00+00:00	Foggy	rain	4.916667	4.916667	0.93	4.7978	ξ
5369	2006-03-02 16:00:00+00:00	Foggy	snow	-0.944444	-5.644444	0.98	15.4399	1
74149	2014-07-23 14:00:00+00:00	Overcast	rain	21.227778	21.227778	0.78	12.4614	35
83001	2015-07-27 10:00:00+00:00	Mostly Cloudy	rain	22.783333	22.783333	0.56	0.3703	18
96164	2016-09-25 21:00:00+00:00	Mostly Cloudy	rain	12.950000	12.950000	0.81	5.9731	
44704	2011-08-14 14:00:00+00:00	Partly Cloudy	rain	29.444444	29.244444	0.42	6.0858	31
53739	2012-08-25 01:00:00+00:00	Clear	rain	20.538889	20.538889	0.57	0.0000	
28965	2009-02-25 20:00:00+00:00	Mostly Cloudy	rain	0.555556	0.555556	0.75	3.2200	33
46207	2011-02-15 06:00:00+00:00	Mostly Cloudy	snow	-2.627778	-7.388889	0.70	13.8299	3
77683	2014-10-18 21:00:00+00:00	Foggy	rain	8.427778	8.427778	0.99	1.3846	19
9970	2007-08-27 08:00:00+00:00	Clear	rain	26.161111	26.161111	0.39	8.7101	\$
31929	2009-03-29 08:00:00+00:00	Overcast	rain	12.222222	12.222222	0.57	24.1500	1€
85371	2015-05-05 05:00:00+00:00	Mostly Cloudy	rain	15.972222	15.972222	0.88	9.5634	1 1

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Win Bearin (degree:
7617	2006-10-21 08:00:00+00:00	Mostly Cloudy	rain	12.755556	12.755556	0.69	11.1251	14
75292	2014-03-11 06:00:00+00:00	Partly Cloudy	snow	-0.138889	-2.411111	0.77	6.6332	34
24151	2008-11-01 07:00:00+00:00	Partly Cloudy	rain	17.038889	17.038889	0.68	24.4720	13
43578	2010-09-26 16:00:00+00:00	Mostly Cloudy	rain	16.833333	16.833333	0.82	22.7815	22
71889	2014-12-02 08:00:00+00:00	Foggy	rain	2.177778	2.177778	0.93	0.8694	7
63062	2013-12-17 13:00:00+00:00	Overcast	rain	0.050000	-2.661111	0.85	7.9856	13
5863	2006-05-10 06:00:00+00:00	Partly Cloudy	rain	14.855556	14.855556	0.78	9.8532	3

→

In [10]:

#Checking Wheather this dataset has Duplicate Values or not sum(df.duplicated())

Out[10]:

24

Explortary Data Analysis(EDA):

Data Cleaning

In [11]:

#Number of Distinct Observation df.nunique()

Out[11]:

Formatted Date	96429
Summary	27
Precip Type	2
Temperature (C)	7574
Apparent Temperature (C)	8984
Humidity	90
Wind Speed (km/h)	2484
Wind Bearing (degrees)	360
Visibility (km)	949
Pressure (millibars)	4979
Daily Summary	214
dtype: int64	

In [12]:

#DataFrame for Duplicate Values
df_duplicated = df[df.duplicated()]
df_duplicated

Out[12]:

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Be (deg	•
36072	2010-08-01 22:00:00+00:00	Clear	rain	18.800000	18.800000	0.93	6.2790		
36073	2010-08-01 23:00:00+00:00	Clear	rain	18.222222	18.222222	0.97	6.2790		
36074	2010-08-02 00:00:00+00:00	Clear	rain	18.072222	18.072222	0.98	11.2700		
36075	2010-08-02 01:00:00+00:00	Clear	rain	16.622222	16.622222	0.99	6.4400		
36076	2010-08-02 02:00:00+00:00	Clear	rain	16.094444	16.094444	0.99	3.0751		
36077	2010-08-02 03:00:00+00:00	Clear	rain	15.955556	15.955556	0.99	3.8801		
36078	2010-08-02 04:00:00+00:00	Clear	rain	17.088889	17.088889	1.00	6.4400		
36079	2010-08-02 05:00:00+00:00	Clear	rain	20.822222	20.822222	0.87	3.2200		

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Be (deg
36080	2010-08-02 06:00:00+00:00	Clear	rain	23.405556	23.405556	0.74	1.8837	
36081	2010-08-02 07:00:00+00:00	Clear	rain	26.050000	26.050000	0.59	1.5939	
36082	2010-08-02 08:00:00+00:00	Clear	rain	27.688889	28.077778	0.50	0.2254	
36083	2010-08-02 09:00:00+00:00	Clear	rain	28.561111	29.588889	0.54	3.2039	
36084	2010-08-02 10:00:00+00:00	Clear	rain	28.816667	29.338889	0.49	1.6100	
36085	2010-08-02 11:00:00+00:00	Partly Cloudy	rain	28.866667	29.044444	0.46	2.8175	
36086	2010-08-02 12:00:00+00:00	Partly Cloudy	rain	29.827778	30.338889	0.47	8.0339	
36087	2010-08-02 13:00:00+00:00	Partly Cloudy	rain	30.072222	30.527778	0.46	2.5921	
36088	2010-08-02 14:00:00+00:00	Partly Cloudy	rain	31.066667	31.627778	0.44	1.3846	
36089	2010-08-02 15:00:00+00:00	Partly Cloudy	rain	30.861111	31.361111	0.44	2.2540	

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Be (deg
36090	2010-08-02 16:00:00+00:00	Partly Cloudy	rain	29.950000	30.416667	0.46	3.0751	
36091	2010-08-02 17:00:00+00:00	Clear	rain	28.811111	30.616667	0.59	3.2039	
36092	2010-08-02 18:00:00+00:00	Clear	rain	25.250000	25.250000	0.75	2.4955	
36093	2010-08-02 19:00:00+00:00	Clear	rain	22.172222	22.172222	0.87	1.6100	
36094	2010-08-02 20:00:00+00:00	Partly Cloudy	rain	21.061111	21.061111	0.90	0.0000	
36095	2010-08-02 21:00:00+00:00	Clear	rain	20.255556	20.255556	0.92	1.0787	_
4								•

In [13]:

 ${\tt df_duplicated.shape}$

Out[13]:

(24, 11)

Observation:

• There are no Duplicate Values in this Dataset.

In [14]:

#DataFrame for only NaN Values for exploration.
df_null = df[df.isna().any(axis=1)]
df_null.head(20)

Out[14]:

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Be (deg
52672	2012-04-11 14:00:00+00:00	Mostly Cloudy	NaN	19.016667	19.016667	0.26	14.8764	
52674	2012 - 04-11 16:00:00+00:00	Mostly Cloudy	NaN	17.850000	17.850000	0.28	13.7977	
52675	2012-04-11 17:00:00+00:00	Mostly Cloudy	NaN	16.322222	16.322222	0.32	10.8192	
52677	2012-04-11 19:00:00+00:00	Mostly Cloudy	NaN	12.566667	12.566667	0.43	9.0160	
52678	2012 - 04 - 11 20:00:00+00:00	Mostly Cloudy	NaN	12.927778	12.927778	0.47	17.6295	
52680	2012-04-11 22:00:00+00:00	Mostly Cloudy	NaN	10.100000	10.100000	0.61	11.3666	
52681	2012-04-11 23:00:00+00:00	Partly Cloudy	NaN	10.183333	10.183333	0.64	7.2128	
52683	2012-04-12 01:00:00+00:00	Mostly Cloudy	NaN	10.700000	10.700000	0.70	14.6993	
52684	2012-04-12 02:00:00+00:00	Partly Cloudy	NaN	9.744444	8.022222	0.73	11.9623	
52687	2012-04-12 05:00:00+00:00	Mostly Cloudy	NaN	10.055556	10.055556	0.76	13.9748	
52692	2012-04-12 10:00:00+00:00	Partly Cloudy	NaN	17.177778	17.177778	0.42	25.3414	
52693	2012-04-12 11:00:00+00:00	Partly Cloudy	NaN	16.183333	16.183333	0.41	28.2877	
52695	2012-04-12 13:00:00+00:00	Partly Cloudy	NaN	16.183333	16.183333	0.47	24.2466	
52696	2012-04-12 14:00:00+00:00	Partly Cloudy	NaN	16.244444	16.244444	0.43	29.8816	
52698	2012-04-12 16:00:00+00:00	Partly Cloudy	NaN	15.183333	15.183333	0.50	26.8548	

		Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Be (deg	
	52699	2012-04-12 17:00:00+00:00	Partly Cloudy	NaN	12.455556	12.455556	0.52	17.6134		
	52701	2012-04-12 19:00:00+00:00	Partly Cloudy	NaN	7.983333	7.983333	0.70	1.7388		
	52702	2012-04-12 20:00:00+00:00	Partly Cloudy	NaN	6.905556	6.905556	0.77	2.1091		
	52704	2012-04-12 22:00:00+00:00	Partly Cloudy	NaN	5.283333	5.283333	0.86	3.7996		
	52705	2012-04-12 23:00:00+00:00	Partly Cloudy	NaN	4.688889	3.133333	0.83	6.7781		~
4									•	

In [15]:

df_null.tail(20)

Out[15]:

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Win Bearin (degree:
95569	2016-10-31 03:00:00+00:00	Clear	NaN	1.905556	1.905556	0.88	4.7656	24
95570	2016-10-31 04:00:00+00:00	Clear	NaN	1.894444	0.088889	0.89	6.2146	24
95571	2016-10-31 05:00:00+00:00	Clear	NaN	1.983333	-0.133333	0.91	7.1645	25
95572	2016-10-31 06:00:00+00:00	Clear	NaN	2.494444	0.227778	0.90	7.9373	25
95573	2016-10-31 07:00:00+00:00	Clear	NaN	3.766667	1.561111	0.86	8.5169	25
95574	2016-10-31 08:00:00+00:00	Clear	NaN	5.688889	3.783333	0.80	8.7262	25
95575	2016-10-31 09:00:00+00:00	Clear	NaN	7.688889	6.177778	0.75	8.5813	26
95576	2016-10-31 10:00:00+00:00	Clear	NaN	9.594444	7.972222	0.71	11.1251	27
95577	2016-10-31 11:00:00+00:00	Clear	NaN	11.494444	11.494444	0.69	14.0553	28
95578	2016-10-31 12:00:00+00:00	Clear	NaN	12.766667	12.766667	0.67	22.5400	28

	Formatted Date	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Win Bearin (degree:
95579	2016-10-31 13:00:00+00:00	Partly Cloudy	NaN	13.022222	13.022222	0.68	14.1680	29
95580	2016-10-31 14:00:00+00:00	Partly Cloudy	NaN	12.538889	12.538889	0.71	10.9641	25
95581	2016-10-31 15:00:00+00:00	Mostly Cloudy	NaN	11.755556	11.755556	0.74	8.3237	29
95582	2016-10-31 16:00:00+00:00	Mostly Cloudy	NaN	10.688889	10.688889	0.78	7.5187	29
95583	2016-10-31 17:00:00+00:00	Mostly Cloudy	NaN	9.394444	8.377778	0.82	7.6797	29
95584	2016-10-31 18:00:00+00:00	Mostly Cloudy	NaN	8.322222	7.044444	0.85	8.0339	29
95585	2016-10-31 19:00:00+00:00	Mostly Cloudy	NaN	7.627778	6.183333	0.87	8.2271	29
95586	2016-10-31 20:00:00+00:00	Mostly Cloudy	NaN	7.111111	5.511111	0.89	8.5008	29
95587	2016-10-31 21:00:00+00:00	Partly Cloudy	NaN	6.672222	4.961111	0.90	8.6457	29
95588	2016-10-31 22:00:00+00:00	Mostly Cloudy	NaN	6.322222	4.588889	0.91	8.4686	29
								>

```
In [16]:
#Droping NaN(Not a Number)
df_target = df.dropna()
df_target.shape
Out[16]:
(95936, 11)
In [17]:
df target.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 95936 entries, 0 to 96452
Data columns (total 11 columns):
 #
     Column
                                 Non-Null Count Dtype
     Formatted Date
                                 95936 non-null datetime64[ns, UTC]
 0
 1
     Summary
                                 95936 non-null object
 2
                                 95936 non-null object
     Precip Type
 3
     Temperature (C)
                                 95936 non-null float64
 4
     Apparent Temperature (C) 95936 non-null float64
 5
     Humidity
                                 95936 non-null float64
 6
     Wind Speed (km/h)
                                 95936 non-null float64
 7
     Wind Bearing (degrees)
                                 95936 non-null int64
                                 95936 non-null float64
 8
     Visibility (km)
 9
     Pressure (millibars)
                                 95936 non-null float64
 10 Daily Summary
                                 95936 non-null object
dtypes: datetime64[ns, UTC](1), float64(6), int64(1), object(3)
memory usage: 8.8+ MB
In [18]:
df_target.columns
Out[18]:
Index(['Formatted Date', 'Summary', 'Precip Type', 'Temperature (C)',
       'Apparent Temperature (C)', 'Humidity', 'Wind Speed (km/h)', 'Wind Bearing (degrees)', 'Visibility (km)', 'Pressure (millibars)',
```

'Daily Summary'], dtype='object')

In [19]:

```
df_target = df_target.set_index("Formatted Date")
df_target
```

Out[19]:

	Summary	Precip Type	Temperature (C)	Apparent Temperature (C)	Humidity	Wind Speed (km/h)	Wind Bearing (degrees)	Visi
Formatted Date								
2006-03-31 22:00:00+00:00	Partly Cloudy	rain	9.472222	7.388889	0.89	14.1197	251	15.
2006-03-31 23:00:00+00:00	Partly Cloudy	rain	9.355556	7.227778	0.86	14.2646	259	15.
2006-04-01 00:00:00+00:00	Mostly Cloudy	rain	9.377778	9.377778	0.89	3.9284	204	14.
2006-04-01 01:00:00+00:00	Partly Cloudy	rain	8.288889	5.944444	0.83	14.1036	269	15.
2006-04-01 02:00:00+00:00	Mostly Cloudy	rain	8.755556	6.977778	0.83	11.0446	259	15.
2016-09-09 17:00:00+00:00	Partly Cloudy	rain	26.016667	26.016667	0.43	10.9963	31	16.
2016-09-09 18:00:00+00:00	Partly Cloudy	rain	24.583333	24.583333	0.48	10.0947	20	15.
2016-09-09 19:00:00+00:00	Partly Cloudy	rain	22.038889	22.038889	0.56	8.9838	30	16.
2016-09-09 20:00:00+00:00	Partly Cloudy	rain	21.522222	21.522222	0.60	10.5294	20	16.
2016-09-09 21:00:00+00:00	Partly Cloudy	rain	20.438889	20.438889	0.61	5.8765	39	15.

Given:

The Null Hypothesis H0 is "Has the Apparent temperature and humidity compared monthly across 10 years of the data indicate an increase due to Global warming".

The Alternative Hypothesis H1 is "Has the Apparent temperature and humidity compared monthly across 10 years of the data not indicate an increase due to Global warming".

In [20]:

```
#Creating new DataFrame only for Apparent Temperature and Humidity

df_column = ['Apparent Temperature (C)', 'Humidity']

df_monthly_mean = df_target[df_column].resample("MS").mean() #MS-Month Starting

df_monthly_mean.head()
```

Out[20]:

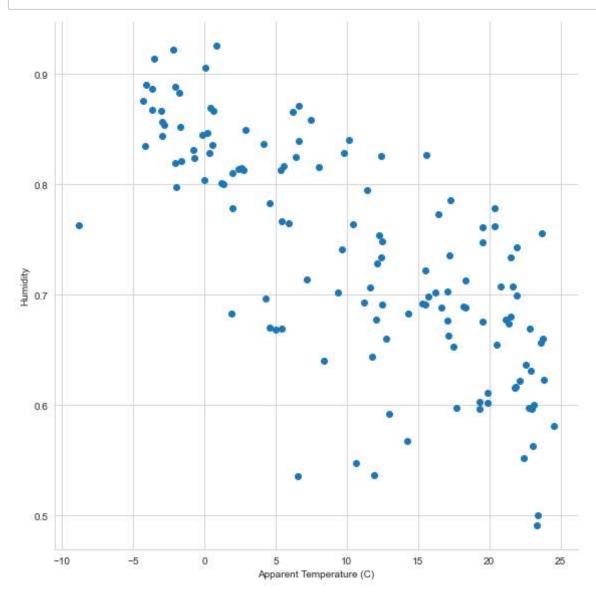
Apparent Temperature (C) Humidity

Formatted Date

2005-12-01 00:00:00+00:00	-4.050000	0.890000
2006-01-01 00:00:00+00:00	-4.173708	0.834610
2006-02-01 00:00:00+00:00	-2.990716	0.843467
2006-03-01 00:00:00+00:00	1.969780	0.778737
2006-04-01 00:00:00+00:00	12.098827	0.728625

In [21]:

```
sns.set_style("whitegrid")
sns.FacetGrid(df_monthly_mean, height=8).map(plt.scatter, "Apparent Temperature (C)", "Humi
plt.show()
```

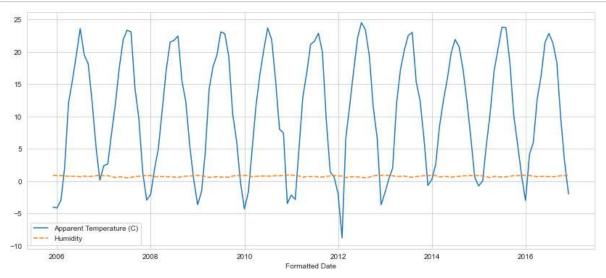


Observation:

• There can be or may be Linear Relationship between "Humidity" and "Apparent Temperature (C)" with negative slope.

In [22]:

```
plt.figure(figsize=(14,6))
sns.lineplot(data = df_monthly_mean)
plt.show()
```



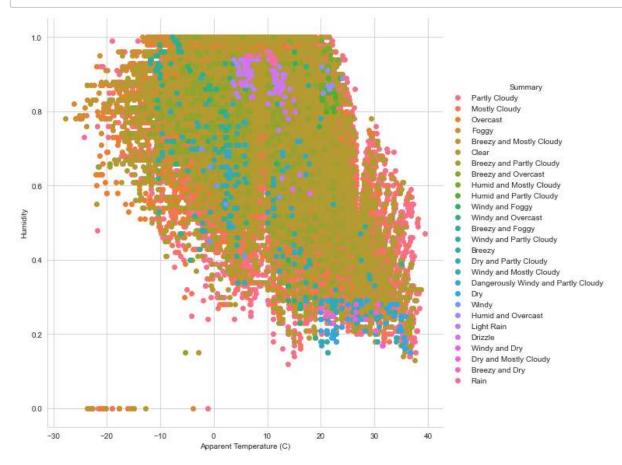
Observation:

- "Humidity" is remain constant from 2006 2016
- But "Apparent Temperature (C)" is frequently changes from 2006 2016.

2D Scatter Plot with Color Coding for each Summary type.

In [23]:

```
sns.set_style("whitegrid")
sns.FacetGrid(df_target, hue="Summary", height=8).map(plt.scatter, "Apparent Temperature (C
plt.show()
```



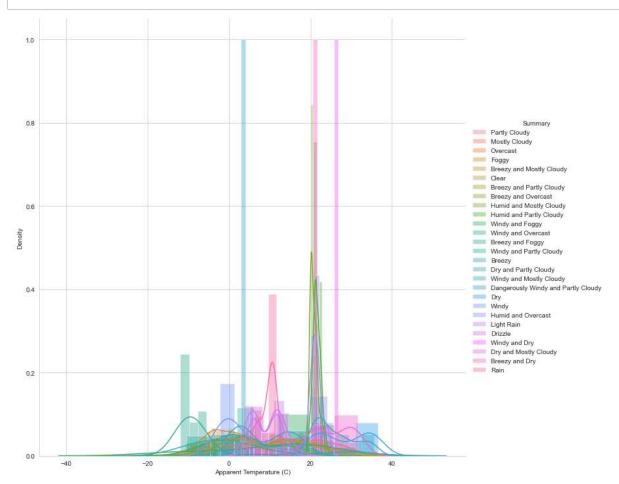
Observation:

- There are very few outlier.
- Mostly Weather is Clear or Partly Cloudy/Rain in Finland.
- Only few days there has a Light Rain or Dry or Dangerously Windy and Partly Cloudy.

Univariate Analysis using PDF:

In [24]:

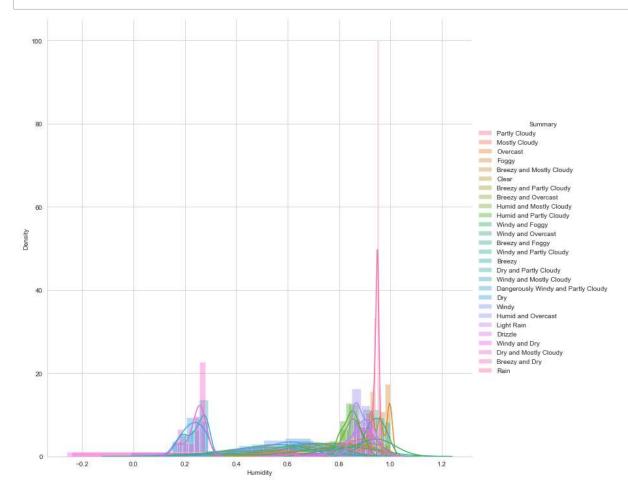
For Apparent Temperature (C)
sns.FacetGrid(df_target, hue="Summary", height=10).map(sns.distplot, "Apparent Temperature
plt.show()



In [25]:

For Humidity

sns.FacetGrid(df_target, hue="Summary", height=10).map(sns.distplot, "Humidity").add_legend
plt.show()



Observation:

• "Humidity" is better Feature than "Apparent Temperature (C)".

^ 1	
CONC	lusion:
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H0 is not accepted	because there is	s no change i	n Humidity from	2006–2016.	So, we will	accept the H1.

In []:			