The analysis of UK smoking data aims to examine patterns and trends in smoking prevalence. By analyzing factors such as age, gender, and Education Qualification, the study aims to gain insights into smoking behaviors.

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In [1]:	# Importing Neccessary Libraries													
In [2]:	<pre>import pandas as pd import numpy as np import matplotlib.pylab as plt import seaborn as sns</pre>													
In [3]:	# reading csv file													
In [4]:	<pre>df = pd.read_csv('C:/Users/sanjith/Desktop/smoking.csv', encoding="unicode_escape")</pre>													
In [5]:	df.head(10)													
Out[5]:	Unnamed:	gender	age	marital_status	highest_qualification	nationality	ethnicity	gross_income						
	0 1	Male	38	Divorced	No Qualification	British	White	2,600 to 5,200						
	1 2	Female	42	Single	No Qualification	British	White	Under 2,600						
	2 3	Male	40	Married	Degree	English	White	28,600 to 36,400						
	3 4	Female	40	Married	Degree	English	White	10,400 to 15,600						
	4 5	Female	39	Married	GCSE/O Level	British	White	2,600 to 5,200						
	5 6	Female	37	Married	GCSE/O Level	British	White	15,600 to 20,800						
	6 7	Male	53	Married	Degree	British	White	Above 36,400						
	7 8	Male	44	Single	Degree	English	White	10,400 to 15,600						
	8 9	Male	40	Single	GCSE/CSE	English	White	2,600 to 5,200						
	9 10	Female	41	Married	No Qualification	English	White	5,200 to 10,400						
(•						
In [6]:	df.tail(10)													
[-] .		ui.taii(10)												

localhost:8888/nbconvert/html/UK-Smoking Data analysis.ipynb?download=false

Out[6]:	Uni	named: 0	gender	age	marital_status	highest_qualification	nationality	ethnicity	gross_incor		
	1681	1682	Male	53	Single	No Qualification	Scottish	White	20,800 28,6		
	1682	1683	Female	63	Married	No Qualification	British	White	Refus		
	1683	1684	Male	35	Married	No Qualification	Scottish	White	10,400 15,6		
	1684	1685	Male	78	Widowed	No Qualification	Scottish	White	Refus		
	1685	1686	Female	31	Single	Other/Sub Degree	Scottish	White	5,200 10,4		
	1686	1687	Male	22	Single	No Qualification	Scottish	White	2,600 to 5,2		
	1687	1688	Female	49	Divorced	Other/Sub Degree	English	White	2,600 to 5,2		
	1688	1689	Male	45	Married	Other/Sub Degree	Scottish	White	5,200 10,4		
	1689	1690	Female	51	Married	No Qualification	English	White	2,600 to 5,2		
	1690	1691	Male	31	Married	Degree	Scottish	White	10,400 15,6		
									>		
In [7]:	df.shape										
Out[7]:	(1691, 1	3)									
In [8]:	df.colum	ins									
Out[8]:	<pre>Index(['Unnamed: 0', 'gender', 'age', 'marital_status',</pre>										
In [9]:	# Data I	nquiry	Questic	ons :							
	# 1) What is the distribution of smoking prevalence among different genders in the dat # 2) How does smoking behavior vary across different age groups? # 3) Is there any correlation between marital status and smoking habits ? # 4) What is the relation between the highest level of education attained and smoking # 5) Are there any regional differences in smoking prevalence in the UK?										
In [10]:	# Dropping Unneccessary columns										
In [11]:	drop_col	umns =	['ethni	city	', 'gross_inc	ome', 'region', 'a	mt_weekday	s', 'type	','Unnamed		
	<pre>df.drop(drop_columns, inplace=True, axis= 1)</pre>										
In [12]:	df.head(- \									

O Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Ye 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 3]: df= df.rename(columns ={ 'gender': 'Gender' , 'age': 'Age'} 4]: df.head(10) 4]: Gender Age Marital Status Education Nationality Status 0 Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British No 6 Male 53 Married Degree British No 6 Male 53 Married Degree English No 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/OSE English Yes 9 Female 41 Married No Qualification English Yes 9 Female 41 Married No Qualification English Yes 15]: df.isna().sum() 55]: Gender @Age @Marital Status @Age @Age @Age @Age @Age @Age @Age @Age							J	•	
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2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 4 Female 39 Married GCSE/O Level British No 3]: df= df.rename(columns ={ 'gender':'Gender' , 'age':'Age'} 4]: df.head(10) 4]: Gender Age Marital Status Education Nationality Status 0 Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British No 6 Male 53 Married Degree English No 8 Male 40 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 15]: df.isna().sum() 55]: df.isna().sum() 55]: df.isna().sum() 55]: df.isna().sum() 55]: df.isna().sum() 55]: df.isna().sum()		0	Male	38	Divorced	No Qualifica	ation Bri	tish	No
3 Female 40 Married Degree English No. 4 Female 39 Married GCSE/O Level British No. 3]: df= df.rename(columns ={ 'gender':'Gender' , 'age':'Age'} 4]: df.head(10) 4]: Gender Age Marital Status Education Nationality Status 0 Male 38 Divorced No Qualification British No. 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No. 3 Female 40 Married Degree English No. 4 Female 39 Married GCSE/O Level British No. 5 Female 37 Married GCSE/O Level British No. 6 Male 53 Married GCSE/O Level British No. 6 Male 53 Married Degree English No. 8 Male 40 Single Degree English No. 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 15]: df.isna().sum() 5]: Gender 0 Age 0 Marrital Status 0 Education 9 Nationality 0 Status 0 dtype: int64		1	Female	42	Single	No Qualifica	ation Bri	tish	Yes
4 Female 39 Married GCSE/O Level British No df= df.rename(columns ={ 'gender': 'Gender' , 'age': 'Age'} df.head(10) 4]: Gender Age Marital Status Education Nationality Status 0 Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 15]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		2	Male	40	Married	De	gree Eng	ılish	No
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4]: df.head(10) 4]: Gender Age Marital Status Education Nationality Status 0 Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 15]: df.isna().sum() 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 Harried No Qualification English Yes 9 Status 0 Harried No Qualification English Yes 4 Status 0 Harried No Qualification English Yes									
Gender Age Marital Status Education Nationality Status O Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 51 Gender O Age O Martial Status O Education Nationality O Status O O O O Nationality O Status O O O O O Nationality O Status O O O O O O O O O	[13]:		df= d	f.ren	ame(columns :	={ 'gender':'G	ender' , 'a	ge':'Ag	e',
<pre>0 Male 38 Divorced No Qualification British No 1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: df.isna().sum() 5]: df.isna().sum()</pre>	[14]:	df	head(1	0)					
1 Female 42 Single No Qualification British Yes 2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 4 Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64	[14]:		Gender	Age	Marital Status	Education	Nationality	Status	
2 Male 40 Married Degree English No 3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		0	Male	38	Divorced	No Qualification	British	No	
3 Female 40 Married Degree English No 4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 4 Female 39 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: df.isna().sum() 5]: Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		1	Female	42	Single	No Qualification	British	Yes	
4 Female 39 Married GCSE/O Level British No 5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		2	Male	40	Married	Degree	English	No	
5 Female 37 Married GCSE/O Level British No 6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender		3	Female	40	Married	Degree	English	No	
6 Male 53 Married Degree British Yes 7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education English Status 0 Nationality 0 Status 0 dtype: int64		4	Female	39	Married	GCSE/O Level	British	No	
7 Male 44 Single Degree English No 8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		5	Female	37	Married	GCSE/O Level	British	No	
8 Male 40 Single GCSE/CSE English Yes 9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		6	Male	53	Married	Degree	British	Yes	
9 Female 41 Married No Qualification English Yes 5]: df.isna().sum() 5]: Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		7	Male	44	Single	Degree	English	No	
df.isna().sum() Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		8	Male	40	Single	GCSE/CSE	English	Yes	
Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64		9	Female	41	Married	No Qualification	English	Yes	
Gender 0 Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64									
Age 0 Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64	[15]:	df	·isna()	.sum()				
Marital Status 0 Education 0 Nationality 0 Status 0 dtype: int64	[15]:	Age Marital Status							
Nationality 0 Status 0 dtype: int64				0					
dtype: int64									
				t64	0				
	n [16]:				\				

```
False
Out[16]:
          1
                  False
          2
                  False
                  False
                  False
                  . . .
          1686
                  False
          1687
                  False
          1688
                  False
          1689
                   True
          1690
                  False
          Length: 1691, dtype: bool
          df.drop_duplicates()
```

In [17]:

Out[17]:		Gender	Age	Marital Status	Education	Nationality	Status
	0	Male	38	Divorced	No Qualification	British	No
	1	Female	42	Single	No Qualification	British	Yes
	2	Male	40	Married	Degree	English	No
	3	Female	40	Married	Degree	English	No
	4	Female	39	Married	GCSE/O Level	British	No
	•••			***		•••	
	1685	Female	31	Single	Other/Sub Degree	Scottish	No
	1686	Male	22	Single	No Qualification	Scottish	No
	1687	Female	49	Divorced	Other/Sub Degree	English	Yes
	1688	Male	45	Married	Other/Sub Degree	Scottish	No
	1690	Male	31	Married	Degree	Scottish	No

1436 rows × 6 columns

```
# # 1) What is the distribution of smoking prevalence among different genders in the a
In [18]:
In [19]:
           df['Gender'].unique()
         array(['Male', 'Female'], dtype=object)
Out[19]:
In [20]:
          df['Gender'].value_counts()
         Female
                    965
Out[20]:
         Male
                    726
         Name: Gender, dtype: int64
         df['Status'].value_counts()
In [21]:
                 1270
Out[21]:
         Yes
                 421
         Name: Status, dtype: int64
In [22]:
          total_male = len(df[(df['Gender'] == 'Male') & (df['Status'] == 'Yes')])
```

```
total_female = len(df[(df['Gender'] == 'Female') & (df['Status'] == 'Yes')])
print( total_male)
print ( total_female)

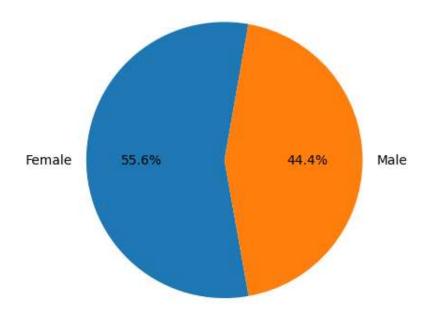
187
234

In [23]: gender_counts = [234, 187]
gender_labels = ['Female', 'Male']

plt.pie(gender_counts, labels=gender_labels, autopct='%1.1f%%', startangle=80)

plt.title('Distribution of Smoking Habits Among Males and Females in United Kingdom')
plt.show()
```

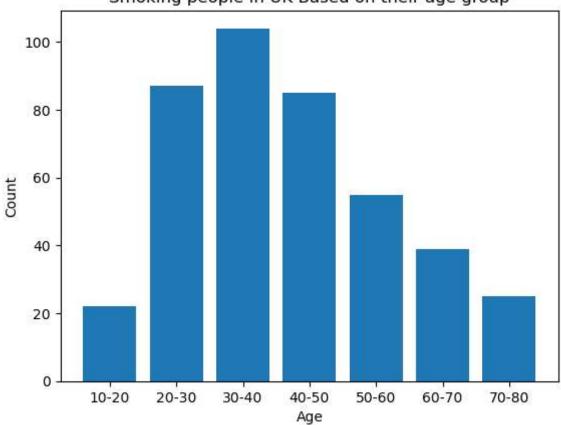
Distribution of Smoking Habits Among Males and Females in United Kingdom



```
In [24]: # 2) How does smoking behavior vary across different age groups?
In [25]: df['Age'] = df['Age'].astype(float).astype(int)
bins = [10, 20, 30, 40, 50, 60, 70, 80]
labels = ['10-20', '20-30', '30-40', '40-50', '50-60','60-70','70-80']
df['Age'] = pd.cut(df['Age'], bins=bins, labels=labels)
In [26]: df.head(5)
```

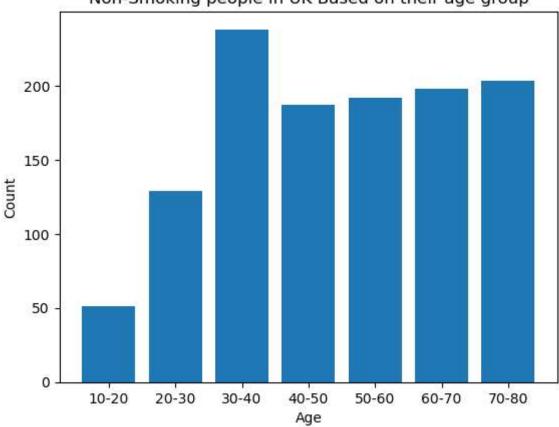
```
Out[26]:
             Gender
                      Age Marital Status
                                              Education Nationality Status
          0
               Male 30-40
                                Divorced No Qualification
                                                             British
                                                                       No
          1
             Female 40-50
                                   Single
                                         No Qualification
                                                             British
                                                                       Yes
               Male 30-40
          2
                                 Married
                                                 Degree
                                                            English
                                                                       No
             Female 30-40
                                 Married
                                                 Degree
          3
                                                            English
                                                                       No
             Female 30-40
                                 Married
                                            GCSE/O Level
                                                             British
                                                                       No
          smokers_by_age_group = df[df['Status'] == 'Yes'].groupby('Age').size()
In [27]:
          print(smokers_by_age_group)
          Age
          10-20
                     22
          20-30
                     87
          30-40
                    104
          40-50
                     85
          50-60
                     55
          60-70
                     39
          70-80
                     25
          dtype: int64
In [28]:
          plt.bar(smokers_by_age_group.index, smokers_by_age_group.values)
          plt.xlabel('Age')
          plt.ylabel('Count')
          plt.title('Smoking people in UK Based on their age group')
          plt.show()
```

Smoking people in UK Based on their age group



```
In [29]:
          non_smokers_by_age_group = df[df['Status'] == 'No'].groupby('Age').size()
          print(non_smokers_by_age_group)
         Age
         10-20
                    51
         20-30
                   129
         30-40
                   238
         40-50
                   187
         50-60
                   192
         60-70
                   198
         70-80
                   203
         dtype: int64
In [30]:
          plt.bar(non_smokers_by_age_group.index, non_smokers_by_age_group.values )
          plt.xlabel('Age')
          plt.ylabel('Count')
          plt.title('Non-Smoking people in UK Based on their age group')
          plt.show()
```

Non-Smoking people in UK Based on their age group



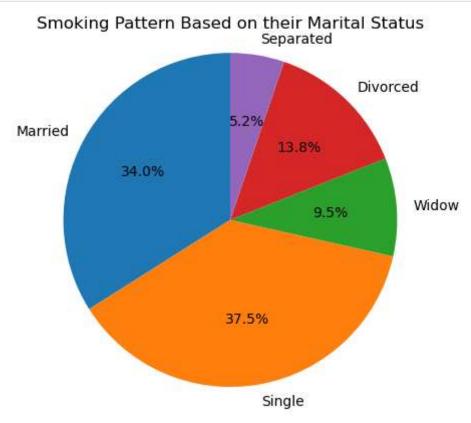
```
# 3) Is there any correlation between marital status and smoking habits?
In [31]:
In [32]:
          df['Marital Status'].value counts()
                       812
         Married
Out[32]:
         Single
                       427
         Widowed
                       223
         Divorced
                       161
         Separated
                        68
         Name: Marital Status, dtype: int64
          smoking_married = len(df[(df['Marital Status'] == 'Married') & (df['Status'] == 'Yes')
In [33]:
          smoking_single = len(df[(df['Marital Status'] == 'Single') & (df['Status'] == 'Yes')])
          smoking widowed = len(df[(df['Marital Status'] == 'Widowed') & (df['Status'] == 'Yes')
          smoking_divorced = len(df[(df['Marital Status'] == 'Divorced') & (df['Status'] == 'Yes
          smoking_separated = len(df[(df['Marital Status'] == 'Separated') & (df['Status'] == 'National Status']
          print (smoking_married)
          print(smoking_single)
          print(smoking widowed)
          print(smoking_divorced)
          print(smoking_separated)
         143
         158
         40
         58
         22
```

```
In [34]:
    labels = ['Married', 'Single', 'Widow', 'Divorced', 'Separated']
    sizes = [smoking_married, smoking_single, smoking_widowed, smoking_divorced, smoking_s
    plt.pie(sizes, labels=labels, autopct='%1.1f%%', startangle=90)

    plt.title('Smoking Pattern Based on their Marital Status')

    plt.axis('equal')

    plt.show()
```



```
# What is the relation between the highest level of education attained and smoking rat
In [35]:
         df['Education'].value_counts()
In [36]:
         No Qualification
                               586
Out[36]:
         GCSE/O Level
                               308
         Degree
                               262
         Other/Sub Degree
                               127
         Higher/Sub Degree
                               125
         A Levels
                               105
         GCSE/CSE
                               102
         ONC/BTEC
         Name: Education, dtype: int64
         df[df['Status'] == 'Yes']['Education'].value_counts()
In [37]:
```

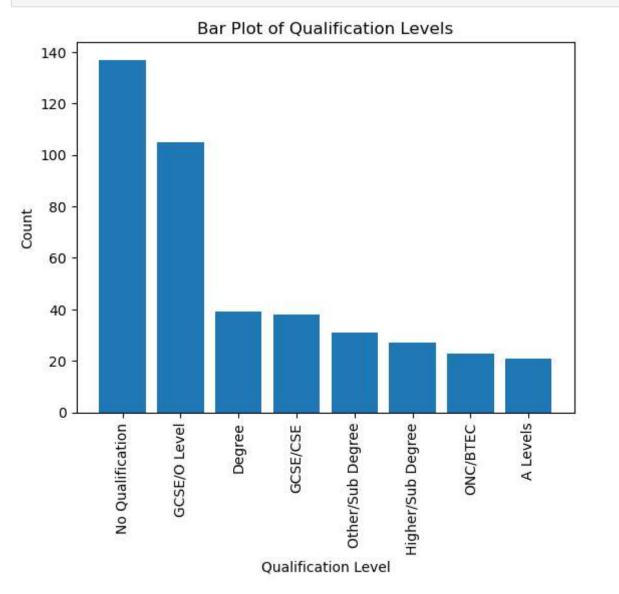
```
No Qualification
                                137
Out[37]:
         GCSE/O Level
                                105
         Degree
                                 39
         GCSE/CSE
                                 38
                                 31
         Other/Sub Degree
                                 27
         Higher/Sub Degree
         ONC/BTEC
                                 23
         A Levels
                                 21
         Name: Education, dtype: int64
```

```
In [38]: import matplotlib.pyplot as plt

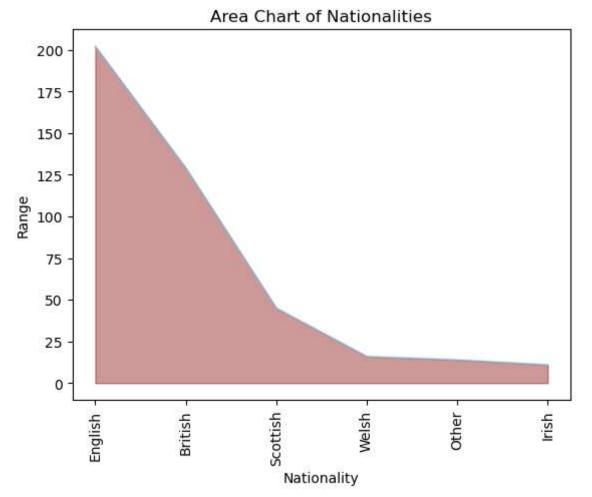
qualification = ['No Qualification', 'GCSE/O Level', 'Degree', 'GCSE/CSE', 'Other/Sub
count = [137, 105, 39, 38, 31, 27, 23, 21]

plt.bar(qualification, count)
plt.xlabel('Qualification Level')
plt.ylabel('Count')
plt.title('Bar Plot of Qualification Levels')
plt.xticks(rotation=90)

plt.show()
```



```
# 5) Are there any regional differences in smoking prevalence in the UK?
In [39]:
          df[df['Status'] == 'Yes']['Nationality'].value_counts()
In [40]:
         English
                      202
Out[40]:
         British
                      129
         Scottish
                       45
         Welsh
                       16
         Other
                       14
         Irish
                       11
         Refused
                        3
         Unknown
                        1
         Name: Nationality, dtype: int64
         nationalities = ['English', 'British', 'Scottish', 'Welsh', 'Other', 'Irish']
In [41]:
          count = [202, 129, 45, 16, 14, 11]
          plt.fill_between(range(len(nationalities)), count, color='maroon', alpha=0.4)
          plt.plot(range(len(nationalities)), count, color='skyblue', alpha=0.6)
          plt.xticks(range(len(nationalities)), nationalities, rotation=90)
          plt.xlabel('Nationality')
          plt.ylabel('Range')
          plt.title('Area Chart of Nationalities')
          plt.show()
```



In []: