

In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
```

In [2]:

```
import io
%cd "C:\Users\deepe\OneDrive\Desktop\Python Datasets\Projects\Capstone"
```

C:\Users\deepe\OneDrive\Desktop\Python Datasets\Projects\Capstone

In [3]:

```
crop=pd.read_csv("Crop_recommendation.csv")
```

In [4]:

```
crop.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2200 entries, 0 to 2199
Data columns (total 8 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   N               2200 non-null   int64
 1   P               2200 non-null   int64
 2   K               2200 non-null   int64
 3   temperature     2200 non-null   float64
 4   humidity        2200 non-null   float64
 5   ph              2200 non-null   float64
 6   rainfall        2200 non-null   float64
 7   label           2200 non-null   object
dtypes: float64(4), int64(3), object(1)
memory usage: 137.6+ KB
```

In [5]:

```
crop.tail()
```

Out[5]:

	N	P	K	temperature	humidity	ph	rainfall	label
<b>2195</b>	107	34	32	26.774637	66.413269	6.780064	177.774507	coffee
<b>2196</b>	99	15	27	27.417112	56.636362	6.086922	127.924610	coffee
<b>2197</b>	118	33	30	24.131797	67.225123	6.362608	173.322839	coffee
<b>2198</b>	117	32	34	26.272418	52.127394	6.758793	127.175293	coffee
<b>2199</b>	104	18	30	23.603016	60.396475	6.779833	140.937041	coffee

In [6]:

```
crop.describe()
```

Out[6]:

	N	P	K	temperature	humidity	ph	rainfall
count	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000
mean	50.551818	53.362727	48.149091	25.616244	71.481779	6.469480	103.420000
std	36.917334	32.985883	50.647931	5.063749	22.263812	0.773938	54.500000
min	0.000000	5.000000	5.000000	8.825675	14.258040	3.504752	20.000000
25%	21.000000	28.000000	20.000000	22.769375	60.261953	5.971693	64.500000
50%	37.000000	51.000000	32.000000	25.598693	80.473146	6.425045	94.500000
75%	84.250000	68.000000	49.000000	28.561654	89.948771	6.923643	124.500000
max	140.000000	145.000000	205.000000	43.675493	99.981876	9.935091	298.500000

In [7]:

```
crop.label.describe()
```

Out[7]:

```
count      2200
unique         22
top         rice
freq         100
Name: label, dtype: object
```

In [8]:

```
crop.columns
```

Out[8]:

```
Index(['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall', 'label'], dtype='object')
```

In [9]:

```
crop.shape
```

Out[9]:

```
(2200, 8)
```

In [10]:

```
crop.isnull().sum()
```

Out[10]:

```
N          0
P          0
K          0
temperature 0
humidity    0
ph          0
rainfall    0
label       0
dtype: int64
```

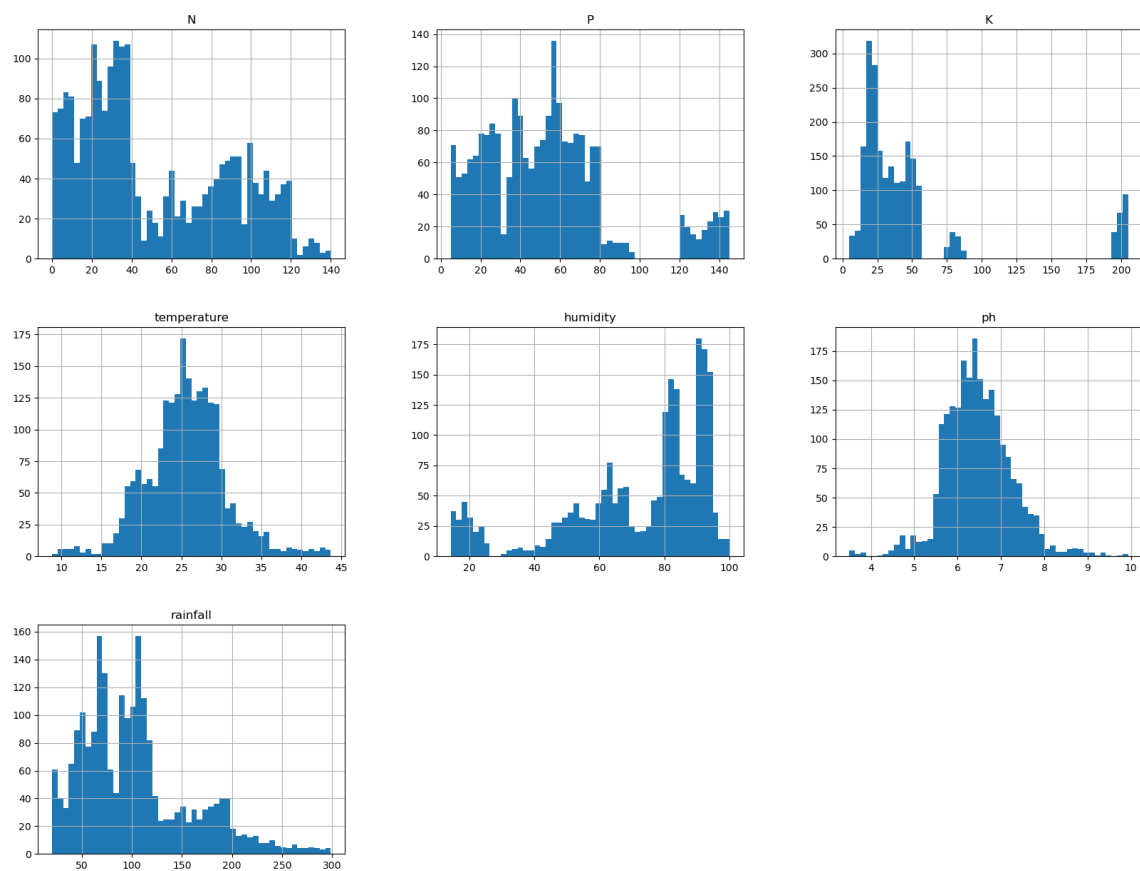
In [11]:

```
print(crop.isnull().sum())
```

```
N          0
P          0
K          0
temperature 0
humidity    0
ph          0
rainfall    0
label       0
dtype: int64
```

In [12]:

```
crop.hist(bins=50,figsize=(20,15))  
plt.show()
```



In [13]:

```
crop.apply(lambda x:len(x.unique()))
```

Out[13]:

N	137
P	117
K	73
temperature	2200
humidity	2200
ph	2200
rainfall	2200
label	22
dtype:	int64

In [14]:

```
crop['label'].value_counts(dropna=False)
```

Out[14]:

rice	100
maize	100
jute	100
cotton	100
coconut	100
papaya	100
orange	100
apple	100
muskmelon	100
watermelon	100
grapes	100
mango	100
banana	100
pomegranate	100
lentil	100
blackgram	100
mungbean	100
mothbeans	100
pigeonpeas	100
kidneybeans	100
chickpea	100
coffee	100

Name: label, dtype: int64

In [15]:

```
objcols=crop[['label']]
```

In [16]:

```
objcols.head(2205)
```

Out[16]:

	label
0	rice
1	rice
2	rice
3	rice
4	rice
...	...
2195	coffee
2196	coffee
2197	coffee
2198	coffee
2199	coffee

2200 rows × 1 columns

In [17]:

```
numcols=crop[['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall']]
```

In [18]:

```
numcols.isnull().sum().sort_values(ascending=False)/numcols.shape[0]
```

Out[18]:

N	0.0
P	0.0
K	0.0
temperature	0.0
humidity	0.0
ph	0.0
rainfall	0.0
dtype:	float64

In [19]:

```
for col in numcols.columns:
    numcols[col]=numcols[col].fillna(numcols[col].median())
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\1213558515.py:2: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy) ([https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy))

```
numcols[col]=numcols[col].fillna(numcols[col].median())
```

In [20]:

```
numcols.head()
```

Out[20]:

	N	P	K	temperature	humidity	ph	rainfall
0	90	42	43	20.879744	82.002744	6.502985	202.935536
1	85	58	41	21.770462	80.319644	7.038096	226.655537
2	60	55	44	23.004459	82.320763	7.840207	263.964248
3	74	35	40	26.491096	80.158363	6.980401	242.864034
4	78	42	42	20.130175	81.604873	7.628473	262.717340

In [21]:

```
numcols.describe()
```

Out[21]:

	N	P	K	temperature	humidity	ph	rainfall
count	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000
mean	50.551818	53.362727	48.149091	25.616244	71.481779	6.469480	103.481779
std	36.917334	32.985883	50.647931	5.063749	22.263812	0.773938	54.551818
min	0.000000	5.000000	5.000000	8.825675	14.258040	3.504752	20.130175
25%	21.000000	28.000000	20.000000	22.769375	60.261953	5.971693	64.551818
50%	37.000000	51.000000	32.000000	25.598693	80.473146	6.425045	94.817779
75%	84.250000	68.000000	49.000000	28.561654	89.948771	6.923643	124.250000
max	140.000000	145.000000	205.000000	43.675493	99.981876	9.935091	298.717340

In [22]:

```
crop.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2200 entries, 0 to 2199
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype
---  -
0   N                2200 non-null   int64
1   P                2200 non-null   int64
2   K                2200 non-null   int64
3   temperature      2200 non-null   float64
4   humidity         2200 non-null   float64
5   ph               2200 non-null   float64
6   rainfall         2200 non-null   float64
7   label            2200 non-null   object
dtypes: float64(4), int64(3), object(1)
memory usage: 137.6+ KB
```

In [23]:

```
crop.head()
```

Out[23]:

	N	P	K	temperature	humidity	ph	rainfall	label
0	90	42	43	20.879744	82.002744	6.502985	202.935536	rice
1	85	58	41	21.770462	80.319644	7.038096	226.655537	rice
2	60	55	44	23.004459	82.320763	7.840207	263.964248	rice
3	74	35	40	26.491096	80.158363	6.980401	242.864034	rice
4	78	42	42	20.130175	81.604873	7.628473	262.717340	rice

## EXPLORATORY DATA ANALYSIS - EDA

HISTOGRAM: THIS HIGHLIGHTS SKEWNESS, KURTOSIS AND OUTLIERS. BOX PLOT: THIS IS HIGHLIGHTS SKEWNESS AND OUTLIERS. DENSITY CURVE: THIS IS HIGHLIGHTS SKEWNESS AND KURTOSIS.

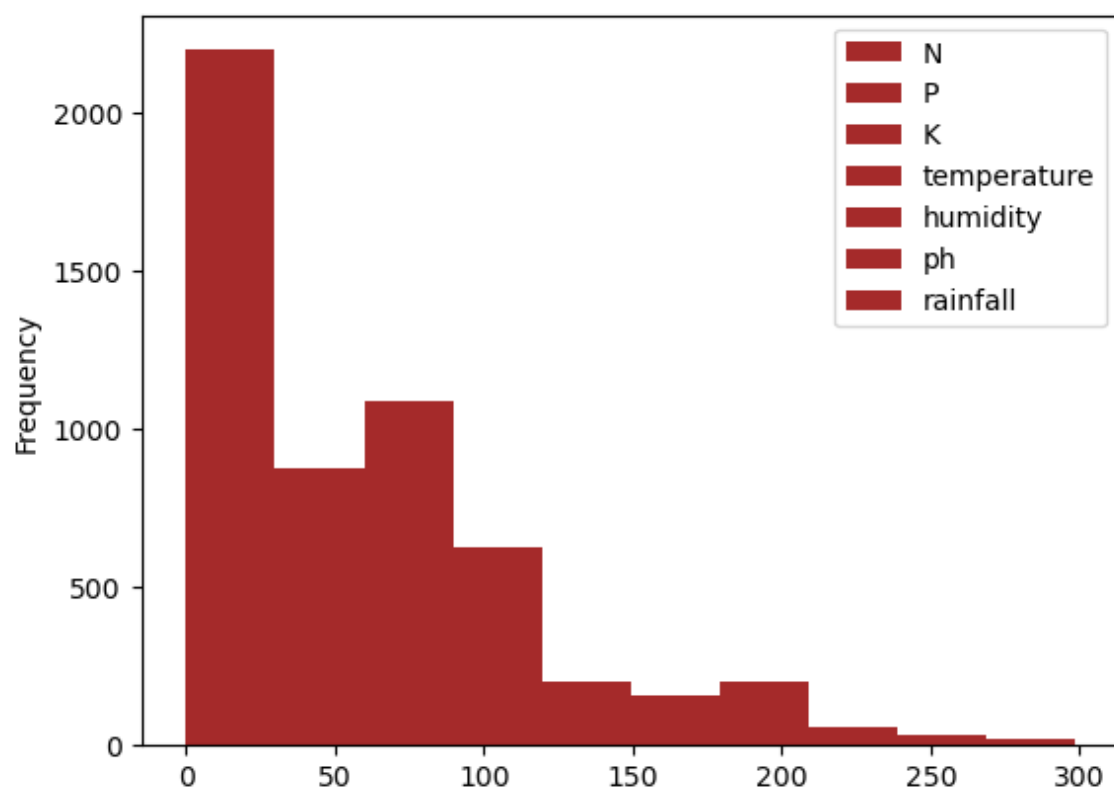


In [24]:

```
crop.plot(kind="hist",color="brown")
```

Out[24]:

<Axes: ylabel='Frequency'>

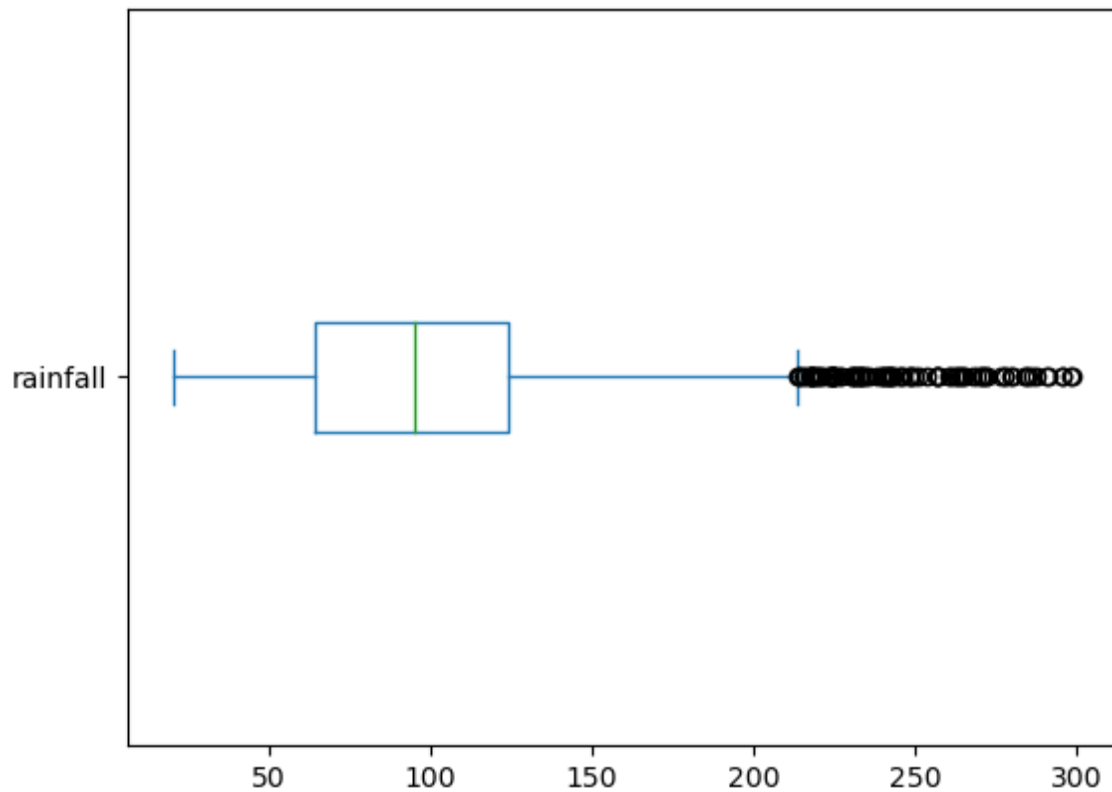


In [25]:

```
crop.rainfall.plot(kind="box", vert=False)
```

Out[25]:

<Axes: >

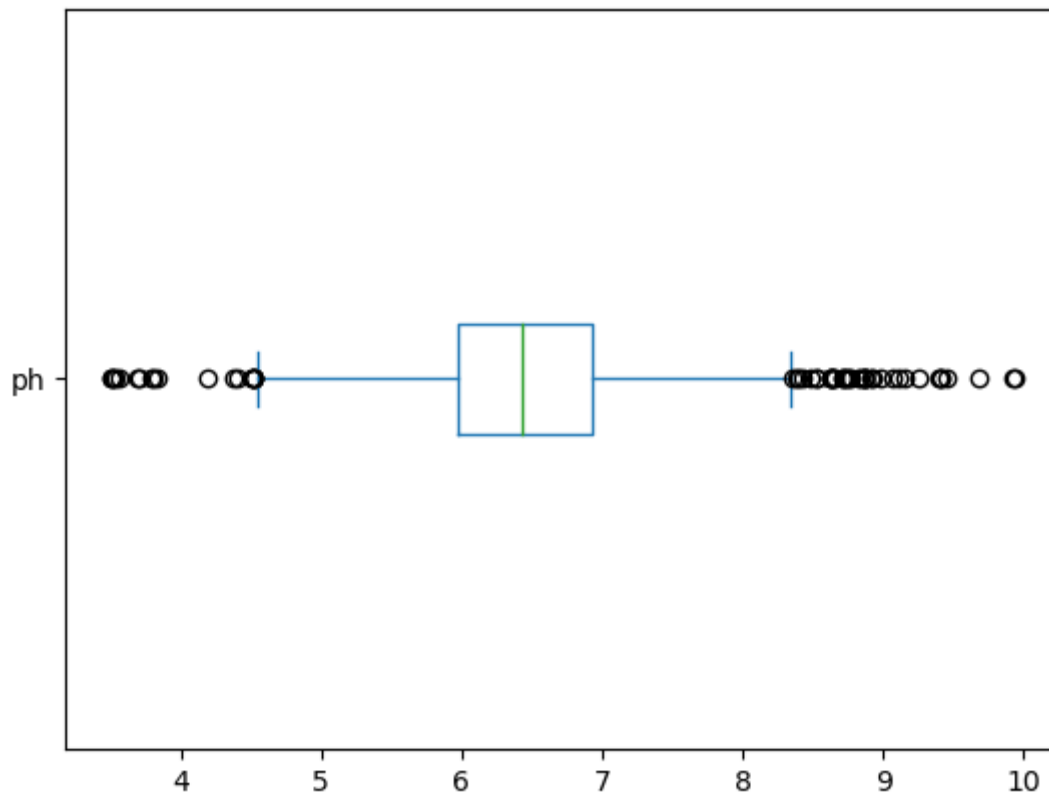


In [26]:

```
crop.ph.plot(kind="box",vert=False)
```

Out[26]:

<Axes: >

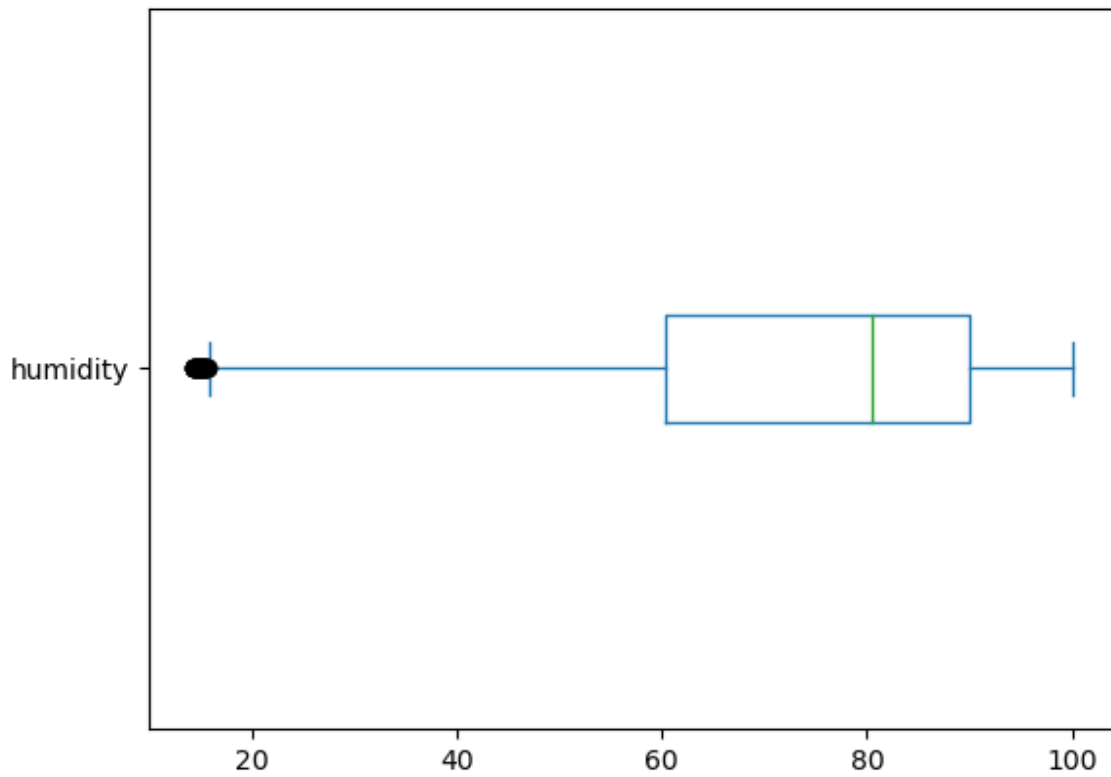


In [27]:

```
crop.humidity.plot(kind="box", vert=False)
```

Out[27]:

<Axes: >

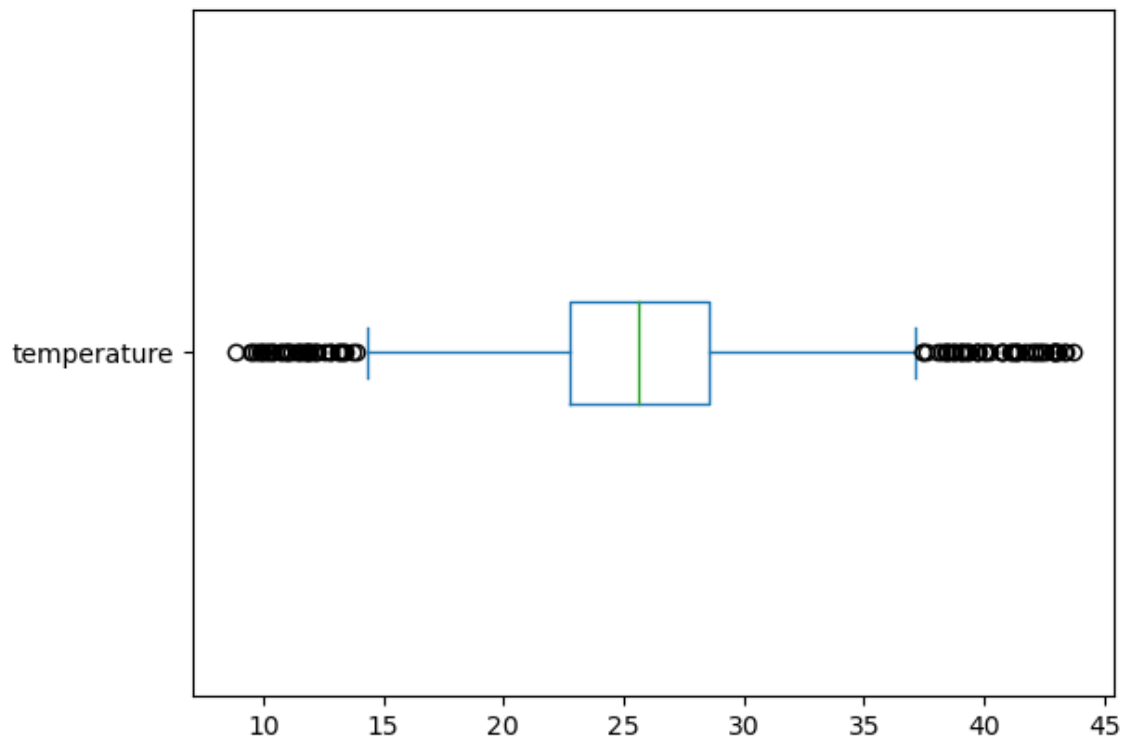


In [28]:

```
crop.temperature.plot(kind="box",vert=False)
```

Out[28]:

<Axes: >

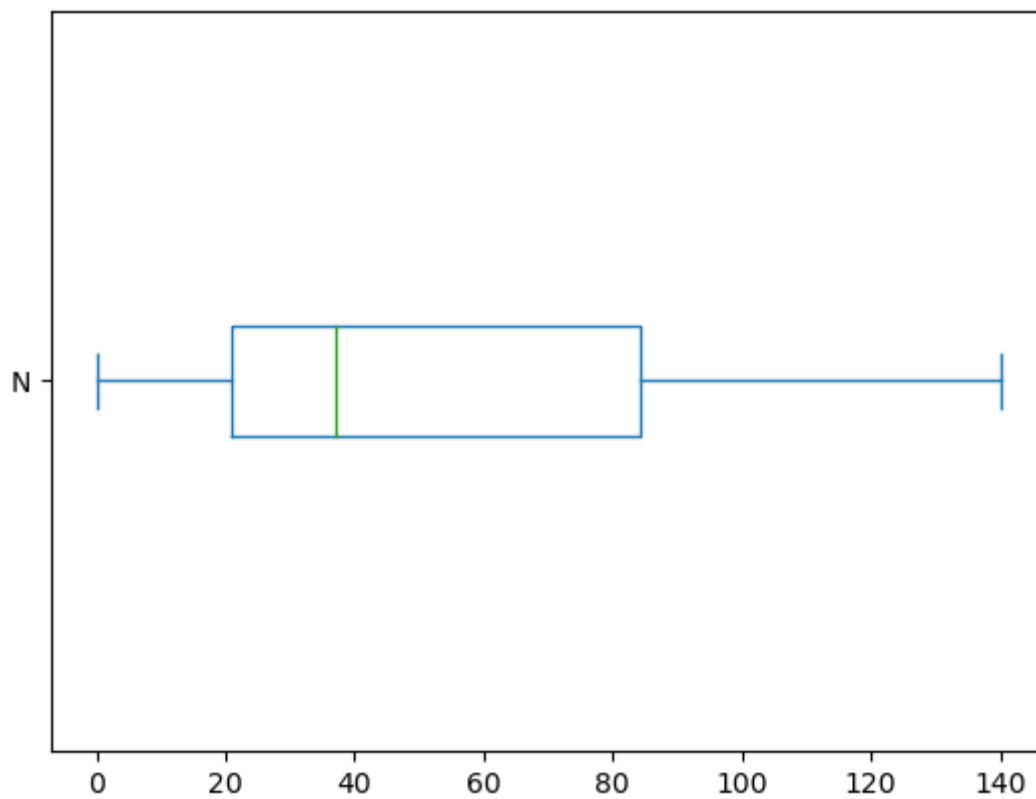


In [29]:

```
crop.N.plot(kind="box",vert=False)
```

Out[29]:

<Axes: >

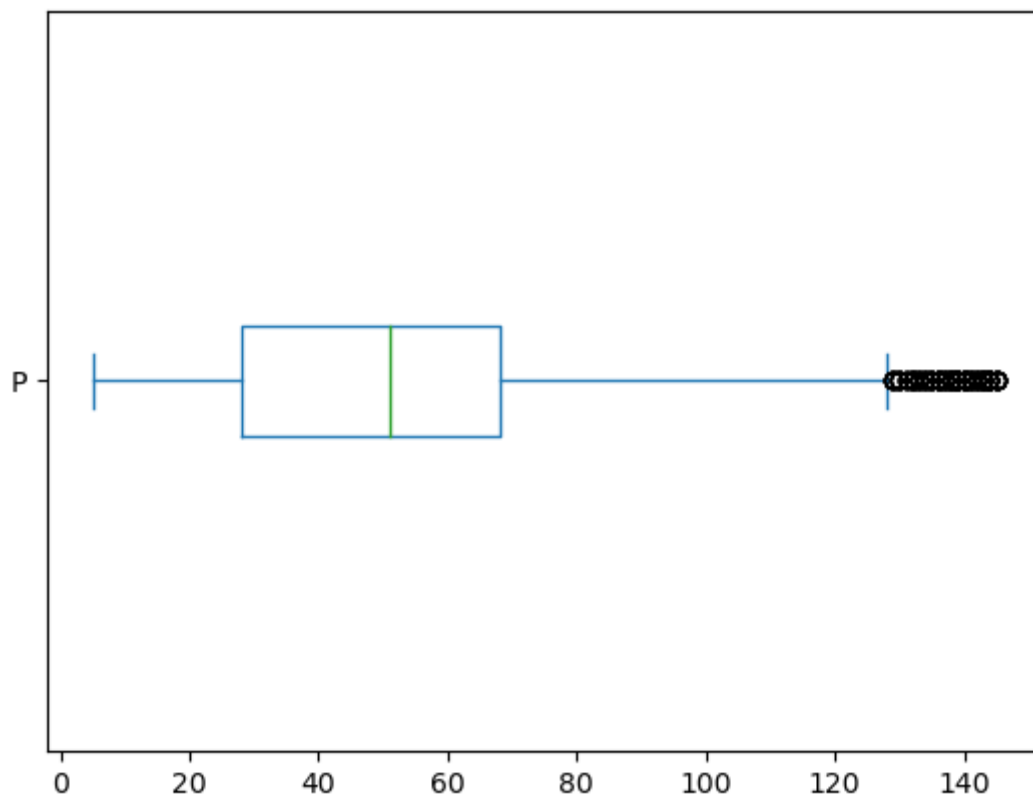


In [30]:

```
crop.P.plot(kind="box",vert=False)
```

Out[30]:

<Axes: >

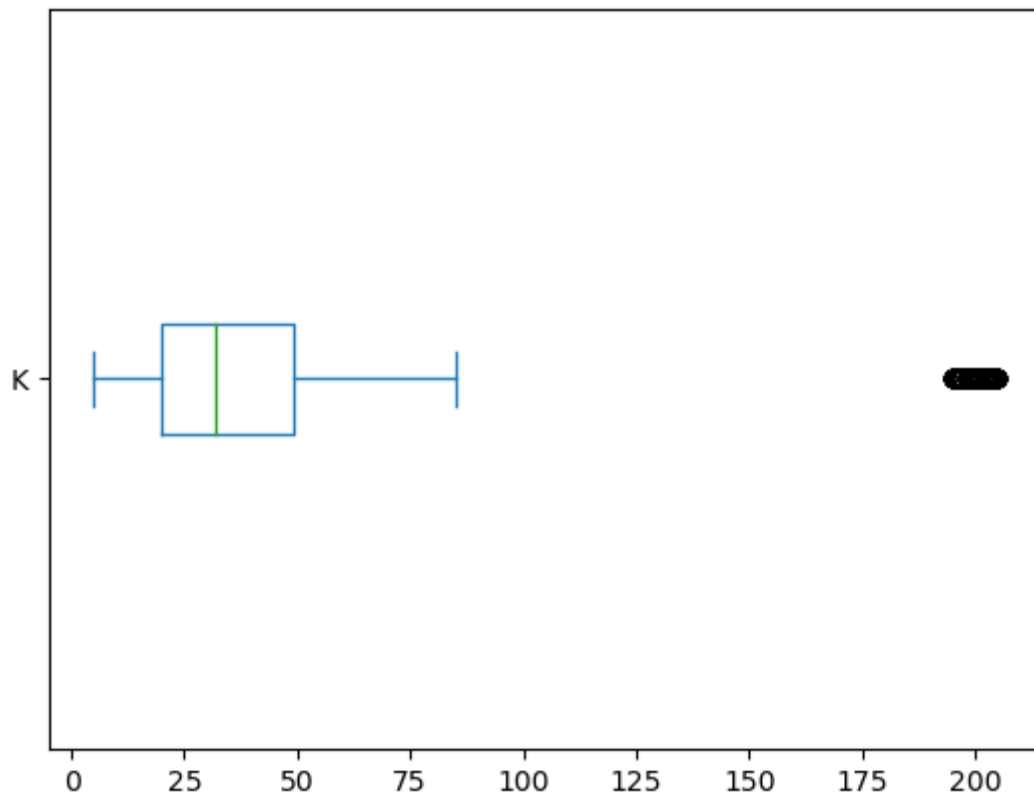


In [31]:

```
crop.K.plot(kind="box",vert=False)
```

Out[31]:

<Axes: >



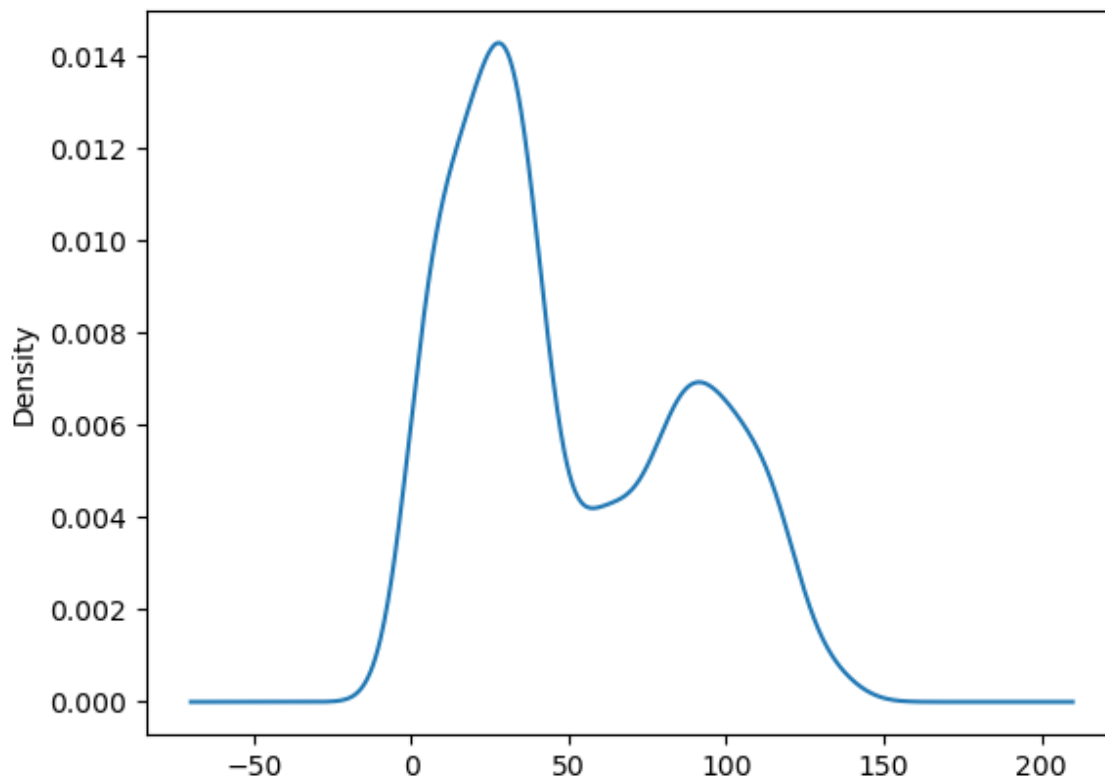


In [32]:

```
crop.N.plot(kind="density")
```

Out[32]:

<Axes: ylabel='Density'>

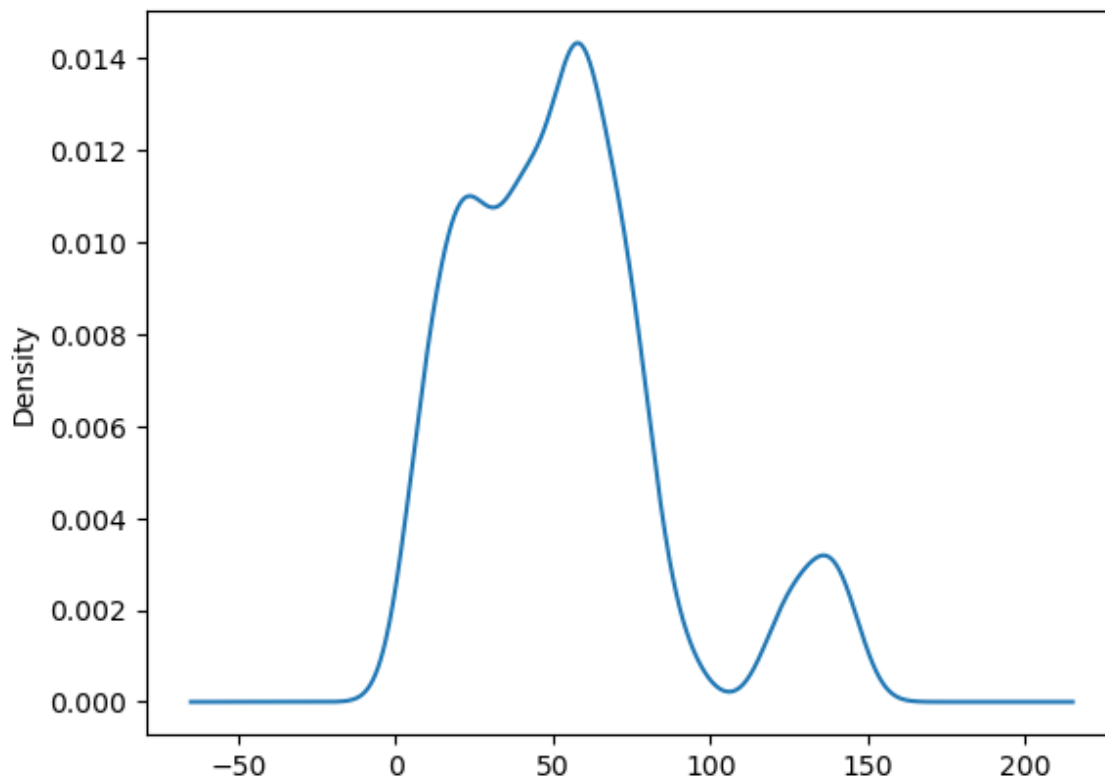


In [33]:

```
crop.P.plot(kind="density")
```

Out[33]:

<Axes: ylabel='Density'>

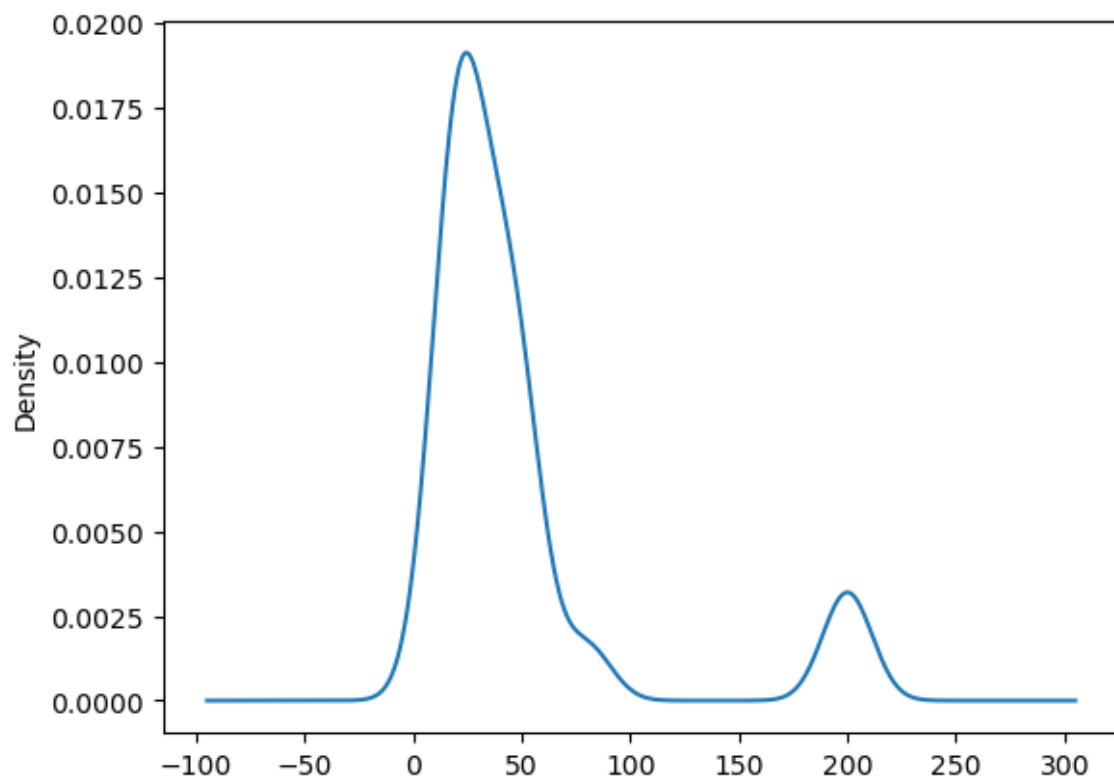


In [34]:

```
crop.K.plot(kind="density")
```

Out[34]:

<Axes: ylabel='Density'>

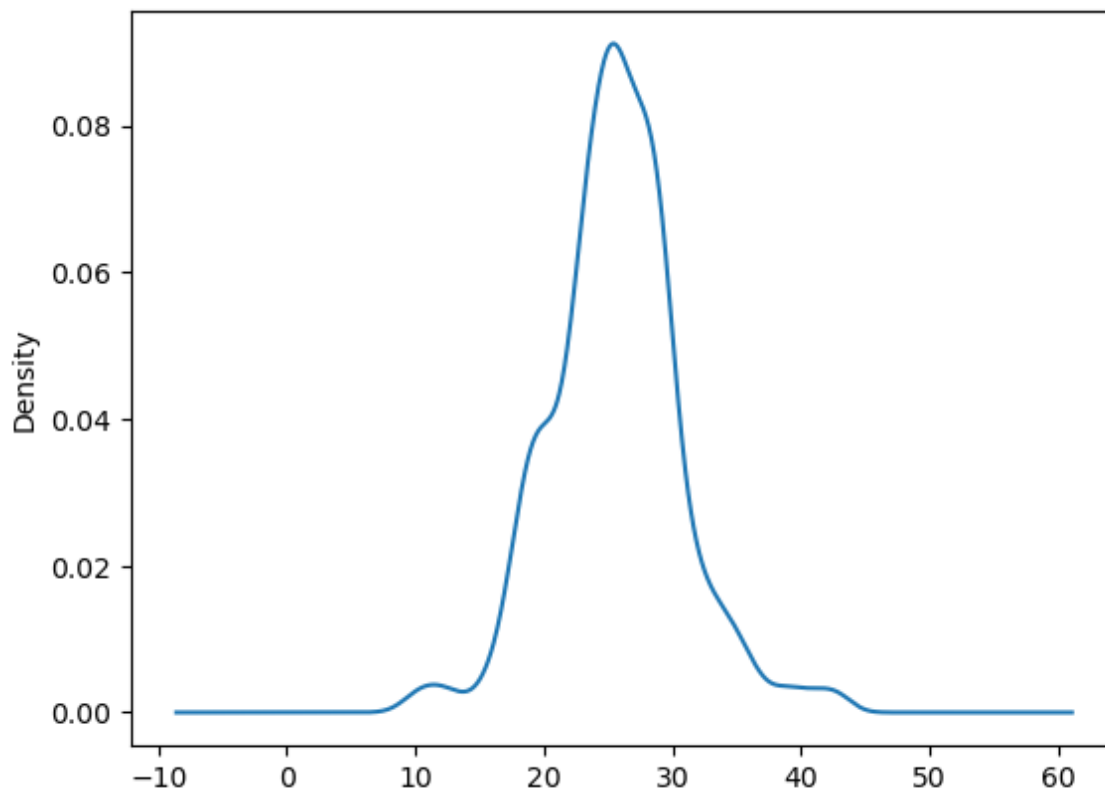


In [35]:

```
crop.temperature.plot(kind="density")
```

Out[35]:

<Axes: ylabel='Density'>

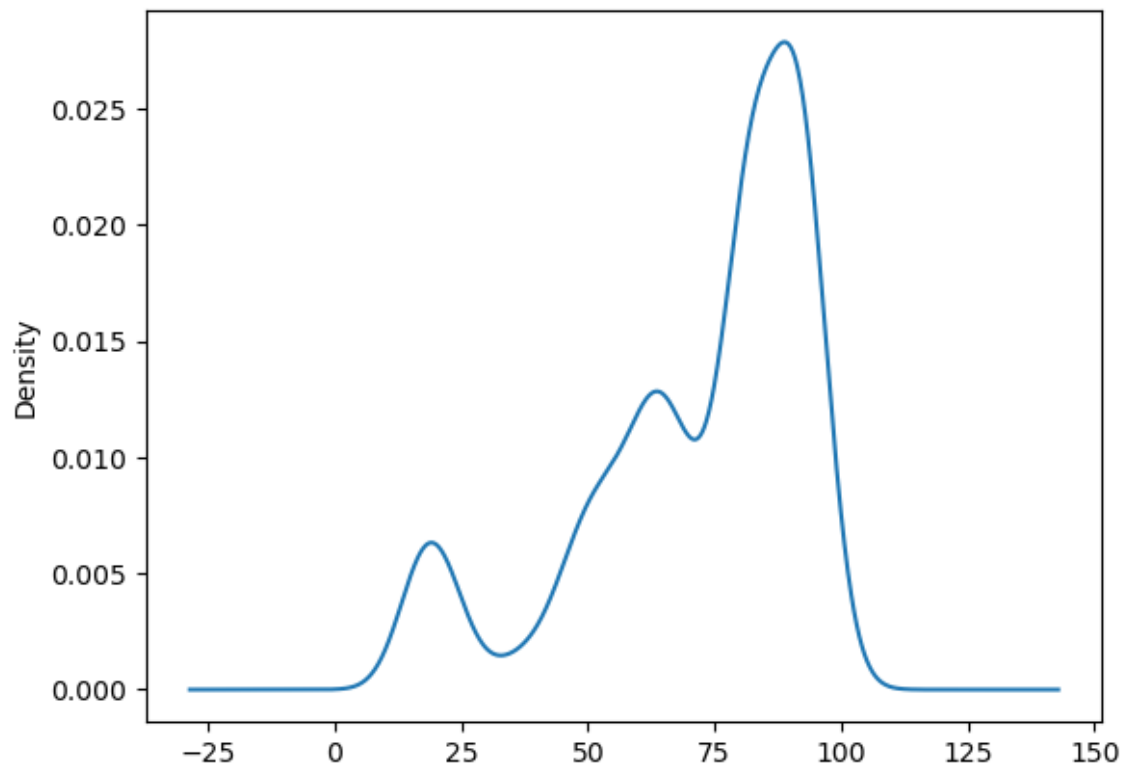


In [36]:

```
crop.humidity.plot(kind="density")
```

Out[36]:

<Axes: ylabel='Density'>

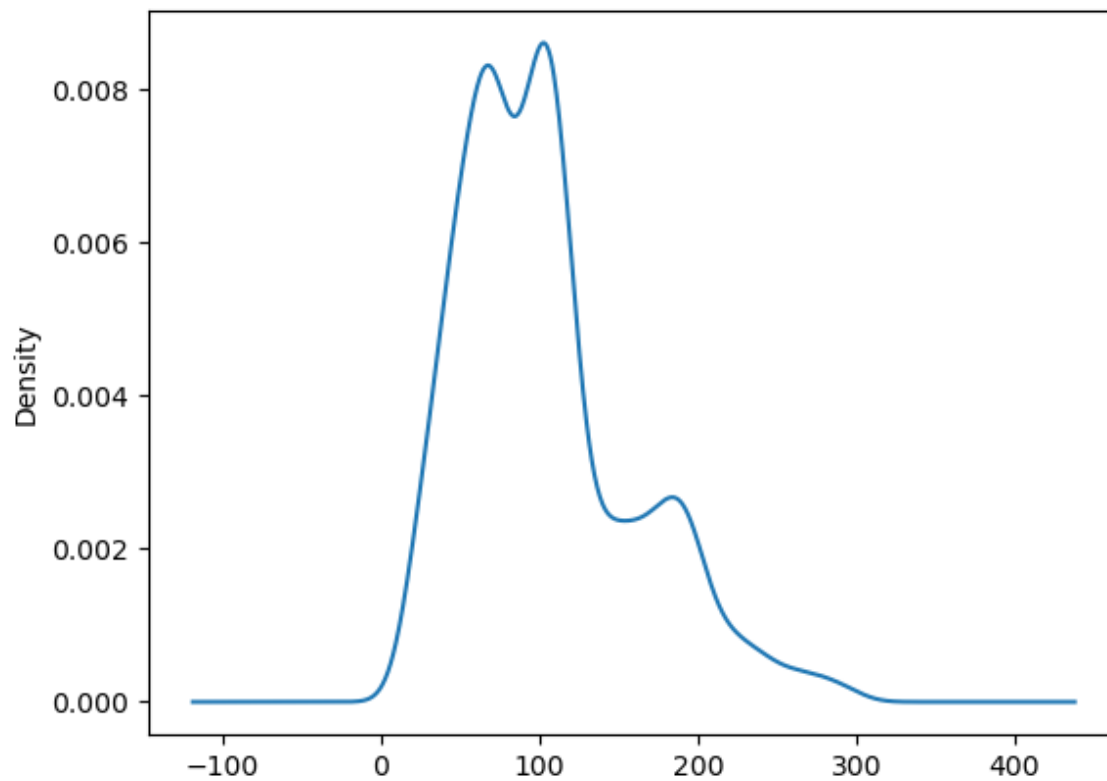


In [37]:

```
crop.rainfall.plot(kind="density")
```

Out[37]:

<Axes: ylabel='Density'>

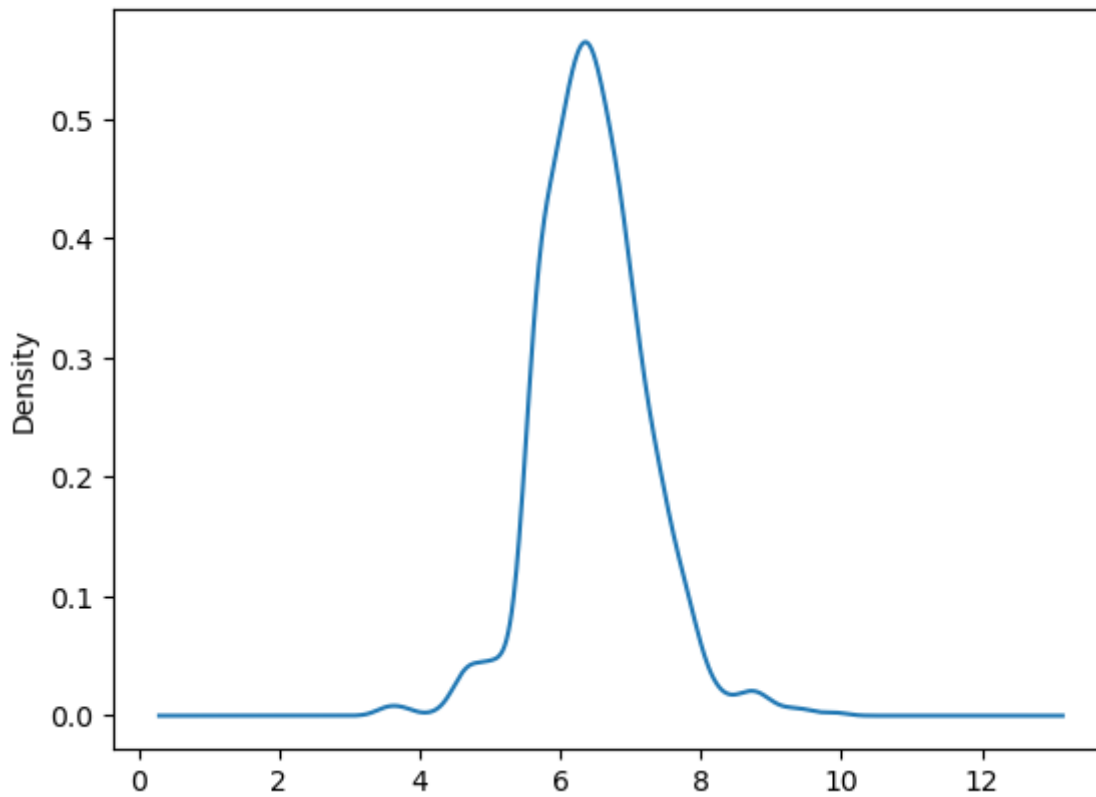


In [38]:

```
crop.ph.plot(kind="density")
```

Out[38]:

<Axes: ylabel='Density'>



In [39]:

```
import seaborn as sns
```

In [40]:

```
f=plt.figure(figsize=(15,5))
ax=f.add_subplot(121)
sns.distplot(crop['N'],color='red',ax=ax)

ax=f.add_subplot(122)
sns.distplot(crop['P'],color='green',ax=ax)
plt.tight_layout()
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\890166929.py:3: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['N'],color='red',ax=ax)
```

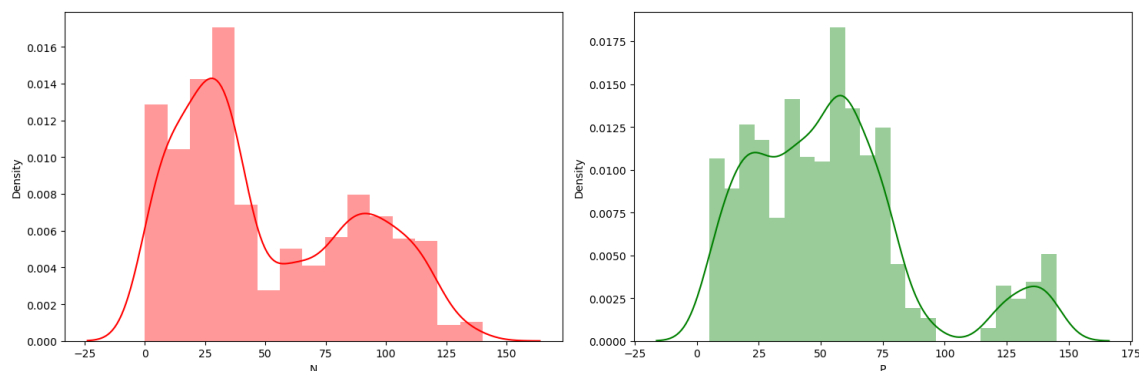
C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\890166929.py:6: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['P'],color='green',ax=ax)
```





In [41]:

```
f=plt.figure(figsize=(15,5))
ax=f.add_subplot(121)
sns.distplot(crop['K'],color='red',ax=ax)

ax=f.add_subplot(122)
sns.distplot(crop['temperature'],color='green',ax=ax)
plt.tight_layout()
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\2857048422.py:3: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['K'],color='red',ax=ax)
```

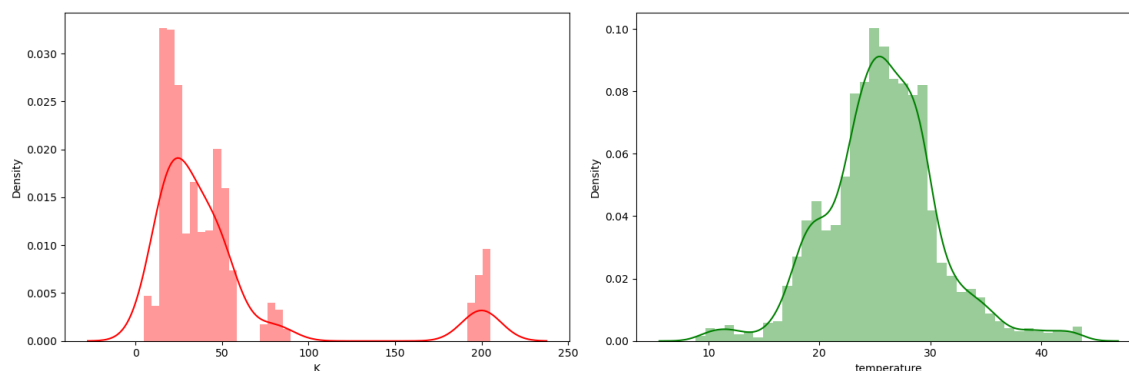
C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\2857048422.py:6: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['temperature'],color='green',ax=ax)
```



In [42]:

```
f=plt.figure(figsize=(15,5))
ax=f.add_subplot(121)
sns.distplot(crop['humidity'],color='red',ax=ax)

ax=f.add_subplot(122)
sns.distplot(crop['ph'],color='green',ax=ax)
plt.tight_layout()
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\392488001.py:3: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['humidity'],color='red',ax=ax)
```

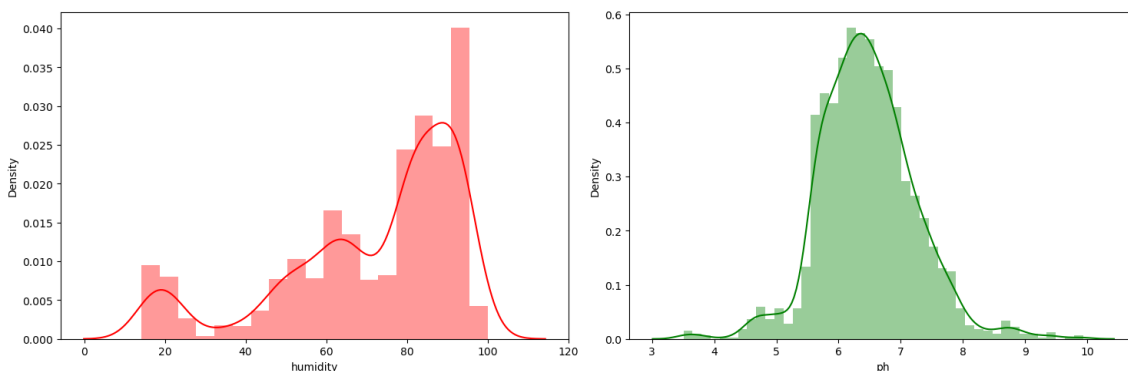
C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\392488001.py:6: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['ph'],color='green',ax=ax)
```



In [43]:

```
sns.distplot(crop['rainfall'],color ='red')
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\2804088430.py:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

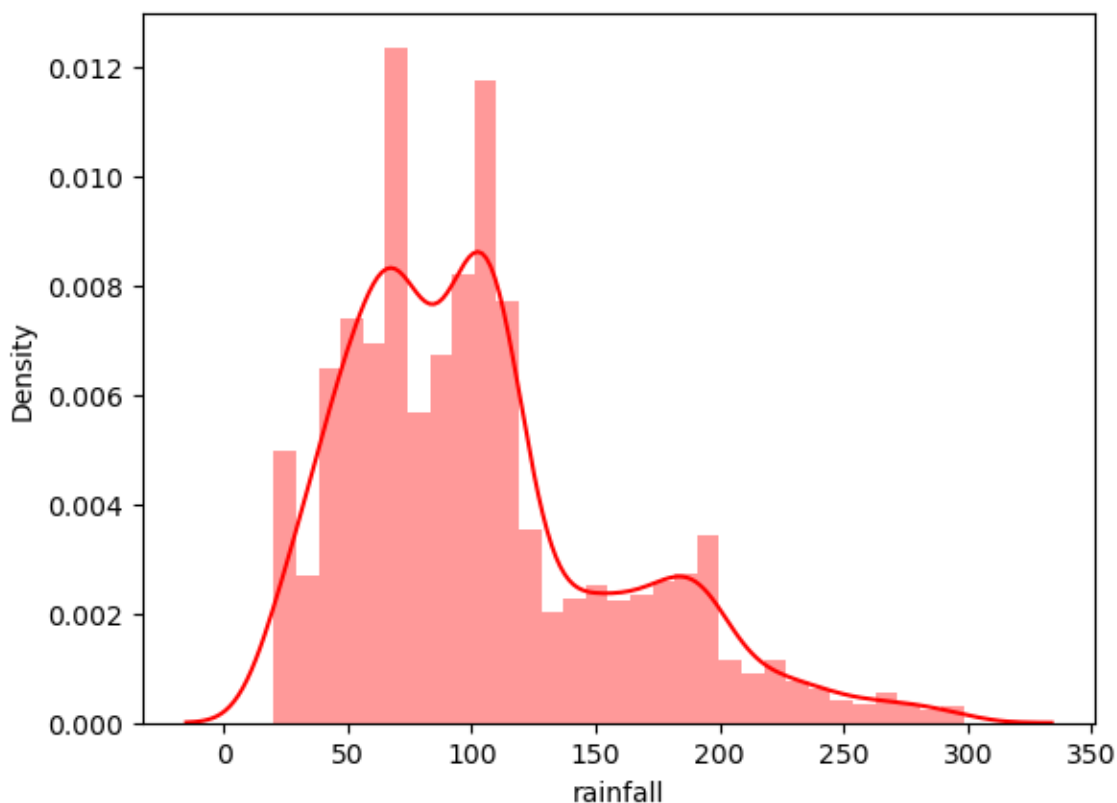
Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(crop['rainfall'],color ='red')
```

Out[43]:

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

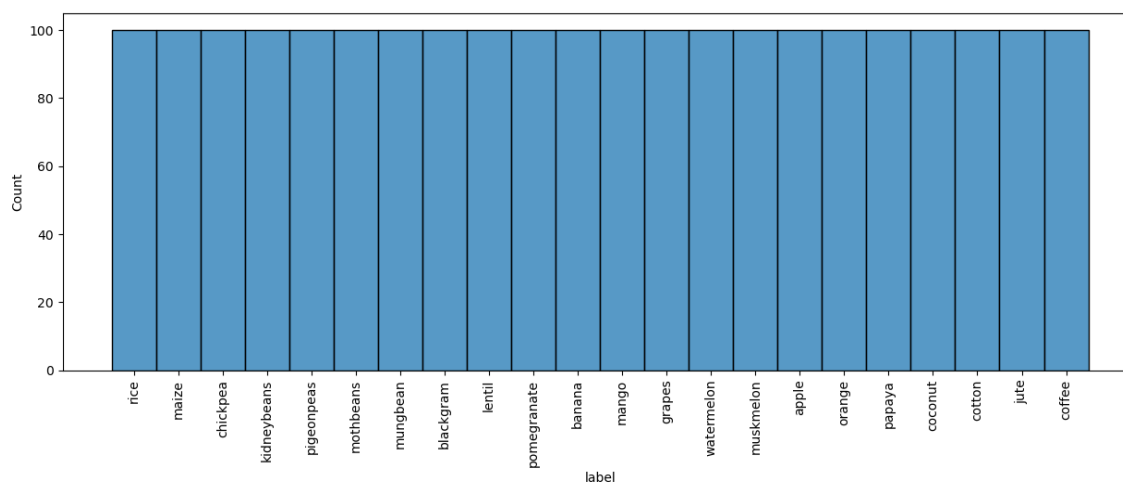


In [44]:

```
f=plt.figure(figsize=(15,5))
sns.histplot(crop['label'],palette='Spectral')
plt.xticks(rotation=90)
plt.show()
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\805242828.py:2: UserWarning: Ignoring `palette` because no `hue` variable has been assigned.

```
sns.histplot(crop['label'],palette='Spectral')
```



In [45]:

```
crop.label.value_counts()
```

Out[45]:

```
rice          100
maize         100
jute          100
cotton        100
coconut       100
papaya        100
orange        100
apple         100
muskmelon     100
watermelon    100
grapes        100
mango         100
banana        100
pomegranate   100
lentil        100
blackgram     100
mungbean      100
mothbeans     100
pigeonpeas    100
kidneybeans   100
chickpea      100
coffee       100
Name: label, dtype: int64
```

## GROUPBY() AND VISUALIZATION

In [46]:

```
crop.temperature.groupby(crop.label).mean()
```

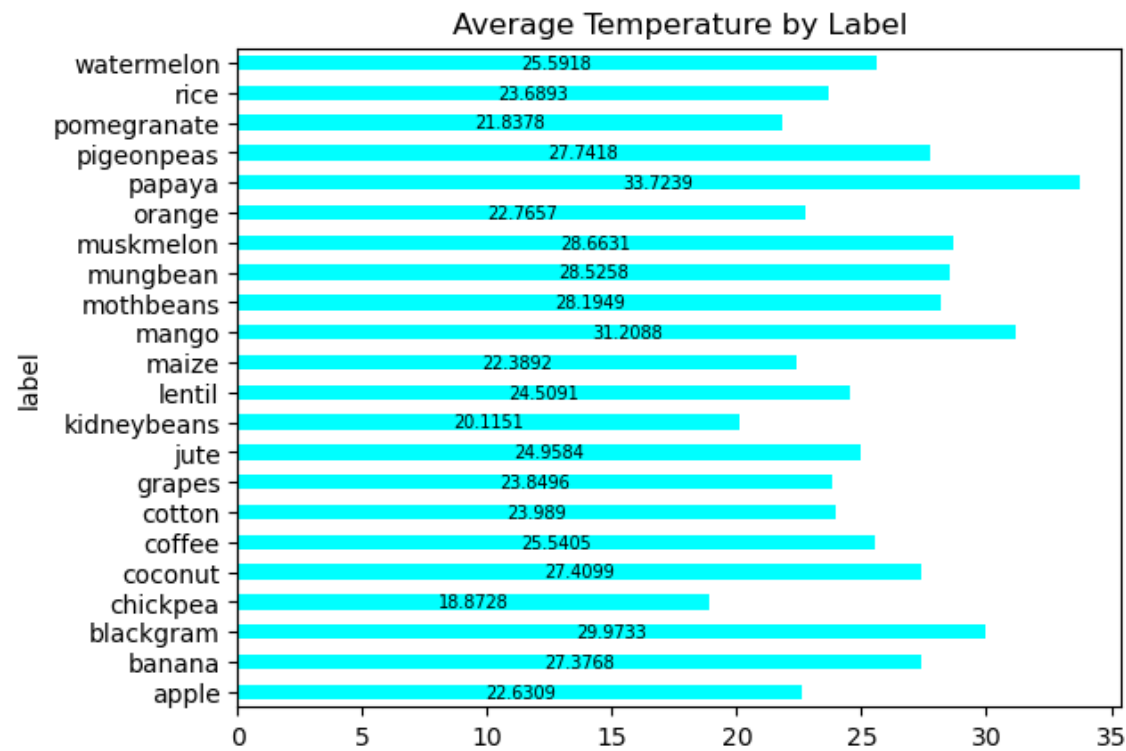
Out[46]:

label	
apple	22.630942
banana	27.376798
blackgram	29.973340
chickpea	18.872847
coconut	27.409892
coffee	25.540477
cotton	23.988958
grapes	23.849575
jute	24.958376
kidneybeans	20.115085
lentil	24.509052
maize	22.389204
mango	31.208770
mothbeans	28.194920
mungbean	28.525775
muskmelon	28.663066
orange	22.765725
papaya	33.723859
pigeonpeas	27.741762
pomegranate	21.837842
rice	23.689332
watermelon	25.591767

Name: temperature, dtype: float64

In [47]:

```
ax=crop.temperature.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Average Temperature by Label")
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [48]:

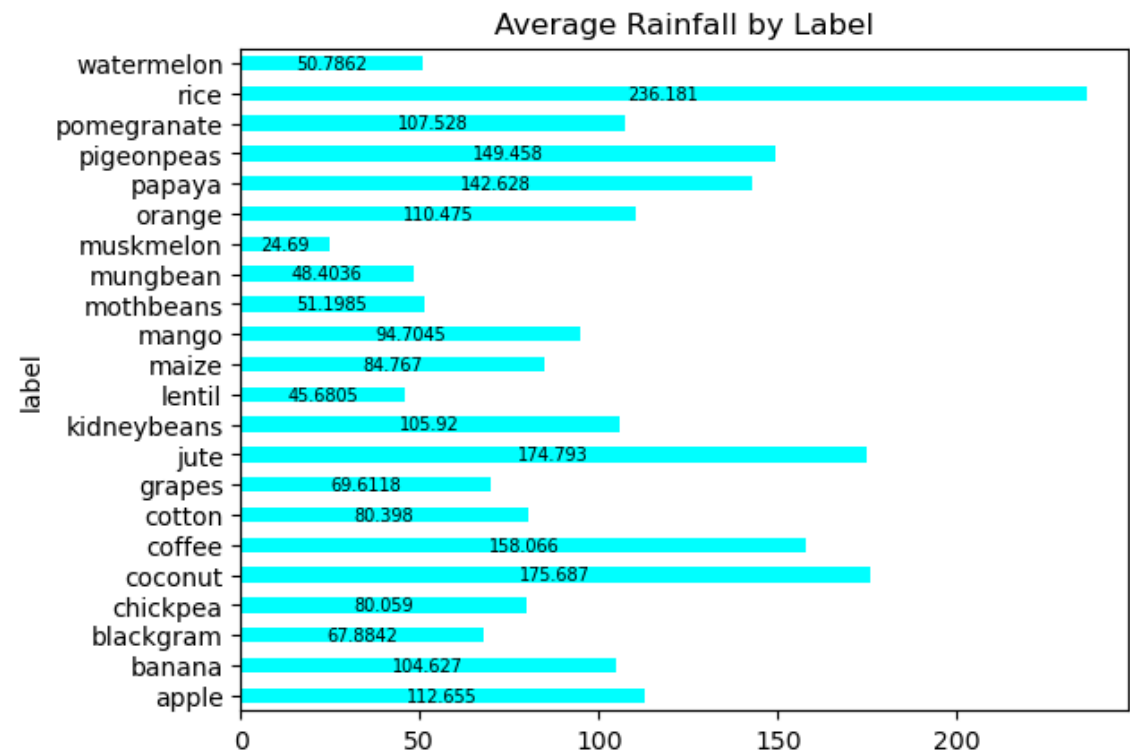
```
crop.rainfall.groupby(crop.label).mean()
```

Out[48]:

label	
apple	112.654779
banana	104.626980
blackgram	67.884151
chickpea	80.058977
coconut	175.686646
coffee	158.066295
cotton	80.398043
grapes	69.611829
jute	174.792798
kidneybeans	105.919778
lentil	45.680454
maize	84.766988
mango	94.704515
mothbeans	51.198487
mungbean	48.403601
muskmelon	24.689952
orange	110.474969
papaya	142.627839
pigeonpeas	149.457564
pomegranate	107.528442
rice	236.181114
watermelon	50.786219
Name: rainfall, dtype: float64	

In [49]:

```
ax=crop.rainfall.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Avera
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [50]:

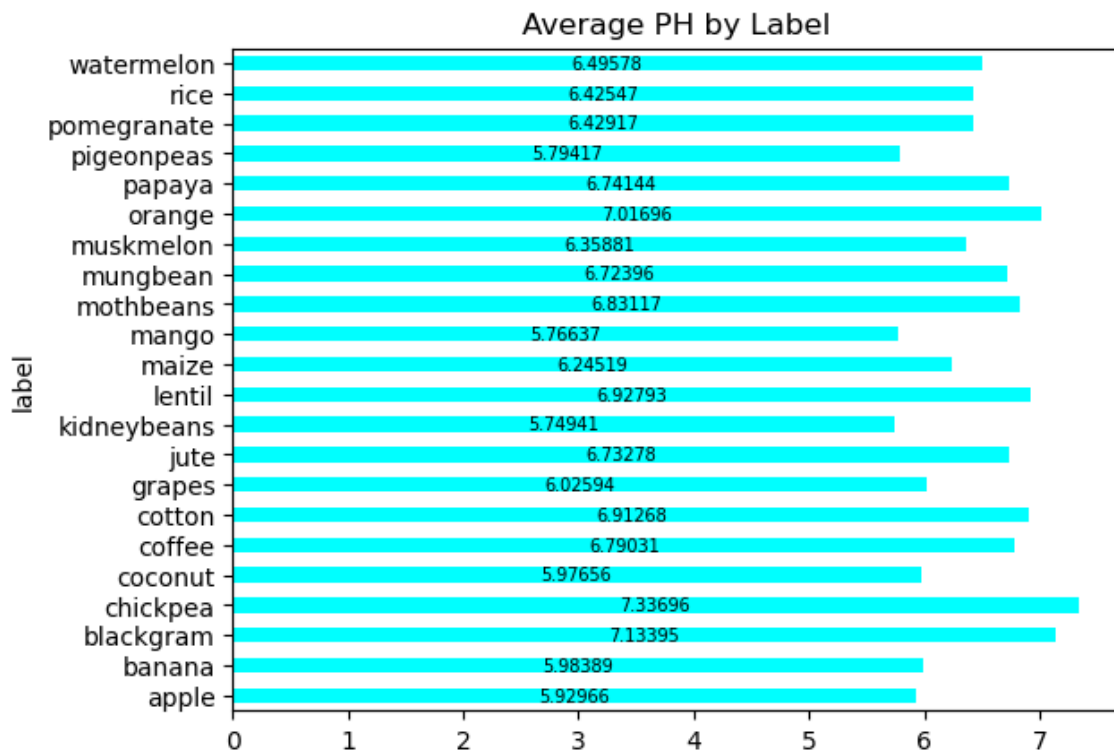
```
crop.ph.groupby(crop.label).mean()
```

Out[50]:

label	
apple	5.929663
banana	5.983893
blackgram	7.133952
chickpea	7.336957
coconut	5.976562
coffee	6.790308
cotton	6.912675
grapes	6.025937
jute	6.732778
kidneybeans	5.749411
lentil	6.927932
maize	6.245190
mango	5.766373
mothbeans	6.831174
mungbean	6.723957
muskmelon	6.358805
orange	7.016957
papaya	6.741442
pigeonpeas	5.794175
pomegranate	6.429172
rice	6.425471
watermelon	6.495778
Name: ph, dtype: float64	

In [51]:

```
ax=crop.ph.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Average PH
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [52]:

```
crop.humidity.groupby(crop.label).mean()
```

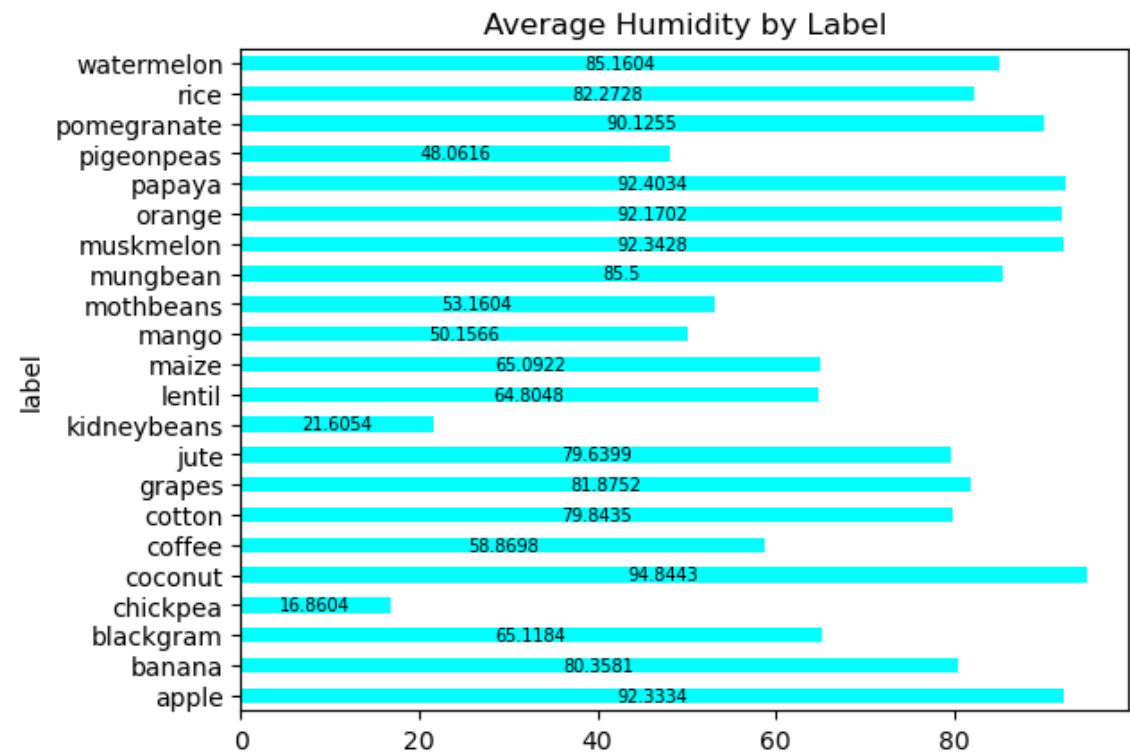
Out[52]:

```
label
apple      92.333383
banana     80.358123
blackgram  65.118426
chickpea   16.860439
coconut    94.844272
coffee     58.869846
cotton     79.843474
grapes     81.875228
jute       79.639864
kidneybeans 21.605357
lentil     64.804785
maize      65.092249
mango      50.156573
mothbeans  53.160418
mungbean   85.499975
muskmelon  92.342802
orange     92.170209
papaya     92.403388
pigeonpeas 48.061633
pomegranate 90.125504
rice       82.272822
watermelon 85.160375
Name: humidity, dtype: float64
```



In [53]:

```
ax=crop.humidity.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Avera
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [54]:

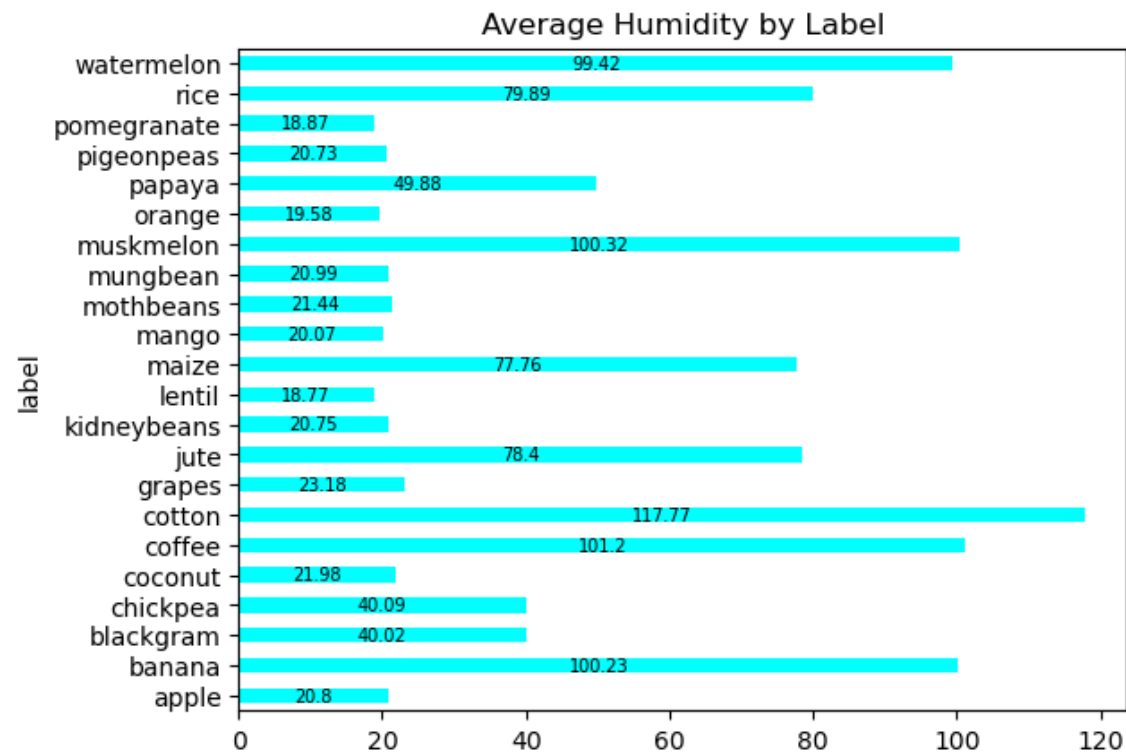
```
crop.N.groupby(crop.label).mean()
```

Out[54]:

```
label
apple      20.80
banana    100.23
blackgram   40.02
chickpea    40.09
coconut     21.98
coffee    101.20
cotton    117.77
grapes     23.18
jute       78.40
kidneybeans 20.75
lentil     18.77
maize     77.76
mango     20.07
mothbeans  21.44
mungbean   20.99
muskmelon 100.32
orange     19.58
papaya     49.88
pigeonpeas 20.73
pomegranate 18.87
rice       79.89
watermelon 99.42
Name: N, dtype: float64
```

In [55]:

```
ax=crop.N.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Average Humi
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [56]:

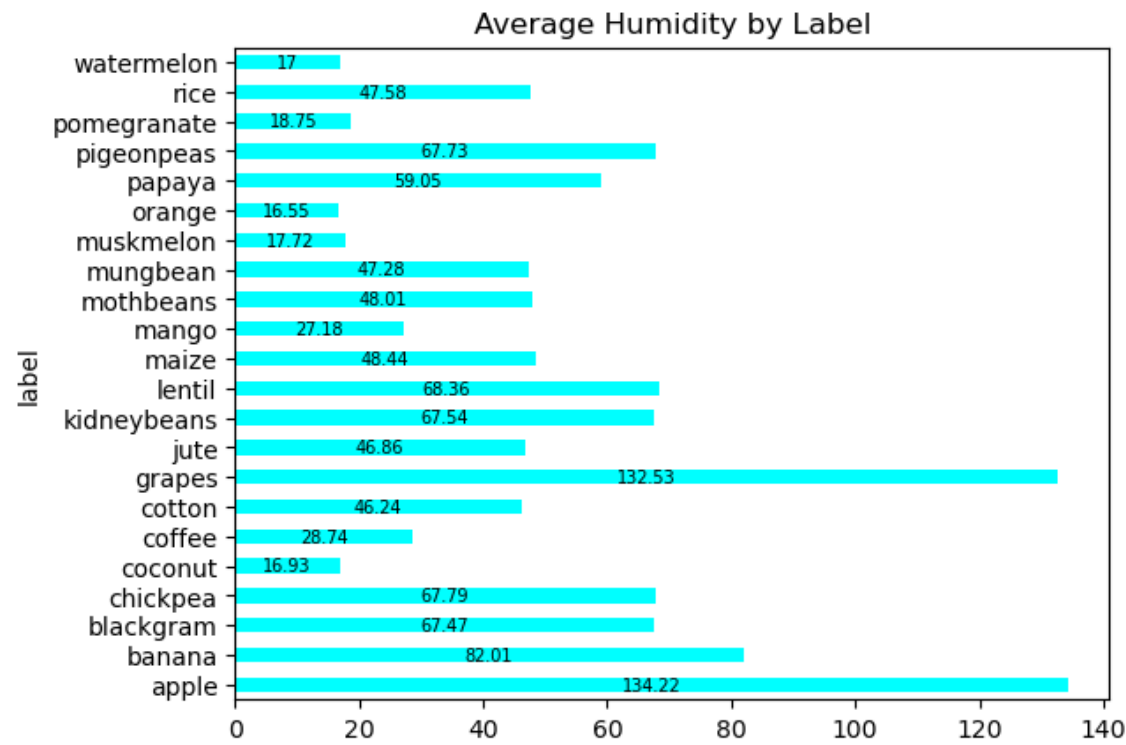
```
crop.P.groupby(crop.label).mean()
```

Out[56]:

label	
apple	134.22
banana	82.01
blackgram	67.47
chickpea	67.79
coconut	16.93
coffee	28.74
cotton	46.24
grapes	132.53
jute	46.86
kidneybeans	67.54
lentil	68.36
maize	48.44
mango	27.18
mothbeans	48.01
mungbean	47.28
muskmelon	17.72
orange	16.55
papaya	59.05
pigeonpeas	67.73
pomegranate	18.75
rice	47.58
watermelon	17.00
Name: P, dtype: float64	

In [57]:

```
ax=crop.P.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Average Humi
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



In [58]:

```
crop.K.groupby(crop.label).mean()
```

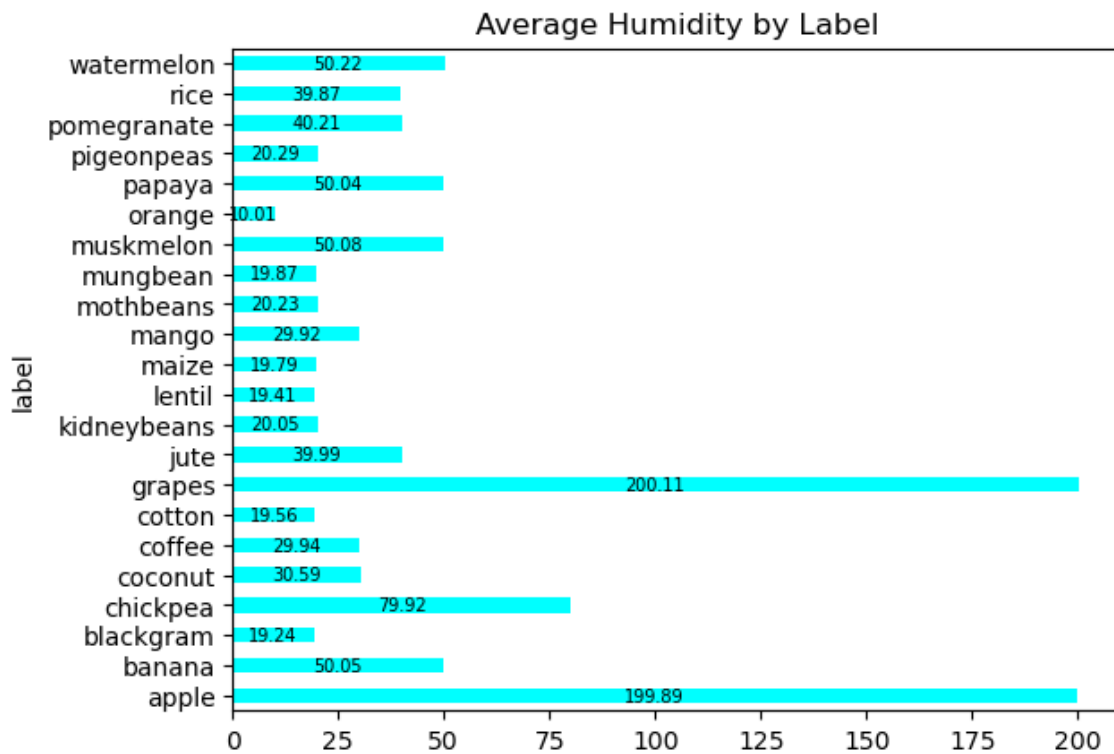
Out[58]:

label	
apple	199.89
banana	50.05
blackgram	19.24
chickpea	79.92
coconut	30.59
coffee	29.94
cotton	19.56
grapes	200.11
jute	39.99
kidneybeans	20.05
lentil	19.41
maize	19.79
mango	29.92
mothbeans	20.23
mungbean	19.87
muskmelon	50.08
orange	10.01
papaya	50.04
pigeonpeas	20.29
pomegranate	40.21
rice	39.87
watermelon	50.22

Name: K, dtype: float64

In [59]:

```
ax=crop.K.groupby(crop.label).mean().plot(kind="barh",color=["cyan"],title="Average Humi
for i in ax.containers:
    ax.bar_label(i,fontsize=7,label_type="center")
```



## SPLITTING TEST DATA AND TRAIN DATA

In [60]:

```
from sklearn.model_selection import train_test_split
train_set,test_set = train_test_split(crop,random_state=0)
```

In [61]:

```
X=train_set.iloc[:,0:7]
print(X.head(5))
print()
```

	N	P	K	temperature	humidity	ph	rainfall
1975	134	56	18	23.808346	83.919026	6.691268	70.973583
1287	29	122	196	41.948657	81.155952	5.638328	73.068630
760	25	68	19	29.399827	64.255107	7.108450	67.476773
267	41	69	82	20.023815	16.632945	6.715587	68.978065
2007	70	43	40	24.355641	88.803910	6.176860	169.116803

In [62]:

```
y=train_set.iloc[:,7]
print(y.head(5))
print()
```

```
1975      cotton
1287      grapes
760      blackgram
267      chickpea
2007       jute
Name: label, dtype: object
```

In [63]:

```
X=test_set.iloc[:,0:7]
print(X.shape)
y=test_set.iloc[:,7]
y= np.array([y]).reshape(550,1)
```

```
(550, 7)
```

In [64]:

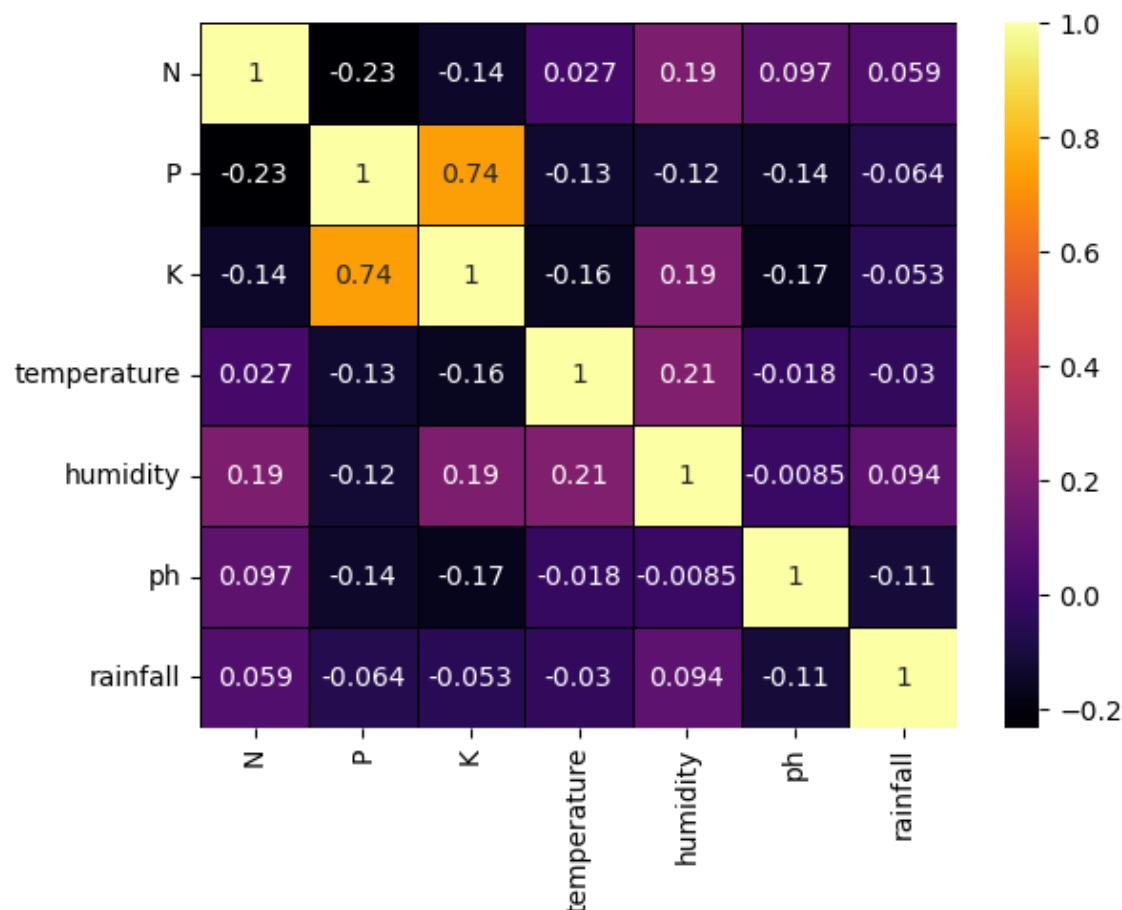
```
sns.heatmap(data=crop.corr(),annot=True,cmap="inferno",linecolor="black",linewidths=0.5)
```

C:\Users\deepe\AppData\Local\Temp\ipykernel\_13088\3497978509.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.

```
sns.heatmap(data=crop.corr(),annot=True,cmap="inferno",linecolor="black",linewidths=0.5)
```

Out[64]:

&lt;Axes: &gt;



## MODEL BUILDINGS

### K Neighbours Classifier

In [65]:

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import cross_val_score
```

In [66]:

```
knn=KNeighborsClassifier(n_neighbors=10)
```

In [67]:

```
knn.fit(X,y)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\neighbors\\_classification.py:215: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().

```
return self._fit(X, y)
```

Out[67]:

```

KNeighborsClassifier
KNeighborsClassifier(n_neighbors=10)

```

In [68]:

```
from sklearn.metrics import classification_report
```

In [69]:

```

knn_predicted = knn.predict(X)
print(classification_report(y,knn_predicted))

```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	24
banana	1.00	1.00	1.00	24
blackgram	0.96	1.00	0.98	26
chickpea	1.00	1.00	1.00	28
coconut	1.00	1.00	1.00	19
coffee	1.00	1.00	1.00	24
cotton	0.88	1.00	0.93	21
grapes	1.00	1.00	1.00	24
jute	0.82	1.00	0.90	28
kidneybeans	0.85	1.00	0.92	23
lentil	0.94	0.94	0.94	17
maize	1.00	0.86	0.93	22
mango	1.00	1.00	1.00	24
mothbeans	0.96	0.93	0.95	29
mungbean	1.00	1.00	1.00	27
muskmelon	1.00	1.00	1.00	27
orange	1.00	1.00	1.00	27
papaya	1.00	0.89	0.94	28
pigeonpeas	1.00	0.85	0.92	27
pomegranate	1.00	1.00	1.00	28
rice	0.96	0.86	0.91	29
watermelon	1.00	1.00	1.00	24
accuracy			0.97	550
macro avg	0.97	0.97	0.97	550
weighted avg	0.97	0.97	0.97	550

In [71]:

```
np.mean([0.95454545, 0.94545455, 0.97272727, 0.94545455, 0.96363636])
```

Out[71]:

```
0.9563636360000001
```

## Desicion Tree Classifier

In [72]:

```
from sklearn.tree import DecisionTreeClassifier
```

In [73]:

```
dtc=DecisionTreeClassifier()
```

In [74]:

```
dtcmodel=dtc.fit(X,y)
```

In [75]:

```
dtcmodel.score(X,y)
```

Out[75]:

```
1.0
```

## LINEAR SUPPORT VECTOR CLASSIFICATION (SVC)

In [76]:

```
from sklearn.svm import LinearSVC
```

In [77]:

```
lsvc=LinearSVC()
```

In [78]:

```
lsvc=LinearSVC(C=5,random_state=67).fit(X,y)
```

```
C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples, ), for example using ravel().
```

```
y = column_or_1d(y, warn=True)
```

```
C:\Users\deepe\anaconda3\lib\site-packages\sklearn\svm\_base.py:1244: ConvergenceWarning: Liblinear failed to converge, increase the number of iterations.
```

```
warnings.warn(
```



In [79]:

```
li_predicted=lsvc.predict(X)
```

In [80]:

```
lsvc.predict(X[200:201])
```

Out[80]:

```
array(['pigeonpeas'], dtype=object)
```

In [81]:

```
print(y[200])
```

```
['coconut']
```

In [82]:

```
print(classification_report(y,li_predicted))
```

	precision	recall	f1-score	support
apple	0.00	0.00	0.00	24
banana	1.00	0.92	0.96	24
blackgram	1.00	0.19	0.32	26
chickpea	1.00	1.00	1.00	28
coconut	1.00	0.89	0.94	19
coffee	1.00	1.00	1.00	24
cotton	1.00	0.71	0.83	21
grapes	1.00	0.54	0.70	24
jute	0.00	0.00	0.00	28
kidneybeans	0.96	1.00	0.98	23
lentil	1.00	0.82	0.90	17
maize	0.42	1.00	0.59	22
mango	1.00	0.96	0.98	24
mothbeans	0.77	0.79	0.78	29
mungbean	0.96	1.00	0.98	27
muskmelon	1.00	1.00	1.00	27
orange	1.00	1.00	1.00	27
papaya	0.50	0.82	0.62	28
pigeonpeas	0.74	0.96	0.84	27
pomegranate	1.00	0.71	0.83	28
rice	0.19	0.28	0.22	29
watermelon	1.00	0.75	0.86	24
accuracy				0.74
macro avg				0.80
weighted avg				0.79

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\metrics\\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\metrics\\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\metrics\\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

```
_warn_prf(average, modifier, msg_start, len(result))
```

In [83]:

```
from sklearn.model_selection import cross_val_score
```

In [84]:

```
from sklearn.svm import SVC
```

In [85]:

```
svc=SVC().fit(X,y)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

In [86]:

```
svc.predict(X[233:243])
```

Out[86]:

```
array(['kidneybeans', 'pomegranate', 'coconut', 'mango', 'mungbean',  
      'cotton', 'rice', 'chickpea', 'mungbean', 'apple'], dtype=object)
```

In [87]:

```
print(y[233:243])
```

```
[['kidneybeans']  
 ['pomegranate']  
 ['coconut']  
 ['mango']  
 ['mungbean']  
 ['cotton']  
 ['rice']  
 ['chickpea']  
 ['mungbean']  
 ['apple']]
```

In [88]:

```
from sklearn.metrics import classification_report
```

In [89]:

```
svm_predicted=svc.predict(X)
```

In [90]:

```
print(classification_report(y,svm_predicted))
```

	precision	recall	f1-score	support
apple	1.00	1.00	1.00	24
banana	1.00	1.00	1.00	24
blackgram	0.93	0.96	0.94	26
chickpea	1.00	1.00	1.00	28
coconut	1.00	1.00	1.00	19
coffee	1.00	1.00	1.00	24
cotton	0.91	1.00	0.95	21
grapes	1.00	1.00	1.00	24
jute	0.90	1.00	0.95	28
kidneybeans	0.85	1.00	0.92	23
lentil	1.00	0.76	0.87	17
maize	1.00	0.91	0.95	22
mango	1.00	1.00	1.00	24
mothbeans	0.91	1.00	0.95	29
mungbean	1.00	1.00	1.00	27
muskmelon	1.00	1.00	1.00	27
orange	1.00	1.00	1.00	27
papaya	1.00	0.96	0.98	28
pigeonpeas	1.00	0.85	0.92	27
pomegranate	1.00	1.00	1.00	28
rice	0.96	0.90	0.93	29
watermelon	1.00	1.00	1.00	24
accuracy			0.97	550
macro avg	0.98	0.97	0.97	550
weighted avg	0.98	0.97	0.97	550

In [91]:

```
cross_val_score(svc,X,y)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

C:\Users\deepe\anaconda3\lib\site-packages\sklearn\utils\validation.py:114  
3: DataConversionWarning: A column-vector y was passed when a 1d array was  
expected. Please change the shape of y to (n\_samples, ), for example using  
ravel().

```
y = column_or_1d(y, warn=True)
```

Out[91]:

```
array([0.95454545, 0.94545455, 0.97272727, 0.94545455, 0.96363636])
```

In [92]:

```
np.mean([0.95454545, 0.94545455, 0.97272727, 0.94545455, 0.96363636])
```

Out[92]:

```
0.9563636360000001
```