In [1]:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from IPython import get_ipython
import warnings
warnings.filterwarnings("ignore")
```

In [2]:

```
data = pd.read_csv('crop.csv')
```

In [3]:

```
data.head()
```

Out[3]:

	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

In [4]:

```
data.tail()
```

Out[4]:

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

In [5]:

```
1 data.shape
```

Out[5]:

```
(2200, 8)
```

In [6]:

```
1 data.columns
```

Out[6]:

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

In [7]:

```
1 data.duplicated().sum()
```

Out[7]:

```
0
```

In [8]:

```
1 data.isnull().sum()
```

Out[8]:

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

In [9]:

```
1 data.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 [10]:

```
1 data.describe()
```

Out[10]:

	N	P	K	temperature	humidity	ph	ra
count	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.000000	2200.00
mean	50.551818	53.362727	48.149091	25.616244	71.481779	6.469480	103.46
std	36.917334	32.985883	50.647931	5.063749	22.263812	0.773938	54.95
min	0.000000	5.000000	5.000000	8.825675	14.258040	3.504752	20.21
25%	21.000000	28.000000	20.000000	22.769375	60.261953	5.971693	64.55
50%	37.000000	51.000000	32.000000	25.598693	80.473146	6.425045	94.86
75%	84.250000	68.000000	49.000000	28.561654	89.948771	6.923643	124.26
max	140.000000	145.000000	205.000000	43.675493	99.981876	9.935091	298.56

In [11]:

```
1 data.nunique()
```

Out[11]:

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

In [12]:

```
1 data['label'].unique()
```

Out[12]:

```
array(['rice', 'maize', 'chickpea', 'kidneybeans', 'pigeonpeas',
       'mothbeans', 'mungbean', 'blackgram', 'lentil', 'pomegranate',
       'banana', 'mango', 'grapes', 'watermelon', 'muskmelon', 'apple',
       'orange', 'papaya', 'coconut', 'cotton', 'jute', 'coffee'],
      dtype=object)
```

In [13]:

```
1 data['label'].value_counts()
```

Out[13]:

```
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 [14]:

```
1 crop_summary = pd.pivot_table(data,index=['label'],
2                               aggfunc='mean')
```

In [15]:

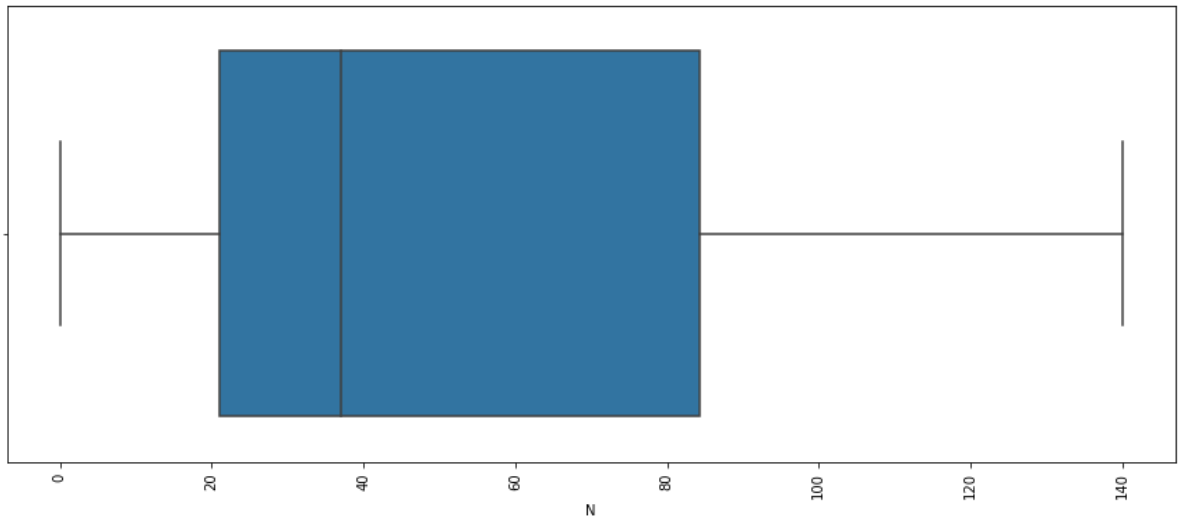
```
1 crop_summary
```

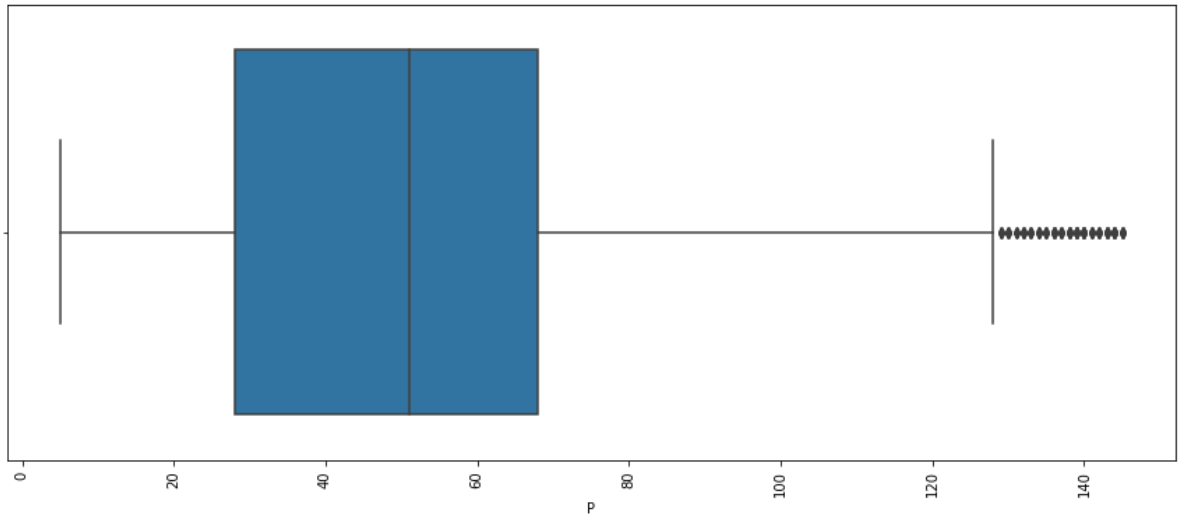
Out[15]:

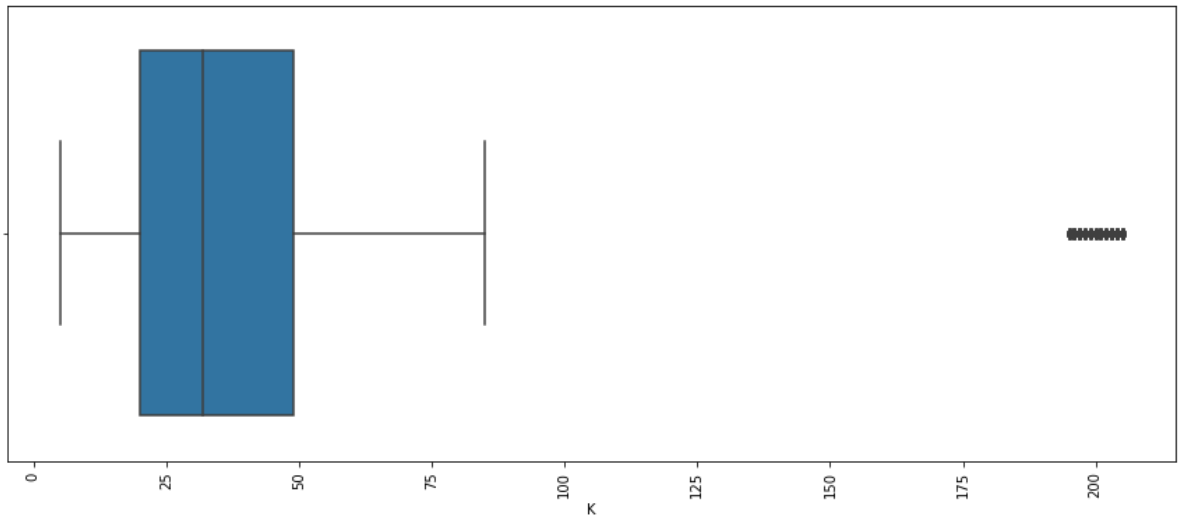
label	K	N	P	humidity	ph	rainfall	temperature
apple	199.89	20.80	134.22	92.333383	5.929663	112.654779	22.630942
banana	50.05	100.23	82.01	80.358123	5.983893	104.626980	27.376798
blackgram	19.24	40.02	67.47	65.118426	7.133952	67.884151	29.973340
chickpea	79.92	40.09	67.79	16.860439	7.336957	80.058977	18.872847
coconut	30.59	21.98	16.93	94.844272	5.976562	175.686646	27.409892
coffee	29.94	101.20	28.74	58.869846	6.790308	158.066295	25.540477
cotton	19.56	117.77	46.24	79.843474	6.912675	80.398043	23.988958
grapes	200.11	23.18	132.53	81.875228	6.025937	69.611829	23.849575
jute	39.99	78.40	46.86	79.639864	6.732778	174.792798	24.958376
kidneybeans	20.05	20.75	67.54	21.605357	5.749411	105.919778	20.115085
lentil	19.41	18.77	68.36	64.804785	6.927932	45.680454	24.509052
maize	19.79	77.76	48.44	65.092249	6.245190	84.766988	22.389204
mango	29.92	20.07	27.18	50.156573	5.766373	94.704515	31.208770
mothbeans	20.23	21.44	48.01	53.160418	6.831174	51.198487	28.194920
mungbean	19.87	20.99	47.28	85.499975	6.723957	48.403601	28.525775
muskmelon	50.08	100.32	17.72	92.342802	6.358805	24.689952	28.663066
orange	10.01	19.58	16.55	92.170209	7.016957	110.474969	22.765725
papaya	50.04	49.88	59.05	92.403388	6.741442	142.627839	33.723859
pigeonpeas	20.29	20.73	67.73	48.061633	5.794175	149.457564	27.741762
pomegranate	40.21	18.87	18.75	90.125504	6.429172	107.528442	21.837842
rice	39.87	79.89	47.58	82.272822	6.425471	236.181114	23.689332
watermelon	50.22	99.42	17.00	85.160375	6.495778	50.786219	25.591767

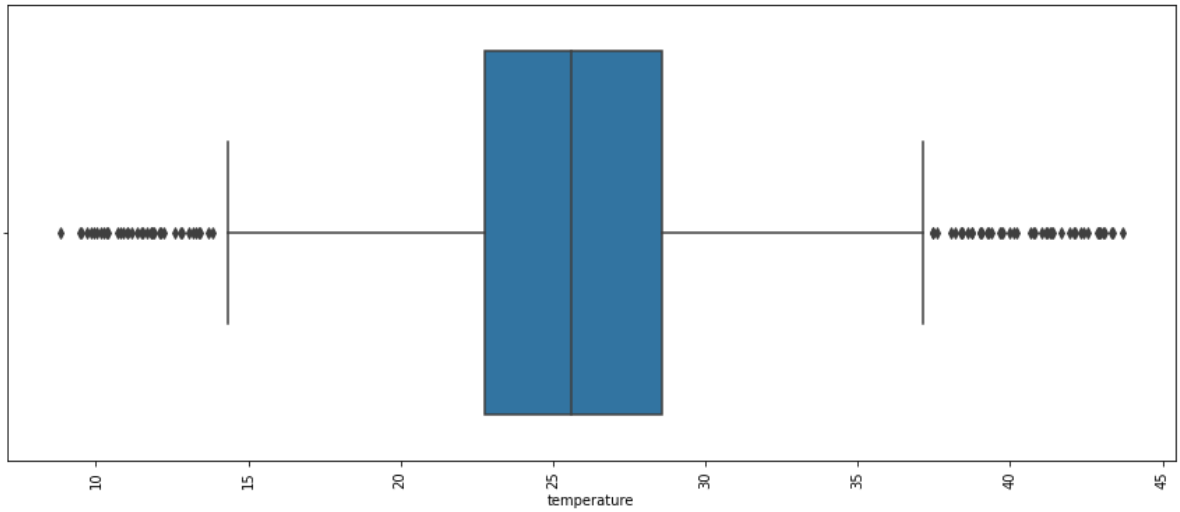
In [18]:

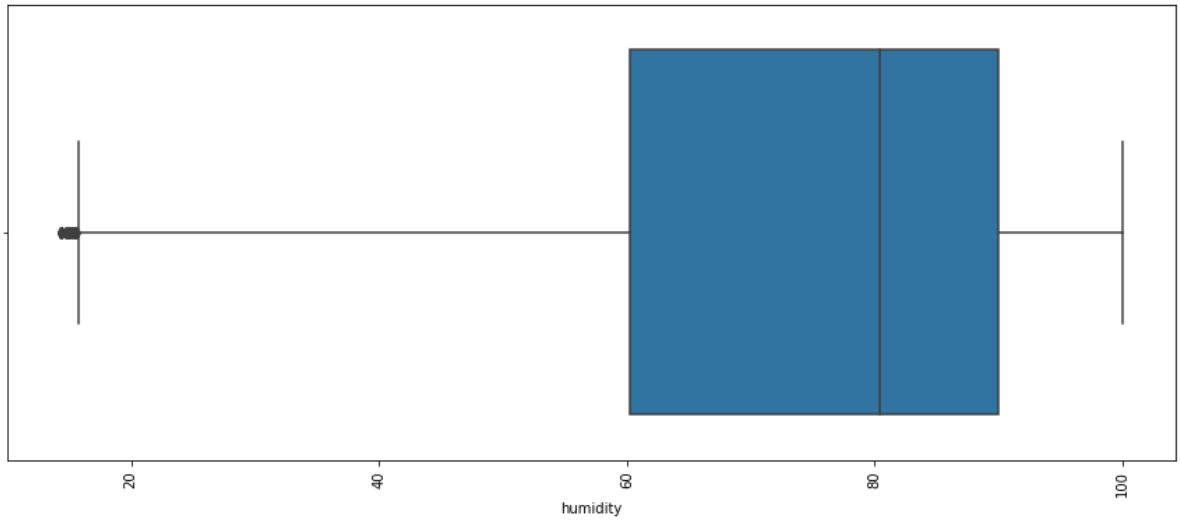
```
data1 = data[['N', 'P', 'K', 'temperature', 'humidity', 'ph', 'rainfall']]
for i in data1.columns:
            plt.figure(figsize=(15,6))
            sns.boxplot(data1[i])
            plt.xticks(rotation = 90)
            plt.show()
```

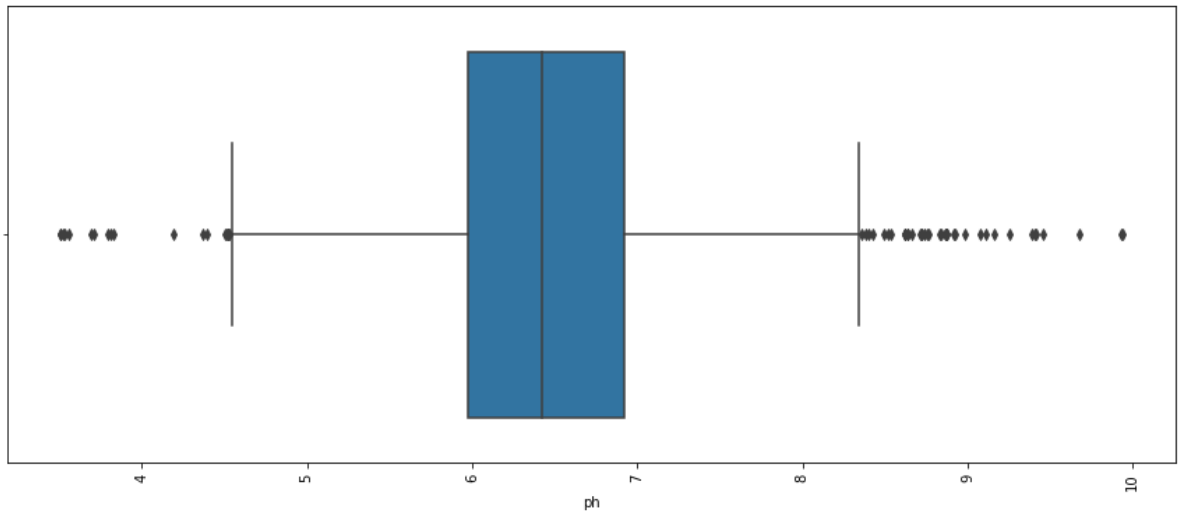


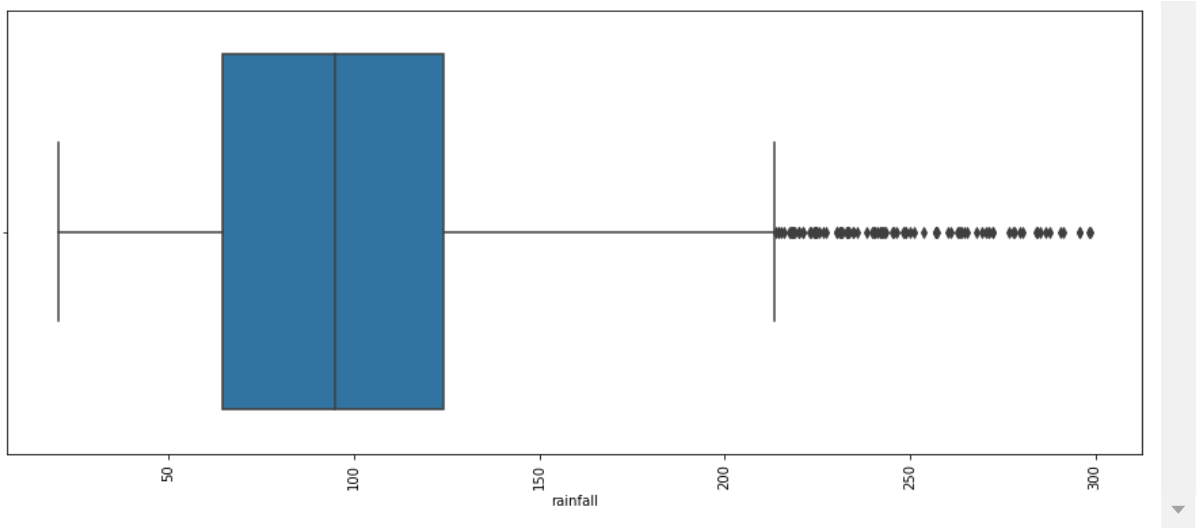



0
25
50
75
100
125
150
175
200
K


10
15
20
25
30
35
40
45
temperature


20
40
60
80
100
humidity


4
5
6
7
8
9
10
ph



In [31]:

```
1 crop_summary_new = crop_summary.reset_index()
```

In [32]:

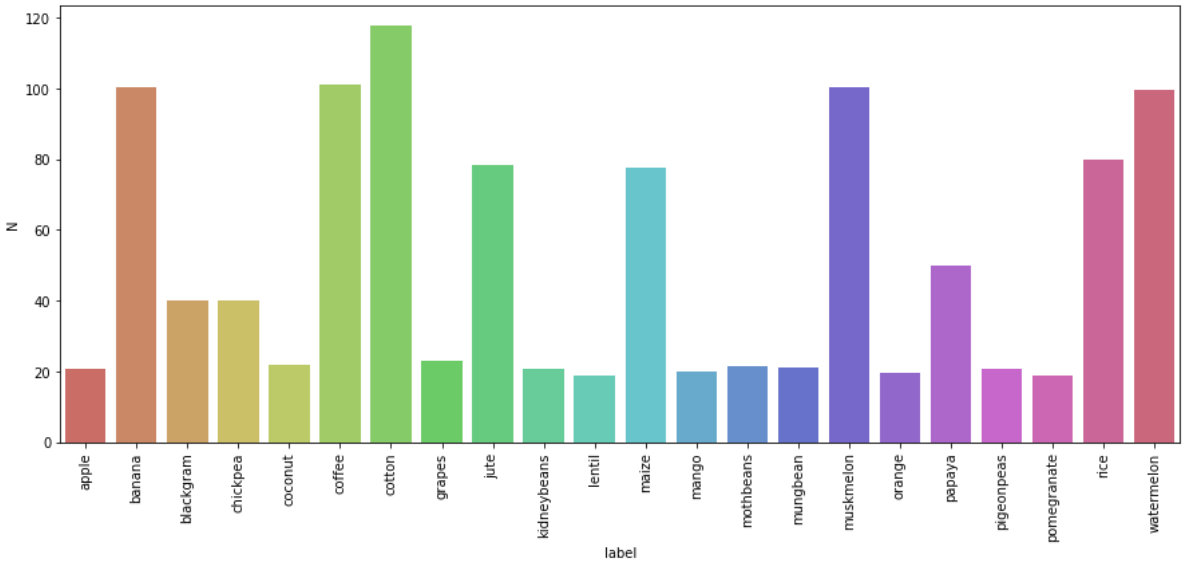
```
1 crop_summary_new
```

Out[32]:

	label	K	N	P	humidity	ph	rainfall	temperature
0	apple	199.89	20.80	134.22	92.333383	5.929663	112.654779	22.630942
1	banana	50.05	100.23	82.01	80.358123	5.983893	104.626980	27.376798
2	blackgram	19.24	40.02	67.47	65.118426	7.133952	67.884151	29.973340
3	chickpea	79.92	40.09	67.79	16.860439	7.336957	80.058977	18.872847
4	coconut	30.59	21.98	16.93	94.844272	5.976562	175.686646	27.409892
5	coffee	29.94	101.20	28.74	58.869846	6.790308	158.066295	25.540477
6	cotton	19.56	117.77	46.24	79.843474	6.912675	80.398043	23.988958
7	grapes	200.11	23.18	132.53	81.875228	6.025937	69.611829	23.849575
8	jute	39.99	78.40	46.86	79.639864	6.732778	174.792798	24.958376
9	kidneybeans	20.05	20.75	67.54	21.605357	5.749411	105.919778	20.115085
10	lentil	19.41	18.77	68.36	64.804785	6.927932	45.680454	24.509052
11	maize	19.79	77.76	48.44	65.092249	6.245190	84.766988	22.389204
12	mango	29.92	20.07	27.18	50.156573	5.766373	94.704515	31.208770
13	mothbeans	20.23	21.44	48.01	53.160418	6.831174	51.198487	28.194920
14	mungbean	19.87	20.99	47.28	85.499975	6.723957	48.403601	28.525775
15	muskmelon	50.08	100.32	17.72	92.342802	6.358805	24.689952	28.663066
16	orange	10.01	19.58	16.55	92.170209	7.016957	110.474969	22.765725
17	papaya	50.04	49.88	59.05	92.403388	6.741442	142.627839	33.723859
18	pigeonpeas	20.29	20.73	67.73	48.061633	5.794175	149.457564	27.741762
19	pomegranate	40.21	18.87	18.75	90.125504	6.429172	107.528442	21.837842
20	rice	39.87	79.89	47.58	82.272822	6.425471	236.181114	23.689332
21	watermelon	50.22	99.42	17.00	85.160375	6.495778	50.786219	25.591767

In [33]:

```
plt.figure(figsize=(15,6))
sns.barplot(y = 'N', x = 'label', data=crop_summary_new, palette = 'hls')
plt.xticks(rotation = 90)
plt.show()
```

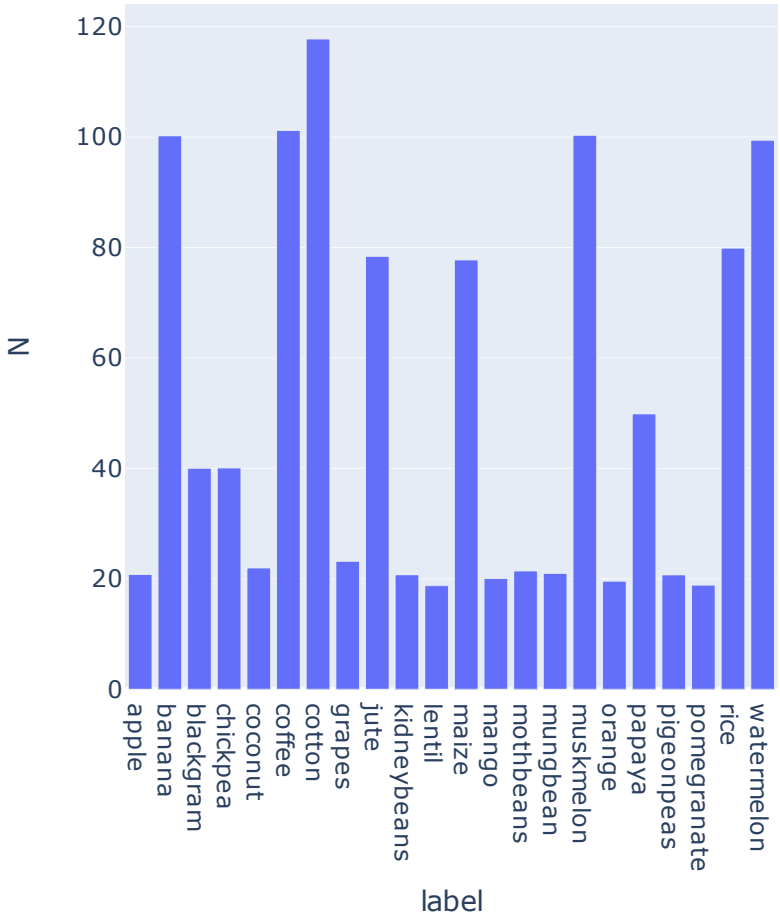


In [34]:

```
import plotly.graph_objects as go
import plotly.express as px
from plotly.subplots import make_subplots
```

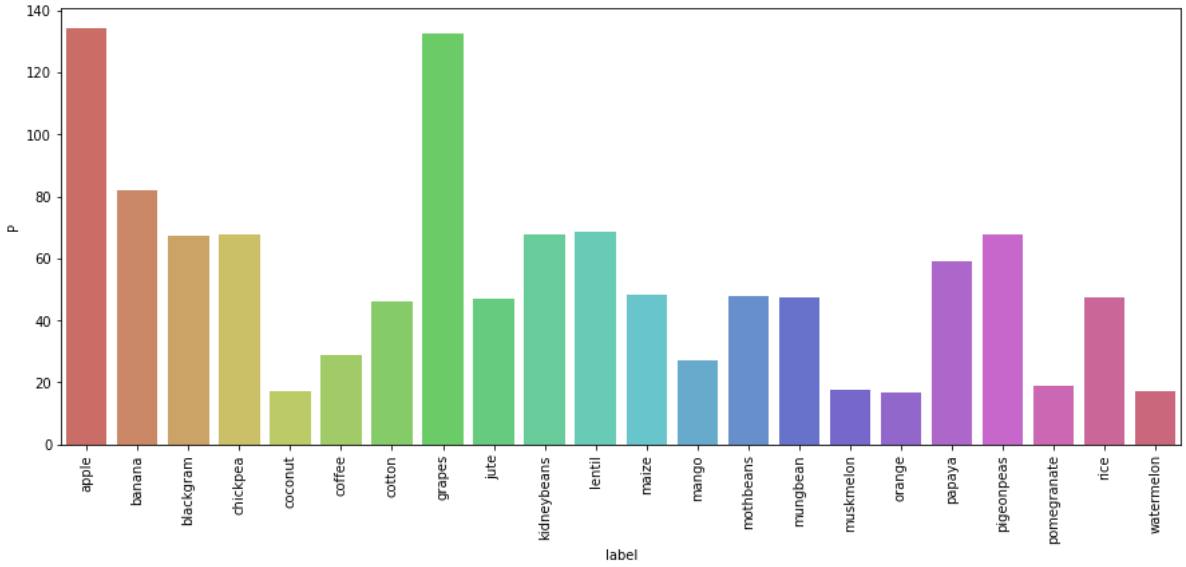
In [35]:

```
fig1 = px.bar(crop_summary_new, x='label', y='N')
fig1.show()
```



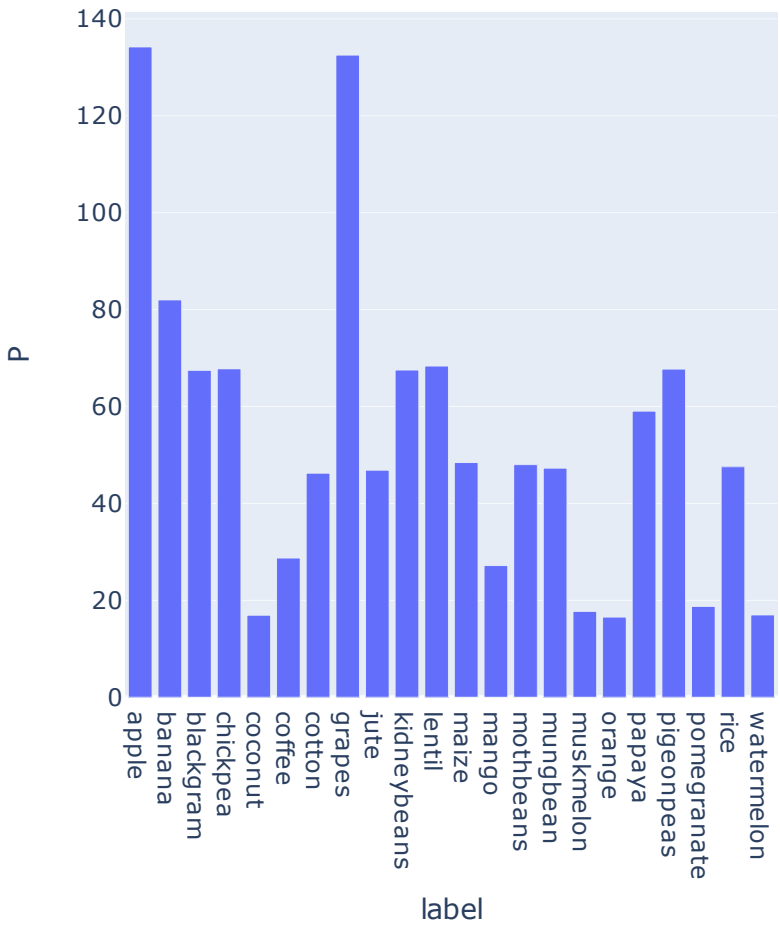
In [36]:

```
plt.figure(figsize=(15,6))
sns.barplot(y = 'P', x = 'label', data=crop_summary_new, palette = 'hls')
plt.xticks(rotation = 90)
plt.show()
```



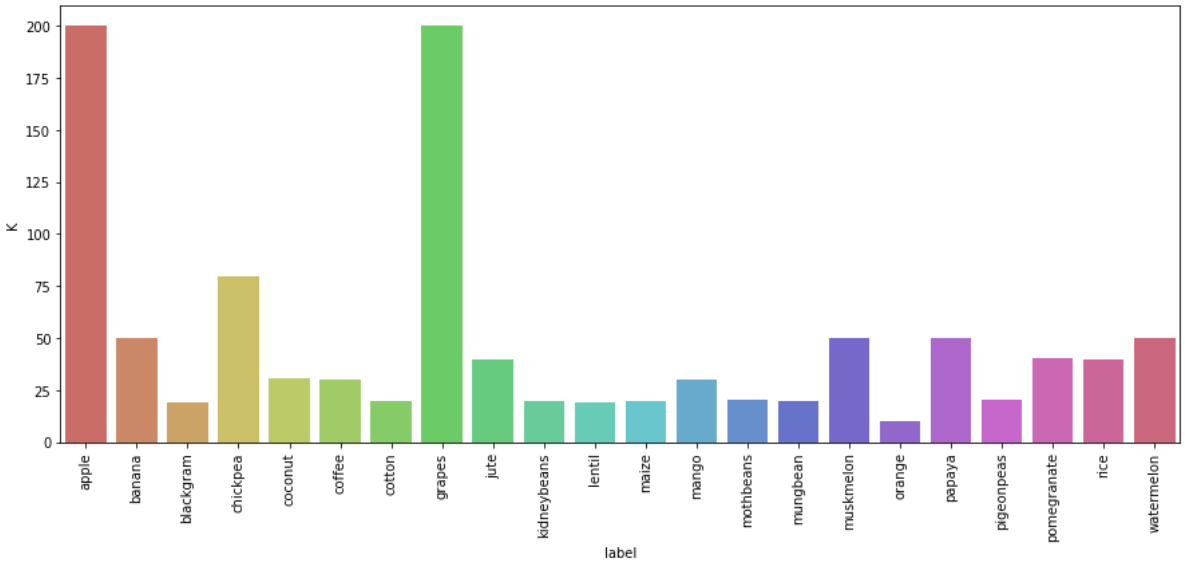
In [37]:

```
1 fig2 = px.bar(crop_summary_new, x='label', y='P')
2 fig2.show()
```



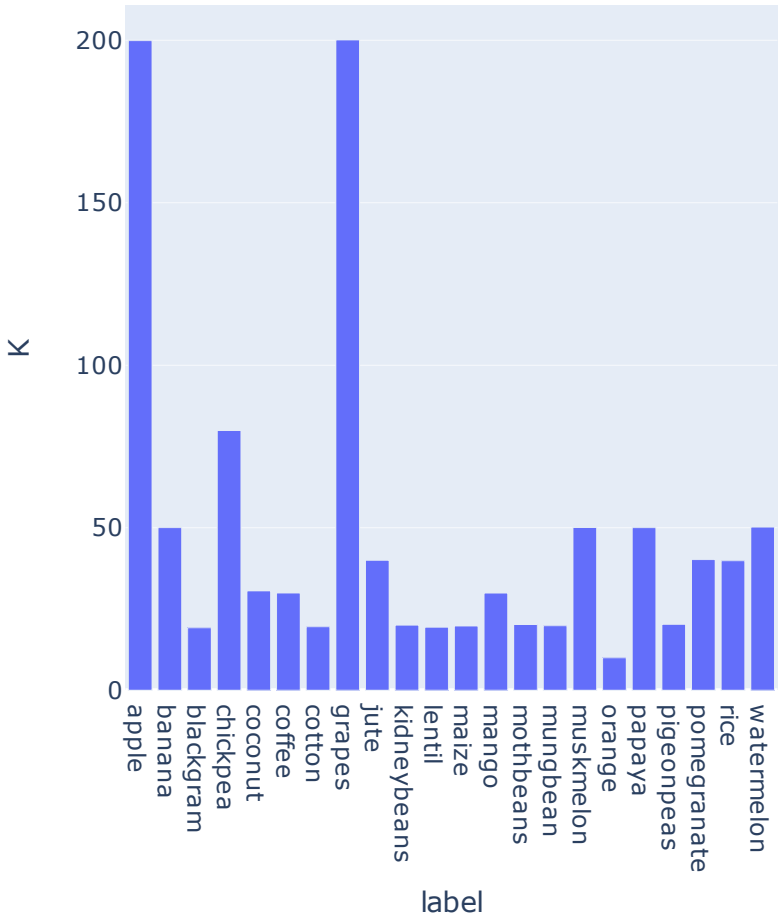
In [38]:

```
plt.figure(figsize=(15,6))
sns.barplot(y = 'K', x = 'label', data=crop_summary_new, palette = 'hls')
plt.xticks(rotation = 90)
plt.show()
```



In [39]:

```
fig3 = px.bar(crop_summary_new, x='label', y='K')
fig3.show()
```



In [40]:

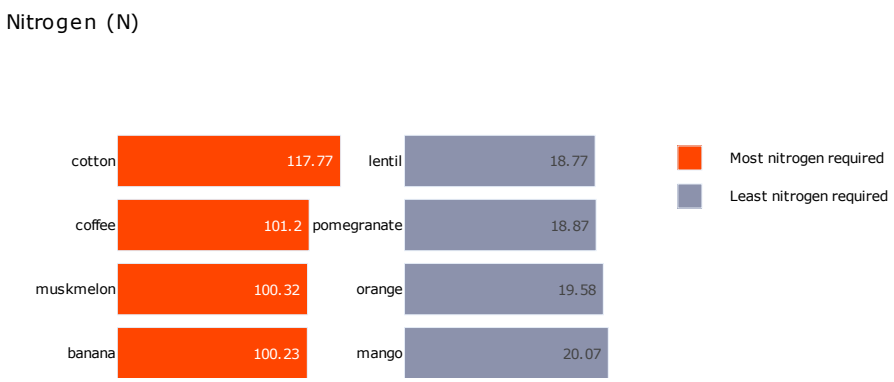
```
colorarr = ['#0592D0','#Cd7f32', '#E97451', '#Bdb76b', '#954535', '#C2b280', '#8080
            '#32cd32','#39ff14','#00ff7f', '#008080', '#36454f', '#F88379', '#Ff4500
            '#Faf0e6', '#8c92ac', '#Dbd7d2','#A7a6ba', '#B38b6d']
```

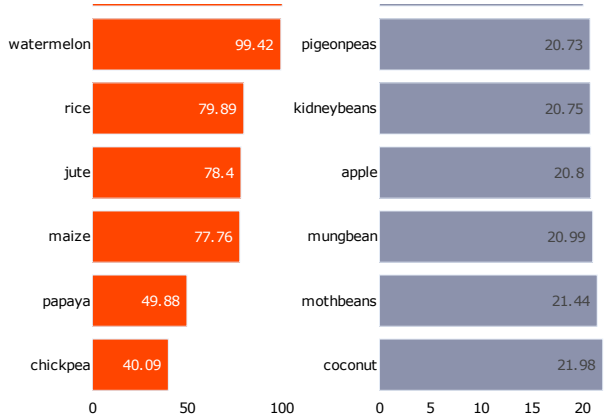
In [41]:

```
import random
```

In [42]:

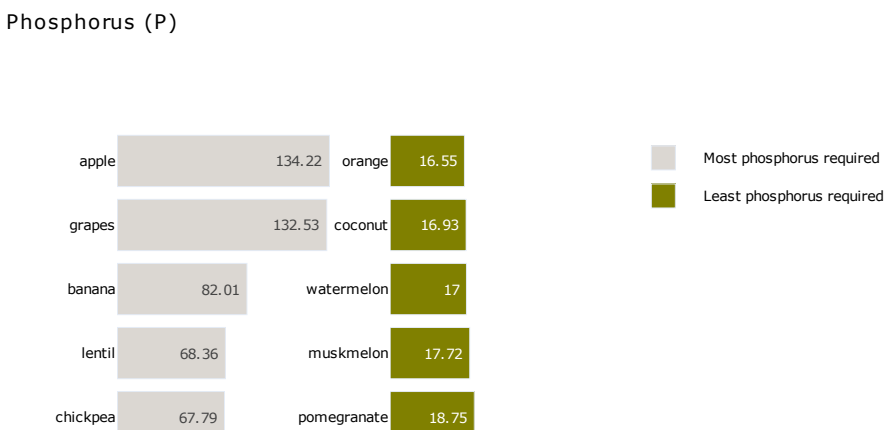
```
crop_summary_N = crop_summary.sort_values(by='N',
                                          ascending=False)

fig = make_subplots(rows=1, cols=2)

top = {
    'y' : crop_summary_N['N'][0:10].sort_values().index,
    'x' : crop_summary_N['N'][0:10].sort_values()
}

last = {
    'y' : crop_summary_N['N'][-10:].index,
    'x' : crop_summary_N['N'][-10:]
}

fig.add_trace(
    go.Bar(top,
           name="Most nitrogen required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=top['x']),

    row=1, col=1
)

fig.add_trace(
    go.Bar(last,
           name="Least nitrogen required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=last['x']),
    row=1, col=2
)
fig.update_traces(texttemplate='%{text}', textposition='inside')
fig.update_layout(title_text="Nitrogen (N)",
                  plot_bgcolor='white',
                  font_size=7,
                  font_color='black',
                 height=500)

fig.update_xaxes(showgrid=False)
fig.update_yaxes(showgrid=False)
fig.show()
```

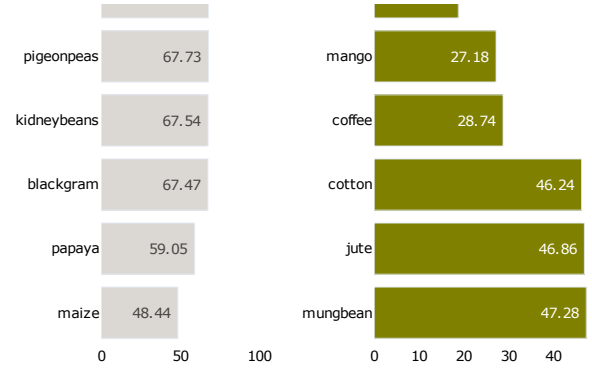



watermelon
99.42
rice
79.89
jute
78.4
maize
77.76
papaya
49.88
chickpea
40.09
0
50
100
pigeonpeas
20.73
kidneybeans
20.75
apple
20.8
mungbean
20.99
mothbeans
21.44
coconut
21.98
0
5
10
15
20

In [43]:

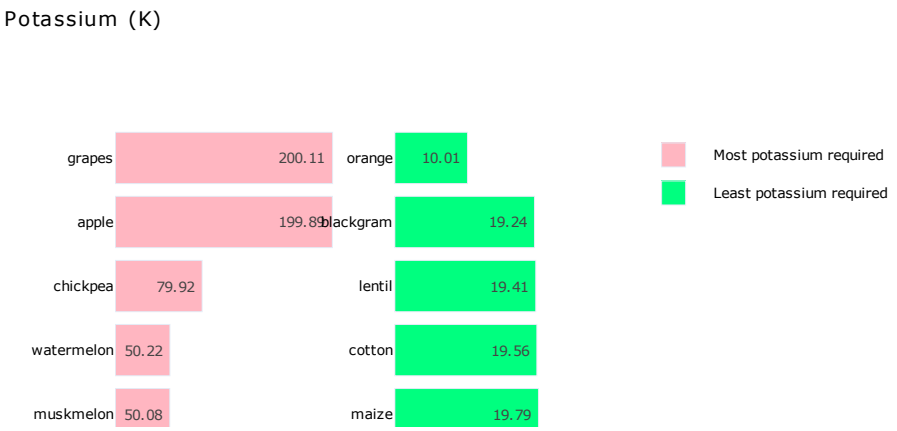
```
crop_summary_P = crop_summary.sort_values(by='P', ascending=False)

fig = make_subplots(rows=1, cols=2)

top = {
    'y' : crop_summary_P['P'][0:10].sort_values().index,
    'x' : crop_summary_P['P'][0:10].sort_values()
}

last = {
    'y' : crop_summary_P['P'][-10:].index,
    'x' : crop_summary_P['P'][-10:]
}

fig.add_trace(
    go.Bar(top,
           name="Most phosphorus required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=top['x']),

    row=1, col=1
)

fig.add_trace(
    go.Bar(last,
           name="Least phosphorus required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=last['x']),
    row=1, col=2
)
fig.update_traces(texttemplate='%{text}', textposition='inside')
fig.update_layout(title_text="Phosphorus (P)",
                  plot_bgcolor='white',
                  font_size=7,
                  font_color='black',
                 height=500)

fig.update_xaxes(showgrid=False)
fig.update_yaxes(showgrid=False)
fig.show()
```

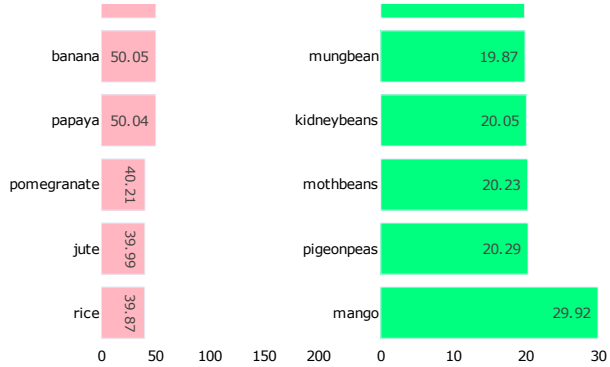



pigeonpeas
67.73
kidneybeans
67.54
blackgram
67.47
papaya
59.05
maize
48.44
0
50
100
mango
27.18
coffee
28.74
cotton
46.24
jute
46.86
mungbean
47.28
0
10
20
30
40

In [44]:

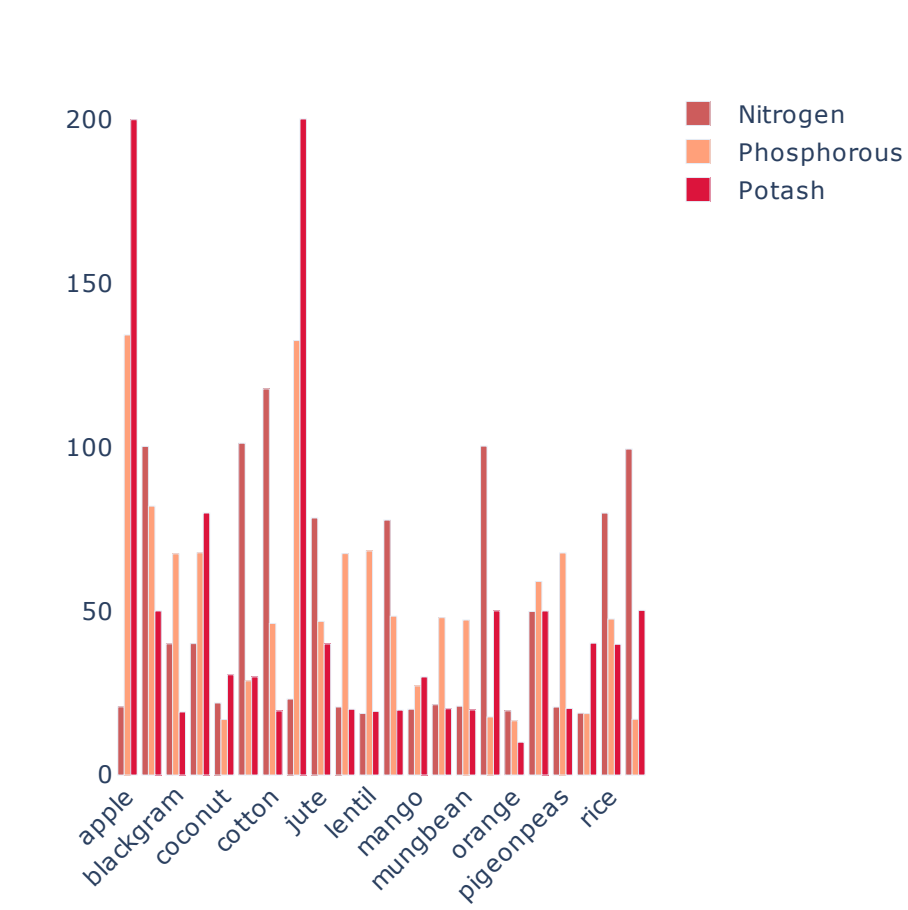
```
crop_summary_K = crop_summary.sort_values(by='K', ascending=False)

fig = make_subplots(rows=1, cols=2)

top = {
    'y' : crop_summary_K['K'][0:10].sort_values().index,
    'x' : crop_summary_K['K'][0:10].sort_values()
}

last = {
    'y' : crop_summary_K['K'][-10:].index,
    'x' : crop_summary_K['K'][-10:]
}

fig.add_trace(
    go.Bar(top,
           name="Most potassium required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=top['x']),

    row=1, col=1
)

fig.add_trace(
    go.Bar(last,
           name="Least potassium required",
           marker_color=random.choice(colorarr),
           orientation='h',
          text=last['x']),
    row=1, col=2
)
fig.update_traces(texttemplate='%{text}', textposition='inside')
fig.update_layout(title_text="Potassium (K)",
                  plot_bgcolor='white',
                  font_size=7,
                  font_color='black',
                  height=500)

fig.update_xaxes(showgrid=False)
fig.update_yaxes(showgrid=False)
fig.show()
```




banana
papaya
pomegranate
jute
rice
50.05
50.04
40.21
39.99
39.87
0
50
100
150
200
mungbean
kidneybeans
mothbeans
pigeonpeas
mango
19.87
20.05
20.23
20.29
29.92
0
10
20
30

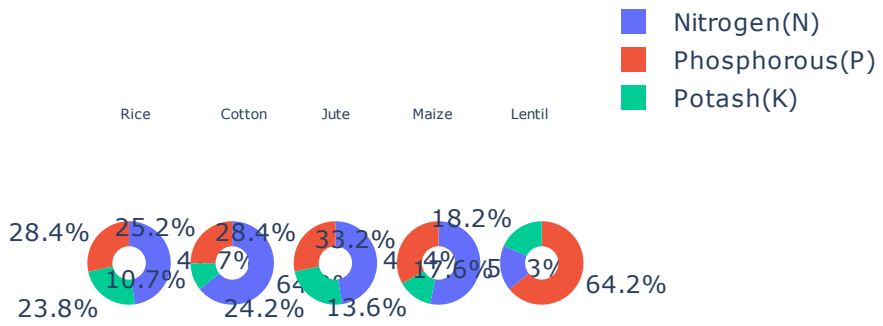
In [45]:

```
fig = go.Figure()
fig.add_trace(go.Bar(
    x=crop_summary.index,
    y=crop_summary['N'],
    name='Nitrogen',
    marker_color='indianred'
))
fig.add_trace(go.Bar(
    x=crop_summary.index,
    y=crop_summary['P'],
    name='Phosphorous',
    marker_color='lightsalmon'
))
fig.add_trace(go.Bar(
    x=crop_summary.index,
    y=crop_summary['K'],
    name='Potash',
    marker_color='crimson'
))

fig.update_layout(title="N, P, K values comparision between crops",
                  plot_bgcolor='white',
                  barmode='group',
                  xaxis_tickangle=-45)

fig.show()
```



In [46]:

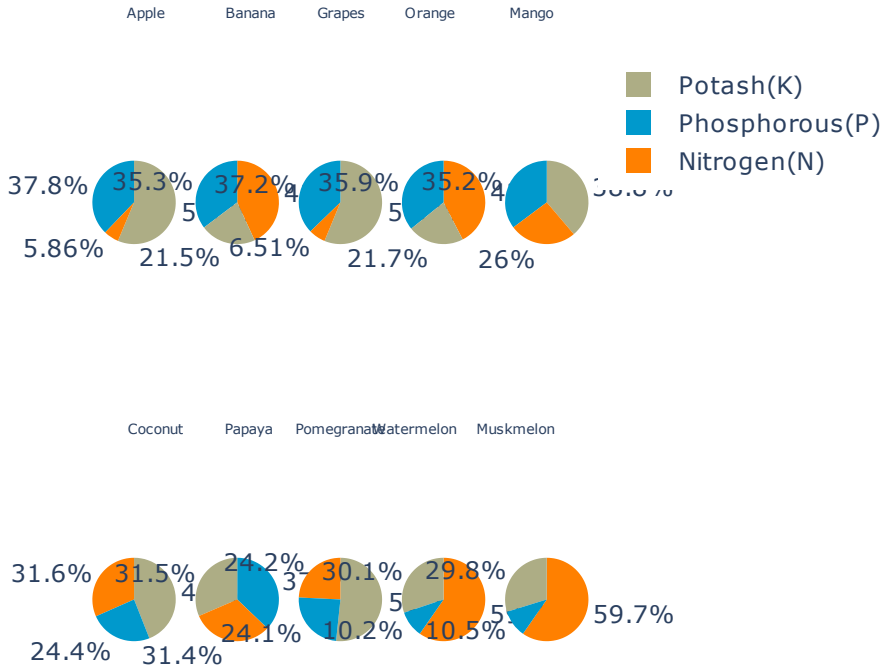
```
labels = ['Nitrogen(N)','Phosphorous(P)','Potash(K)']
fig = make_subplots(rows=1, cols=5, specs=[[{'type':'domain'}, {'type':'domain'},
                                            {'type':'domain'}, {'type':'domain'},
                                            {'type':'domain'}]])

rice_npk = crop_summary[crop_summary.index=='rice']
values = [rice_npk['N'][0], rice_npk['P'][0], rice_npk['K'][0]]
fig.add_trace(go.Pie(labels=labels, values=values,name="Rice"),1, 1)

cotton_npk = crop_summary[crop_summary.index=='cotton']
values = [cotton_npk['N'][0], cotton_npk['P'][0], cotton_npk['K'][0]]
fig.add_trace(go.Pie(labels=labels, values=values,name="Cotton"),1, 2)

jute_npk = crop_summary[crop_summary.index=='jute']
values = [jute_npk['N'][0], jute_npk['P'][0], jute_npk['K'][0]]
fig.add_trace(go.Pie(labels=labels, values=values,name="Jute"),1, 3)

maize_npk = crop_summary[crop_summary.index=='maize']
values = [maize_npk['N'][0], maize_npk['P'][0], maize_npk['K'][0]]
fig.add_trace(go.Pie(labels=labels, values=values,name="Maize"),1, 4)

lentil_npk = crop_summary[crop_summary.index=='lentil']
values = [lentil_npk['N'][0], lentil_npk['P'][0], lentil_npk['K'][0]]
fig.add_trace(go.Pie(labels=labels, values=values,name="Lentil"),1, 5)

fig.update_traces(hole=.4, hoverinfo="label+percent+name")
fig.update_layout(
    title_text="NPK ratio for rice, cotton, jute, maize, lentil",
    annotations=[dict(text='Rice',x=0.06,y=0.8, font_size=7, showarrow=False),
                 dict(text='Cotton',x=0.26,y=0.8, font_size=7, showarrow=False),
                 dict(text='Jute',x=0.50,y=0.8, font_size=7, showarrow=False),
                 dict(text='Maize',x=0.74,y=0.8, font_size=7, showarrow=False),
                dict(text='Lentil',x=0.94,y=0.8, font_size=7, showarrow=False)])
fig.show()
```



In [48]:

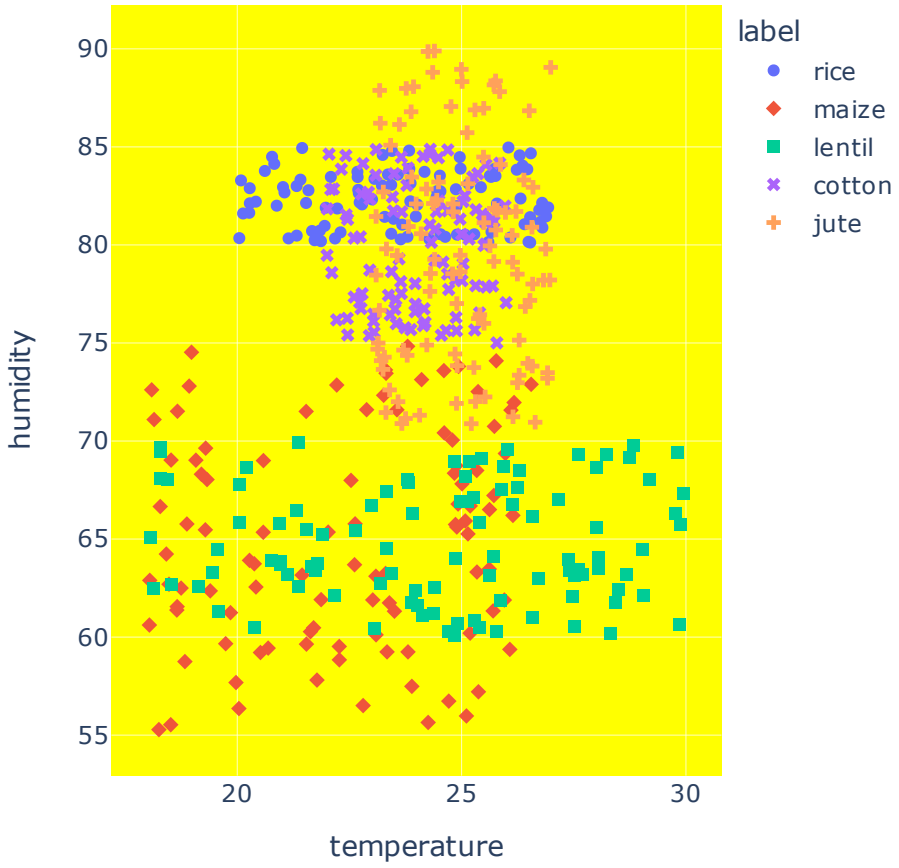
```
Nitrogen(N)','Phosphorous(P)','Potash(K)']
['type':'domain'}, {'type':'domain'}, {'type':'domain'}, {'type':'domain'}, {'type':'doma
type':'domain'}, {'type':'domain'}, {'type':'domain'}, {'type':'domain'}, {'type':'domai
_subplots(rows=2, cols=5, specs=specs)
s =  ['rgb(255, 128, 0)', 'rgb(0, 153, 204)', 'rgb(173, 173, 133)']

= crop_summary[crop_summary.index=='apple']
apple_npk['N'][0], apple_npk['P'][0], apple_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Apple", marker_colors=cafe_colors),1, 1)

= crop_summary[crop_summary.index=='banana']
banana_npk['N'][0], banana_npk['P'][0], banana_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Banana", marker_colors=cafe_colors),1, 2)

= crop_summary[crop_summary.index=='grapes']
grapes_npk['N'][0], grapes_npk['P'][0], grapes_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Grapes", marker_colors=cafe_colors),1, 3)

= crop_summary[crop_summary.index=='orange']
orange_npk['N'][0], orange_npk['P'][0], orange_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Orange", marker_colors=cafe_colors),1, 4)

= crop_summary[crop_summary.index=='mango']
mango_npk['N'][0], mango_npk['P'][0], mango_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Mango", marker_colors=cafe_colors),1, 5)

= crop_summary[crop_summary.index=='coconut']
coconut_npk['N'][0], coconut_npk['P'][0], coconut_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Coconut", marker_colors=cafe_colors),2, 1)

= crop_summary[crop_summary.index=='papaya']
papaya_npk['N'][0], papaya_npk['P'][0], papaya_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Papaya", marker_colors=cafe_colors),2, 2)

e_npk = crop_summary[crop_summary.index=='pomegranate']
pomegranate_npk['N'][0], pomegranate_npk['P'][0], pomegranate_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Pomegranate", marker_colors=cafe_colors),2,

_npk = crop_summary[crop_summary.index=='watermelon']
watermelon_npk['N'][0], watermelon_npk['P'][0], watermelon_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Watermelon", marker_colors=cafe_colors),2,

npk = crop_summary[crop_summary.index=='muskmelon']
muskmelon_npk['N'][0], muskmelon_npk['P'][0], muskmelon_npk['K'][0]]
ace(go.Pie(labels=labels, values=values,name="Muskmelon", marker_colors=cafe_colors),2, 5

_layout(
ext="NPK ratio for fruits",
ions=[dict(text='Apple',x=0.06,y=1.08, font_size=7, showarrow=False),
      dict(text='Banana',x=0.26,y=1.08, font_size=7, showarrow=False),
      dict(text='Grapes',x=0.50,y=1.08, font_size=7, showarrow=False),
      dict(text='Orange',x=0.74,y=1.08, font_size=7, showarrow=False),
     dict(text='Mango',x=0.94,y=1.08, font_size=7, showarrow=False),
     dict(text='Coconut',x=0.06,y=0.46, font_size=7, showarrow=False),
      dict(text='Papaya',x=0.26,y=0.46, font_size=7, showarrow=False),
      dict(text='Pomegranate',x=0.50,y=0.46, font_size=7, showarrow=False),
      dict(text='Watermelon',x=0.74,y=0.46, font_size=7, showarrow=False),
     dict(text='Muskmelon',x=0.94,y=0.46, font_size=7, showarrow=False)])
```

NPK ratio for fruits



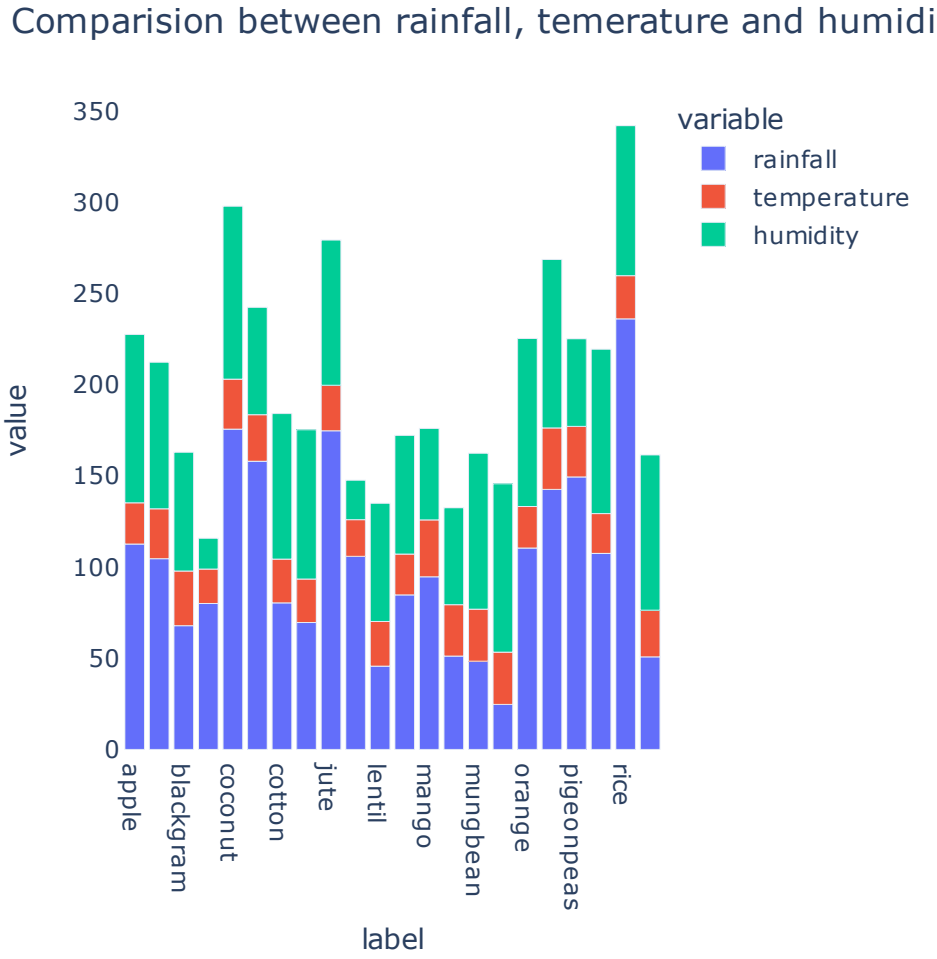
In [51]:

```
crop_scatter = data[(data['label']=='rice') |
                    (data['label']=='jute') |
                    (data['label']=='cotton') |
                   (data['label']=='maize') |
                   (data['label']=='lentil')]

fig = px.scatter(crop_scatter, x="temperature", y="humidity", color="label", symbol
fig.update_layout(plot_bgcolor='yellow')
fig.update_xaxes(showgrid=True)
fig.update_yaxes(showgrid=True)

fig.show()
```



In [52]:

```
fig = px.bar(crop_summary, x=crop_summary.index, y=["rainfall", "temperature", "hum
fig.update_layout(title_text="Comparision between rainfall, temerature and humidity
                  plot_bgcolor='white',
                 height=500)

fig.update_xaxes(showgrid=False)
fig.update_yaxes(showgrid=False)
fig.show()
```



In [53]:

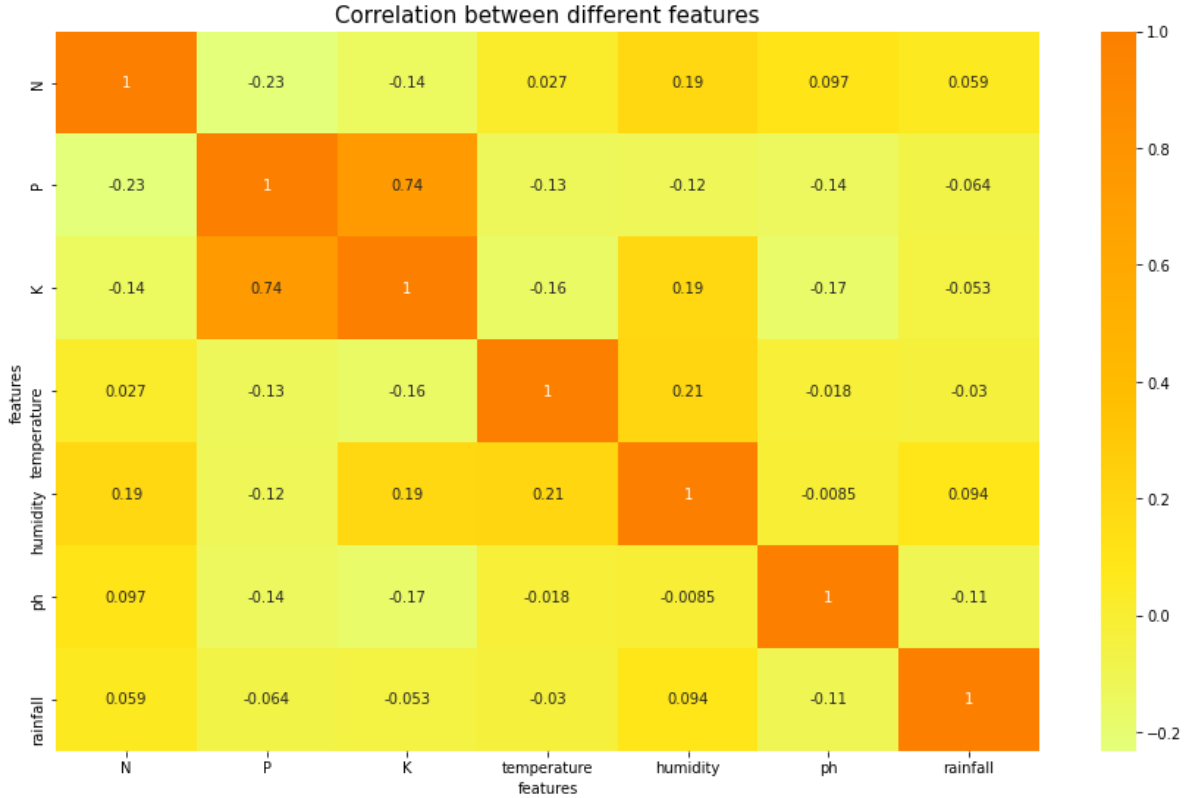
```
1 data.corr()
```

Out[53]:

	N	P	K	temperature	humidity	ph	rainfall
N	1.000000	-0.231460	-0.140512	0.026504	0.190688	0.096683	0.059020
P	-0.231460	1.000000	0.736232	-0.127541	-0.118734	-0.138019	-0.063839
K	-0.140512	0.736232	1.000000	-0.160387	0.190859	-0.169503	-0.053461
temperature	0.026504	-0.127541	-0.160387	1.000000	0.205320	-0.017795	-0.030084
humidity	0.190688	-0.118734	0.190859	0.205320	1.000000	-0.008483	0.094423
ph	0.096683	-0.138019	-0.169503	-0.017795	-0.008483	1.000000	-0.109069
rainfall	0.059020	-0.063839	-0.053461	-0.030084	0.094423	-0.109069	1.000000

In [54]:

```
1 fig, ax = plt.subplots(1, 1, figsize=(15, 9))
2 sns.heatmap(data.corr(), annot=True,cmap='Wistia')
3 ax.set(xlabel='features')
4 ax.set(ylabel='features')
5
6 plt.title('Correlation between different features', fontsize = 15, c='black')
7 plt.show()
```



In [55]:

```
X = data.drop('label', axis=1)
y = data['label']
```

In [56]:

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.33,
                                                    shuffle = True, random_state =
```

In [57]:

```
import lightgbm as lgb
model = lgb.LGBMClassifier()
model.fit(X_train, y_train)
```

Out[57]:

```
LGBMClassifier()
```

In [58]:

```
y_pred=model.predict(X_test)
```

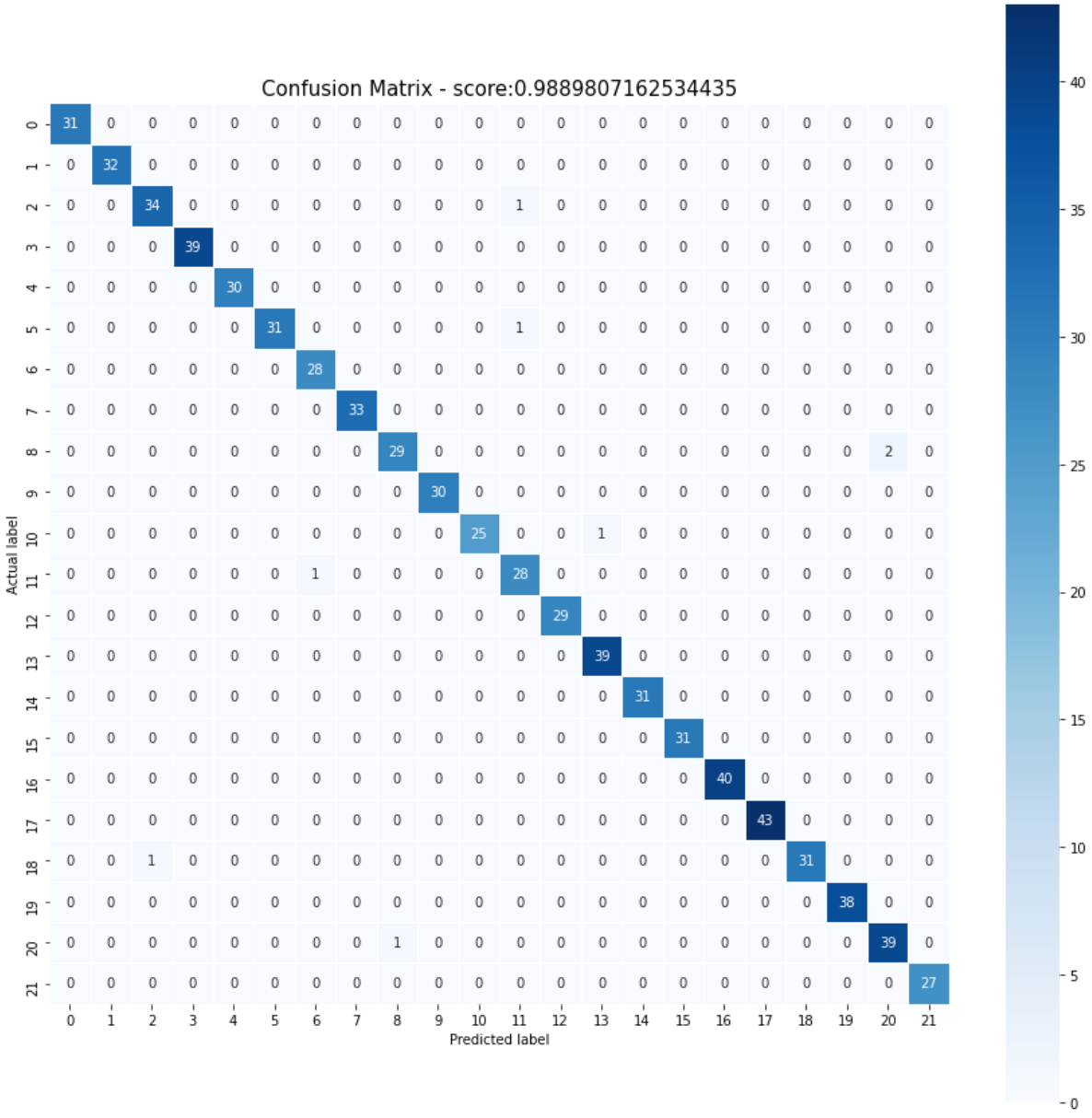
In [59]:

```
from sklearn.metrics import accuracy_score
accuracy=accuracy_score(y_pred, y_test)
print('LightGBM Model accuracy score: {0:0.4f}'.format(accuracy_score(y_test, y_pre
```

```
LightGBM Model accuracy score: 0.9890
```

In [60]:

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(15,15))
sns.heatmap(cm, annot=True, fmt=".0f", linewidths=.5, square = True, cmap = 'Blues'
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Confusion Matrix - score:'+str(accuracy_score(y_test,y_pred))
plt.title(all_sample_title, size = 15);
plt.show()
```



In [61]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

       apple       1.00      1.00      1.00        31
      banana       1.00      1.00      1.00        32
   blackgram       0.97      0.97      0.97        35
    chickpea       1.00      1.00      1.00        39
     coconut       1.00      1.00      1.00        30
      coffee       1.00      0.97      0.98        32
      cotton       0.97      1.00      0.98        28
      grapes       1.00      1.00      1.00        33
        jute       0.97      0.94      0.95        31
 kidneybeans       1.00      1.00      1.00        30
      lentil       1.00      0.96      0.98        26
       maize       0.93      0.97      0.95        29
       mango       1.00      1.00      1.00        29
   mothbeans       0.97      1.00      0.99        39
    mungbean       1.00      1.00      1.00        31
   muskmelon       1.00      1.00      1.00        31
      orange       1.00      1.00      1.00        40
      papaya       1.00      1.00      1.00        43
  pigeonpeas       1.00      0.97      0.98        32
 pomegranate       1.00      1.00      1.00        38
        rice       0.95      0.97      0.96        40
  watermelon       1.00      1.00      1.00        27

    accuracy                           0.99       726
   macro avg       0.99      0.99      0.99       726
weighted avg       0.99      0.99      0.99       726
```

In [80]:

```
from sklearn.tree import DecisionTreeClassifier
classifier= DecisionTreeClassifier(criterion='entropy', random_state=0)
classifier.fit(X_train, y_train)
```

Out[80]:

```
DecisionTreeClassifier(criterion='entropy', random_state=0)
```

In [81]:

```
y_pred=classifier.predict(X_test)
```

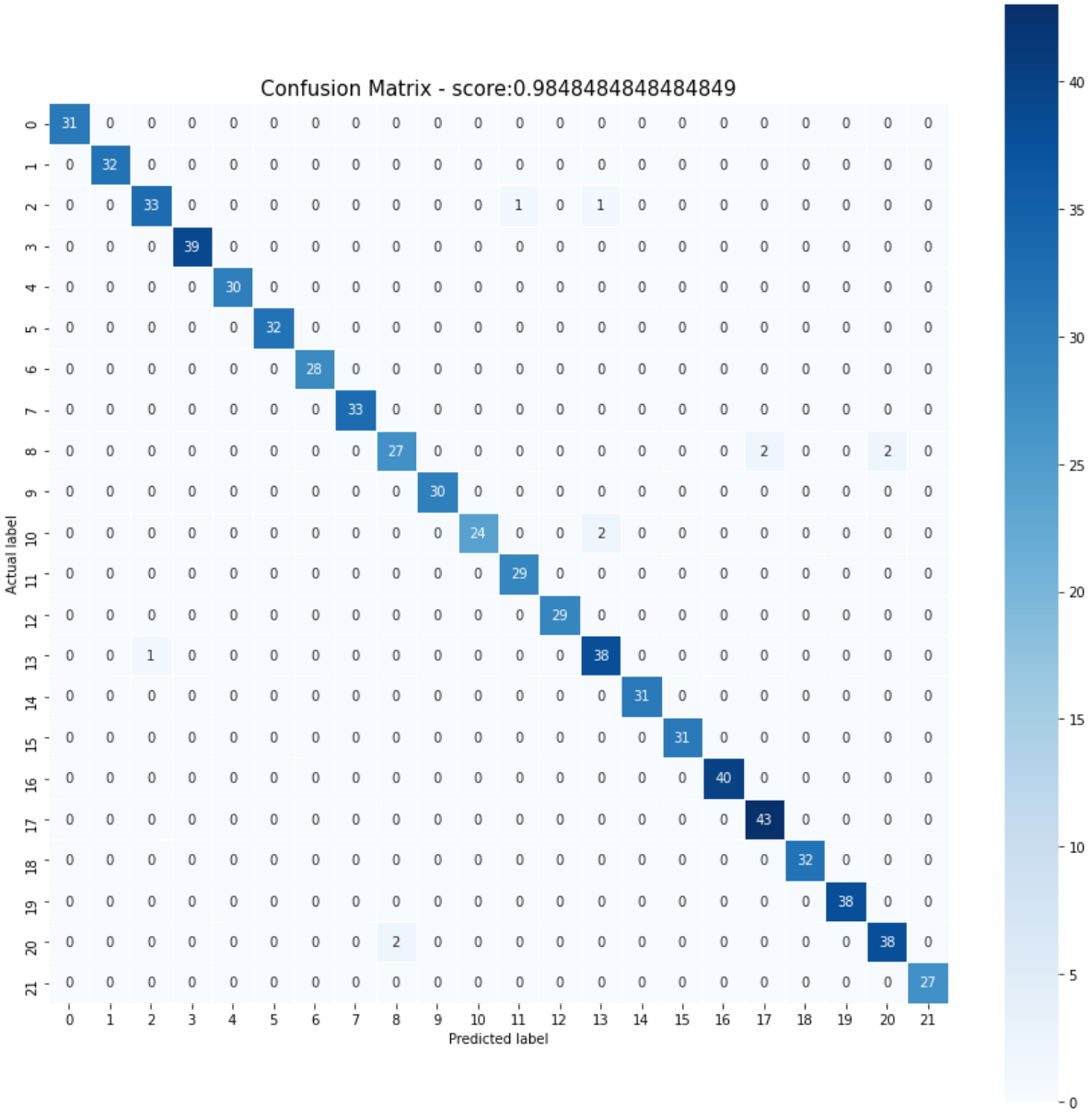
In [82]:

```
from sklearn.metrics import accuracy_score
accuracy=accuracy_score(y_pred, y_test)
print('Decision Tree Model accuracy score: {0:0.4f}'.format(accuracy_score(y_test,
```

```
Decision Tree Model accuracy score: 0.9848
```

In [83]:

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(15,15))
sns.heatmap(cm, annot=True, fmt=".0f", linewidths=.5, square = True, cmap = 'Blues'
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Confusion Matrix - score:'+str(accuracy_score(y_test,y_pred))
plt.title(all_sample_title, size = 15);
plt.show()
```



In [84]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

       apple       1.00      1.00      1.00        31
      banana       1.00      1.00      1.00        32
   blackgram       0.97      0.94      0.96        35
    chickpea       1.00      1.00      1.00        39
     coconut       1.00      1.00      1.00        30
      coffee       1.00      1.00      1.00        32
      cotton       1.00      1.00      1.00        28
      grapes       1.00      1.00      1.00        33
        jute       0.93      0.87      0.90        31
 kidneybeans       1.00      1.00      1.00        30
      lentil       1.00      0.92      0.96        26
       maize       0.97      1.00      0.98        29
       mango       1.00      1.00      1.00        29
   mothbeans       0.93      0.97      0.95        39
    mungbean       1.00      1.00      1.00        31
   muskmelon       1.00      1.00      1.00        31
      orange       1.00      1.00      1.00        40
      papaya       0.96      1.00      0.98        43
  pigeonpeas       1.00      1.00      1.00        32
 pomegranate       1.00      1.00      1.00        38
        rice       0.95      0.95      0.95        40
  watermelon       1.00      1.00      1.00        27

    accuracy                           0.98       726
   macro avg       0.99      0.98      0.99       726
weighted avg       0.98      0.98      0.98       726
```

In [85]:

```
from sklearn.ensemble import RandomForestClassifier
classifier_rf= RandomForestClassifier(n_estimators= 10, criterion="entropy")
classifier_rf.fit(X_train, y_train)
```

Out[85]:

```
RandomForestClassifier(criterion='entropy', n_estimators=10)
```

In [86]:

```
y_pred= classifier_rf.predict(X_test)
```

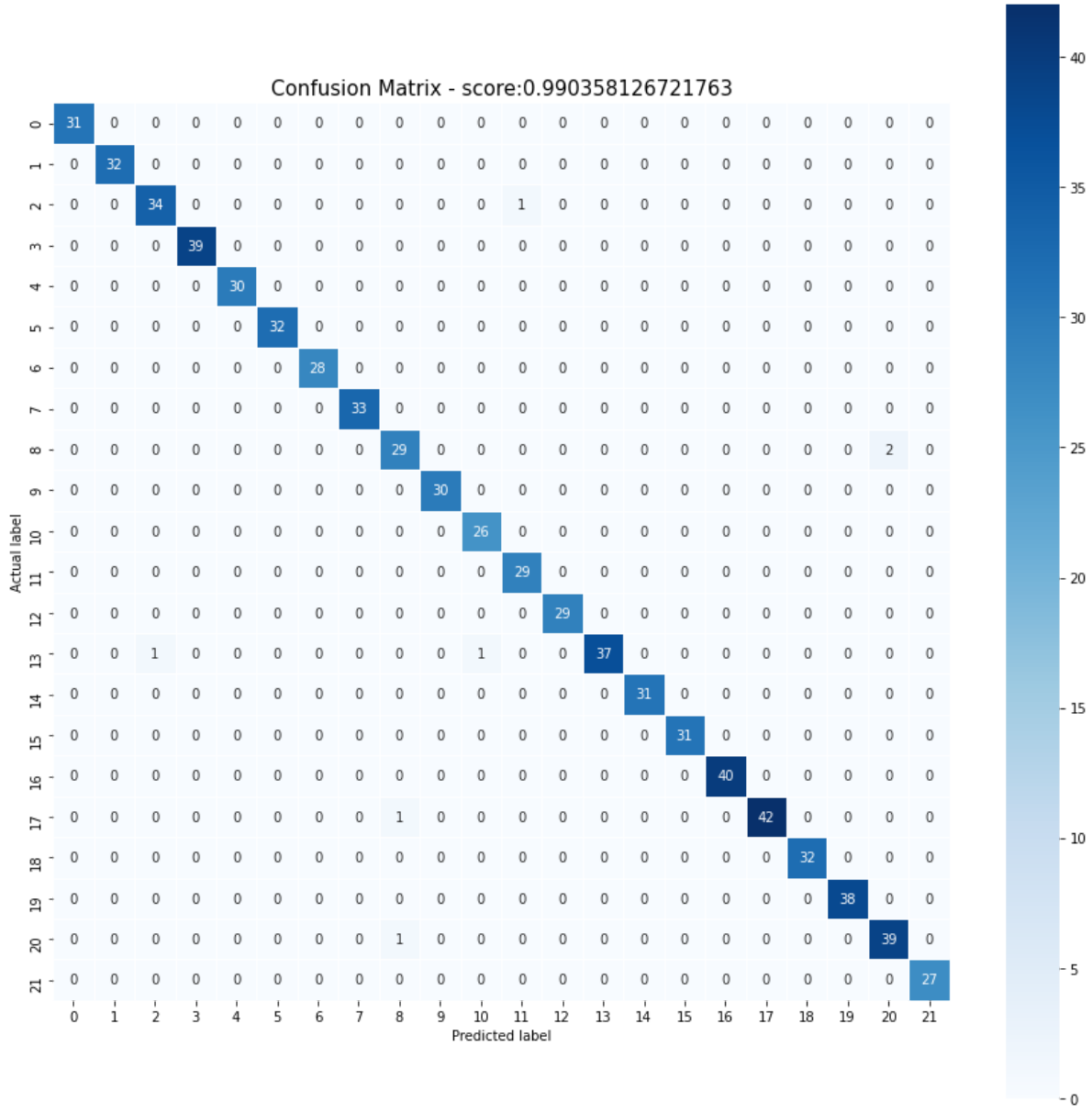
In [87]:

```
from sklearn.metrics import accuracy_score
accuracy=accuracy_score(y_pred, y_test)
print('Random Forest Model accuracy score: {0:0.4f}'.format(accuracy_score(y_test,
```

```
Random Forest Model accuracy score: 0.9904
```

In [88]:

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(15,15))
sns.heatmap(cm, annot=True, fmt=".0f", linewidths=.5, square = True, cmap = 'Blues'
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Confusion Matrix - score:'+str(accuracy_score(y_test,y_pred))
plt.title(all_sample_title, size = 15);
plt.show()
```



In [89]:

```
from sklearn.metrics import classification_report
print(classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

       apple       1.00      1.00      1.00        31
      banana       1.00      1.00      1.00        32
   blackgram       0.97      0.97      0.97        35
    chickpea       1.00      1.00      1.00        39
     coconut       1.00      1.00      1.00        30
      coffee       1.00      1.00      1.00        32
      cotton       1.00      1.00      1.00        28
      grapes       1.00      1.00      1.00        33
        jute       0.94      0.94      0.94        31
 kidneybeans       1.00      1.00      1.00        30
      lentil       0.96      1.00      0.98        26
       maize       0.97      1.00      0.98        29
       mango       1.00      1.00      1.00        29
   mothbeans       1.00      0.95      0.97        39
    mungbean       1.00      1.00      1.00        31
   muskmelon       1.00      1.00      1.00        31
      orange       1.00      1.00      1.00        40
      papaya       1.00      0.98      0.99        43
  pigeonpeas       1.00      1.00      1.00        32
 pomegranate       1.00      1.00      1.00        38
        rice       0.95      0.97      0.96        40
  watermelon       1.00      1.00      1.00        27

    accuracy                           0.99       726
   macro avg       0.99      0.99      0.99       726
weighted avg       0.99      0.99      0.99       726
```

In [79]:

```
from sklearn.linear_model import LogisticRegression
classifier_lr = LogisticRegression(random_state = 0)
classifier_lr.fit(X_train, y_train)
```

Out[79]:

```
LogisticRegression(random_state=0)
```

In [75]:

```
y_pred = classifier_lr.predict(X_test)
```

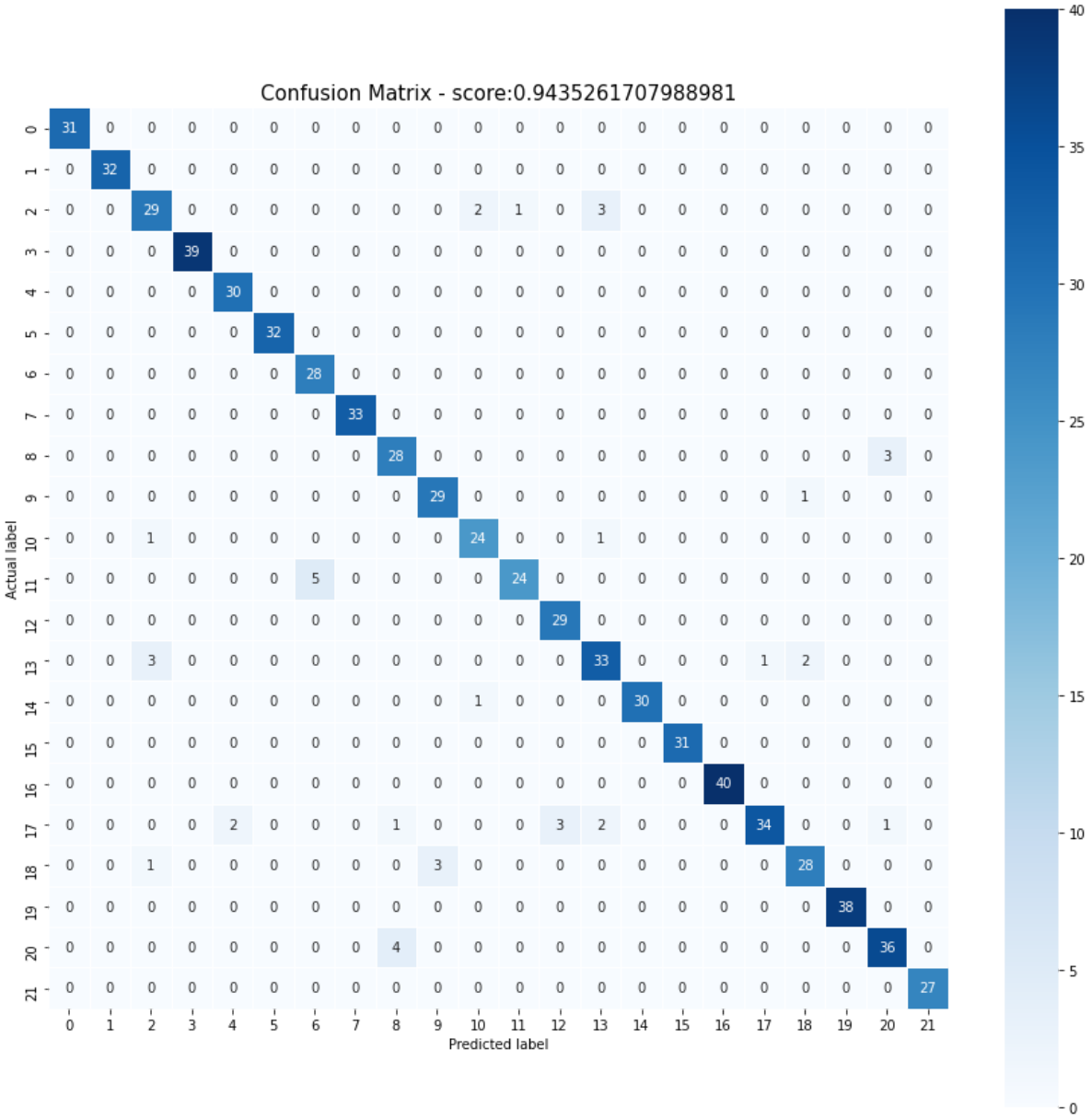
In [76]:

```
from sklearn.metrics import accuracy_score
accuracy=accuracy_score(y_pred, y_test)
print('Logistic Regression Model accuracy score: {0:0.4f}'.format(accuracy_score(y_
```

```
Logistic Regression Model accuracy score: 0.9435
```

In [78]:

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
plt.figure(figsize=(15,15))
sns.heatmap(cm, annot=True, fmt=".0f", linewidths=.5, square = True, cmap = 'Blues'
plt.ylabel('Actual label');
plt.xlabel('Predicted label');
all_sample_title = 'Confusion Matrix - score:'+str(accuracy_score(y_test,y_pred))
plt.title(all_sample_title, size = 15);
plt.show()
```



In [77]:

```
1 from sklearn.metrics import classification_report
2 print(classification_report(y_test, y_pred))
```

```
              precision    recall  f1-score   support

       apple       1.00      1.00      1.00        31
      banana       1.00      1.00      1.00        32
   blackgram       0.85      0.83      0.84        35
    chickpea       1.00      1.00      1.00        39
     coconut       0.94      1.00      0.97        30
      coffee       1.00      1.00      1.00        32
      cotton       0.85      1.00      0.92        28
      grapes       1.00      1.00      1.00        33
        jute       0.85      0.90      0.88        31
 kidneybeans       0.91      0.97      0.94        30
      lentil       0.89      0.92      0.91        26
       maize       0.96      0.83      0.89        29
       mango       0.91      1.00      0.95        29
   mothbeans       0.85      0.85      0.85        39
    mungbean       1.00      0.97      0.98        31
   muskmelon       1.00      1.00      1.00        31
      orange       1.00      1.00      1.00        40
      papaya       0.97      0.79      0.87        43
  pigeonpeas       0.90      0.88      0.89        32
 pomegranate       1.00      1.00      1.00        38
        rice       0.90      0.90      0.90        40
  watermelon       1.00      1.00      1.00        27

    accuracy                           0.94       726
   macro avg       0.94      0.95      0.94       726
weighted avg       0.95      0.94      0.94       726
```