

```
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	West
Bengal						
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	West
Bengal						

	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, satttu, 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						
8	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						
10	0.377578	0.000000	0.000000	0.000000	0.377578	0.000000
0.000000						
11	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						
13	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
0.460391						
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
0.000000						
17	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						
19	0.000000	0.336520	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						
21	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						
23	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							

7	0.346459	0.000000	0.000000	...	0.000000	0.000000	0.000000
8	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.441473
9	0.000000	0.000000	0.000000	...	0.000000	0.439448	0.000000
10	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
11	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
12	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
14	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
15	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
17	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
19	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
20	0.000000	0.000000	0.000000	...	0.000000	0.000000	0.000000
21	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
23	0.000000	0.000000	0.351162	...	0.000000	0.000000	0.000000

	urad	water	wheat	white	whole	yogurt
0	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
2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
3	0.251635	0.000000	0.202164	0.222696	0.186238	0.000000
4	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
6	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
8	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
9	0.000000	0.000000	0.000000	0.000000	0.325241	0.000000
10	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
11	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
12	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
13	0.000000	0.000000	0.000000	0.000000	0.000000	0.315331

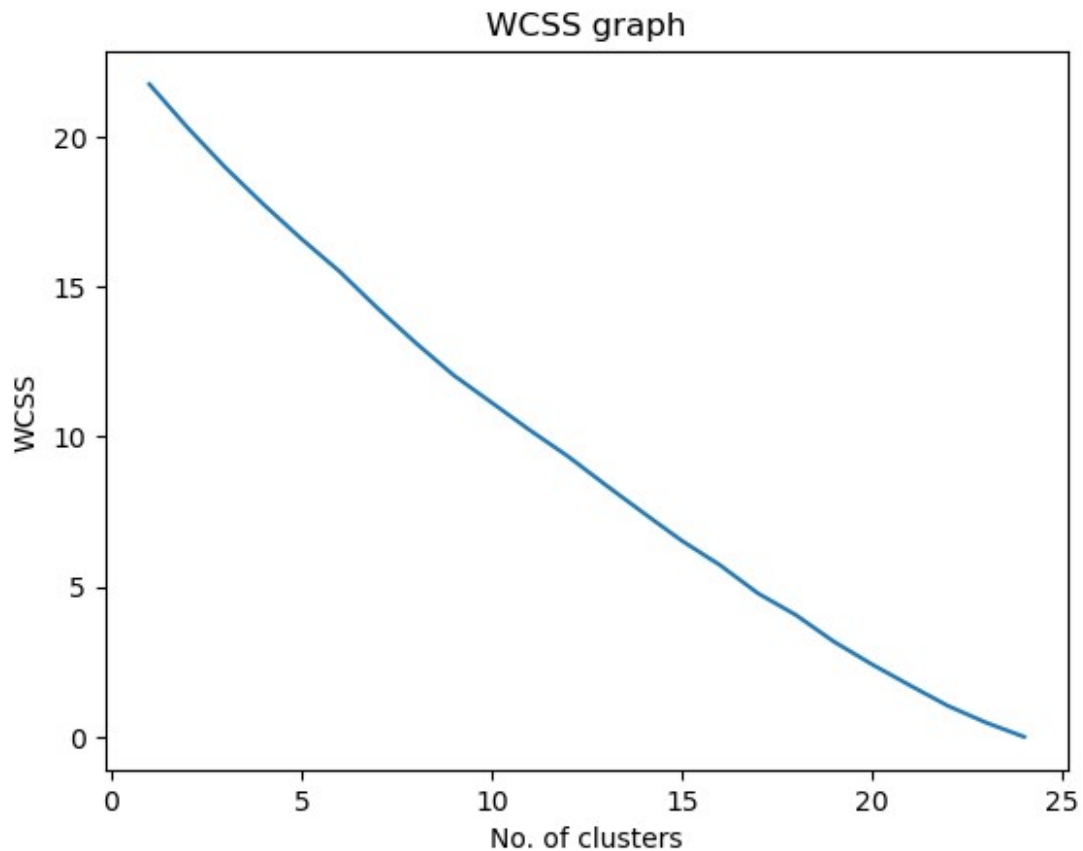
14	0.000000	0.407445	0.000000	0.000000	0.000000	0.000000
15	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
16	0.000000	0.000000	0.000000	0.000000	0.000000	0.305667
17	0.000000	0.000000	0.294096	0.000000	0.270928	0.000000
18	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
19	0.000000	0.336520	0.000000	0.336520	0.000000	0.000000
20	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
22	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

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

2259244.214, 8720421.528...  
Tripura  
2818339.599, 10260273.8...  
Uttar Pradesh  
3555885.598, 8637654.287...  
Uttarakhand  
3692833.282, 8802083.049...  
West Bengal  
3151090.311, 9800377.779...

POLYGON ((10260260.337  
POLYGON ((8637489.997  
POLYGON ((8801802.136  
POLYGON ((9800305.279

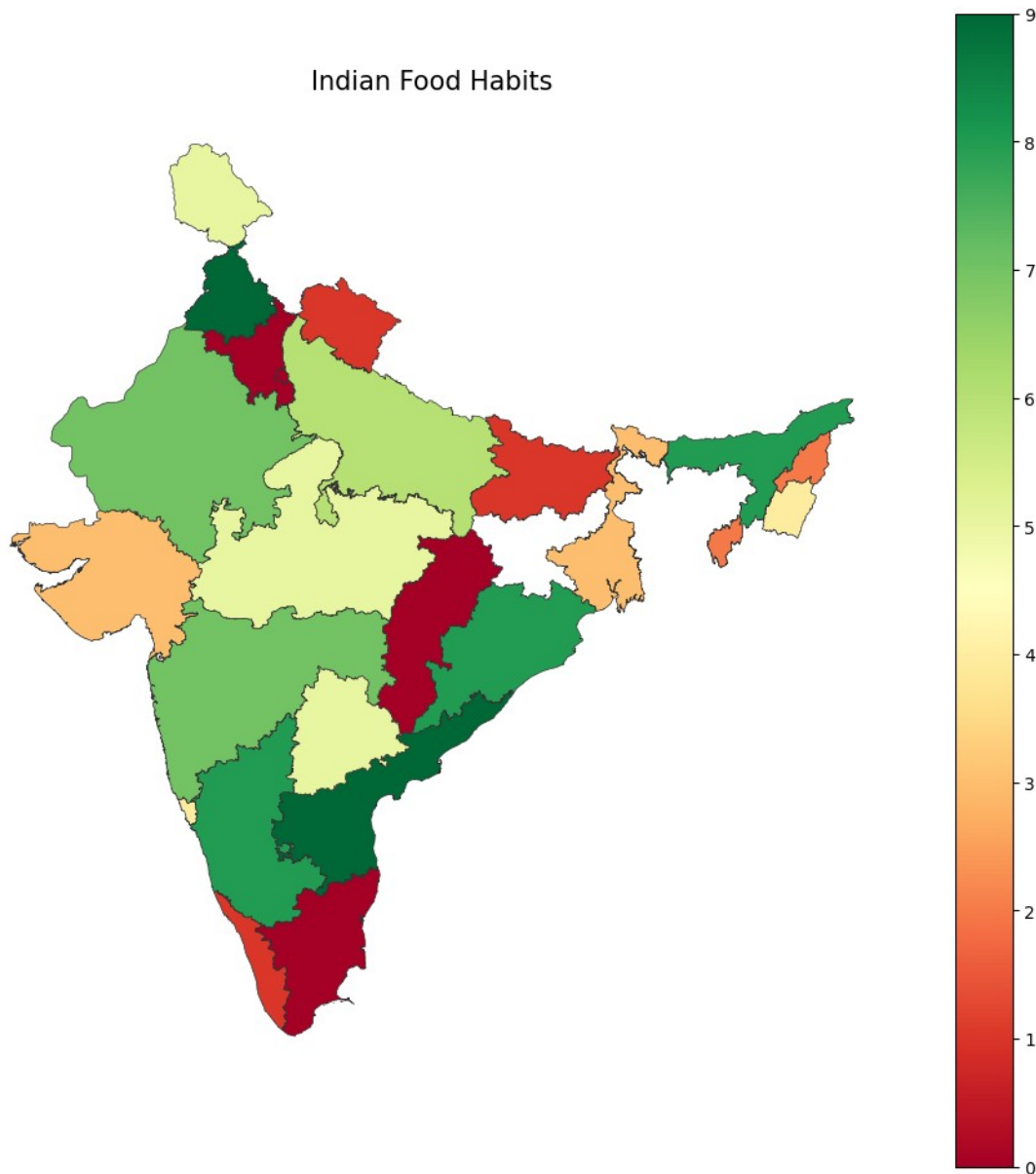
Andaman & Nicobar  
Andhra Pradesh  
Arunachal Pradesh  
Assam  
Bihar  
Chandigarh  
Chhattisgarh  
Daman and Diu and Dadra and Nagar Haveli  
Delhi  
Goa  
Gujarat  
Haryana  
Himachal Pradesh  
Jammu & Kashmir  
Jharkhand  
Karnataka  
Kerala  
Ladakh  
Lakshadweep  
Madhya Pradesh  
Maharashtra  
Manipur  
Meghalaya  
Mizoram  
Nagaland  
Odisha  
Puducherry  
Puducherry  
Punjab  
Rajasthan  
Sikkim  
Tamil Nadu  
Telangana  
Tripura  
Uttar Pradesh  
Uttarakhand  
West Bengal

cluster  
NaN  
9.0  
NaN  
8.0  
1.0  
NaN  
0.0  
NaN  
0.0  
4.0  
3.0  
0.0  
NaN  
5.0  
NaN  
8.0  
1.0  
NaN  
NaN  
5.0  
7.0  
4.0  
NaN  
NaN  
2.0  
8.0  
NaN  
NaN  
9.0  
7.0  
NaN  
0.0  
5.0  
2.0  
6.0  
1.0  
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)

```



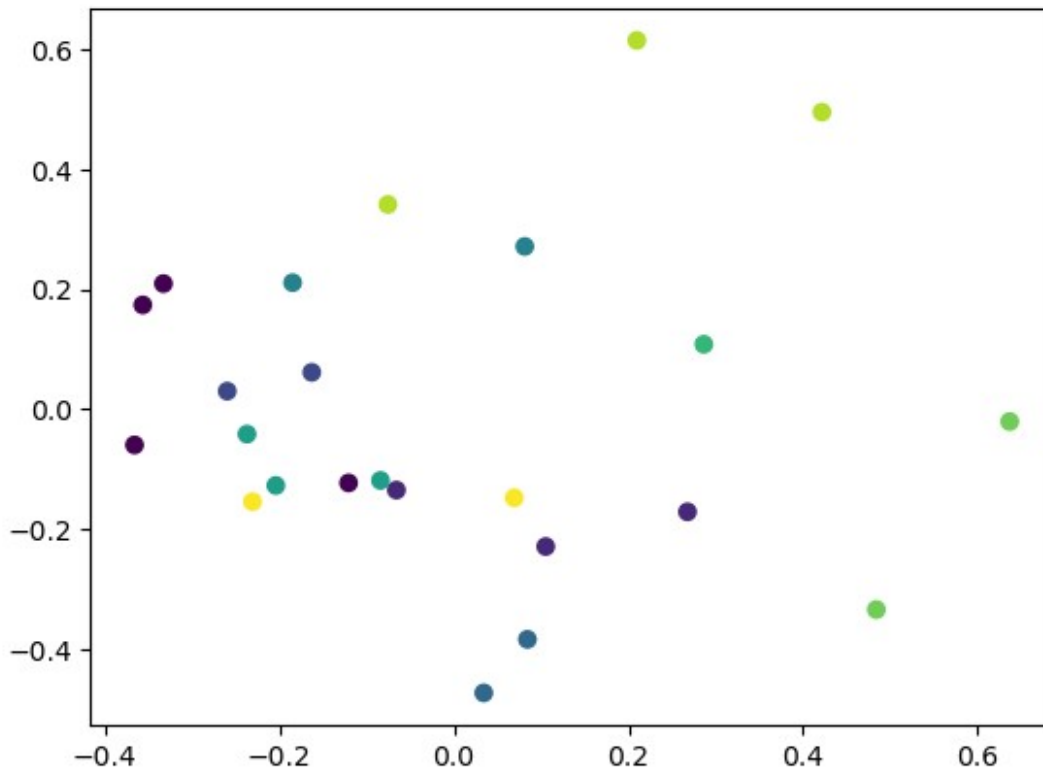
```

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

