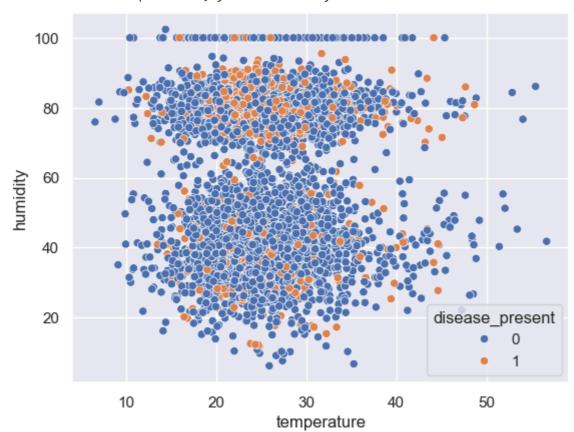
this data is pretty simple, and even humans can read it. so we can use simple models like decision tree classifiers to analise and predict them



<Axes: xlabel='temperature', ylabel='humidity'>

Humidity and Disease Presence

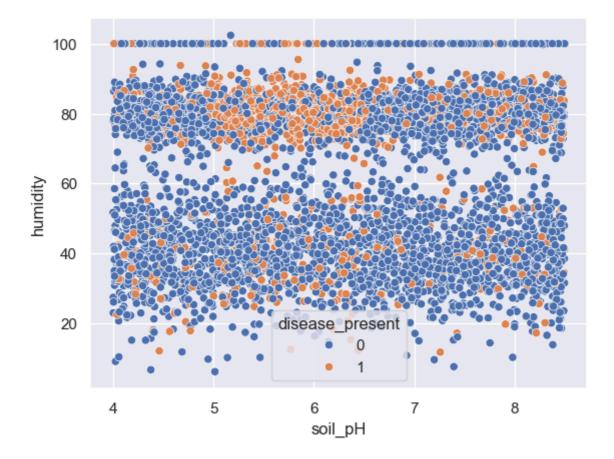
Humidity plays a significant role in the presence of plant disease. The scatter plot reveals **two distinct clusters** based on humidity:

- low humidity (~40%)
- and high humidity (~80%)

The **high humidity cluster** shows a much higher rate of disease presence.

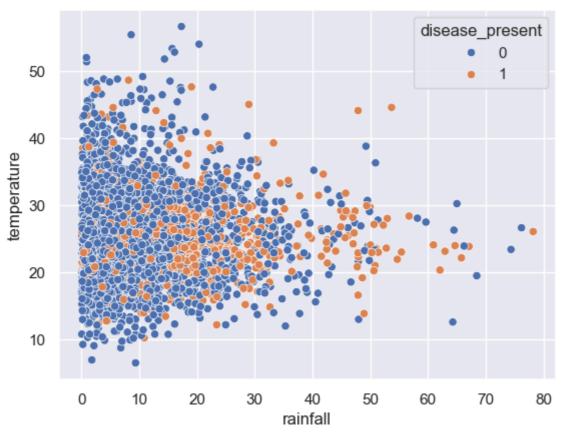
This might suggests that **elevated humidity** creates favorable conditions for disease

<Axes: xlabel='soil_pH', ylabel='humidity'>



- the area with ph level between **5 and 6.5** in the high humidity cluster
- has a pretty high chance of disease presence

<Axes: xlabel='rainfall', ylabel='temperature'>



rainfall kinda behaves similarly to humidity, sample size is small, but anything over **20mm** has high **disease** risk it's a good idea to get more data on that range.

well i think it's time to write some models, we'll train a decision tree classifier on the data

	temperature	humidity	rainfall	soil_pH	disease_present			
6252	38.189409	81.236450	1.821797	5.189183	0			
4684	19.814670	32.527782	8.650504	5.983636	1			
1731	23.253416	64.457696	8.218361	4.022305	0			
4742	25.008095	44.829299	2.177695	8.153636	0			
4521	30.371665	41.536491	14.861256	7.288489	0			
Accuracy: 0.77								

well i was wrong, this data seems to be a little tougher than i expected we're goonna need bigger guns, a **random forest** model might do the trick

Accuracy with	Random Forest: 0.86			
	precision	recall	f1-score	support
0	0.87	0.96	0.91	1147
1	0.79	0.54	0.64	353
accuracy			0.86	1500
macro avg	0.83	0.75	0.78	1500
weighted avg	0.85	0.86	0.85	1500

way better results, but the model has high false negative rates,