

Notebook for transforming raw cpdata to Mergable data

Filter cpdata.csv to MergeFileCrop.cv

Filter fertilizer.csv to MergerFileFert.csv

```
import pandas as pd
```

```
# Reading the data
```

```
crop_data_path = '../Data-raw/cpdata.csv'  
fertilizer_data_path = '../Data-raw/Fertilizer.csv'
```

```
crop = pd.read_csv(crop_data_path)  
fert = pd.read_csv(fertilizer_data_path)
```

```
crop.head()
```

	temperature	humidity	ph	rainfall	label
0	20.879744	82.002744	6.502985	202.935536	rice
1	21.770462	80.319644	7.038096	226.655537	rice
2	23.004459	82.320763	7.840207	263.964248	rice
3	26.491096	80.158363	6.980401	242.864034	rice
4	20.130175	81.604873	7.628473	262.717340	rice

```
fert.head()
```

	Unnamed: 0	Crop	N	P	K	pH
0	0	Rice	80	40	40	5.5
1	1	Jowar(Sorghum)	80	40	40	5.5
2	2	Barley(JAV)	70	40	45	5.5

	Unnamed: 0	Crop	N	P	K	pH
3	3	Maize	80	40	20	5.5
4	4	Ragi(naachnnii)	50	40	20	5.5

In [5]:

```
# Function for lowering the cases
def change_case(i):
    i = i.replace(" ", "")
    i = i.lower()
    return i
```

In [6]:

```
fert['Crop'] = fert['Crop'].apply(change_case)
crop['label'] = crop['label'].apply(change_case)
```

In [7]:

```
#make some changes in ferttilizer dataset

fert['Crop'] = fert['Crop'].replace('mungbeans','mungbean')
fert['Crop'] = fert['Crop'].replace('lentils(masoordal)','lentil')
fert['Crop'] = fert['Crop'].replace('pigeonpeas(toordal)','pigeonpeas')
fert['Crop'] = fert['Crop'].replace('mothbean(matki)','mothbeans')
fert['Crop'] = fert['Crop'].replace('chickpeas(channa)','chickpea')
```

In [8]:

```
crop.head()
```

Out[8]:

	temperature	humidity	ph	rainfall	label
0	20.879744	82.002744	6.502985	202.935536	rice
1	21.770462	80.319644	7.038096	226.655537	rice
2	23.004459	82.320763	7.840207	263.964248	rice
3	26.491096	80.158363	6.980401	242.864034	rice
4	20.130175	81.604873	7.628473	262.717340	rice

In [9]:

```
crop.tail()
```

Out[9]:

	temperature	humidity	ph	rainfall	label
3095	25.287846	89.636679	6.765095	58.286977	watermelon

	temperature	humidity	ph	rainfall	label
3096	26.638386	84.695469	6.189214	48.324286	watermelon
3097	25.331045	84.305338	6.904242	41.532187	watermelon
3098	26.897502	83.892415	6.463271	43.971937	watermelon
3099	26.986037	89.413849	6.260839	58.548767	watermelon

In [10]:

```
crop_names = crop['label'].unique()
crop_names
```

Out[10]:

```
array(['rice', 'wheat', 'mungbean', 'tea', 'millet', 'maize', 'lentil',
      'jute', 'coffee', 'cotton', 'groundnut', 'peas', 'rubber',
      'sugarcane', 'tobacco', 'kidneybeans', 'mothbeans', 'coconut',
      'blackgram', 'adzukibeans', 'pigeonpeas', 'chickpea', 'banana',
      'grapes', 'apple', 'mango', 'muskmelon', 'orange', 'papaya',
      'pomegranate', 'watermelon'], dtype=object)
```

In [11]:

```
fert.head()
```

Out[11]:

	Unnamed: 0	Crop	N	P	K	pH
0	0	rice	80	40	40	5.5
1	1	jowar(sorghum)	80	40	40	5.5
2	2	barley(jav)	70	40	45	5.5
3	3	maize	80	40	20	5.5
4	4	ragi(naachnnii)	50	40	20	5.5

In [12]:

```
del fert['Unnamed: 0']
```

In [13]:

```
crop_names_from_fert = fert['Crop'].unique()
crop_names_from_fert
```

Out[13]:

```
array(['rice', 'jowar(sorghum)', 'barley(jav)', 'maize',
      'ragi(naachnnii)', 'chickpea', 'frenchbeans(farasbi)',
      'favabeans(papdi-val)', 'limabeans(pavta)', 'clusterbeans(gavar)',
      'soyabean', 'blackeyedbeans(chawli)', 'kidneybeans', 'pigeonpeas',
```

```

'mothbeans', 'mungbean', 'greenpeas', 'horsegram(kulthi)',
'blackgram', 'rapeseed(mohri)', 'corianderseeds', 'mustardseeds',
'sesameseed', 'cuminseeds', 'lentil', 'brinjal', 'beetroot',
'bittergourd', 'bottlegourd', 'capsicum', 'cabbage', 'carrot',
'cauliflower', 'cucumber', 'corianderleaves', 'curryleaves',
'drumstick-moringa', 'chili', 'ladyfinger', 'mushroom', 'onion',
'potato', 'pumpkin', 'radish', 'olive', 'sweetpotato',
'fenugreekleaf(methi)', 'spinach', 'ridgegourd',
'gooseberry(amlam)', 'jambun(syzygiumcumini)',
'ziziphusmauritiana(bor)', 'garciniaindica(kokam)', 'tamarind',
'tapioca(suran)', 'garlic', 'lemon', 'tomato', 'ashgourd',
'pineapple', 'pomegranate', 'banana', 'mango', 'grapes',
'jackfruit', 'guava', 'watermelon', 'muskmelon', 'apricot',
'apple', 'chickoo', 'custardapple', 'dates', 'figs', 'orange',
'papaya', 'aniseed', 'asafoetida', 'bayleaf', 'blackpepper',
'cardamom', 'cinnamon', 'cloves', 'jaiphal(nutmeg)', 'ginger',
'turmeric', 'cashewnuts', 'raisins', 'coconut', 'almondnut',
'arecanut', 'pistachionut', 'lemongrass', 'cotton', 'jute',
'coffee', 'sunflower'], dtype=object)

```

In [14]:

```

for i in crop_names_from_fert:
    print(crop[crop['label'] == i])

    temperature    humidity        ph    rainfall    label
0      20.879744    82.002744    6.502985    202.935536    rice
1      21.770462    80.319644    7.038096    226.655537    rice
2      23.004459    82.320763    7.840207    263.964248    rice
3      26.491096    80.158363    6.980401    242.864034    rice
4      20.130175    81.604873    7.628473    262.717340    rice
..      ...      ...      ...      ...      ...
95     22.683191    83.463583    6.604993    194.265172    rice
96     21.533463    82.140041    6.500343    295.924880    rice
97     21.408658    83.329319    5.935745    287.576694    rice
98     26.543481    84.673536    7.072656    183.622266    rice
99     23.359054    83.595123    5.333323    188.413665    rice

[100 rows x 5 columns]
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []

    temperature    humidity        ph    rainfall    label
500     22.613600    63.690706    5.749914     87.759539    maize
501     26.100184    71.574769    6.931757    102.266244    maize
502     23.558821    71.593514    6.657965     66.719955    maize
503     19.972160    57.682729    6.596061     60.651715    maize
504     18.478913    62.695039    5.970458     65.438354    maize
..      ...      ...      ...      ...      ...
595     18.928519    72.800861    6.158860     82.341629    maize
596     23.305468    63.246480    6.385684    108.760300    maize
597     18.748267    62.498785    6.417820     70.234016    maize
598     19.742133    59.662631    6.381202     65.508614    maize
599     25.730444    70.747393    6.877869     98.737713    maize

```

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

	temperature	humidity	ph	rainfall	label
2100	17.024985	16.988612	7.485996	88.551231	chickpea
2101	19.020613	17.131591	6.920251	79.926981	chickpea
2102	17.887765	15.405897	5.996932	68.549329	chickpea
2103	18.868056	15.658092	6.391174	88.510490	chickpea
2104	18.369526	19.563810	7.152811	79.263577	chickpea
...
2195	17.341502	18.756263	8.861480	67.954543	chickpea
2196	17.437327	14.338474	7.861128	73.092670	chickpea
2197	18.897802	19.761829	7.452671	69.095125	chickpea
2198	18.591908	14.779596	7.168096	89.609825	chickpea
2199	18.315615	15.361435	7.263119	81.787105	chickpea

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

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Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

	temperature	humidity	ph	rainfall	label
1500	17.136928	20.595417	5.685972	128.256862	kidneybeans
1501	19.634743	18.907056	5.759237	106.359818	kidneybeans
1502	22.913502	21.339531	5.873172	109.225556	kidneybeans
1503	16.433403	24.240459	5.926677	140.371781	kidneybeans
1504	22.139747	23.022511	5.955617	76.641283	kidneybeans
...
1595	20.109938	23.223238	5.595032	73.363865	kidneybeans
1596	23.605066	21.905396	5.525905	100.597873	kidneybeans
1597	19.731369	24.894874	5.819404	84.063541	kidneybeans
1598	20.934099	21.189301	5.562202	133.191442	kidneybeans
1599	18.782263	20.247683	5.630665	104.257072	kidneybeans

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2000	36.512684	57.928872	6.031608	122.653969	pigeonpeas
2001	36.891637	62.731782	5.269085	163.726655	pigeonpeas
2002	29.235405	59.389676	5.985793	103.330180	pigeonpeas
2003	27.335349	43.357960	6.091863	142.330368	pigeonpeas
2004	21.064368	55.469859	5.624731	184.622671	pigeonpeas
...
2095	29.892866	66.353751	6.931925	198.140300	pigeonpeas

2096	29.377356	44.822946	6.842744	172.401680	pigeonpeas
2097	29.650529	42.898332	6.876573	186.922605	pigeonpeas
2098	19.542849	66.347773	6.151029	173.110698	pigeonpeas
2099	20.046118	48.939056	4.567446	122.456420	pigeonpeas

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
1600	27.910952	64.709306	3.692864	32.678919	mothbeans
1601	27.322206	51.278688	4.371746	36.503791	mothbeans
1602	28.660242	59.318912	8.399136	36.926297	mothbeans
1603	29.029553	61.093875	8.840656	72.980166	mothbeans
1604	27.780315	54.650300	8.153023	32.050253	mothbeans
...
1695	29.337434	49.003231	8.914075	42.440543	mothbeans
1696	27.965837	61.349001	8.639586	70.104721	mothbeans
1697	24.868040	48.275320	8.621514	63.918765	mothbeans
1698	25.876823	45.963419	5.838509	38.532547	mothbeans
1699	31.019636	49.976752	3.532009	32.812965	mothbeans

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
200	27.433294	87.805077	7.185301	54.733676	mungbean
201	28.334043	80.772760	7.034214	38.797641	mungbean
202	27.014704	84.342627	6.635969	55.296354	mungbean
203	28.174327	81.045548	6.828187	36.357207	mungbean
204	29.878881	87.327612	6.890780	44.752159	mungbean
..
295	28.727527	89.127604	7.069748	58.529743	mungbean
296	27.956397	83.527060	6.921994	43.257268	mungbean
297	28.174587	83.696593	6.770955	37.246465	mungbean
298	28.776535	86.691340	6.983130	56.124432	mungbean
299	28.438097	83.489914	6.267684	52.554700	mungbean

[100 rows x 5 columns]

```
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []
```

	temperature	humidity	ph	rainfall	label
1800	29.484400	63.199153	7.454532	71.890907	blackgram
1801	26.734340	68.139997	7.040056	67.150964	blackgram
1802	26.272744	62.288149	7.418651	70.232076	blackgram
1803	34.036792	67.211138	6.501869	73.235736	blackgram
1804	28.036441	65.066017	6.814411	72.495077	blackgram
...
1895	33.369844	65.677182	6.874142	64.895175	blackgram
1896	31.434506	62.993035	7.760618	64.776515	blackgram
1897	27.716783	63.291034	6.781842	68.565080	blackgram
1898	32.639187	61.300905	7.326980	61.838761	blackgram
1899	32.747739	67.779546	7.453975	63.377844	blackgram

[100 rows x 5 columns]

```
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []
```

[illegible]

[illegible]


```

Columns: [temperature, humidity, ph, rainfall, label]
Index: []
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Index: []
Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []

```

	temperature	humidity	ph	rainfall	label
2900	24.559816	91.635362	5.922936	111.968462	pomegranate
2901	19.656901	89.937010	5.937650	108.045893	pomegranate
2902	18.783596	87.402477	6.804781	102.518476	pomegranate
2903	24.146963	94.511066	6.424671	110.231663	pomegranate
2904	22.445813	89.901470	6.738016	109.390600	pomegranate
...
2995	20.002190	85.836182	7.116539	112.337046	pomegranate
2996	19.851393	89.807323	6.430163	102.818636	pomegranate
2997	21.254336	92.650589	7.159521	106.278467	pomegranate
2998	23.653741	93.326575	6.431266	109.807618	pomegranate
2999	23.884048	86.206138	6.082572	108.312179	pomegranate

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2200	29.367924	76.249001	6.149934	92.828409	banana
2201	27.333690	83.676752	5.849076	101.049479	banana
2202	27.400536	82.962213	6.276800	104.937800	banana
2203	29.315908	80.115857	5.926825	90.109781	banana
2204	26.054330	79.396545	5.519088	113.229737	banana
...
2295	27.359116	84.546250	6.387431	90.812505	banana
2296	28.010680	76.528081	5.891414	103.704078	banana
2297	28.672089	82.207936	5.725419	94.379875	banana
2298	27.345851	78.487383	6.281070	92.155243	banana
2299	29.507046	78.205856	5.507642	98.125658	banana

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2500	29.737700	47.548852	5.954627	90.095869	mango
2501	33.556956	53.729798	4.757115	98.675276	mango
2502	27.003155	47.675254	5.699587	95.851183	mango
2503	33.561502	45.535566	5.977414	95.705259	mango
2504	35.898556	54.259642	6.430139	92.197217	mango
...
2595	31.484517	48.779263	4.525722	93.172220	mango
2596	27.698193	51.415932	5.403908	100.772070	mango
2597	30.412358	52.481006	6.621624	93.923759	mango
2598	32.177520	54.013527	6.207496	91.887661	mango
2599	32.611261	47.749165	5.418475	91.101908	mango

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2300	29.996772	81.541566	6.112306	67.125345	grapes
2301	30.728040	82.426141	6.092242	68.381355	grapes
2302	32.445778	83.885049	5.896343	68.739325	grapes
2303	37.465668	80.659687	6.155261	66.838723	grapes
2304	22.032962	83.743728	5.732454	65.344408	grapes
...
2395	9.851243	80.226317	5.965379	68.428024	grapes
2396	24.972561	82.728287	6.476758	66.700163	grapes
2397	27.237083	82.945733	6.224543	70.425089	grapes
2398	18.706791	83.479529	6.209928	66.596449	grapes
2399	9.949929	82.551390	5.841138	66.008176	grapes

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

	temperature	humidity	ph	rainfall	label
3000	26.473302	80.922544	6.283818	53.657426	watermelon
3001	25.187800	83.446217	6.818261	46.874209	watermelon
3002	25.299547	81.775276	6.376201	57.041471	watermelon
3003	24.746313	88.308663	6.581588	57.958261	watermelon
3004	26.587407	81.325632	6.932740	41.875400	watermelon
...
3095	25.287846	89.636679	6.765095	58.286977	watermelon
3096	26.638386	84.695469	6.189214	48.324286	watermelon
3097	25.331045	84.305338	6.904242	41.532187	watermelon
3098	26.897502	83.892415	6.463271	43.971937	watermelon
3099	26.986037	89.413849	6.260839	58.548767	watermelon

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2600	27.578269	94.118782	6.776533	28.082532	muskmelon
2601	27.820548	93.035552	6.528404	26.324055	muskmelon
2602	29.099104	94.222378	6.750146	22.524973	muskmelon
2603	28.049436	90.831307	6.562833	20.762230	muskmelon
2604	29.916906	94.556956	6.117530	28.160572	muskmelon
...
2695	29.527531	94.574594	6.700338	21.135457	muskmelon
2696	28.504164	93.468065	6.565313	24.200072	muskmelon
2697	28.895786	94.789930	6.286515	23.036250	muskmelon
2698	27.049275	91.382173	6.448062	23.657475	muskmelon
2699	28.960179	91.695322	6.585873	24.745820	muskmelon

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

	temperature	humidity	ph	rainfall	label
2400	22.750888	90.694892	5.521467	110.431786	apple
2401	23.849401	94.348150	6.133221	114.051250	apple
2402	22.608010	94.589006	6.226290	116.039659	apple

2403	21.186674	91.134357	6.321152	122.233323	apple
2404	23.410447	91.699133	5.587906	116.077793	apple
...
2495	23.805938	92.488795	5.889481	119.633555	apple
2496	22.319441	90.851744	5.732758	100.117344	apple
2497	22.144641	93.825674	6.400321	120.631078	apple
2498	23.651676	94.505288	6.496934	115.361127	apple
2499	22.169395	90.271856	6.229499	124.468311	apple

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

	temperature	humidity	ph	rainfall	label
2700	15.781442	92.510777	6.354007	119.035002	orange
2701	26.030973	91.508193	7.511755	101.284774	orange
2702	13.360506	91.356082	7.335158	111.226688	orange
2703	18.879577	92.043045	7.813917	114.665951	orange
2704	29.477417	91.578029	7.129137	111.172750	orange
...
2795	32.717485	90.546083	7.656978	113.328978	orange
2796	25.162966	92.547360	7.105905	114.311720	orange
2797	27.681673	94.473169	7.199106	113.999515	orange
2798	21.350934	90.949297	7.871063	107.086209	orange
2799	11.698946	93.256389	7.566166	103.200599	orange

[100 rows x 5 columns]

	temperature	humidity	ph	rainfall	label
2800	35.214628	91.497251	6.793245	243.074507	papaya
2801	42.394134	90.790281	6.576261	88.466075	papaya
2802	38.419163	91.142204	6.751453	119.265388	papaya
2803	35.332949	92.115086	6.560743	235.613359	papaya
2804	42.923253	90.076005	6.938313	196.240824	papaya
...
2895	40.102077	94.351102	6.979102	149.119999	papaya
2896	38.589545	91.580765	6.825665	102.270823	papaya
2897	41.313301	91.150880	6.617067	239.742755	papaya
2898	37.035519	91.794302	6.551893	188.518142	papaya
2899	23.012402	91.073555	6.598860	208.335798	papaya

[100 rows x 5 columns]

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame

Columns: [temperature, humidity, ph, rainfall, label]

Index: []

Empty DataFrame


```

901    23.095956  84.862757  6.925412  71.295811  cotton
902    23.965635  76.976967  7.633437  90.756167  cotton
903    24.887381  75.621372  6.827355  89.760504  cotton
904    25.362438  83.632761  6.176716  88.436189  cotton
..
995    22.107190  78.583201  6.364730  74.941366  cotton
996    23.038140  76.110215  6.913679  91.496975  cotton
997    24.547953  75.397527  7.766260  63.880799  cotton
998    23.738680  75.775038  7.556064  76.636692  cotton
999    22.318719  83.861300  7.288377  65.357470  cotton

```

[100 rows x 5 columns]

```

      temperature  humidity      ph      rainfall  label
700    25.524690  72.248508  6.002525  151.886997  jute
701    26.591050  82.941641  6.033485  161.247000  jute
702    25.297818  86.887054  7.121934  196.624951  jute
703    25.721009  88.165136  6.207460  175.608670  jute
704    23.584193  72.004608  6.090060  190.424216  jute
..
795    23.874845  86.792613  6.718725  177.514731  jute
796    23.928879  88.071123  6.880205  154.660874  jute
797    24.814412  81.686889  6.861069  190.788639  jute
798    24.447439  82.286484  6.769346  190.968489  jute
799    26.574217  73.819949  7.261581  159.322307  jute

```

[100 rows x 5 columns]

```

      temperature  humidity      ph      rainfall  label
800    26.333780  57.364700  7.261314  191.654941  coffee
801    26.452885  55.322227  7.235070  144.686134  coffee
802    25.708227  52.886671  7.189156  136.732509  coffee
803    24.128325  56.181077  6.431900  147.275782  coffee
804    23.443723  60.395233  6.423211  122.210325  coffee
..
895    26.774637  66.413269  6.780064  177.774507  coffee
896    27.417112  56.636362  6.086922  127.924610  coffee
897    24.131797  67.225123  6.362608  173.322839  coffee
898    26.272418  52.127394  6.758793  127.175293  coffee
899    23.603016  60.396475  6.779833  140.937041  coffee

```

[100 rows x 5 columns]

```

Empty DataFrame
Columns: [temperature, humidity, ph, rainfall, label]
Index: []

```

```
crop['label']
```

```

0          rice
1          rice
2          rice
3          rice
4          rice
...
3095  watermelon
3096  watermelon
3097  watermelon
3098  watermelon

```

In [15]:

Out[15]:

```
3099     watermelon
Name: label, Length: 3100, dtype: object
```

In [16]:

```
extract_labels = []
for i in crop_names_from_fert:
    if i in crop_names:
        extract_labels.append(i)
```

In [17]:

```
# using extract labels on crop to get all the data related to those labels
new_crop = pd.DataFrame(columns = crop.columns)
new_fert = pd.DataFrame(columns = fert.columns)
```

In [18]:

```
for label in extract_labels:
    new_crop = new_crop.append(crop[crop['label'] == label])
```

In [20]:

```
for label in extract_labels:
    new_fert = new_fert.append(fert[fert['Crop'] == label].iloc[0])
```

In [21]:

```
new_crop
```

Out[21]:

	temperature	humidity	ph	rainfall	label
0	20.879744	82.002744	6.502985	202.935536	rice
1	21.770462	80.319644	7.038096	226.655537	rice
2	23.004459	82.320763	7.840207	263.964248	rice
3	26.491096	80.158363	6.980401	242.864034	rice
4	20.130175	81.604873	7.628473	262.717340	rice
...
895	26.774637	66.413269	6.780064	177.774507	coffee
896	27.417112	56.636362	6.086922	127.924610	coffee
897	24.131797	67.225123	6.362608	173.322839	coffee
898	26.272418	52.127394	6.758793	127.175293	coffee
899	23.603016	60.396475	6.779833	140.937041	coffee

2200 rows × 5 columns

```
new_fert
```

In [22]:

Out[22]:

	Crop	N	P	K	pH
0	rice	80	40	40	5.5
3	maize	80	40	20	5.5
5	chickpea	40	60	80	5.5
12	kidneybeans	20	60	20	5.5
13	pigeonpeas	20	60	20	5.5
14	mothbeans	20	40	20	5.5
15	mungbean	20	40	20	5.5
18	blackgram	40	60	20	5.0
24	lentil	20	60	20	5.5
60	pomegranate	20	10	40	5.5
61	banana	100	75	50	6.5
62	mango	20	20	30	5.0
63	grapes	20	125	200	4.0
66	watermelon	100	10	50	5.5
67	muskmelon	100	10	50	5.5
69	apple	20	125	200	6.5
74	orange	20	10	10	4.0

	Crop	N	P	K	pH
75	papaya	50	50	50	6.0
88	coconut	20	10	30	5.0
93	cotton	120	40	20	5.5
94	jute	80	40	40	5.5
95	coffee	100	20	30	5.5

In [23]:

```
new_crop.to_csv('../Data-raw/MergeFileCrop.csv')
new_fert.to_csv('../Data-raw/FertilizerData.csv')
```