

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

df = pd.read_csv('./indian_food.csv')
df.head()

          name
ingredients \
0      Balu shahi           Maida flour, yogurt, oil, sugar
1        Boondi             Gram flour, ghee, sugar
2  Gajar ka halwa       Carrots, milk, sugar, ghee, cashews, raisins
3        Ghevar   Flour, ghee, kewra, milk, clarified butter, su...
4     Gulab jamun Milk powder, plain flour, baking powder, ghee, ...

```

	diet	prep_time	cook_time	flavor_profile	course
state \					
0	vegetarian	45	25	sweet	dessert
Bengal					West
1	vegetarian	80	30	sweet	dessert
Rajasthan					
2	vegetarian	15	60	sweet	dessert
Punjab					
3	vegetarian	15	30	sweet	dessert
Rajasthan					
4	vegetarian	15	40	sweet	dessert
Bengal					West

	region
0	East
1	West
2	North
3	West
4	East

```

df.describe()

    prep_time    cook_time
count  255.000000  255.000000
mean    31.105882  34.529412
std     72.554409  48.265650
min    -1.000000  -1.000000
25%    10.000000  20.000000
50%    10.000000  30.000000

```

```
75%      20.000000  40.000000
max     500.000000  720.000000

df.isna().sum()

name          0
ingredients   0
diet          0
prep_time     0
cook_time     0
flavor_profile 0
course         0
state          0
region         1
dtype: int64

df.loc[df['region'].isna(),'region'] = 'North'

df.loc[df['state']=='-1']

          name                      ingredients
\ 7    Kaju katli        Cashews, ghee, cardamom, sugar
  9    Kheer            Milk, rice, sugar, dried fruits
 10   Laddu           Gram flour, ghee, sugar
 12   Nankhatai      Refined flour, besan, ghee, powdered sugar, yo...
 94   Khichdi       Moong dal, green peas, ginger, tomato, green c...
 96   Kulfi falooda Rose syrup, falooda sev, mixed nuts, saffron, ...
 98   Lauki ki subji Bottle gourd, coconut oil, garam masala, ginge...
 109  Pani puri       Kala chana, mashed potato, boondi, sev, lemon
 111  Papad          Urad dal, sev, lemon juice, chopped tomatoes
 115  Rajma chaval  Red kidney beans, garam masala powder, ginger, ...
 117  Samosa         Potatoes, green peas, garam masala, ginger, dough
 128  Dosa            Chana dal, urad dal, whole urad dal, blend ric...
 130  Idli            Split urad dal, urad dal, idli rice, thick poh...
 144  Masala Dosa   Chana dal, urad dal, potatoes, idli rice, thic...
```

145	Pachadi	Coconut oil, cucumber, curd, curry leaves, mus...
149	Payasam	Rice, cashew nuts, milk, raisins, sugar
154	Rasam	Tomato, curry leaves, garlic, mustard seeds, h...
156	Sambar	Pigeon peas, eggplant, drumsticks, sambar powd...
158	Sevai	Sevai, parboiled rice, steamer
161	Uttapam	Chana dal, urad dal, thick poha, tomato, butter
162	Vada	Urad dal, ginger, curry leaves, green chilies,...
164	Upma	Chana dal, urad dal, ginger, curry leaves, sugar
231	Brown Rice	Brown rice, soy sauce, olive oil
248	Red Rice	Red pepper, red onion, butter, watercress, oli...

	diet	prep_time	cook_time	flavor_profile	course	
state	region					
7	vegetarian	10	20	sweet	dessert	-
1	-1					
9	vegetarian	10	40	sweet	dessert	-
1	-1					
10	vegetarian	10	40	sweet	dessert	-
1	-1					
12	vegetarian	20	30	sweet	dessert	-
1	-1					
94	vegetarian	40	20	spicy	main course	-
1	-1					
96	vegetarian	45	25	sweet	dessert	-
1	-1					
98	vegetarian	10	20	spicy	main course	-
1	-1					
109	vegetarian	15	2	spicy	snack	-
1	-1					
111	vegetarian	5	5	spicy	snack	-
1	-1					
115	vegetarian	15	90	spicy	main course	-
1	North					
117	vegetarian	30	30	spicy	snack	-
1	-1					
128	vegetarian	360	90	spicy	snack	-
1	South					
130	vegetarian	360	90	spicy	snack	-
1	South					

144	vegetarian	360	90	spicy	snack	-
1	South					
145	vegetarian	10	25	-1	main course	-
1	South					
149	vegetarian	15	30	sweet	dessert	-
1	South					
154	vegetarian	10	35	spicy	main course	-
1	South					
156	vegetarian	20	45	spicy	main course	-
1	South					
158	vegetarian	120	30	-1	main course	-
1	South					
161	vegetarian	10	20	spicy	snack	-
1	South					
162	vegetarian	15	20	spicy	snack	-
1	South					
164	vegetarian	10	20	spicy	snack	-
1	-1					
231	vegetarian	15	25	-1	main course	-
1	-1					
248	vegetarian	-1	-1	-1	main course	-
1	-1					

```
x = df.groupby('state')
['ingredients'].apply(list).reset_index(name='ingredients')
```

```
# function to get the unique values from each string
def get_unique_ingred(ingred):
    for i in ingred:
        word = i.lower().split(',')
    return set(word)
```

```
x['ingredients'] = x['ingredients'].apply(get_unique_ingred)
x
```

	state	ingredients
0	-1	{ watercress, butter, red onion, red pepper,...
1	Andhra Pradesh	{ rice flour, green moong beans}
2	Assam	{ glutinous rice, black sesame seeds, gur}
3	Bihar	{ filling, sattu, dough, atta, mustard oil}
4	Chhattisgarh	{chana dal, garam masala powder, arhar dal, ...
5	Goa	{ fennel seeds, ginger powder, grated coconu...}
6	Gujarat	{ sugar, peas, baking soda, grated coconut,...}
7	Haryana	{ gram flour, besan, curry leaves, garam mas...
8	Jammu & Kashmir	{ badam, dried rose petals, cottage cheese, ...}
9	Karnataka	{ black sesame seeds, curry leaves, thin rice...}
10	Kerala	{ masala, sesame oil, whole red beans, tama...
11	Madhya Pradesh	{ arrowroot powder, all purpose flour, milk p...
12	Maharashtra	{ khus khus, gobi, potato, coconut, beans}
13	Manipur	{forbidden black rice, garlic powder, olive ...}
14	NCT of Delhi	{ butter, greek yogurt, chicken, cashew nuts...}

```
15      Nagaland { chillies, axone, water, pork, salt, rice}
16      Odisha  {curd, curry leaves, cooked rice, dry chilli}
17      Punjab { fish fillets, green bell pepper, yogurt, ...}
18      Rajasthan { khus khus, whole wheat flour, dry coconut, ...}
19      Tamil Nadu {cinnamon, meat curry powder, tomato, chick...}
20      Telangana {rose water, almonds, milk, saffron, white...}
21      Tripura { ginger and garlic, onions, boiled pork, ch...}
22      Uttar Pradesh {whole wheat flour, musk melon seeds, edible...}
23      Uttarakhand { coconut, molu leaf, khoa}
24      West Bengal { brinjal, ridge gourd, banana, green beans,...}
```

```
x.drop(0,inplace=True)
```

```
x
```

	state	ingredients
1	Andhra Pradesh	{ rice flour, green moong beans}
2	Assam	{glutinous rice, black sesame seeds, gur}
3	Bihar	{ filling, sattu, dough, atta, mustard oil}
4	Chhattisgarh	{chana dal, garam masala powder, arhar dal, ...}
5	Goa	{ fennel seeds, ginger powder, grated coconu...}
6	Gujarat	{ sugar, peas, baking soda, grated coconut,...}
7	Haryana	{ gram flour, besan, curry leaves, garam mas...
8	Jammu & Kashmir	{ badam, dried rose petals, cottage cheese, ...}
9	Karnataka	{ black sesame seeds, curry leaves, thin rice...}
10	Kerala	{ masala, sesame oil, whole red beans, tama...
11	Madhya Pradesh	{ arrowroot powder, all purpose flour, milk p...
12	Maharashtra	{ khus khus, gobi, potato, coconut, beans}
13	Manipur	{forbidden black rice, garlic powder, olive ...}
14	NCT of Delhi	{ butter, greek yogurt, chicken, cashew nuts...}
15	Nagaland	{ chillies, axone, water, pork, salt, rice}
16	Odisha	{curd, curry leaves, cooked rice, dry chilli}
17	Punjab	{ fish fillets, green bell pepper, yogurt, ...}
18	Rajasthan	{ khus khus, whole wheat flour, dry coconut, ...}
19	Tamil Nadu	{cinnamon, meat curry powder, tomato, chick...}
20	Telangana	{rose water, almonds, milk, saffron, white...}
21	Tripura	{ ginger and garlic, onions, boiled pork, ch...}
22	Uttar Pradesh	{whole wheat flour, musk melon seeds, edible...}
23	Uttarakhand	{ coconut, molu leaf, khoa}
24	West Bengal	{ brinjal, ridge gourd, banana, green beans,...}

```
x['ingredients'] = x['ingredients'].apply(' '.join)
```

```
x
```

	state	ingredients
1	Andhra Pradesh	rice flour green moong beans
2	Assam	glutinous rice black sesame seeds gur
3	Bihar	filling sattu dough atta mustard oil
4	Chhattisgarh	chana dal garam masala powder arhar dal whi...
5	Goa	fennel seeds ginger powder grated coconut ...
6	Gujarat	sugar peas baking soda grated coconut ridg...

```

7          Haryana      gram flour besan  curry leaves garam masala ...
8  Jammu & Kashmir    badam dried rose petals cottage cheese pist...
9       Karnataka      black sesame seeds  curry leaves thin rice fl...
10      Kerala        masala sesame oil whole red beans tamarind...
11  Madhya Pradesh    arrowroot powder all purpose flour milk powd...
12  Maharashtra      khus khus gobi potato coconut beans
13      Manipur        forbidden black rice garlic powder olive oil...
14  NCT of Delhi      butter greek yogurt chicken cashew nuts ga...
15  Nagaland         chillies axone water pork salt rice
16  Odisha           curd curry leaves cooked rice dry chilli
17  Punjab            fish fillets green bell pepper yogurt biry...
18  Rajasthan        khus khus whole wheat flour dry coconut ses...
19  Tamil Nadu       cinnamon meat curry powder tomato chicken c...
20  Telangana         rose water almonds milk saffron white brea...
21  Tripura          ginger and garlic onions boiled pork chillies
22  Uttar Pradesh    whole wheat flour musk melon seeds edible gu...
23  Uttarakhand      coconut molu leaf khoa
24  West Bengal      brinjal ridge gourd banana green beans bit...

```

```

from sklearn.feature_extraction.text import TfidfVectorizer
tfidf = TfidfVectorizer()

# Fit and transform text column to get TF-IDF matrix
vec = tfidf.fit_transform(x['ingredients'])

# Convert sparse matrix to dense matrix and create DataFrame
df_tfidf = pd.DataFrame(vec.todense(),
columns=tfidf.get_feature_names_out())

# Print results
print(df_tfidf)

          all  almonds       and     arhar  arrowroot      atta
axone \
0  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000
1  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000
2  0.000000  0.000000  0.000000  0.000000  0.000000  0.420872
0.000000
3  0.000000  0.000000  0.000000  0.251635  0.000000  0.000000
0.000000
4  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000
5  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000
6  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000
7  0.000000  0.000000  0.000000  0.000000  0.000000  0.000000
0.000000

```

0.000000							
8 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
9 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
10 0.377578	0.000000	0.000000	0.000000	0.377578	0.000000	0.000000	
0.000000							
11 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
12 0.000000	0.332084	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
13 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
14 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.460391							
15 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
16 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
17 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
18 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
19 0.000000	0.336520	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
20 0.000000	0.000000	0.411783	0.000000	0.000000	0.000000	0.000000	
0.000000							
21 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
22 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							
23 0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
0.000000							

	badam	baking	banana	...	sugar	tamarind	thin
tomato \							
0 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							
1 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							
2 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							
3 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							
4 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							
5 0.000000	0.384465	0.000000	...	0.384465	0.000000	0.000000	
0.000000							
6 0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000	
0.000000							

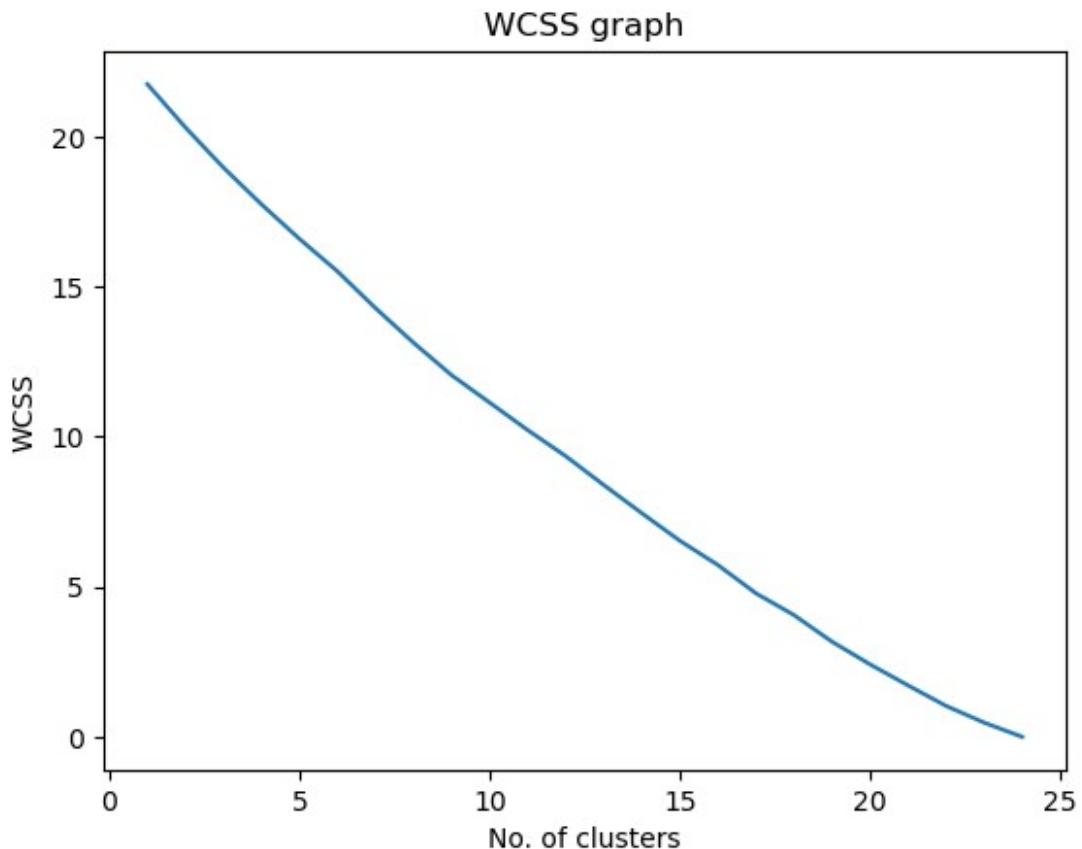

```
14 0.000000 0.407445 0.000000 0.000000 0.000000 0.000000 0.000000
15 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
16 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.305667
17 0.000000 0.000000 0.294096 0.000000 0.270928 0.000000 0.000000
18 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
19 0.000000 0.336520 0.000000 0.336520 0.000000 0.000000 0.000000
20 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
21 0.000000 0.000000 0.261250 0.000000 0.240669 0.000000 0.000000
22 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
23 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000 0.000000
```

[24 rows x 110 columns]

```
from sklearn.cluster import KMeans

wcss = []
for i in range(1,25):
    km = KMeans(n_clusters=i,init='k-means++',random_state=42)
    km.fit(vec)
    wcss.append(km.inertia_)

plt.plot(range(1,25),wcss)
plt.xlabel('No. of clusters')
plt.ylabel('WCSS')
plt.title('WCSS graph')
plt.show()
```



```

no_cluster = 10

kmeans = KMeans(n_clusters=no_cluster, init='k-means++',
random_state=36)
y_pred = kmeans.fit_predict(vec)

x['cluster'] = y_pred
x

      state                               ingredients
\ 1 Andhra Pradesh                   rice flour green moong beans
2   Assam                         glutinous rice black sesame seeds gur
3    Bihar                         filling sattu dough atta mustard oil
4 Chhattisgarh chana dal garam masala powder arhar dal whi...
5     Goa       fennel seeds ginger powder grated coconut ...
6   Gujarat      sugar peas baking soda grated coconut ridg...

```

7 Haryana gram flour besan curry leaves garam masala ...
8 Jammu & Kashmir badam dried rose petals cottage cheese pist...
9 Karnataka black sesame seeds curry leaves thin rice fl...
10 Kerala masala sesame oil whole red beans tamarind...
11 Madhya Pradesh arrowroot powder all purpose flour milk powd...
12 Maharashtra khus khus gobi potato coconut beans
13 Manipur forbidden black rice garlic powder olive oil...
14 NCT of Delhi butter greek yogurt chicken cashew nuts ga...
15 Nagaland chillies axone water pork salt rice
16 Odisha curd curry leaves cooked rice dry chilli
17 Punjab fish fillets green bell pepper yogurt biry...
18 Rajasthan khus khus whole wheat flour dry coconut ses...
19 Tamil Nadu cinnamon meat curry powder tomato chicken c...
20 Telangana rose water almonds milk saffron white brea...
21 Tripura ginger and garlic onions boiled pork chillies
22 Uttar Pradesh whole wheat flour musk melon seeds edible gu...
23 Uttarakhand coconut molu leaf khoa
24 West Bengal brinjal ridge gourd banana green beans bit...

cluster
1 9
2 8
3 1
4 0
5 4
6 3
7 0
8 5
9 8
10 1

```
11      5
12      7
13      4
14      0
15      2
16      8
17      9
18      7
19      0
20      5
21      2
22      6
23      1
24      3
```

```
import geopandas as gpd

out_res = pd.concat([x['state'],x['cluster']], axis=1)
out_res.replace('NCT of Delhi','Delhi',inplace=True)

indian_map = gpd.read_file('./India-State-and-Country-Shapefile-
Updated-Jan-2020-master/India_State_Boundary.shp')

indian_map.replace('Jammu and Kashmir','Jammu & Kashmir',inplace=True)
indian_map.replace('Telengana','Telangana',inplace=True)
indian_map.replace('Tamilnadu','Tamil Nadu',inplace=True)
indian_map.replace('Chhattishgarh','Chhattisgarh',inplace=True)

merged =
indian_map.set_index('State_Name').join(out_res.set_index('state'))
merged

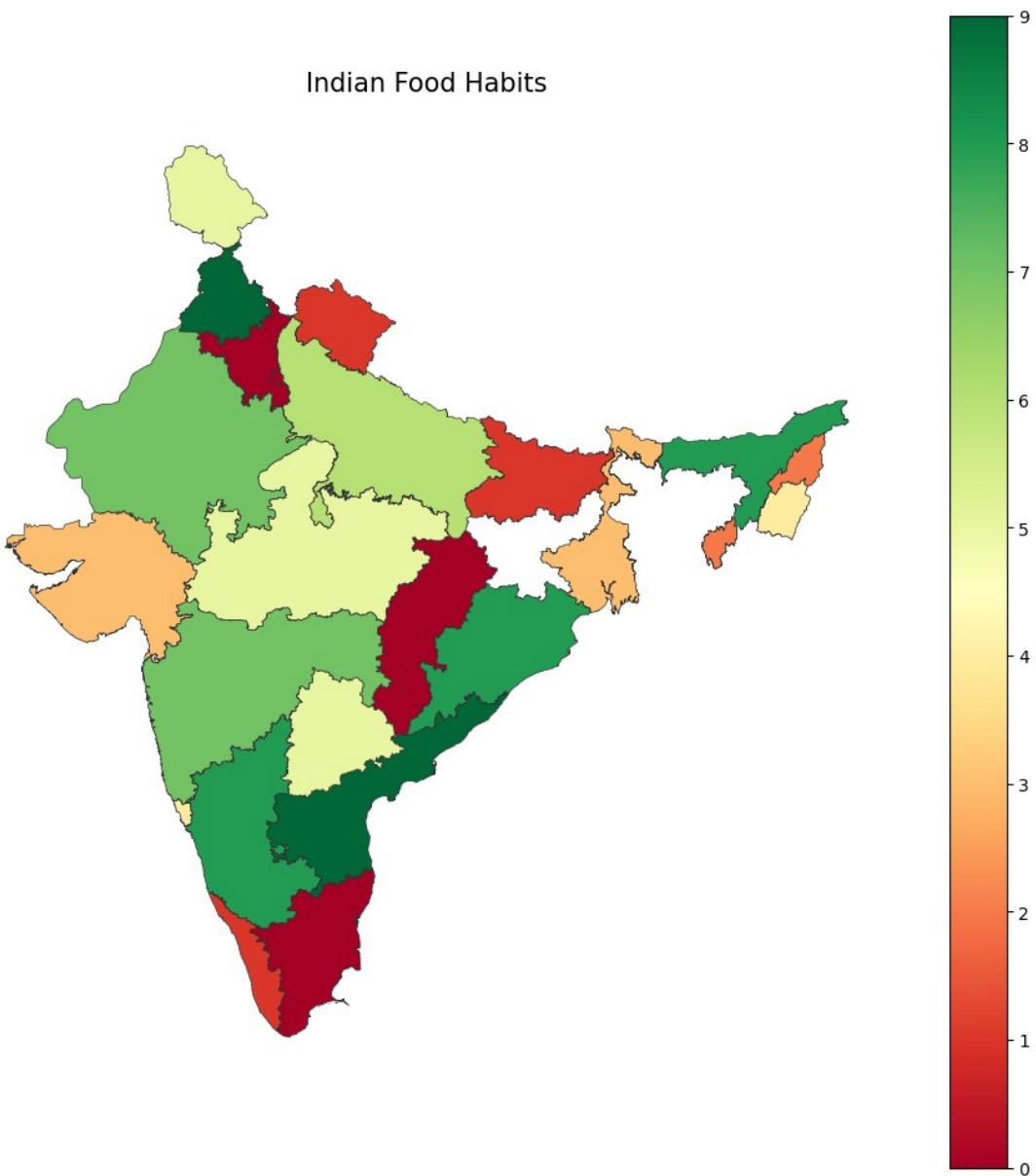
geometry \
Andaman & Nicobar                         MULTIPOLYGON (((10341718.474
1449533.161, 1034...
Andhra Pradesh                                POLYGON ((9426056.496
2174632.352, 9426228.484...
Arunachal Pradesh                            POLYGON ((10696175.277
3434232.650, 10696981.8...
Assam                                         POLYGON ((10380499.251
2872443.723, 10380499.2...
Bihar                                         POLYGON ((9362949.333
3188807.607, 9362966.106...
Chandigarh                                    POLYGON ((8546255.616
3606050.813, 8546315.400...
Chhattisgarh                                  POLYGON ((9275926.808
2765881.317, 9276185.437...
Daman and Diu and Dadra and Nagar Haveli  MULTIPOLYGON (((8122247.822
```

2312434.407, 81223...	
Delhi	POLYGON ((8583390.570
3359116.190, 8583476.212...	
Goa	POLYGON ((8223217.424
1779394.764, 8223279.301...	
Gujarat	POLYGON ((7914780.837
2837315.493, 7915101.603...	
Haryana	POLYGON ((8524318.539
3516490.865, 8524451.392...	
Himachal Pradesh	POLYGON ((8548682.698
3929291.879, 8548760.706...	
Jammu & Kashmir	POLYGON ((8550375.654
3927668.327, 8550332.102...	
Jharkhand	POLYGON ((9762288.285
2772949.712, 9762301.816...	
Karnataka	MULTIPOLYGON (((8608594.474
2090389.205, 86086...	
Kerala	POLYGON ((8347733.191
1436381.747, 8347795.745...	
Ladakh	POLYGON ((8550375.654
3927668.327, 8548619.625...	
Lakshadweep	MULTIPOLYGON (((8135256.290
930182.487, 813526...	
Madhya Pradesh	POLYGON ((8724343.278
3106498.184, 8724579.382...	
Maharashtra	MULTIPOLYGON (((8280974.863
2515416.345, 82809...	
Manipur	POLYGON ((10527945.945
2960789.340, 10528432.7...	
Meghalaya	POLYGON ((10222042.433
3013858.327, 10222165.9...	
Mizoram	POLYGON ((10326423.582
2817021.246, 10326465.4...	
Nagaland	POLYGON ((10596805.532
3126858.281, 10597031.2...	
Odisha	MULTIPOLYGON (((9578537.936
2579790.782, 95786...	
Puducherry	MULTIPOLYGON (((8878474.160
1232399.360, 88784...	
Puducherry	POLYGON ((9150619.186
1888315.624, 9149005.345...	
Punjab	POLYGON ((8442331.679
3830799.529, 8442574.742...	
Rajasthan	POLYGON ((8234599.326
3529026.887, 8234599.327...	
Sikkim	POLYGON ((9864726.992
3265074.341, 9865469.610...	
Tamil Nadu	MULTIPOLYGON (((8939353.702
1513831.235, 89395...	
Telangana	POLYGON ((8720284.876

2259244.214, 8720421.528...	
Tripura	POLYGON ((10260260.337
2818339.599, 10260273.8...	
Uttar Pradesh	POLYGON ((8637489.997
3555885.598, 8637654.287...	
Uttarakhand	POLYGON ((8801802.136
3692833.282, 8802083.049...	
West Bengal	POLYGON ((9800305.279
3151090.311, 9800377.779...	

	cluster
Andaman & Nicobar	NaN
Andhra Pradesh	9.0
Arunachal Pradesh	NaN
Assam	8.0
Bihar	1.0
Chandigarh	NaN
Chhattisgarh	0.0
Daman and Diu and Dadra and Nagar Haveli	NaN
Delhi	0.0
Goa	4.0
Gujarat	3.0
Haryana	0.0
Himachal Pradesh	NaN
Jammu & Kashmir	5.0
Jharkhand	NaN
Karnataka	8.0
Kerala	1.0
Ladakh	NaN
Lakshadweep	NaN
Madhya Pradesh	5.0
Maharashtra	7.0
Manipur	4.0
Meghalaya	NaN
Mizoram	NaN
Nagaland	2.0
Odisha	8.0
Puducherry	NaN
Puducherry	NaN
Punjab	9.0
Rajasthan	7.0
Sikkim	NaN
Tamil Nadu	0.0
Telangana	5.0
Tripura	2.0
Uttar Pradesh	6.0
Uttarakhand	1.0
West Bengal	3.0

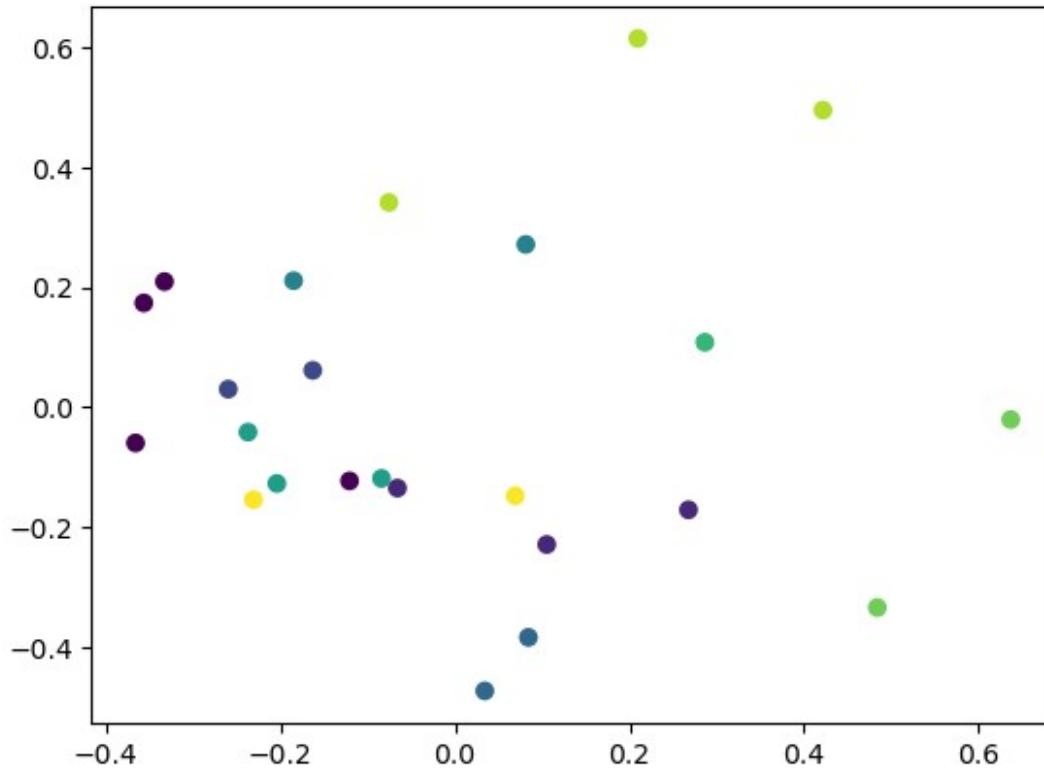
```
fig, ax = plt.subplots(1, figsize=(12, 12))
ax.axis('off')
ax.set_title('Indian Food Habits',
             fontdict={'fontsize': '15', 'fontweight' : '3'})
fig = merged.plot(column='cluster', cmap='RdYlGn', linewidth=0.5,
ax=ax, edgecolor='0.2', legend=True)
```



```
from sklearn.decomposition import PCA

pca = PCA(n_components=10)
pca_matrix = pca.fit_transform(vec.toarray())
labels = kmeans.labels_
# Create scatter plot of data points, using cluster labels as color
```

```
plt.scatter(pca_matrix[:, 0], pca_matrix[:, 1], c=labels)
plt.show()
```



```
# Print the state and cluster for each row
for idx, row in x.iterrows():
    print(f"{row['state']} belongs to cluster {row['cluster']}")

# Visualize the clustering result using a bar chart
cluster_counts = x.groupby('cluster')['state'].count()
plt.bar(cluster_counts.index, cluster_counts.values)
plt.xlabel('Cluster')
plt.ylabel('Number of States')
plt.show()
```

Andhra Pradesh belongs to cluster 9
Assam belongs to cluster 8
Bihar belongs to cluster 1
Chhattisgarh belongs to cluster 0
Goa belongs to cluster 4
Gujarat belongs to cluster 3
Haryana belongs to cluster 0
Jammu & Kashmir belongs to cluster 5
Karnataka belongs to cluster 8
Kerala belongs to cluster 1
Madhya Pradesh belongs to cluster 5
Maharashtra belongs to cluster 7

Manipur belongs to cluster 4
NCT of Delhi belongs to cluster 0
Nagaland belongs to cluster 2
Odisha belongs to cluster 8
Punjab belongs to cluster 9
Rajasthan belongs to cluster 7
Tamil Nadu belongs to cluster 0
Telangana belongs to cluster 5
Tripura belongs to cluster 2
Uttar Pradesh belongs to cluster 6
Uttarakhand belongs to cluster 1
West Bengal belongs to cluster 3

