

In [1]:

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [2]:

```
df = pd.read_csv("Weather Data.csv")
df.head()
```

Out[2]:

	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
0	1/1/2012 0:00	-1.8	-3.9	86	4	8.0	101.24	Fog
1	1/1/2012 1:00	-1.8	-3.7	87	4	8.0	101.24	Fog
2	1/1/2012 2:00	-1.8	-3.4	89	7	4.0	101.26	Freezing Drizzle,Fog
3	1/1/2012 3:00	-1.5	-3.2	88	6	4.0	101.27	Freezing Drizzle,Fog
4	1/1/2012 4:00	-1.5	-3.3	88	7	4.8	101.23	Fog

In [3]:

```
df.tail()
```

Out[3]:

	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
8779	12/31/2012 19:00	0.1	-2.7	81	30	9.7	100.13	Snow
8780	12/31/2012 20:00	0.2	-2.4	83	24	9.7	100.03	Snow
8781	12/31/2012 21:00	-0.5	-1.5	93	28	4.8	99.95	Snow
8782	12/31/2012 22:00	-0.2	-1.8	89	28	9.7	99.91	Snow
8783	12/31/2012 23:00	0.0	-2.1	86	30	11.3	99.89	Snow

In [4]:

```
df.isnull().sum()
```

Out[4]:

Date/Time	0
Temp_C	0
Dew Point Temp_C	0
Rel Hum_%	0
Wind Speed_km/h	0
Visibility_km	0
Press_kPa	0
Weather	0
dtype:	int64

In [5]:

```
df.skew()
```

/tmp/ipykernel\_2960/1665899112.py:1: FutureWarning: The default value of numeric\_only in DataFrame.skew is deprecate  
d. In a future version, it will default to False. In addition, specifying 'numeric\_only=None' is deprecated. Select  
only valid columns or specify the value of numeric\_only to silence this warning.  
df.skew()

Out[5]:

Temp_C	-0.177666
Dew Point Temp_C	-0.318433
Rel Hum_%	-0.323830
Wind Speed_km/h	0.871374
Visibility_km	0.413362
Press_kPa	-0.229925
dtype:	float64

In [6]:

```
df.describe()
```

Out[6]:

	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa
count	8784.000000	8784.000000	8784.000000	8784.000000	8784.000000	8784.000000
mean	8.798144	2.555294	67.431694	14.945469	27.664447	101.051623
std	11.687883	10.883072	16.918881	8.688696	12.622688	0.844005
min	-23.300000	-28.500000	18.000000	0.000000	0.200000	97.520000
25%	0.100000	-5.900000	56.000000	9.000000	24.100000	100.560000
50%	9.300000	3.300000	68.000000	13.000000	25.000000	101.070000
75%	18.800000	11.800000	81.000000	20.000000	25.000000	101.590000
max	33.000000	24.400000	100.000000	83.000000	48.300000	103.650000

In [7]:

```
corr = df.corr()
corr
```

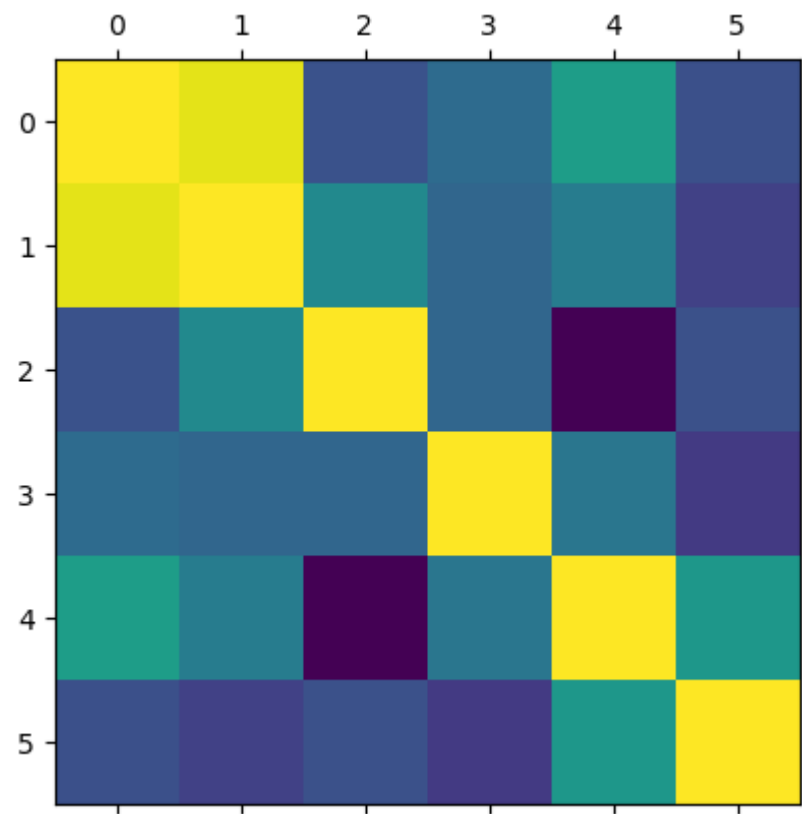
```
/tmp/ipykernel_2960/2438084875.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
corr = df.corr()
```

Out[7]:

	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa
Temp_C	1.000000	0.932714	-0.220182	-0.061876	0.273455	-0.236389
Dew Point Temp_C	0.932714	1.000000	0.139494	-0.095685	0.050813	-0.320616
Rel Hum_%	-0.220182	0.139494	1.000000	-0.092743	-0.633683	-0.231424
Wind Speed_km/h	-0.061876	-0.095685	-0.092743	1.000000	0.004883	-0.356613
Visibility_km	0.273455	0.050813	-0.633683	0.004883	1.000000	0.231847
Press_kPa	-0.236389	-0.320616	-0.231424	-0.356613	0.231847	1.000000

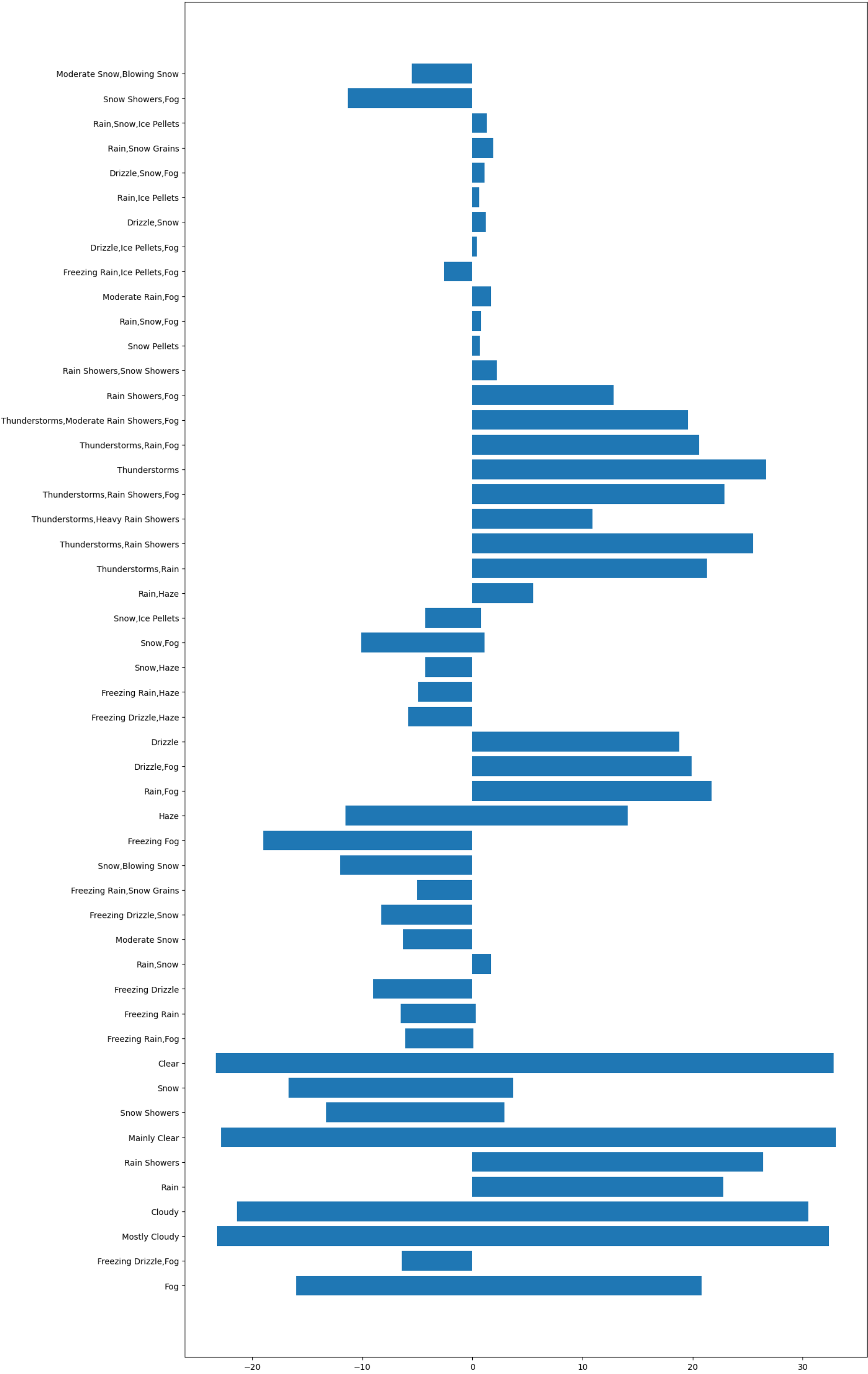
```
In [8]: plt.matshow(corr)
```

Out[8]: <matplotlib.image.AxesImage at 0x7f57bdee8b10>



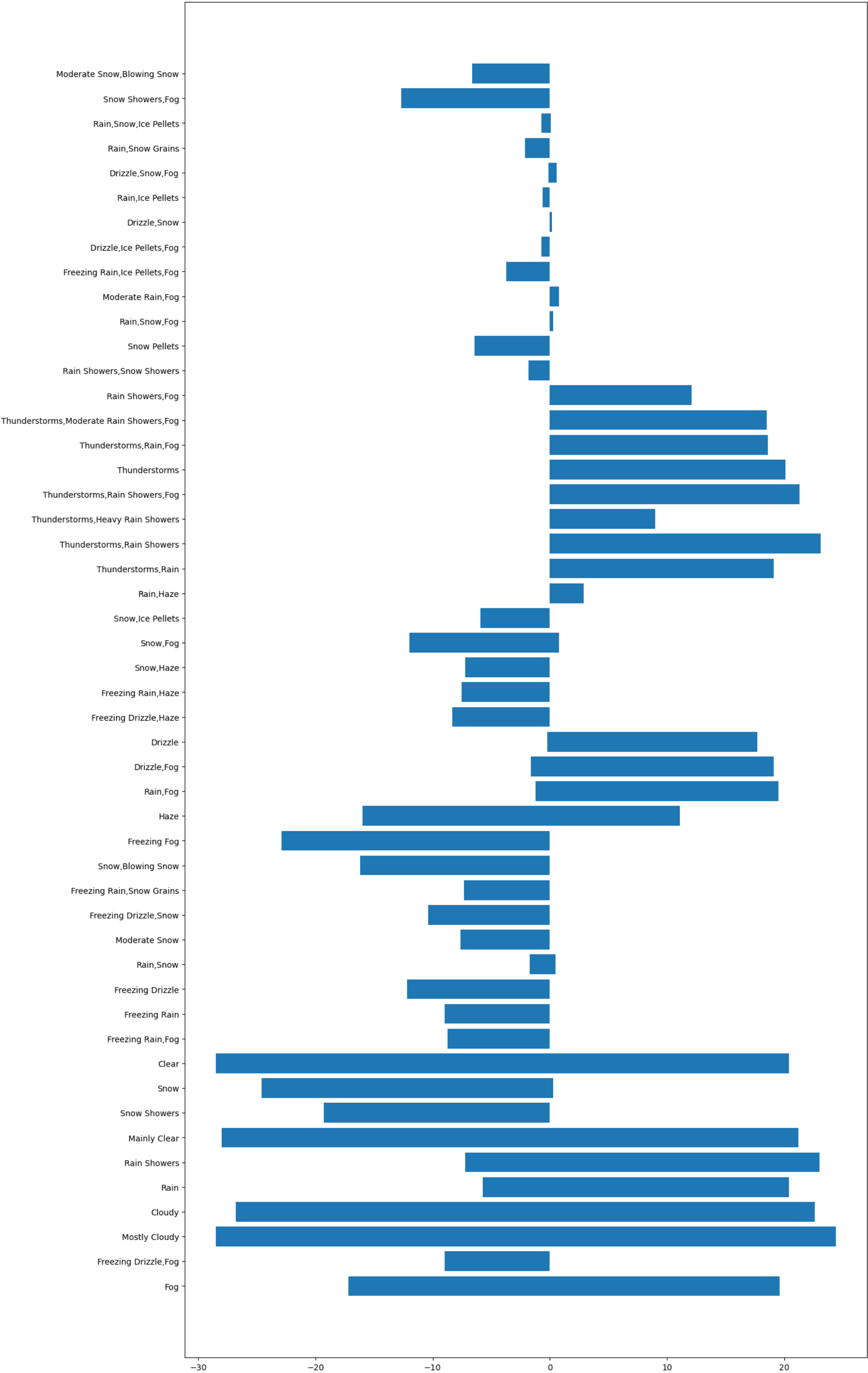
```
In [9]: plt.figure(figsize=(15,30))
plt.barh(df["Weather"],df["Temp_C"])
```

Out[9]: <BarContainer object of 8784 artists>



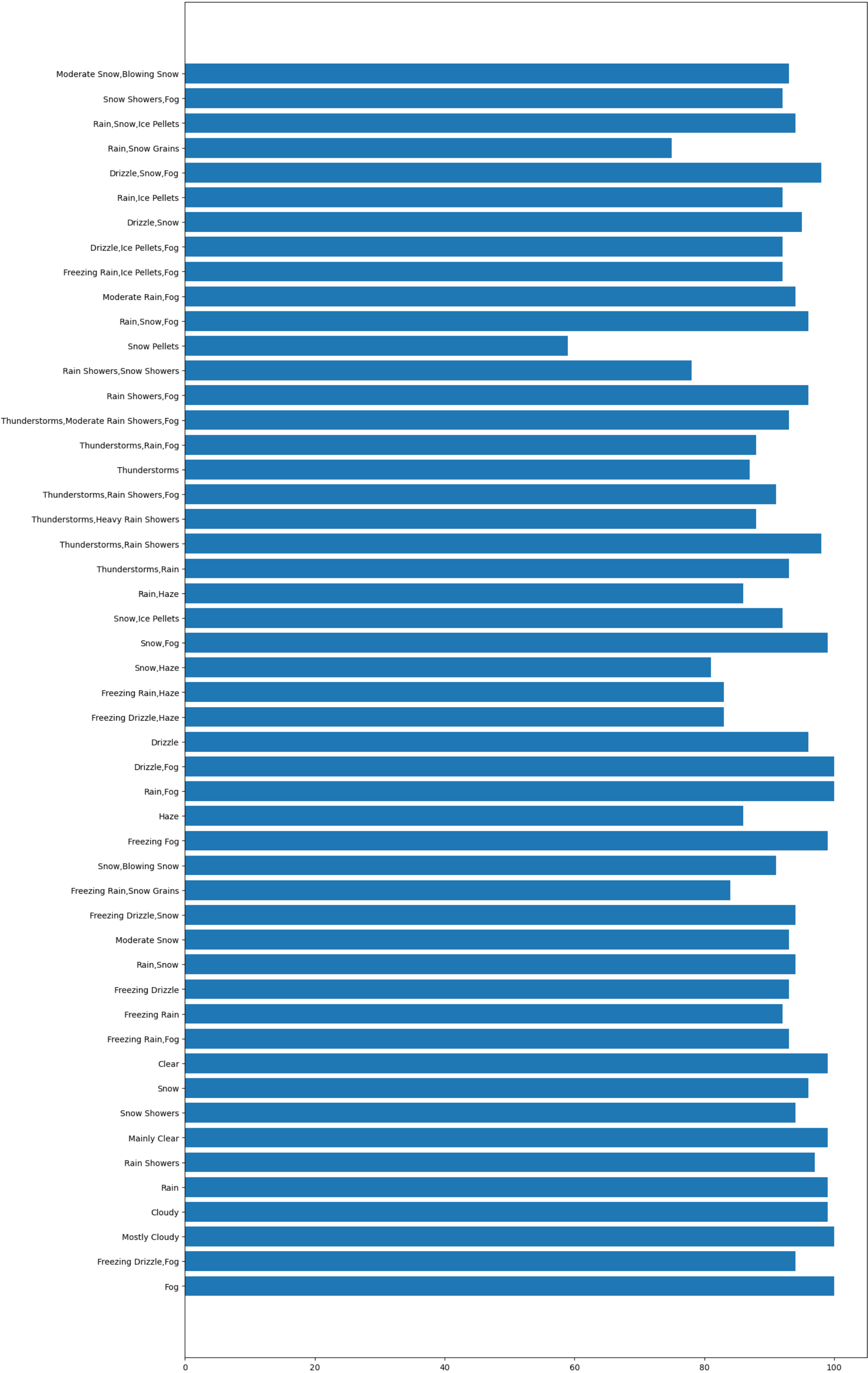
```
In [10]: plt.figure(figsize=(15,30))  
plt.barh(df["Weather"],df["Dew Point Temp_C"])
```

```
Out[10]: <BarContainer object of 8784 artists>
```



```
In [11]: plt.figure(figsize=(15,30))  
plt.barh(df["Weather"],df["Rel Hum_%"])
```

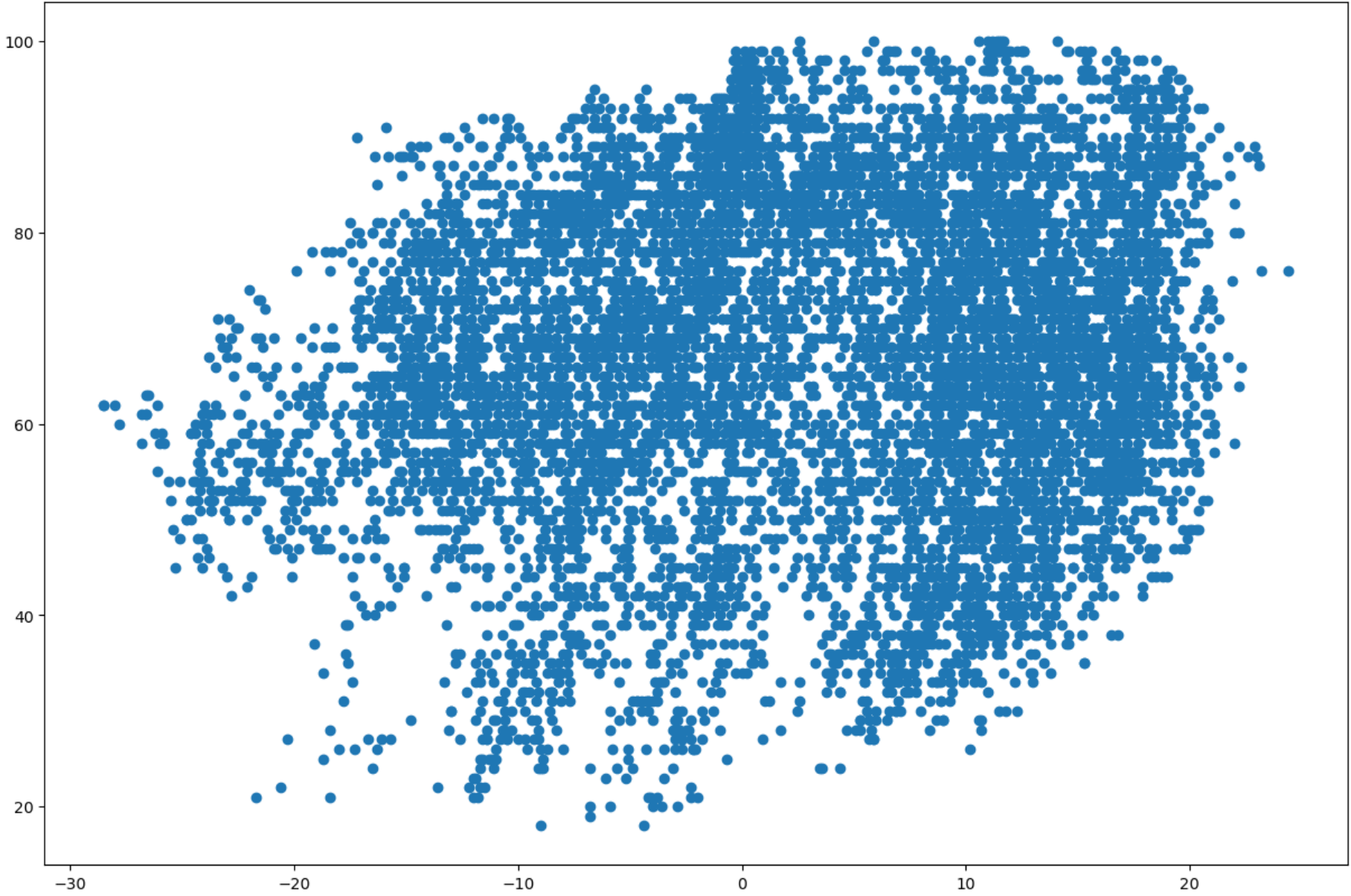
```
Out[11]: <BarContainer object of 8784 artists>
```



## Dew Point Vs Relative Humidity

```
In [12]: plt.figure(figsize=(15,10))
plt.scatter(df["Dew Point Temp_C"],df["Rel Hum_%"])
```

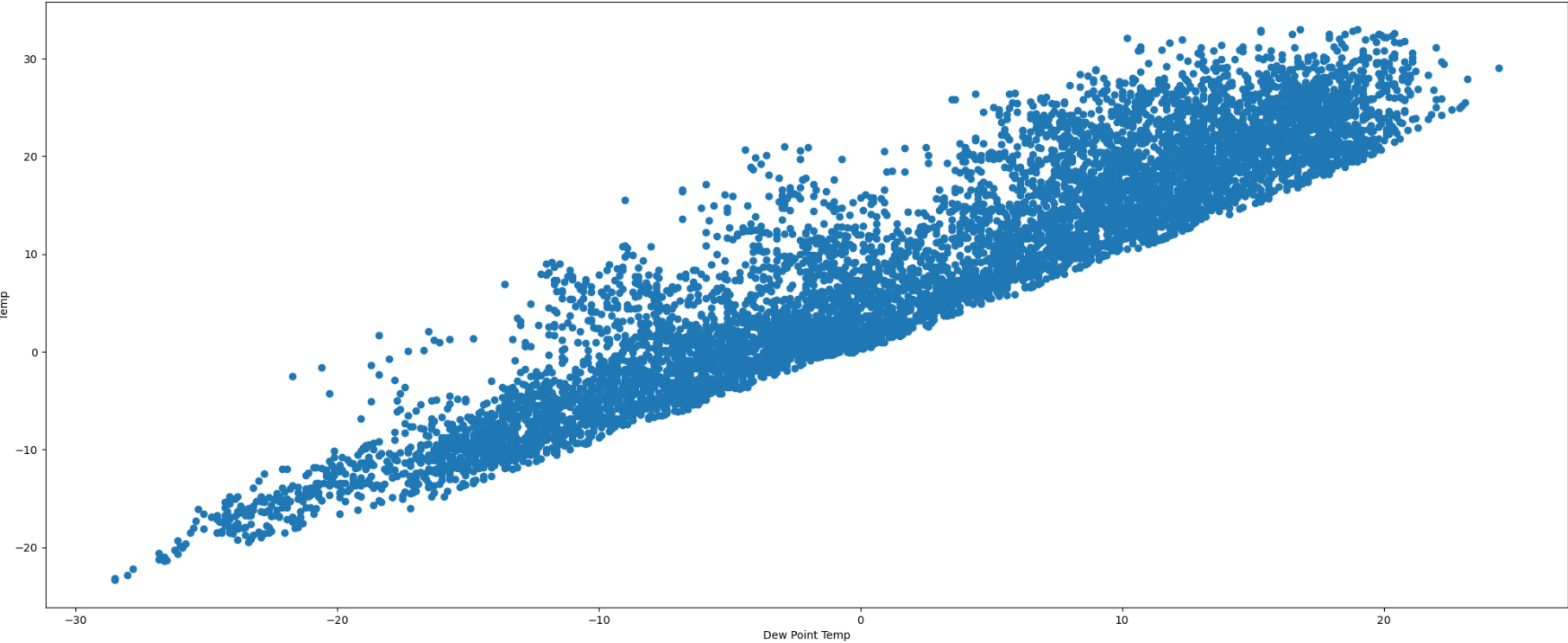
Out[12]: <matplotlib.collections.PathCollection at 0x7f57aaf4da90>



## Dew Point Vs Temp\_C

```
In [13]: plt.figure(figsize=(25,10))
plt.xlabel("Dew Point Temp")
plt.ylabel("Temp")
plt.scatter(df["Dew Point Temp_C"],df["Temp_C"])
```

Out[13]: <matplotlib.collections.PathCollection at 0x7f57aae67a10>



## Analyzing Data When Temperature Was -10 to 0

```
In [14]: result_df = df.loc[df["Temp_C"]> -10]
result_df2 = result_df.loc[result_df["Temp_C"]<=0]
```

In [15]: result\_df2



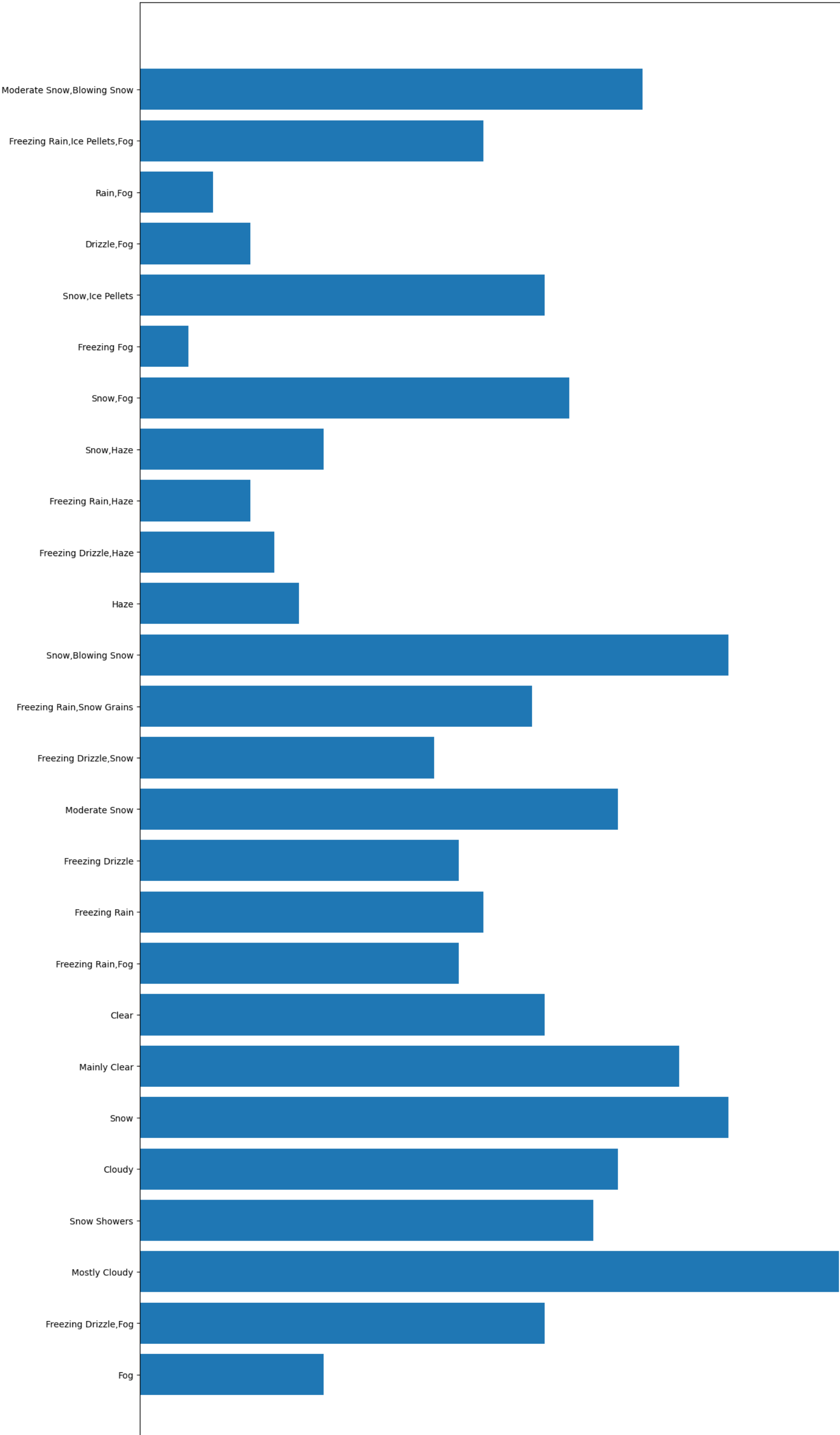
Out[15]:

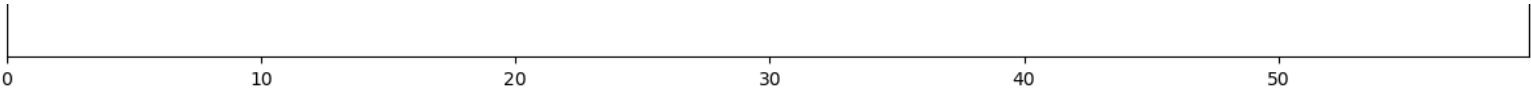
	Date/Time	Temp_C	Dew Point	Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
<b>0</b>	1/1/2012 0:00	-1.8		-3.9	86	4	8.0	101.24	Fog
<b>1</b>	1/1/2012 1:00	-1.8		-3.7	87	4	8.0	101.24	Fog
<b>2</b>	1/1/2012 2:00	-1.8		-3.4	89	7	4.0	101.26	Freezing Drizzle,Fog
<b>3</b>	1/1/2012 3:00	-1.5		-3.2	88	6	4.0	101.27	Freezing Drizzle,Fog
<b>4</b>	1/1/2012 4:00	-1.5		-3.3	88	7	4.8	101.23	Fog
...	...	...		...	...	...	...	...	...
<b>8777</b>	12/31/2012 17:00	-1.1		-3.3	85	19	9.7	100.30	Snow
<b>8778</b>	12/31/2012 18:00	-1.3		-3.1	88	17	9.7	100.19	Snow
<b>8781</b>	12/31/2012 21:00	-0.5		-1.5	93	28	4.8	99.95	Snow
<b>8782</b>	12/31/2012 22:00	-0.2		-1.8	89	28	9.7	99.91	Snow
<b>8783</b>	12/31/2012 23:00	0.0		-2.1	86	30	11.3	99.89	Snow

1675 rows × 8 columns

```
In [16]: plt.figure(figsize=(15,30))
plt.barh(result_df2["Weather"],result_df2["Wind Speed_km/h"])
```

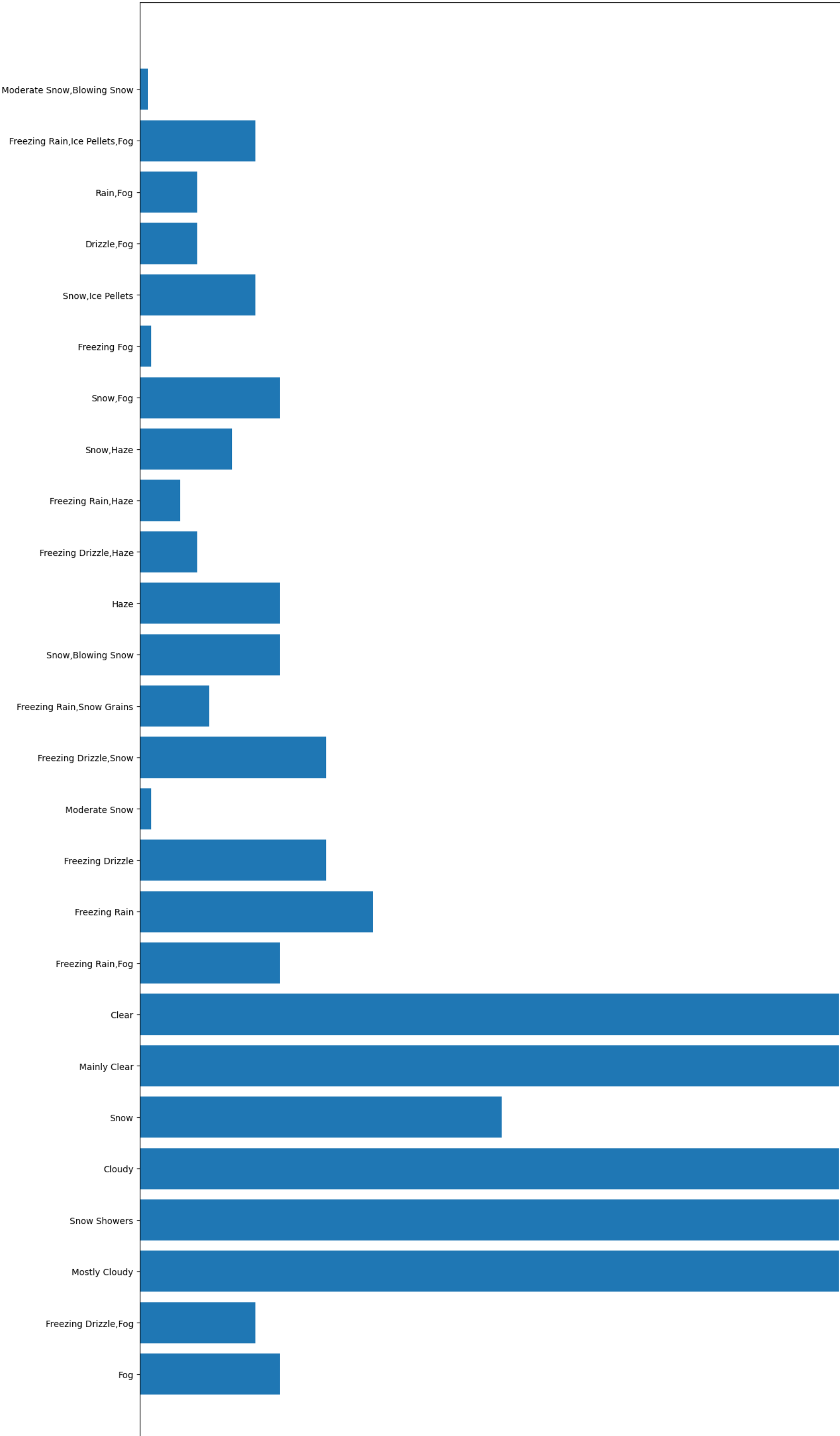
Out[16]: <BarContainer object of 1675 artists>





```
In [17]: plt.figure(figsize=(15,30))
plt.barh(result_df2["Weather"],result_df2["Visibility_km"])
```

Out[17]: <BarContainer object of 1675 artists>



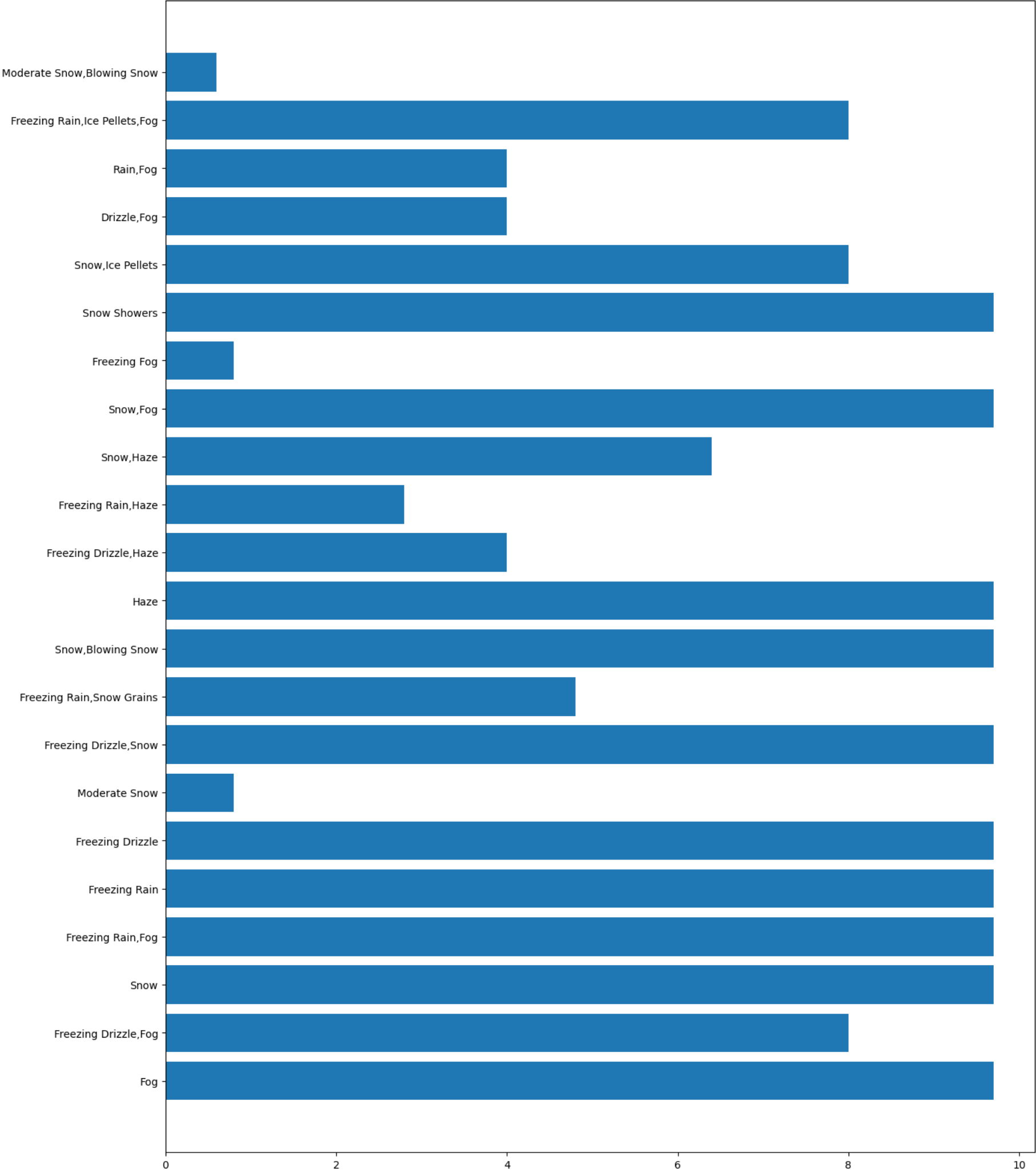


Visibility is said to be good when a person can see farther than about 10 kilometers.

```
In [18]: result_df2_1 = result_df2.loc[result_df2["Visibility_km"]<=10]

plt.figure(figsize=(15,20))
plt.barh(result_df2_1["Weather"],result_df2_1["Visibility_km"])
```

Out[18]: <BarContainer object of 325 artists>

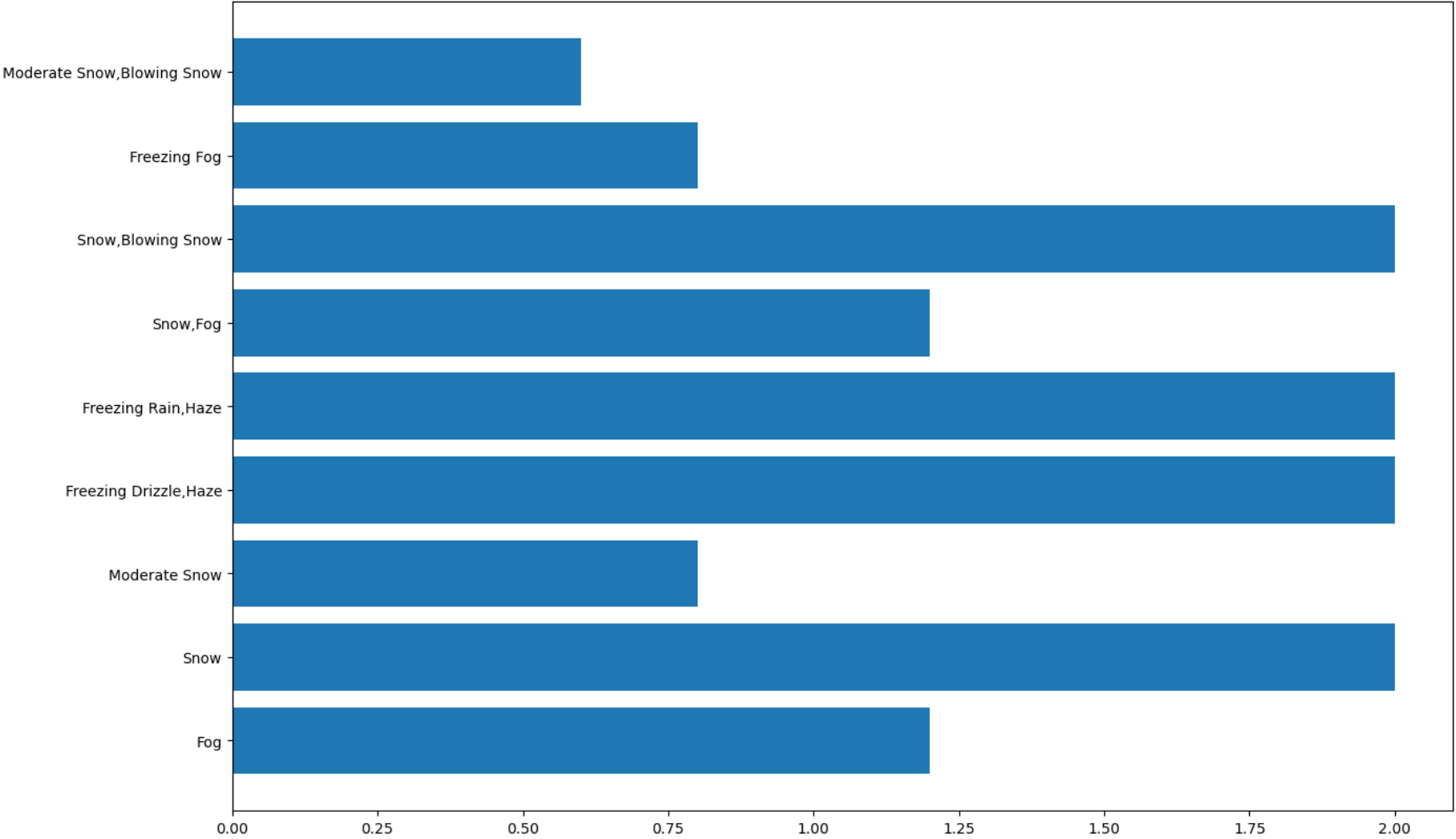


In these Weathers It was barely Visible Specially when it was Moderate Snow and Freezing Fog and Haze

```
In [19]: result_df2_2 = result_df2.loc[result_df2["Visibility_km"]<=2]

plt.figure(figsize=(15,10))
plt.barh(result_df2_2["Weather"],result_df2_2["Visibility_km"])
```

Out[19]: <BarContainer object of 80 artists>



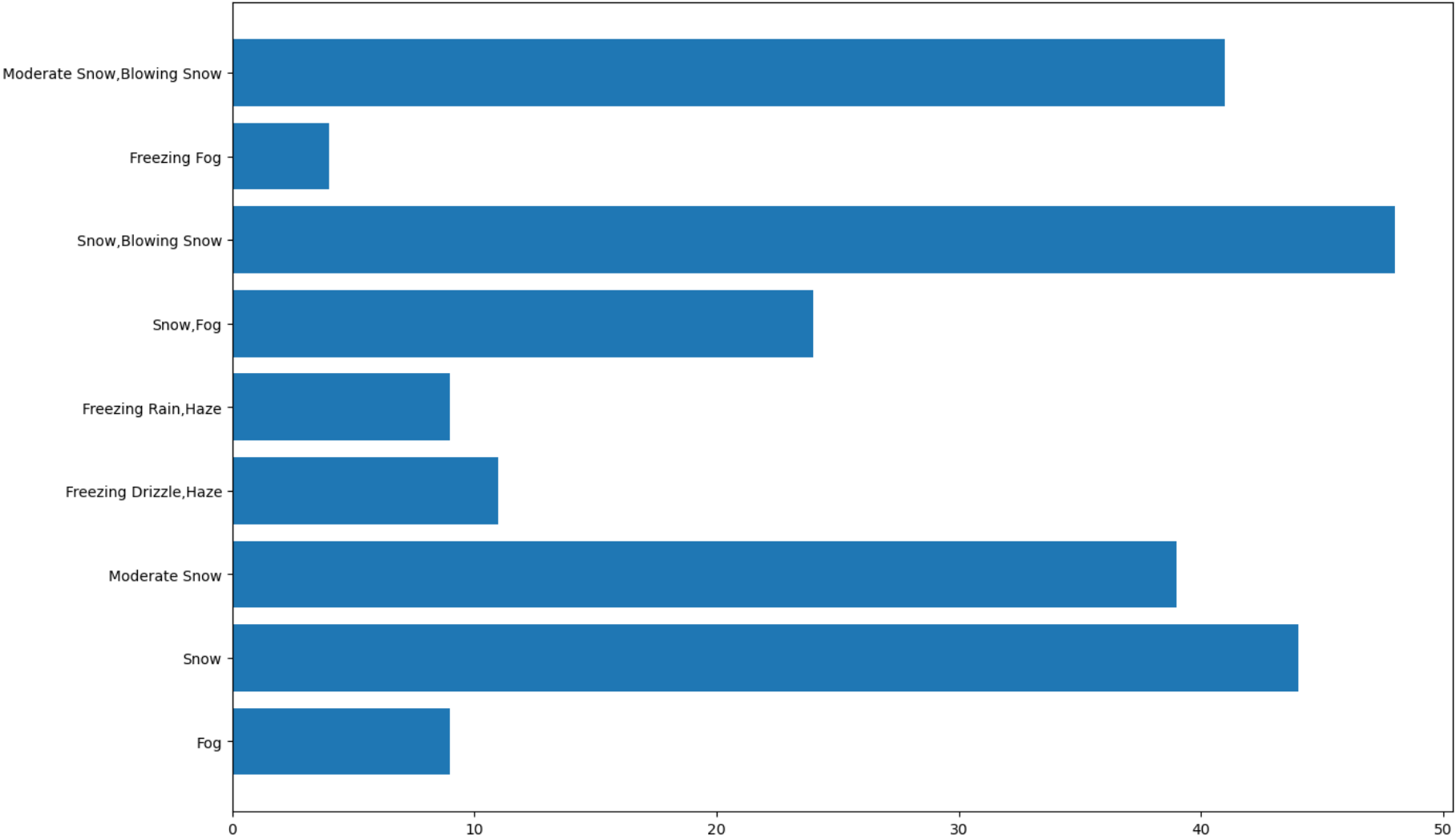
Barely Visble in these conditions Like Blowing Snow Freezing Fog and Moderate snow. These have visiblity less than 750 meters

In [ ]:

Looking At Wind Speed when Visiblity was too low

```
In [20]: plt.figure(figsize=(15,10))
plt.barh(result_df2_2["Weather"],result_df2_2["Wind Speed_km/h"])
```

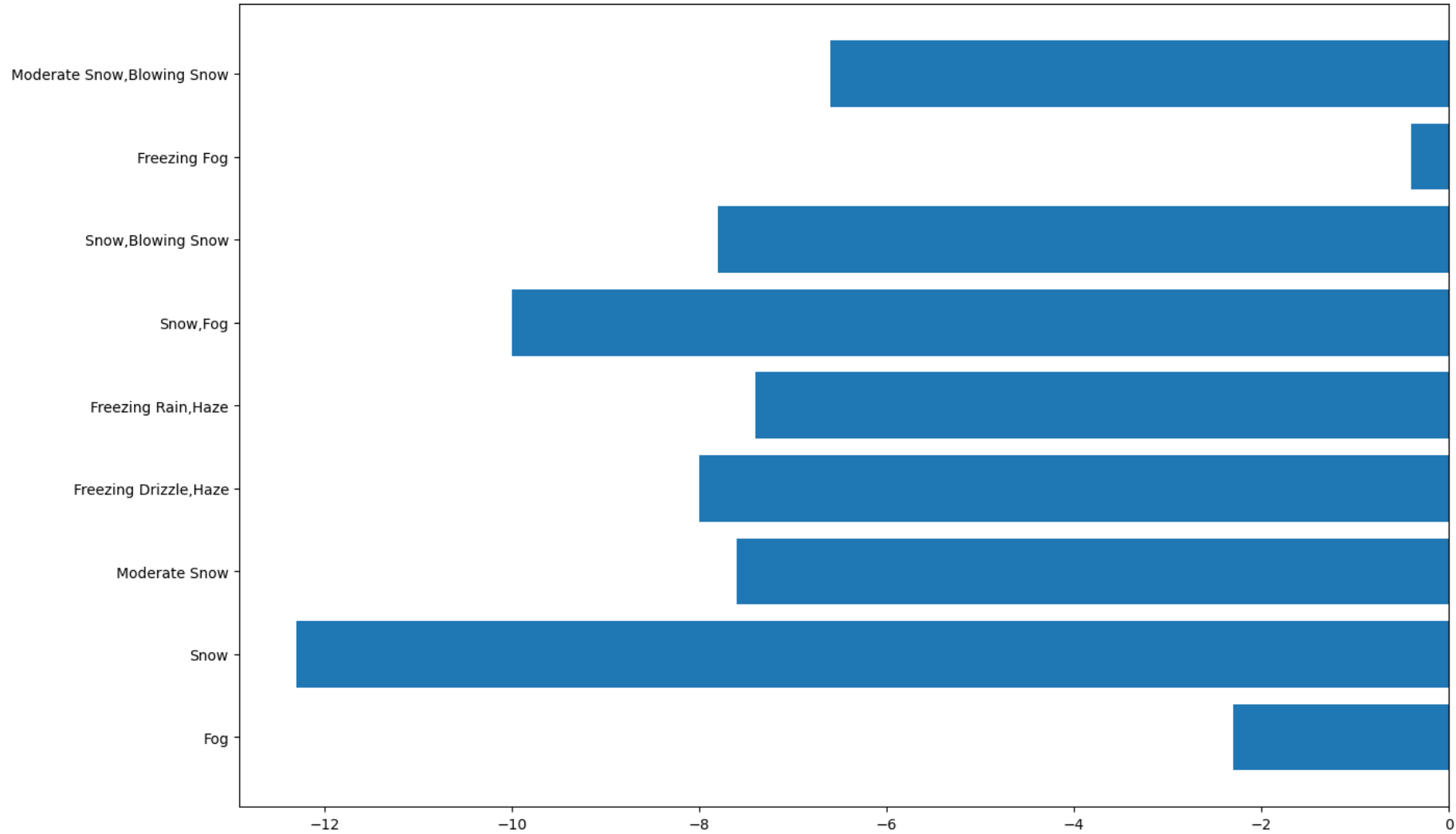
Out[20]: <BarContainer object of 80 artists>



When Weather is Blowing Snow/Moderate Snow Speed was above 40km/hr

```
In [21]: plt.figure(figsize=(15,10))
plt.barh(result_df2_2["Weather"],result_df2_2["Dew Point Temp_C"])
```

Out[21]: <BarContainer object of 80 artists>



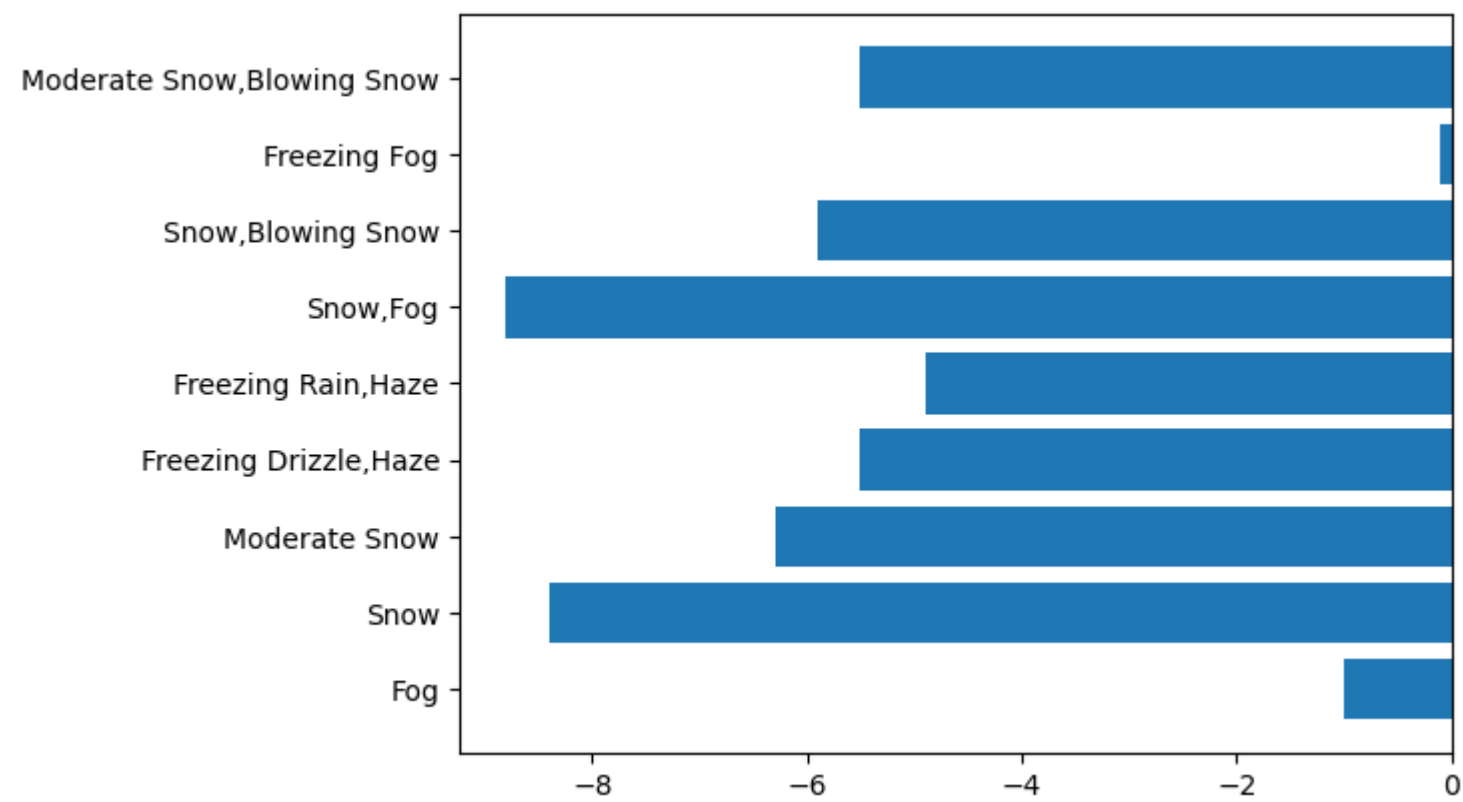
In Weather Conditions Like Moderate Snow,Blowing Snow, Frrexing Rain, Haze and Normal Snow the Dew point went bellow -12

Dew point temperature is a measure of atmospheric moisture. It is the temperature to which air must be cooled in order to reach saturation (assuming air pressure and moisture content are constant).

In [ ]:

In [22]: `plt.barh(result_df2_2["Weather"],result_df2_2["Temp_C"])`

Out[22]: <BarContainer object of 80 artists>

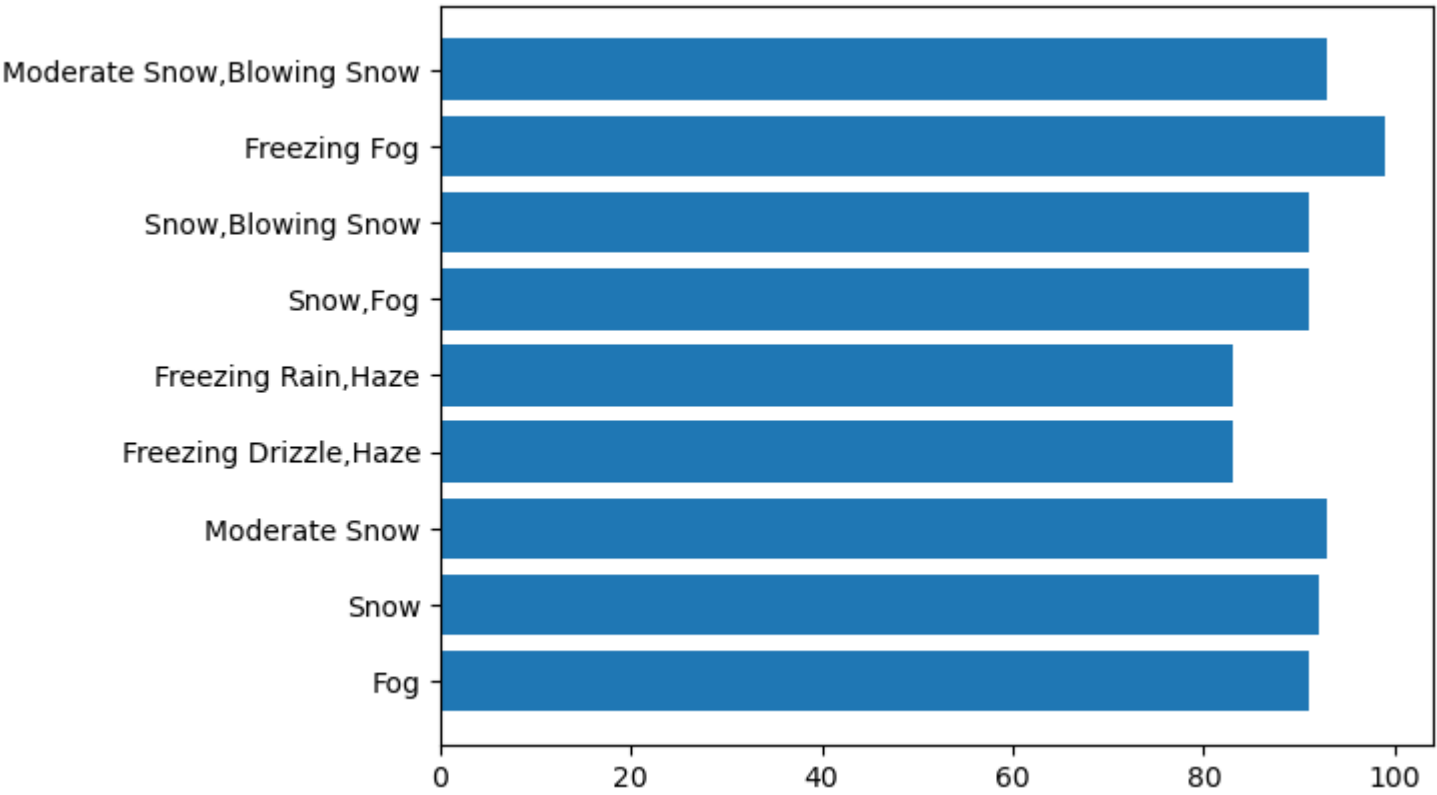


In These weather conditions where Visiblity is so low and Humidity is High the temprature goes bellow -5 Degree Celcius

In [ ]:

In [23]: `plt.barh(result_df2_2["Weather"],result_df2_2["Rel Hum_%"])`

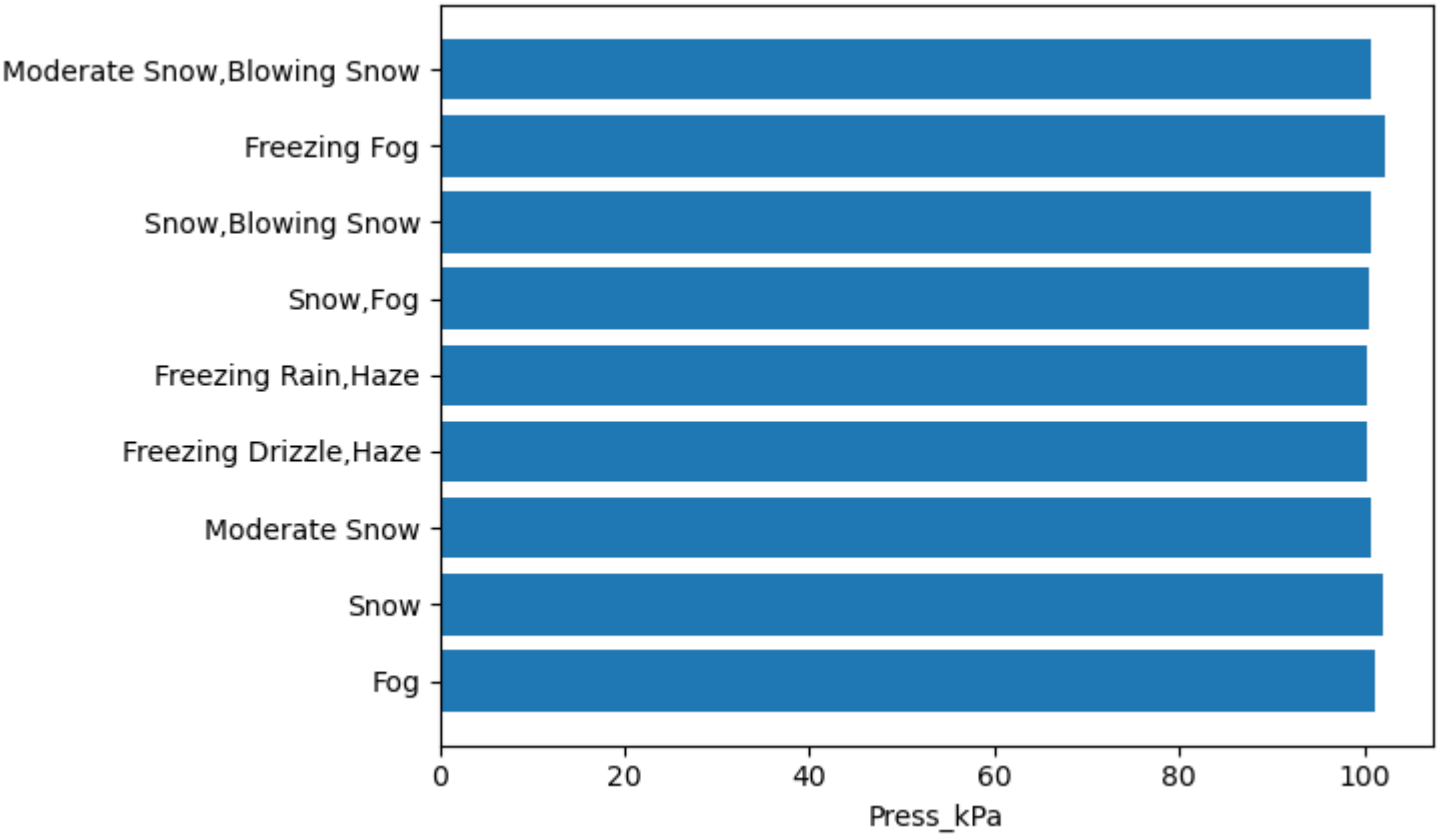
Out[23]: <BarContainer object of 80 artists>



In All Of the Weather Conditions Humidity is above 80

```
In [24]: plt.xlabel("Press_kPa")
plt.barh(result_df2_2["Weather"],result_df2_2["Press_kPa"])
```

Out[24]: <BarContainer object of 80 artists>



```
In [ ]:
```

Now That We analyzed when the Visiblity was low, Lets Analyze the Temprature, Wind Speed, Humidity When Visibility is High

```
In [25]: result_df3 = result_df2.loc[result_df2["Visibility_km"]>=20]
result_df3
```



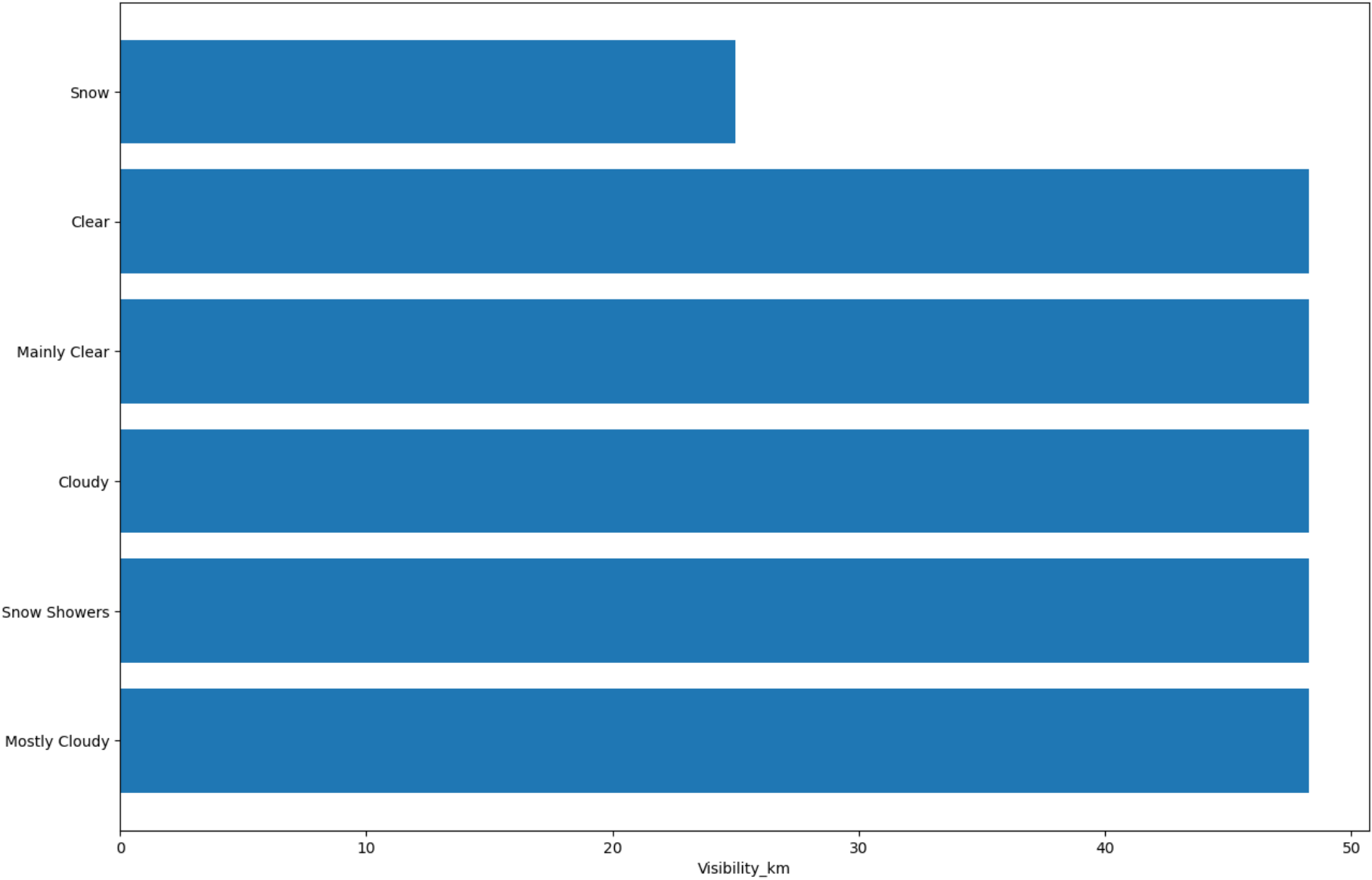
Out[25]:

	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
39	1/2/2012 15:00	0.0	-7.0	59	33	24.1	99.41	Mostly Cloudy
40	1/2/2012 16:00	-0.7	-8.7	55	24	24.1	99.50	Mostly Cloudy
41	1/2/2012 17:00	-2.1	-9.5	57	22	25.0	99.66	Snow Showers
42	1/2/2012 18:00	-4.1	-11.4	57	28	25.0	99.86	Mostly Cloudy
43	1/2/2012 19:00	-4.8	-12.1	57	24	25.0	100.00	Mostly Cloudy
...	...	...	...	...	...	...	...	...
8708	12/28/2012 20:00	-9.8	-12.6	80	20	25.0	101.35	Mainly Clear
8709	12/28/2012 21:00	-9.8	-12.5	81	19	25.0	101.36	Mainly Clear
8710	12/28/2012 22:00	-9.9	-12.5	81	17	25.0	101.38	Mainly Clear
8743	12/30/2012 7:00	-9.9	-13.2	77	19	25.0	100.48	Mostly Cloudy
8766	12/31/2012 6:00	-9.7	-11.7	85	4	25.0	101.23	Cloudy

1182 rows × 8 columns

```
In [26]: plt.figure(figsize=(15,10))
plt.xlabel("Visibility_km")
plt.barh(result_df3["Weather"],result_df3["Visibility_km"])
```

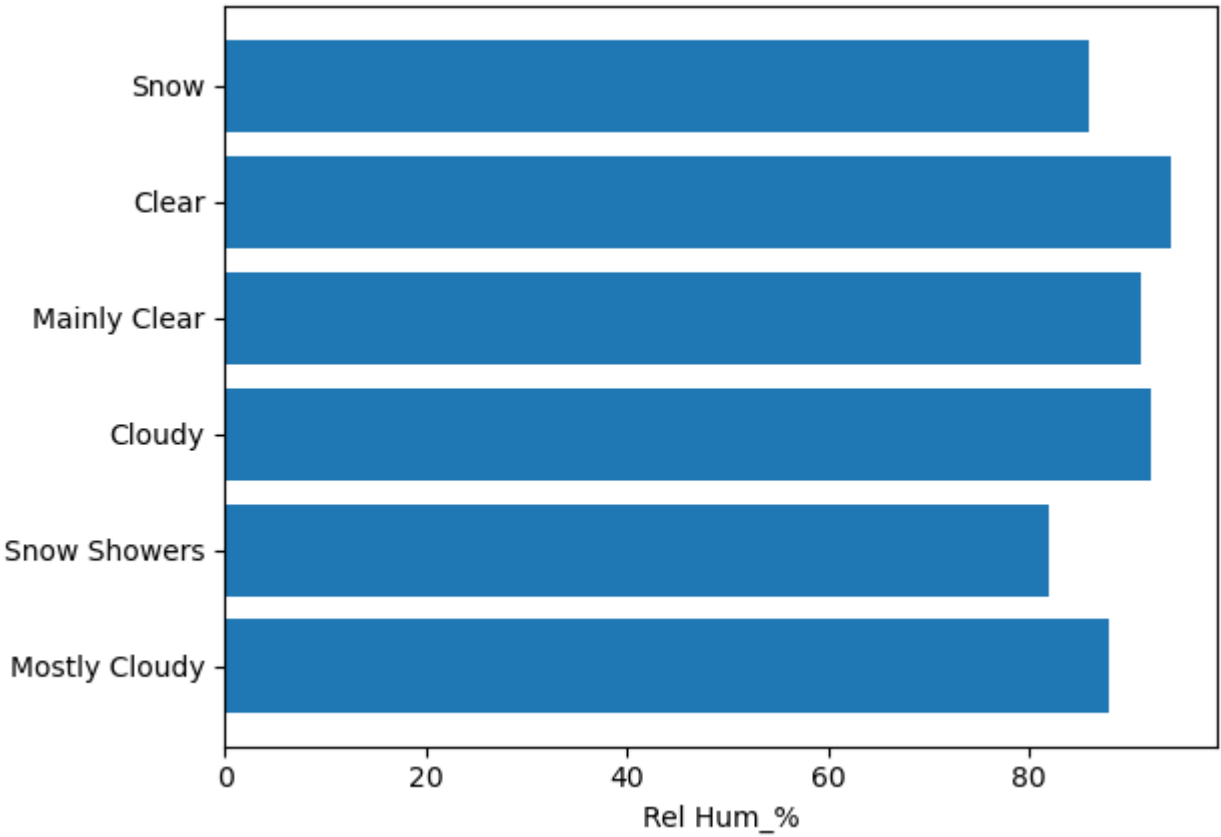
Out[26]: <BarContainer object of 1182 artists>



## Lets Look At the Temperatures and Humidity

```
In [27]: plt.xlabel("Rel Hum_%")
plt.barh(result_df3["Weather"],result_df3["Rel Hum_%"])
```

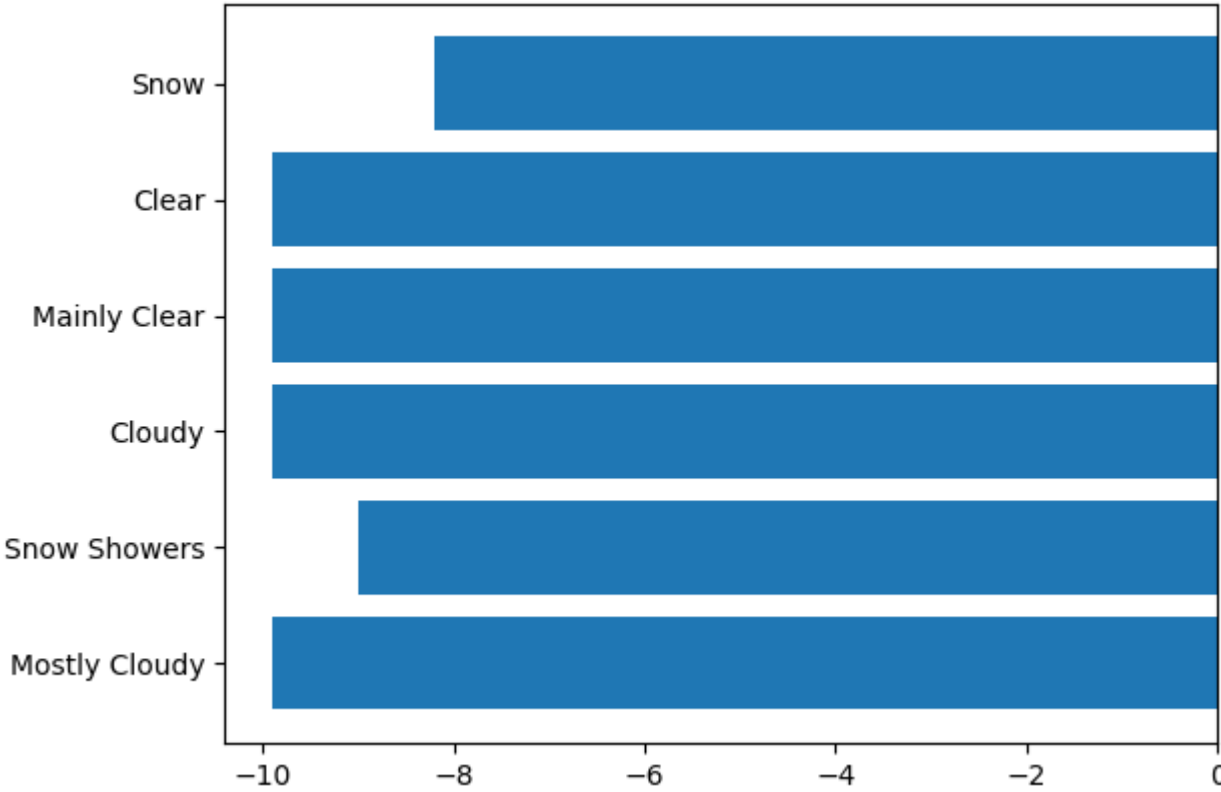
Out[27]: <BarContainer object of 1182 artists>



Though The Visiblity is Better Humidity Is still high

```
In [28]: plt.barh(result_df3["Weather"],result_df3["Temp_C"])
```

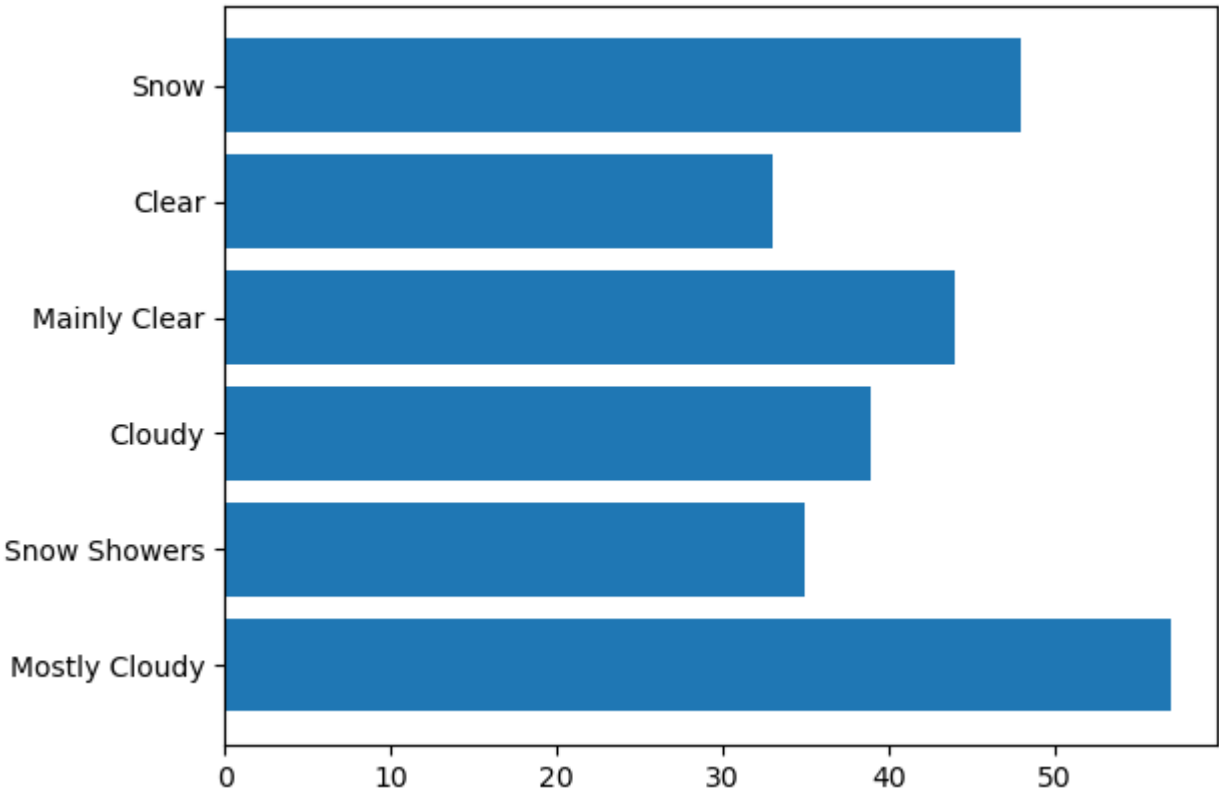
Out[28]: <BarContainer object of 1182 artists>



Temprature Is also Quiet cold... Exeeding -8 Degree Celcius

```
In [29]: plt.barh(result_df3["Weather"],result_df3["Wind Speed_km/h"])
```

Out[29]: <BarContainer object of 1182 artists>



Wind Speed Seems Better,,, Hovering between 30-40 km

Lets See Which season it was when Visiblity was High and Temps were within -10 degree Celcius

1,2,3,4,10,11,12

Jan, Feb, March, April, October, November And December 2012 had high visiblity

Jan, Feb, March, April Had Low Visiblity In 2012

```
In [ ]:
```

Analyzing the data when Temprature was 0 - 20 degree Celcius

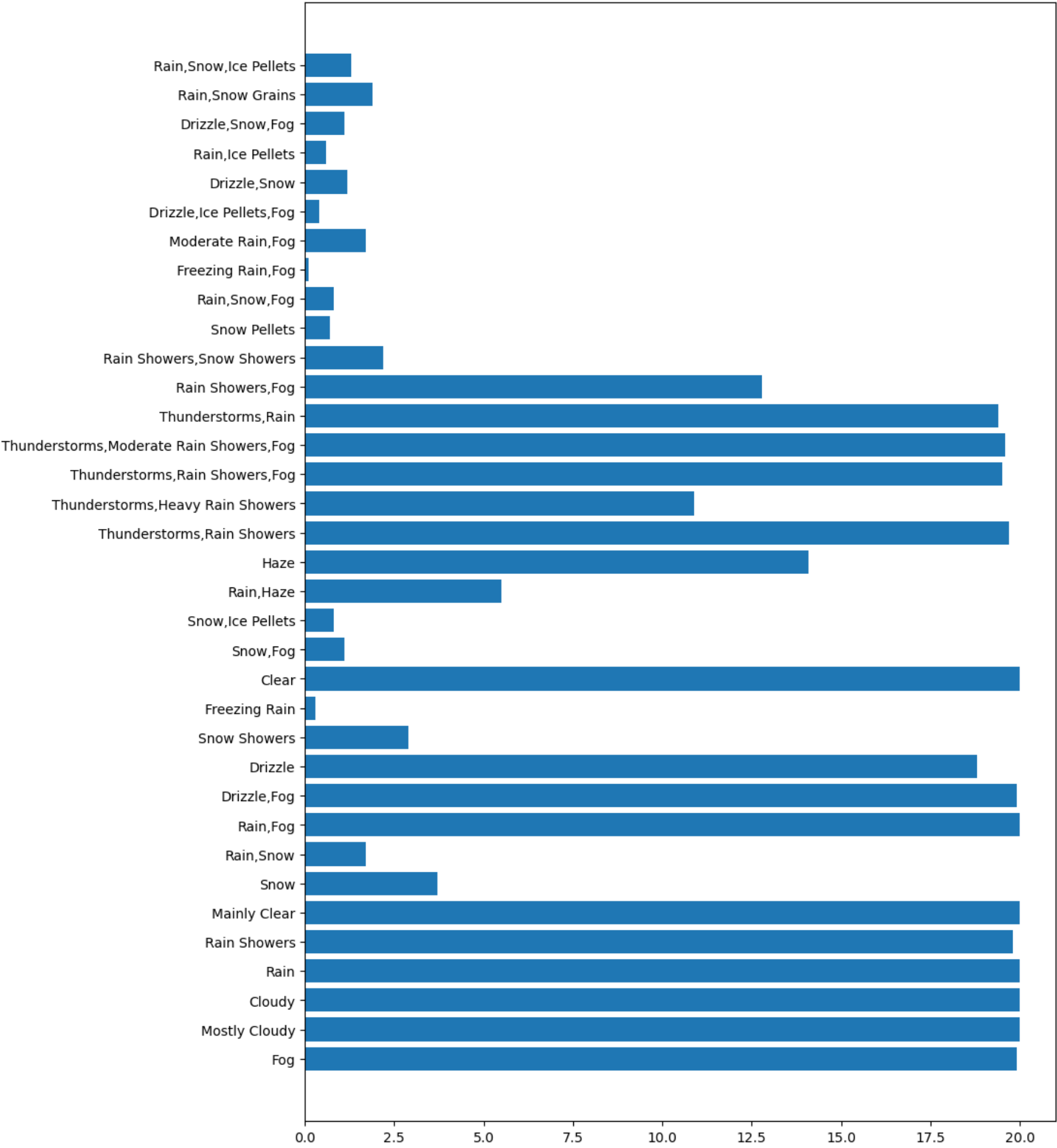
```
In [32]: result_df4 = df.loc[df["Temp_C"]> 0]
result_df5 = result_df4.loc[result_df4["Temp_C"]<=20]
```

```
In [33]: result_df5.head()
```

Out[33]:	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
13	1/1/2012 13:00	0.2	-1.7	87	13	4.8	100.58	Fog
14	1/1/2012 14:00	0.8	-1.1	87	20	4.8	100.31	Fog
15	1/1/2012 15:00	1.8	-0.4	85	22	6.4	100.07	Fog
16	1/1/2012 16:00	2.6	-0.2	82	13	12.9	99.93	Mostly Cloudy
17	1/1/2012 17:00	3.0	0.0	81	13	16.1	99.81	Cloudy

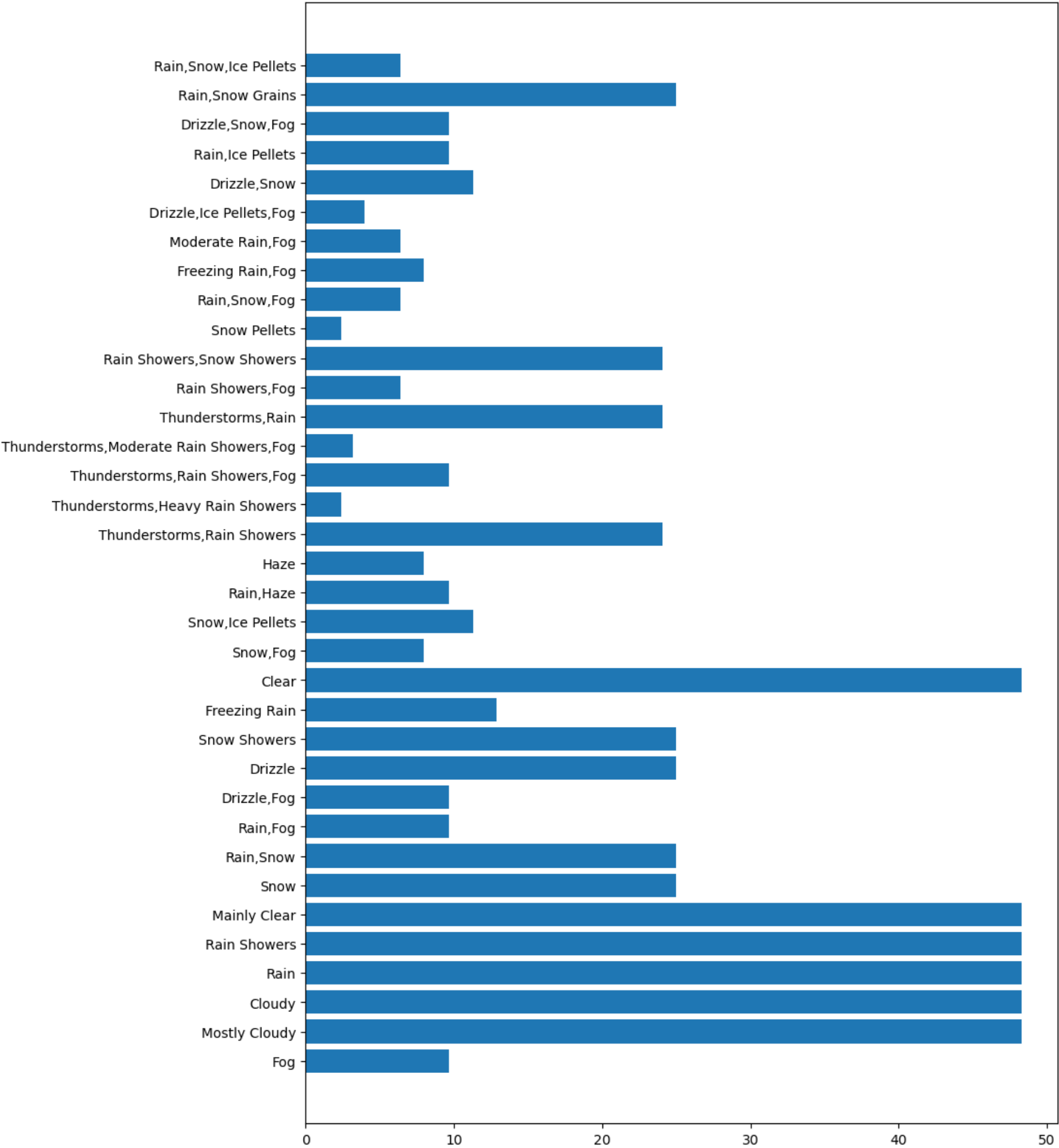
```
In [34]: plt.figure(figsize=(10,15))
plt.barh(result_df5["Weather"],result_df5["Temp_C"])
```

```
Out[34]: <BarContainer object of 4747 artists>
```



```
In [35]: plt.figure(figsize=(10,15))
plt.barh(result_df5["Weather"],result_df5["Visibility_km"])
```

Out[35]: <BarContainer object of 4747 artists>



Analyzing the Data when Visiblity was more than 20km

```
In [36]: result_df5_1 = result_df5.loc[result_df5["Visibility_km"]>20]
```

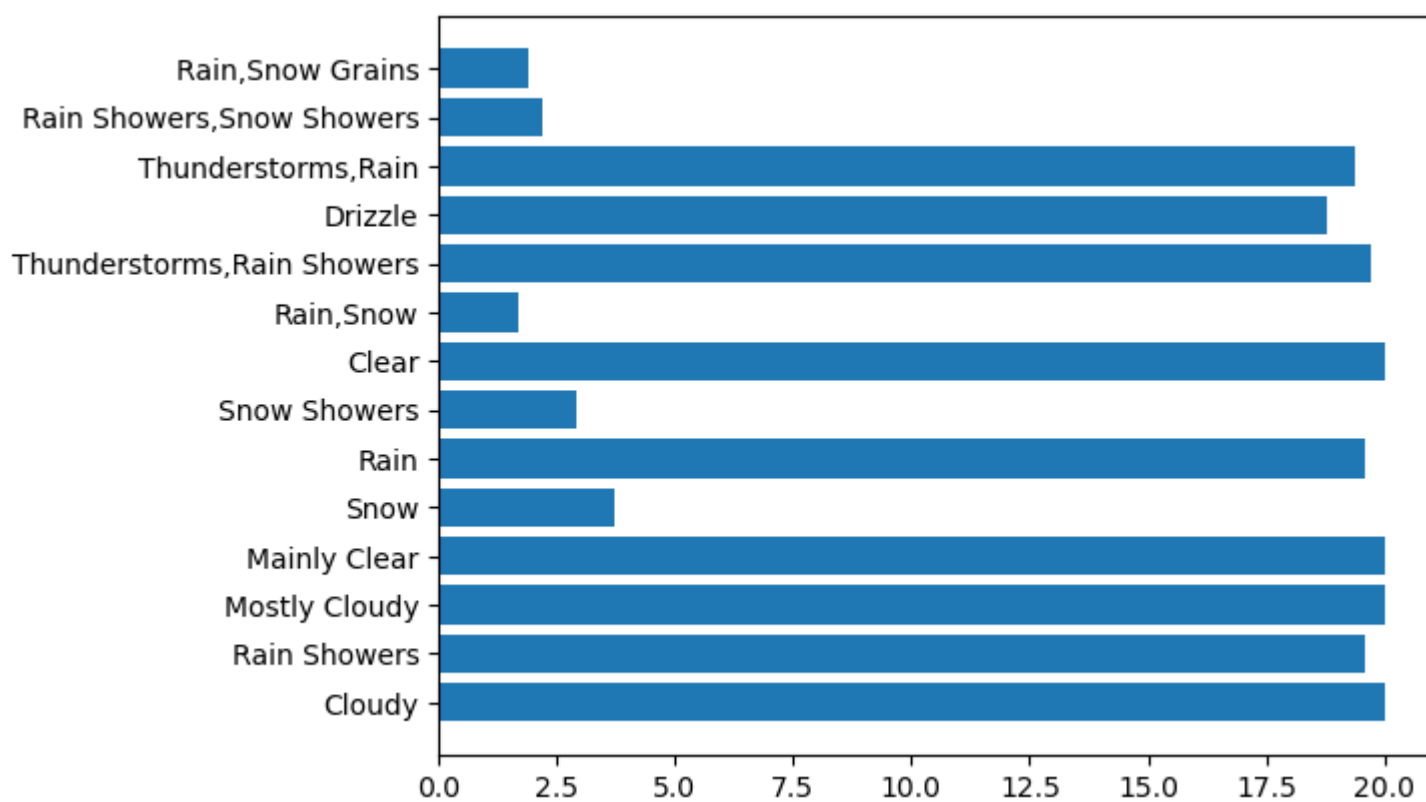
```
In [37]: result_df5_1.head()
```

Out[37]:

	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather
20	1/1/2012 20:00	3.2	1.3	87	19	25.0	99.50	Cloudy
21	1/1/2012 21:00	4.0	1.7	85	20	25.0	99.39	Cloudy
23	1/1/2012 23:00	5.3	2.0	79	30	25.0	99.31	Cloudy
24	1/2/2012 0:00	5.2	1.5	77	35	25.0	99.26	Rain Showers
25	1/2/2012 1:00	4.6	0.0	72	39	25.0	99.26	Cloudy

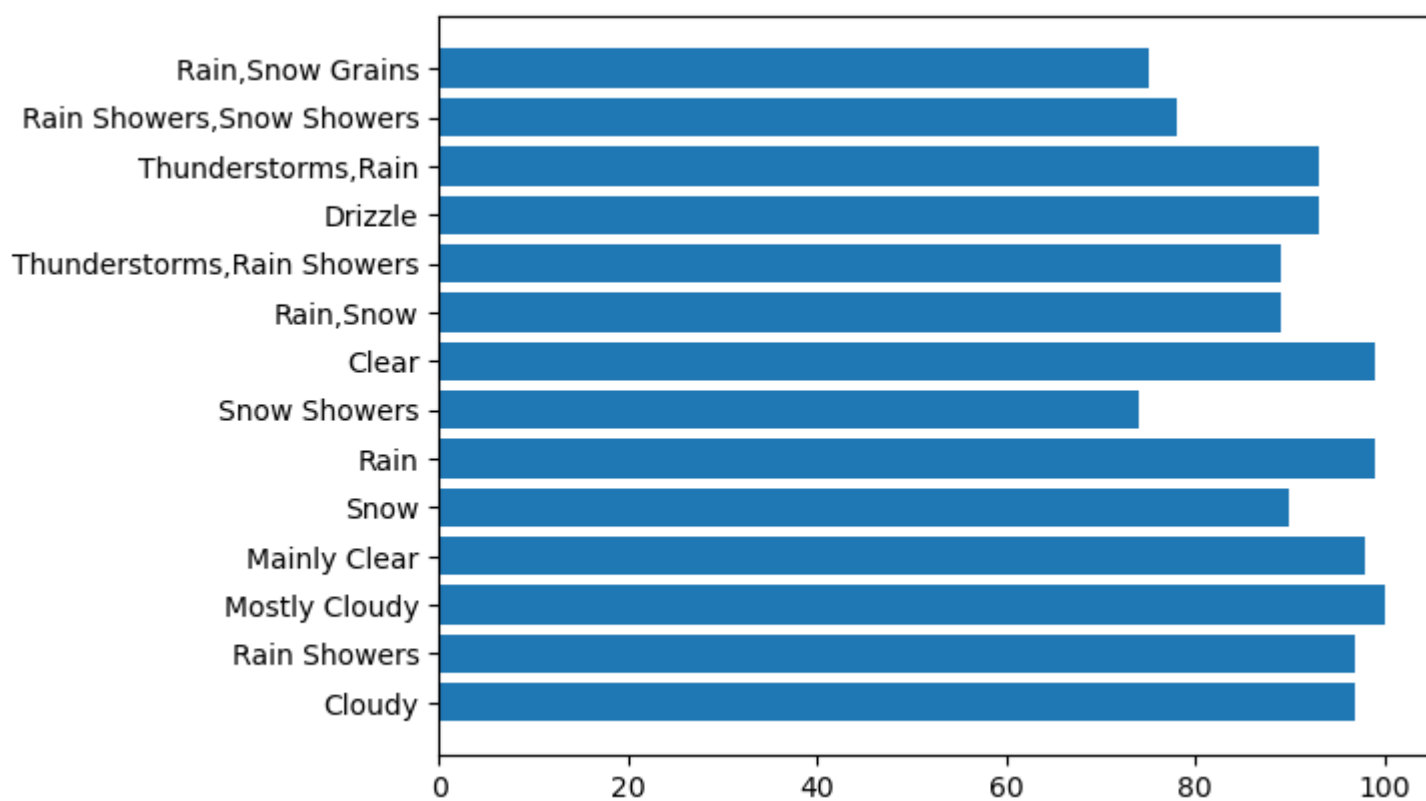
```
In [38]: plt.barh(result_df5_1["Weather"],result_df5_1["Temp_C"])
```

Out[38]: <BarContainer object of 3872 artists>



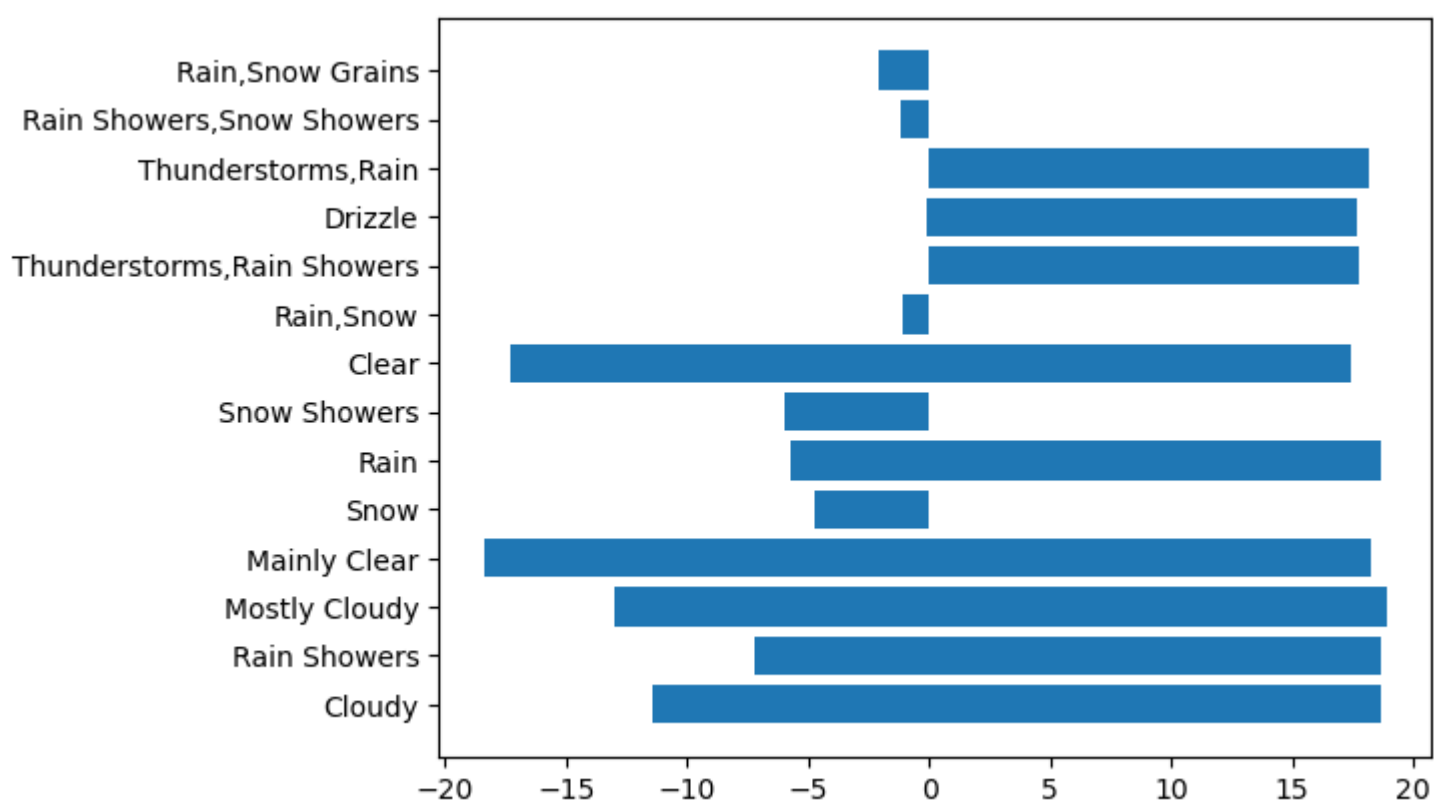
```
In [39]: plt.barh(result_df5_1["Weather"],result_df5_1["Rel Hum_%"])
```

```
Out[39]: <BarContainer object of 3872 artists>
```



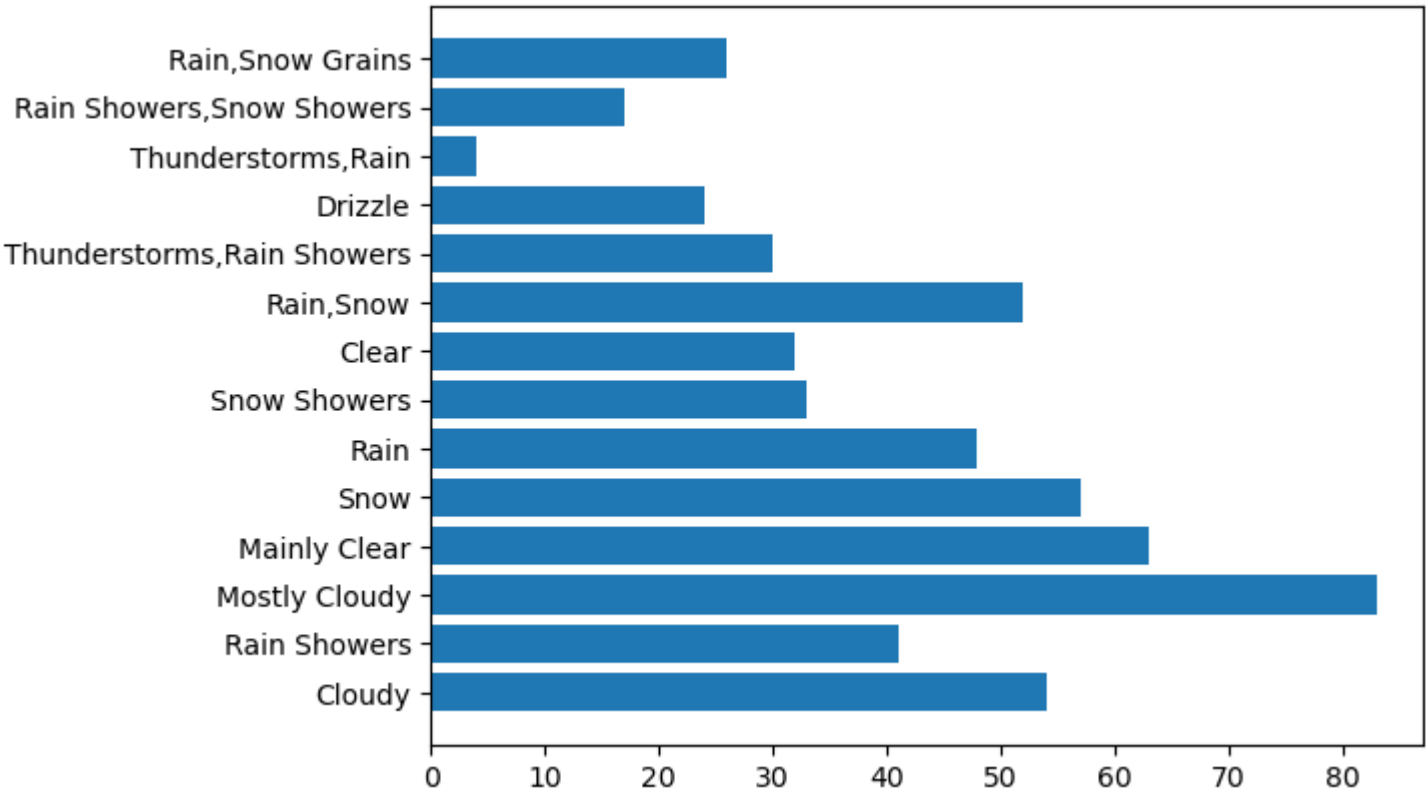
```
In [40]: plt.barh(result_df5_1["Weather"],result_df5_1["Dew Point Temp_C"])
```

```
Out[40]: <BarContainer object of 3872 artists>
```



```
In [41]: plt.barh(result_df5_1["Weather"],result_df5_1["Wind Speed_km/h"])
```

```
Out[41]: <BarContainer object of 3872 artists>
```



In [ ]:

### Classifying Data

```
In [185... X1 = df[["Temp_C", "Dew Point Temp_C", "Rel Hum_%", "Wind Speed_kmh", "Visibility_km", "Press_kPa"]]
Y1 = df["Weather"]

from sklearn.model_selection import train_test_split
```

```
In [195... x_train, x_test, y_train, y_test = train_test_split(X1, Y1, test_size=0.3, random_state=42, shuffle=True)
```

```
In [200... from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import GridSearchCV
from sklearn.neighbors import KNeighborsClassifier
```

```
In [248... model = DecisionTreeClassifier(criterion="gini")
model.fit(x_train, y_train)
```

Out[248... DecisionTreeClassifier ⓘ ?

DecisionTreeClassifier()

```
In [249... pred = model.predict(x_test)
pred
accuracy_score(y_test, pred)
```

Out[249... 0.4393019726858877

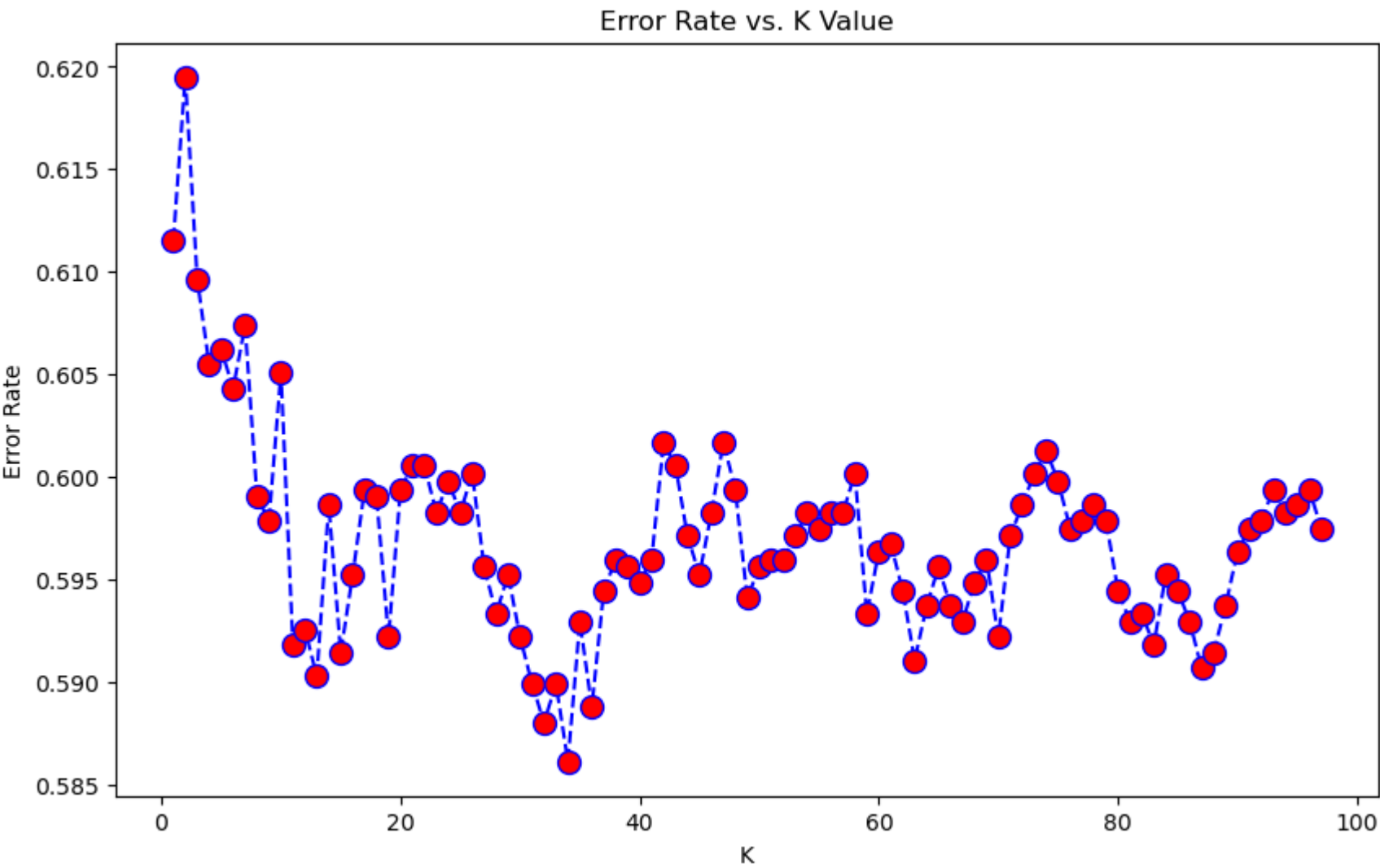
```
In [235... import numpy as np
error_rate = []

# Will take some time
for i in range(1, 98):

    knn = KNeighborsClassifier(n_neighbors=i)
    knn.fit(x_train, y_train)
    pred_i = knn.predict(x_test)
    error_rate.append(np.mean(pred_i != y_test))

plt.figure(figsize=(10, 6))
plt.plot(range(1, 98), error_rate, color='blue',
         linestyle='dashed', marker='o',
         markerfacecolor='red', markersize=10)

plt.title('Error Rate vs. K Value')
plt.xlabel('K')
plt.ylabel('Error Rate')
plt.show()
```



```
In [229... from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report

knn = KNeighborsClassifier(n_neighbors = 34)

knn.fit(x_train, y_train)
pred = knn.predict(x_test)
accuracy_score(y_test, pred)
```

Out[229... 0.41388467374810317

## Forecasting Timeseries Data

```
In [44]: df[['Date', 'Time']] = df['Date/Time'].str.split(' ', 1, expand=True)

/tmp/ipykernel_2960/1918148834.py:1: FutureWarning: In a future version of pandas all arguments of StringMethods.split
it except for the argument 'pat' will be keyword-only.
  df[['Date', 'Time']] = df['Date/Time'].str.split(' ', 1, expand=True)

In [45]: df
```



Out[45]:

	Date/Time	Temp_C	Dew Point Temp_C	Rel Hum_%	Wind Speed_km/h	Visibility_km	Press_kPa	Weather	Date	Time
0	1/1/2012 0:00	-1.8	-3.9	86	4	8.0	101.24	Fog	1/1/2012	0:00
1	1/1/2012 1:00	-1.8	-3.7	87	4	8.0	101.24	Fog	1/1/2012	1:00
2	1/1/2012 2:00	-1.8	-3.4	89	7	4.0	101.26	Freezing Drizzle,Fog	1/1/2012	2:00
3	1/1/2012 3:00	-1.5	-3.2	88	6	4.0	101.27	Freezing Drizzle,Fog	1/1/2012	3:00
4	1/1/2012 4:00	-1.5	-3.3	88	7	4.8	101.23	Fog	1/1/2012	4:00
...	...	...	...	...	...	...	...	...	...	...
8779	12/31/2012 19:00	0.1	-2.7	81	30	9.7	100.13	Snow	12/31/2012	19:00
8780	12/31/2012 20:00	0.2	-2.4	83	24	9.7	100.03	Snow	12/31/2012	20:00
8781	12/31/2012 21:00	-0.5	-1.5	93	28	4.8	99.95	Snow	12/31/2012	21:00
8782	12/31/2012 22:00	-0.2	-1.8	89	28	9.7	99.91	Snow	12/31/2012	22:00
8783	12/31/2012 23:00	0.0	-2.1	86	30	11.3	99.89	Snow	12/31/2012	23:00

8784 rows × 10 columns

```
In [46]: resp = pd.read_csv("Data.csv")
```

```
In [47]: resp = resp[["Date","Temp_C","Dew Point Temp_C"]]
resp
```

Out[47]:

	Date	Temp_C	Dew Point Temp_C
0	1/1/2012	1.8	-0.4
1	1/2/2012	0.0	-7.0
2	1/3/2012	-14.8	-22.2
3	1/4/2012	-10.2	-16.3
4	1/5/2012	-4.3	-12.0
...	...	...	...
361	12/27/2012	-5.0	-6.2
362	12/28/2012	-6.8	-10.3
363	12/29/2012	-8.8	-10.0
364	12/30/2012	-11.3	-15.6
365	12/31/2012	-2.3	-4.6

366 rows × 3 columns

```
In [48]: resp.to_csv("Data.csv")
```

```
In [49]: resp["Date"] = pd.to_datetime(resp["Date"])
```

```
In [50]: resp.dtypes
```

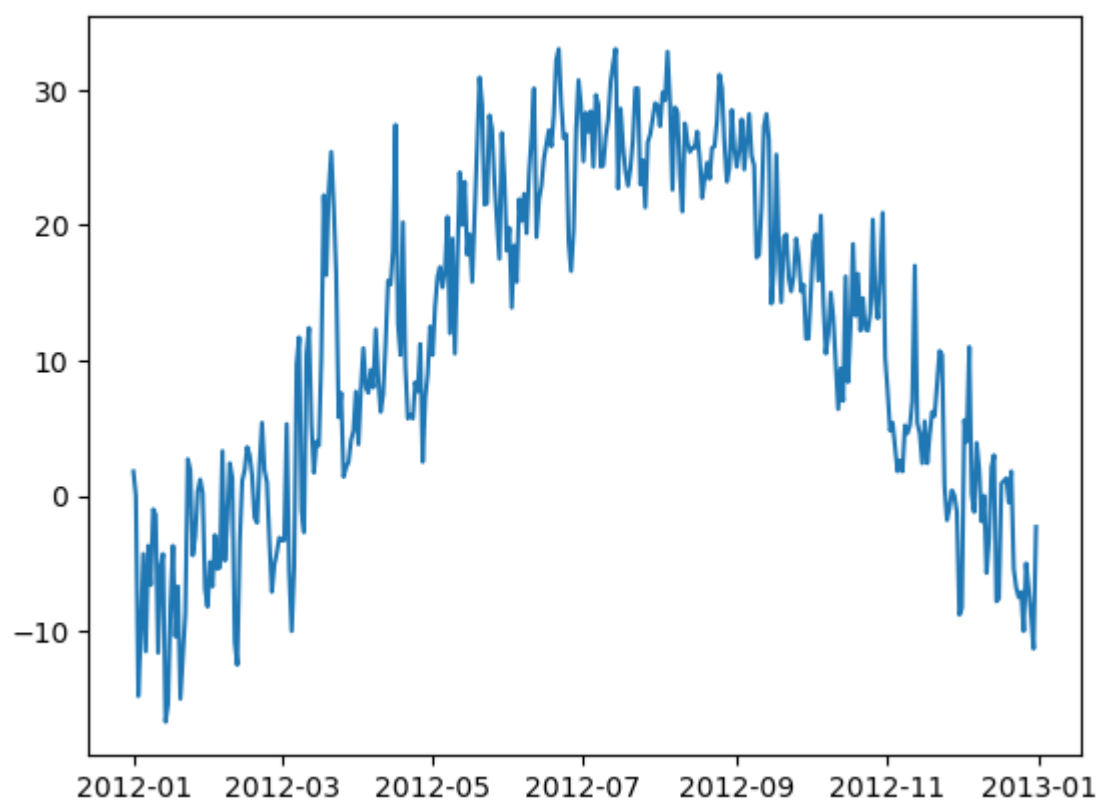
Out[50]: Date                   datetime64[ns]  
Temp\_C                     float64  
Dew Point Temp\_C           float64  
dtype: object

```
In [51]: X = resp["Date"]
Y = resp["Temp_C"]
```

```
In [52]: from statsmodels.tsa.seasonal import seasonal_decompose
```

```
In [53]: plt.plot(X,Y)
```

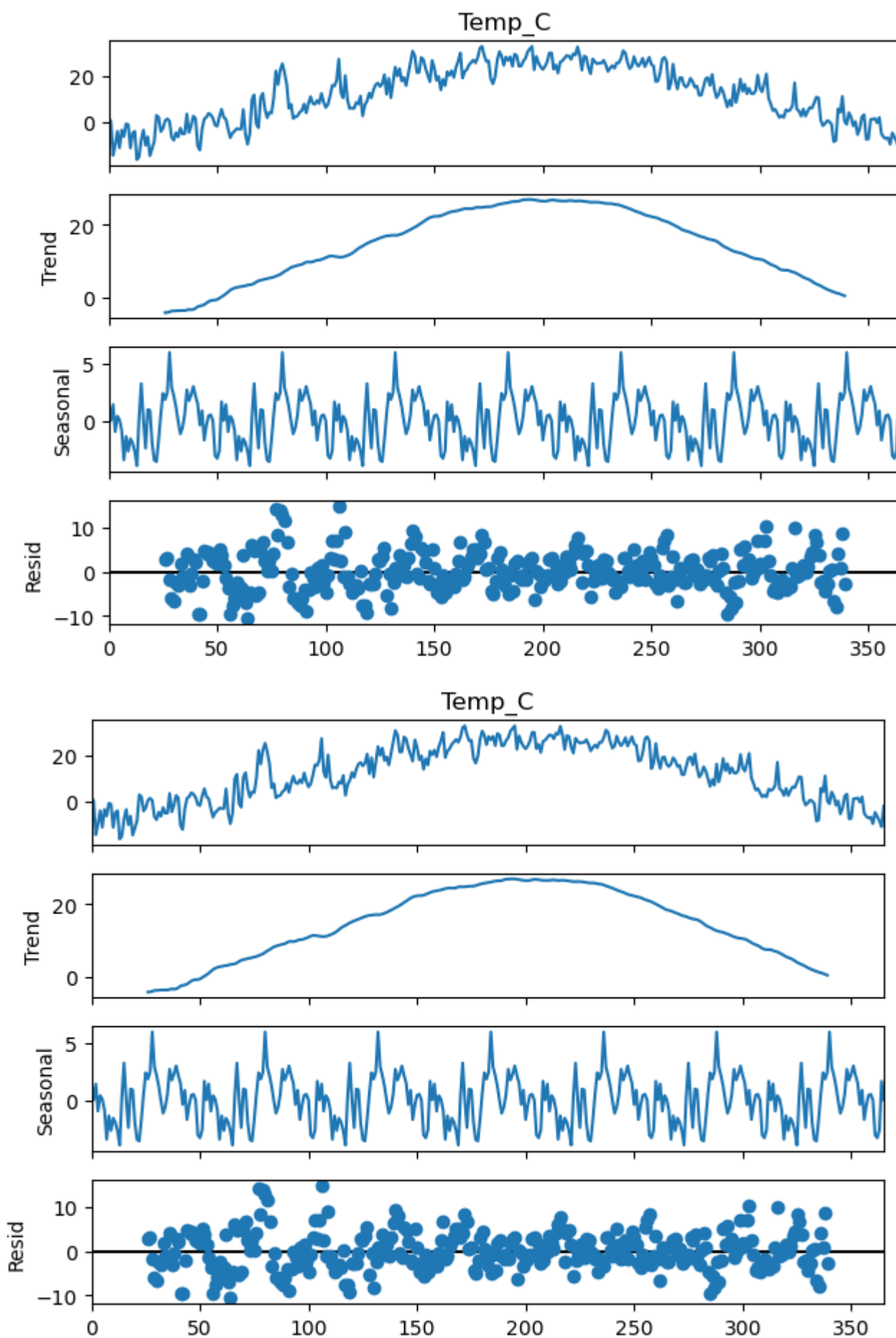
Out[53]: [<matplotlib.lines.Line2D at 0x7f57a872da50>]



```
In [54]: decomp = seasonal_decompose(Y,model="additive",period=52)
```

```
In [55]: decomp.plot()
```

Out[55]:



```
In [56]: train = resp.iloc[:int(resp.shape[0]*0.70)]
test = resp.iloc[int(resp.shape[0]*0.70):]
```

```
In [57]: train.head()
```

Out[57]:

	Date	Temp_C	Dew Point Temp_C
0	2012-01-01	1.8	-0.4
1	2012-01-02	0.0	-7.0
2	2012-01-03	-14.8	-22.2
3	2012-01-04	-10.2	-16.3
4	2012-01-05	-4.3	-12.0

```
In [58]: test.head()
```

Out[58]:

	Date	Temp_C	Dew Point Temp_C
256	2012-09-13	28.2	13.9
257	2012-09-14	26.3	16.4
258	2012-09-15	14.2	11.4
259	2012-09-16	16.7	3.6
260	2012-09-17	25.2	10.8

```
In [59]: import statsmodels.api as sm
```

```
In [60]: exponential_smoothing = sm.tsa.ExponentialSmoothing(train["Temp_C"],trend="additive",seasonal="additive",seasonal_p
```

```
In [61]: model = exponential_smoothing.fit()
forecast = model.forecast(len(test))
forecast
```

Out[61]:

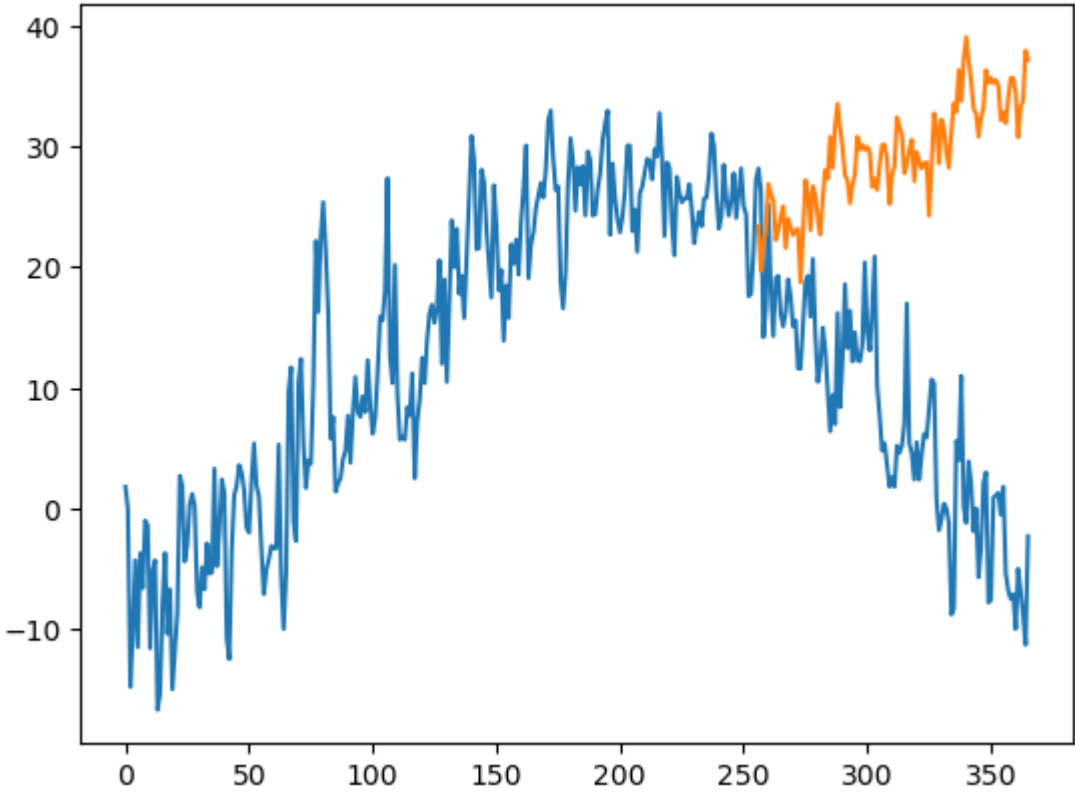
256	23.421722
257	19.722149
258	22.169364
259	22.823211
260	26.910514
	...
361	30.787608
362	33.234822
363	33.888670
364	37.975972
365	37.214722

Length: 110, dtype: float64

```
In [ ]:
```

```
In [62]: plt.plot(resp['Temp_C'])
plt.plot(forecast)
plt.plot
```

Out[62]: <function matplotlib.pyplot.plot(\*args: 'float | ArrayLike | str', scalex: 'bool' = True, scaley: 'bool' = True, d  
ata=None, \*\*kwargs) -> 'list[Line2D]'



```
In [63]: model = sm.tsa.ExponentialSmoothing(test["Temp_C"],trend='additive',seasonal="additive",seasonal_periods=52)
tes_model = model.fit()
forecast1 = tes_model.forecast(8)
```

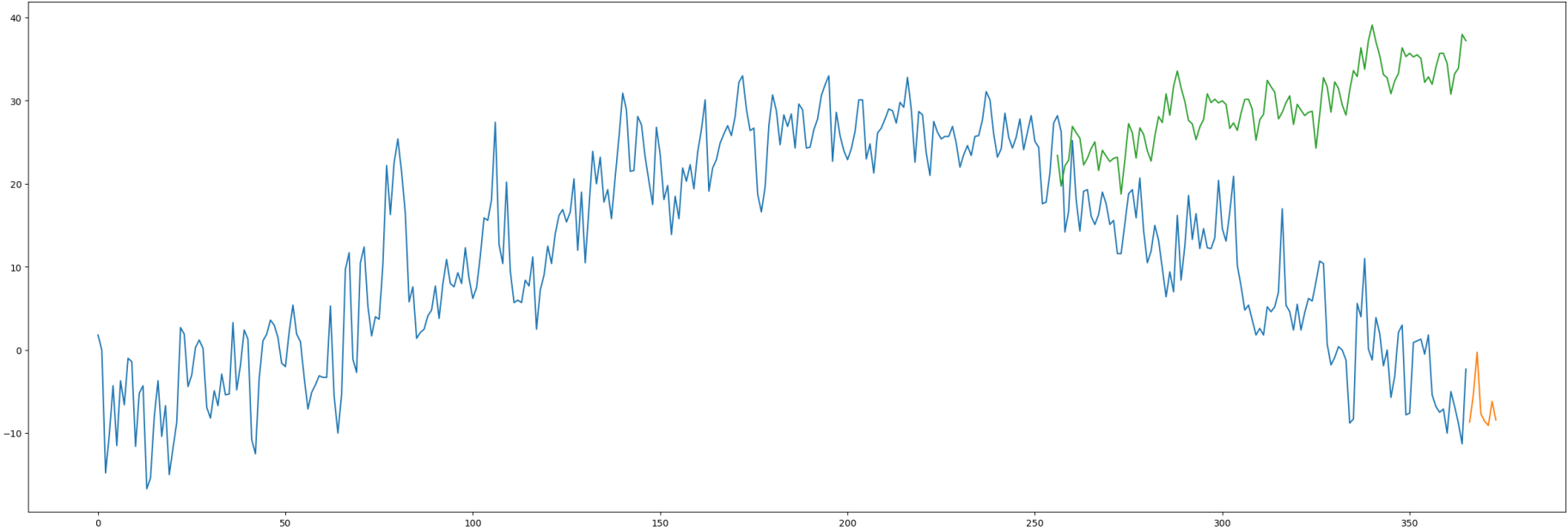
```
In [64]: forecast1
```

Out[64]: 366 -8.677518  
367 -5.377505  
368 -0.277631  
369 -7.677573  
370 -8.577687  
371 -9.077681  
372 -6.177736  
373 -8.427788  
dtype: float64

```
In [65]: plt.figure(figsize=(30,10))
plt.plot(resp['Temp_C'])
plt.plot(forecast1)
plt.plot(forecast)

plt.plot
```

Out[65]: <function matplotlib.pyplot.plot(\*args: 'float | ArrayLike | str', scalex: 'bool' = True, scaley: 'bool' = True, data=None, \*\*kwargs) -> 'list[Line2D]'



```
In [ ]:
```