Loading Libraries

In [1]:
```
# Importing libraries
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

Loading Dataset

In [2]:
```
# Load the dataset
data = pd.read_csv("Hospital.csv")
```

Display First 5 Rows

In [ ]:
```
# Display first 5 rows
data.head()
```

Out[ ]:

	OrganisationID	OrganisationCode	OrganisationType	SubType	Sector	Organisati
0	17970	NDA07	Hospital	Hospital	Independent Sector	
1	17981	NDA18	Hospital	Hospital	Independent Sector	
2	18102	NLT02	Hospital	Hospital	NHS Sector	
3	18138	NMP01	Hospital	Hospital	Independent Sector	
4	18142	NMV01	Hospital	Hospital	Independent Sector	

5 rows × 22 columns

Display Last 5 Rows

In [ ]:
```
# Display last 5 rows
data.tail()
```

Out[ ]:

	OrganisationID	OrganisationCode	OrganisationType	SubType	Sector	Orga
1206	10956142	U7P1U	Hospital	UNKNOWN	Independent Sector	
1207	10956143	B6Q2K	Hospital	UNKNOWN	Independent Sector	
1208	10956150	K5E9C	Hospital	UNKNOWN	Independent Sector	
1209	10956151	L4S0G	Hospital	UNKNOWN	Independent Sector	
1210	10956153	P3P8S	Hospital	UNKNOWN	Independent Sector	

5 rows × 22 columns

Display Data Types Of Columns

In [ ]:
```
# Display data types of columns
data.dtypes
```

Out[ ]:

	0
OrganisationID	int64
OrganisationCode	object
OrganisationType	object
SubType	object
Sector	object
OrganisationStatus	object
IsPimsManaged	bool
OrganisationName	object
Address1	object
Address2	object
Address3	object
City	object
County	object
Postcode	object
Latitude	float64
Longitude	float64
ParentODSCode	object
ParentName	object
Phone	object
Email	object
Website	object
Fax,,,	object

dtype: object

Display Five Point Summary

In [ ]:
```
# Five point summary
data.describe(include='all')
```

Out[ ]:

	OrganisationID	OrganisationCode	OrganisationType	SubType	Sector	Organisati
count	1.211000e+03	1211	1211	1211	1211	
unique	NaN	1211	1	3	2	
top	NaN	NDA07	Hospital	Hospital	NHS Sector	
freq	NaN	1	1211	961	743	
mean	1.375611e+06	NaN	NaN	NaN	NaN	
std	3.024986e+06	NaN	NaN	NaN	NaN	
min	1.797000e+04	NaN	NaN	NaN	NaN	
25%	4.064900e+04	NaN	NaN	NaN	NaN	
50%	4.311000e+04	NaN	NaN	NaN	NaN	
75%	7.610700e+04	NaN	NaN	NaN	NaN	
max	1.095615e+07	NaN	NaN	NaN	NaN	

11 rows × 22 columns

Removing Outlier

In [3]:
```
from scipy.stats import zscore
z_scores = data.select_dtypes(include=['number']).apply(zscore)
data_no_outliers = data[(z_scores < 3).all(axis=1)]
```

Display Total Number Of Rows & Columns

In [ ]:
```
# Display total number of rows and columns
data.shape
```

Out[ ]:
```
(1211, 22)
```

Finding Duplicates Values

In [ ]:
```
# Finding duplicate values count
data.duplicated().sum()
```

Out[ ]:
```
0
```

Finding Missing Values

In [ ]:
```
# Finding missing values summary
data.isnull().sum()
```

Out[ ]:

	0
OrganisationID	0
OrganisationCode	0
OrganisationType	0
SubType	0
Sector	0
OrganisationStatus	0
IsPimsManaged	0
OrganisationName	0
Address1	328
Address2	484
Address3	1064
City	15
County	238
Postcode	1
Latitude	2
Longitude	2
ParentODSCode	0
ParentName	0
Phone	250
Email	789
Website	358
Fax,,,	2

dtype: int64

Dropping Missing Values

In [ ]:
```
# Dropping Missing Values
data = data.dropna(axis=1)
```

Finding Uniques Values For Each Columns

In [ ]:
```
# Finding unique values for each column
data.nunique()
```

Out[ ]:

	0
OrganisationID	1211
OrganisationCode	1211
OrganisationType	1
SubType	3
Sector	2
OrganisationStatus	1
IsPimsManaged	2
OrganisationName	1189
ParentODSCode	307
ParentName	307

dtype: int64

Finding object columns

In [ ]:
```
# Finding object columns
object_cols = data.select_dtypes(include=['object']).columns.tolist()
object_cols
```

Out[ ]:
```
['OrganisationCode',
 'OrganisationType',
 'SubType',
 'Sector',
 'OrganisationStatus',
 'OrganisationName',
 'ParentODSCode',
 'ParentName']
```

Finding numerical columns

In [ ]:
```
# Finding numerical columns
numerical_cols = data.select_dtypes(include=['number']).columns.tolist()
numerical_cols
```

Out[ ]:
```
['OrganisationID']
```

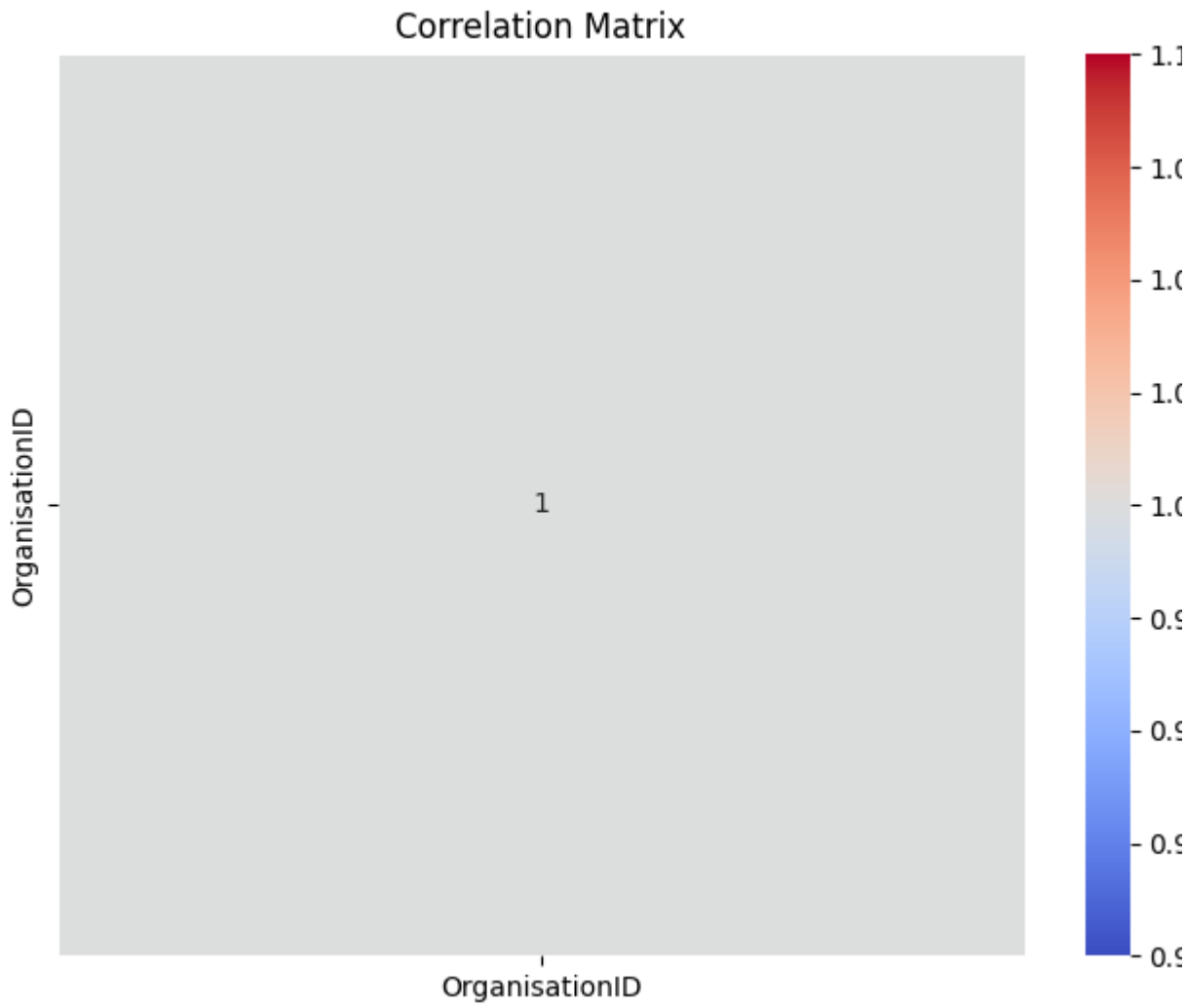
Calculate & Display the Correlation Matrix for t Numeric Columns

In [ ]:
```
# Step 1: Select only numeric columns
numeric_cols = data.select_dtypes(include=['number']).columns

# Step 2: Compute the correlation matrix for numeric columns
correlation_matrix = data[numeric_cols].corr()

# Step 3: Display the correlation matrix
print(correlation_matrix)
```

```
                OrganisationID
OrganisationID             1.0
```

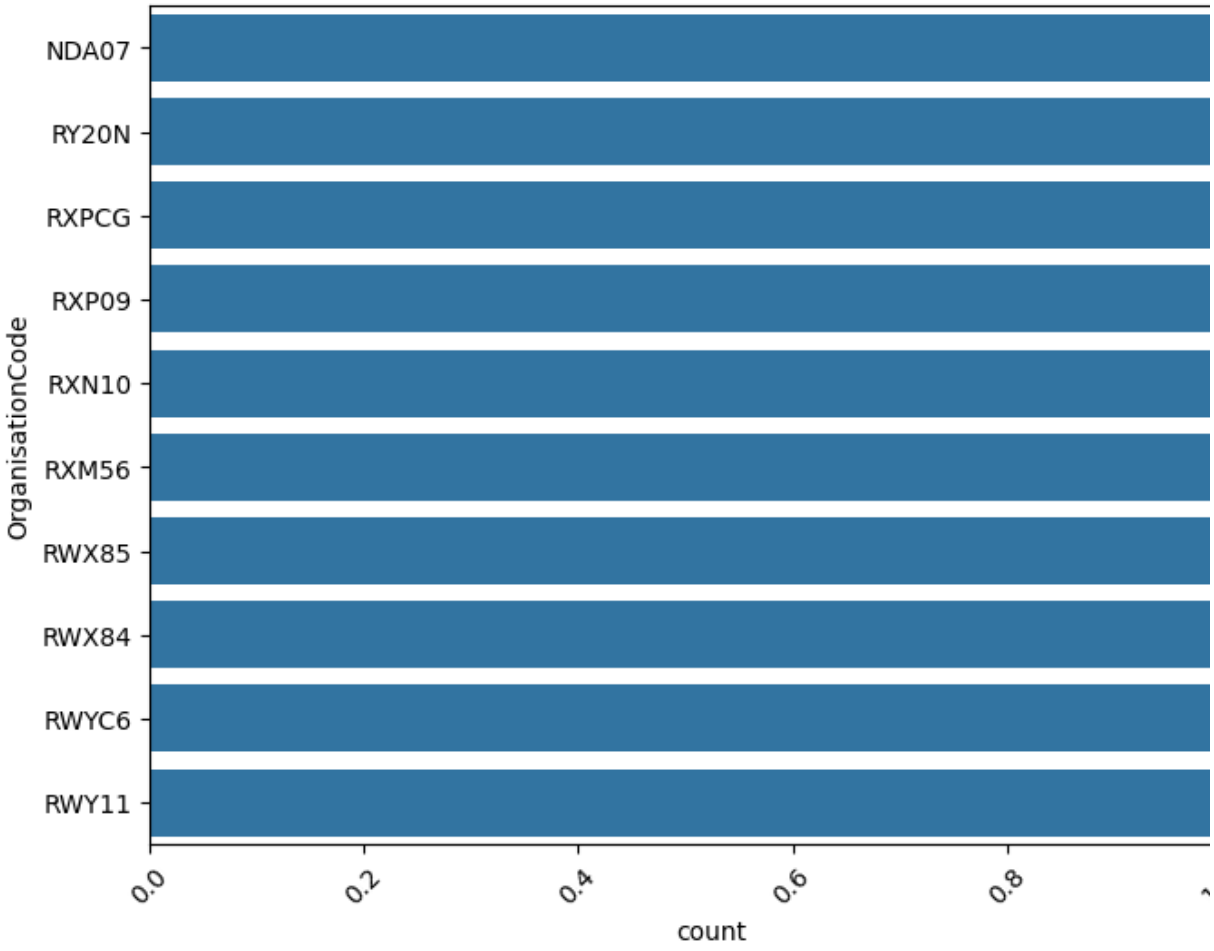
Display Heatmap for Numeric Columns

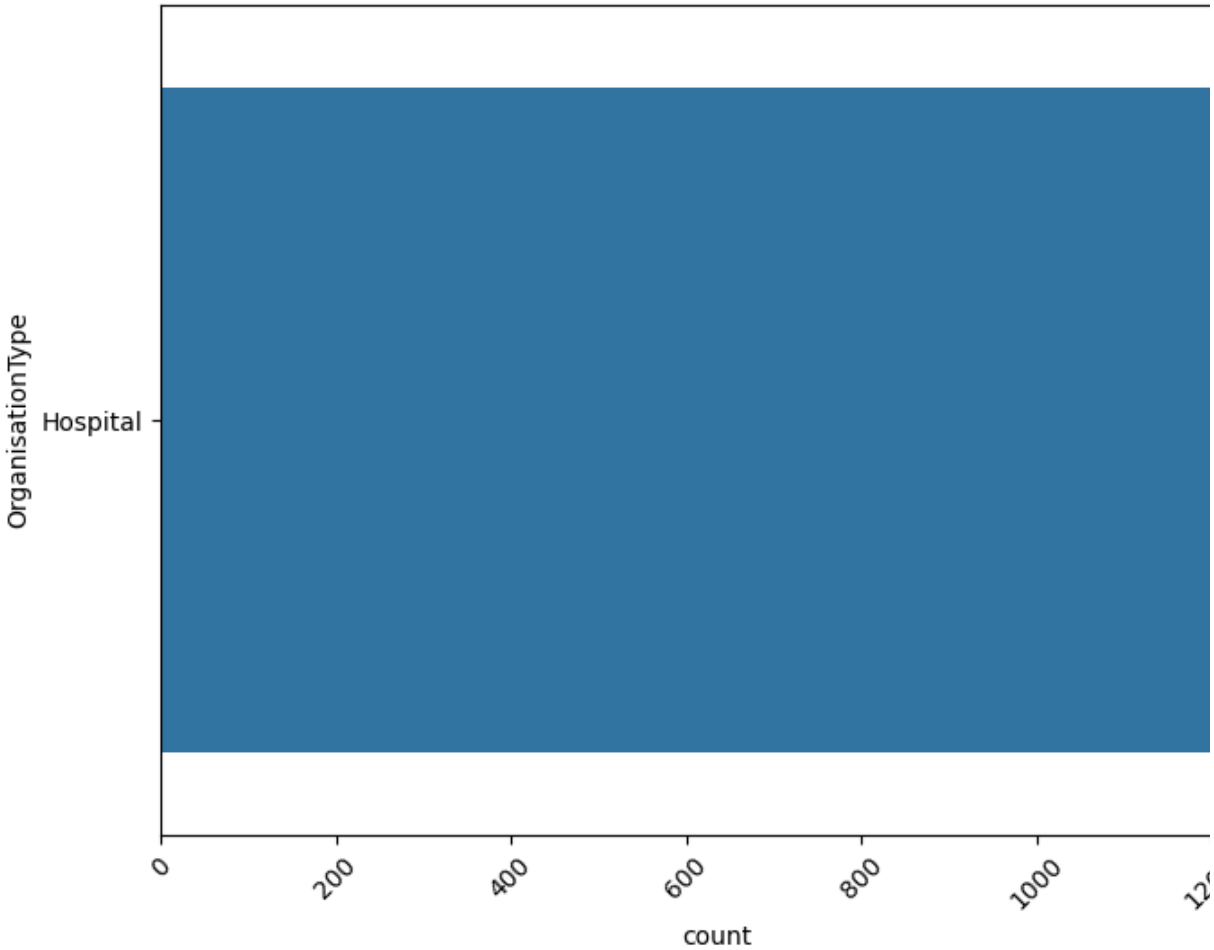
In [ ]:
```
# Select only numeric columns before calculating correlations
numeric_data = data.select_dtypes(include=['number'])

# Calculate the correlation matrix
corr_matrix = numeric_data.corr()

# Plot the correlation matrix
plt.figure(figsize=(8, 6))
sns.heatmap(corr_matrix, annot=True, cmap='coolwarm', linewidths=0.5)
plt.title("Correlation Matrix")
plt.show()
```

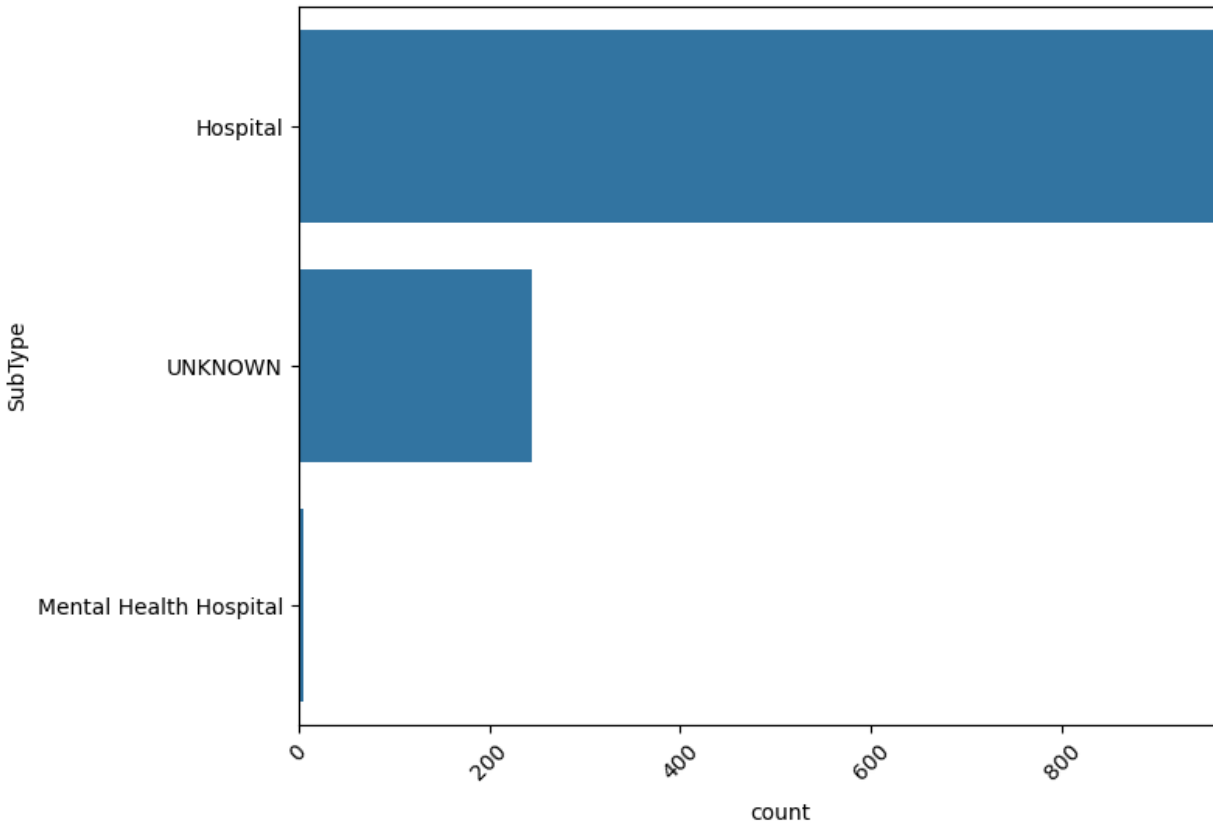


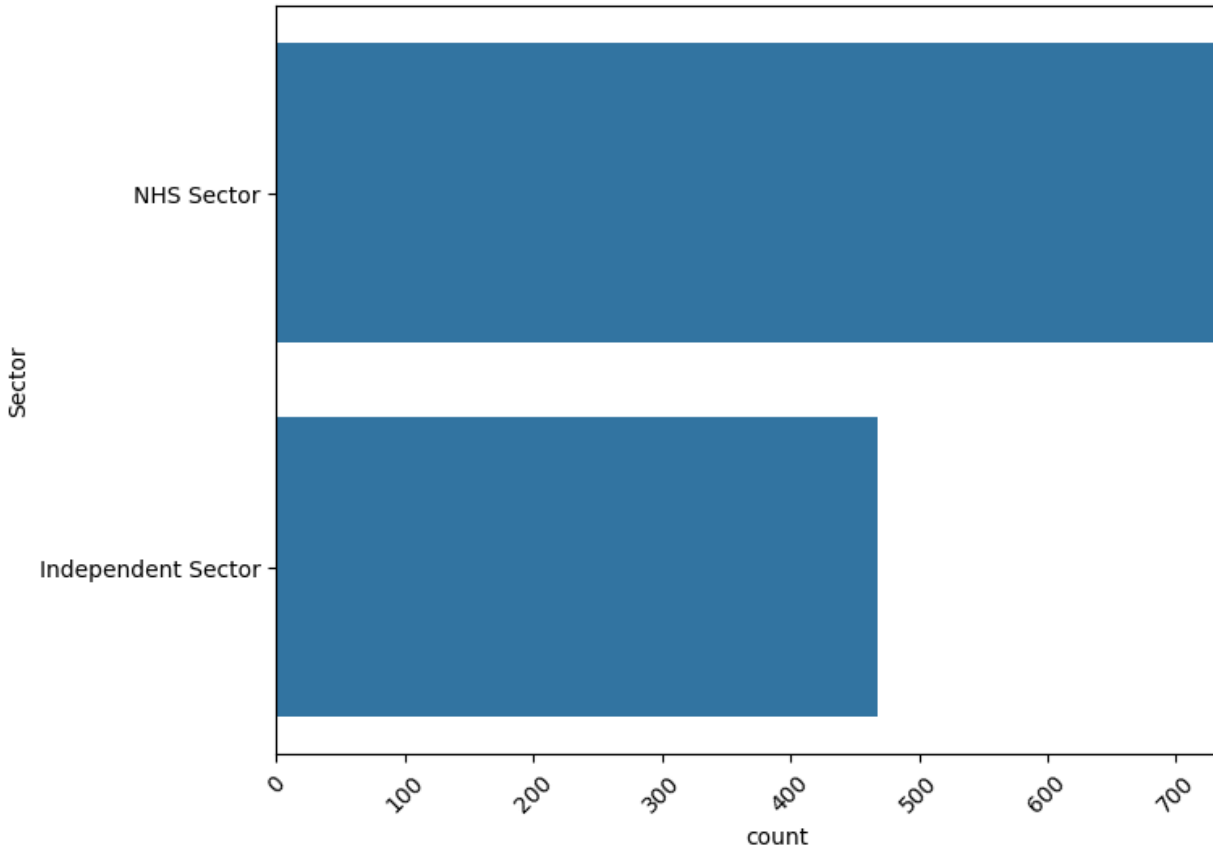
Count Plots For Each Object Type Columns

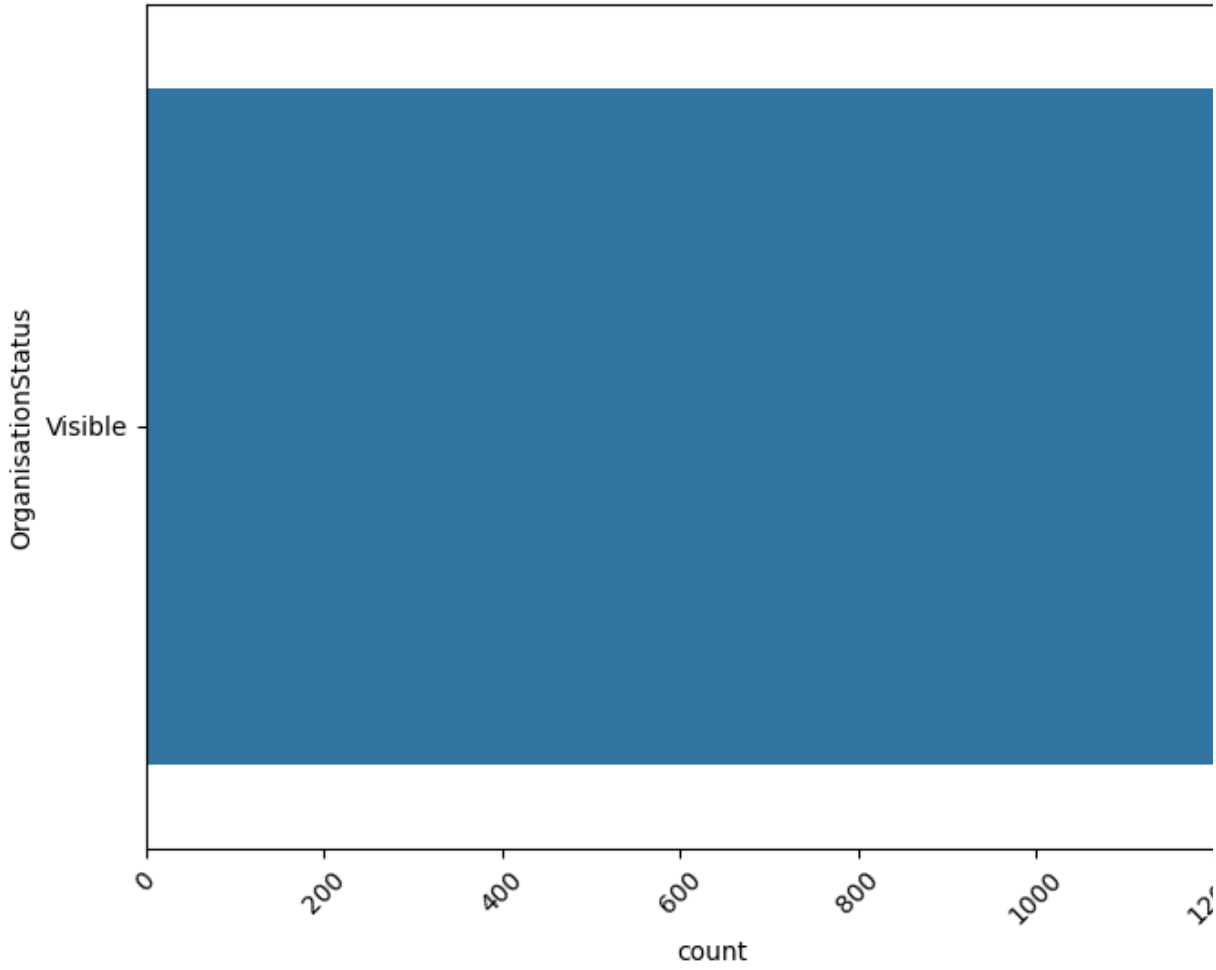
In [ ]:
```
import seaborn as sns
import matplotlib.pyplot as plt

# Assuming 'data' is your DataFrame and 'object_columns' is a list of cate
for col in object_columns:
    plt.figure(figsize=(8, 6))  # Increase plot size
    top_categories = data[col].value_counts().nlargest(10).index  # Limit
    sns.countplot(y=col, data=data, order=top_categories)  # Use order fo
    plt.xticks(rotation=45)  # Rotate x-axis labels for better readabilit
    plt.show()
```

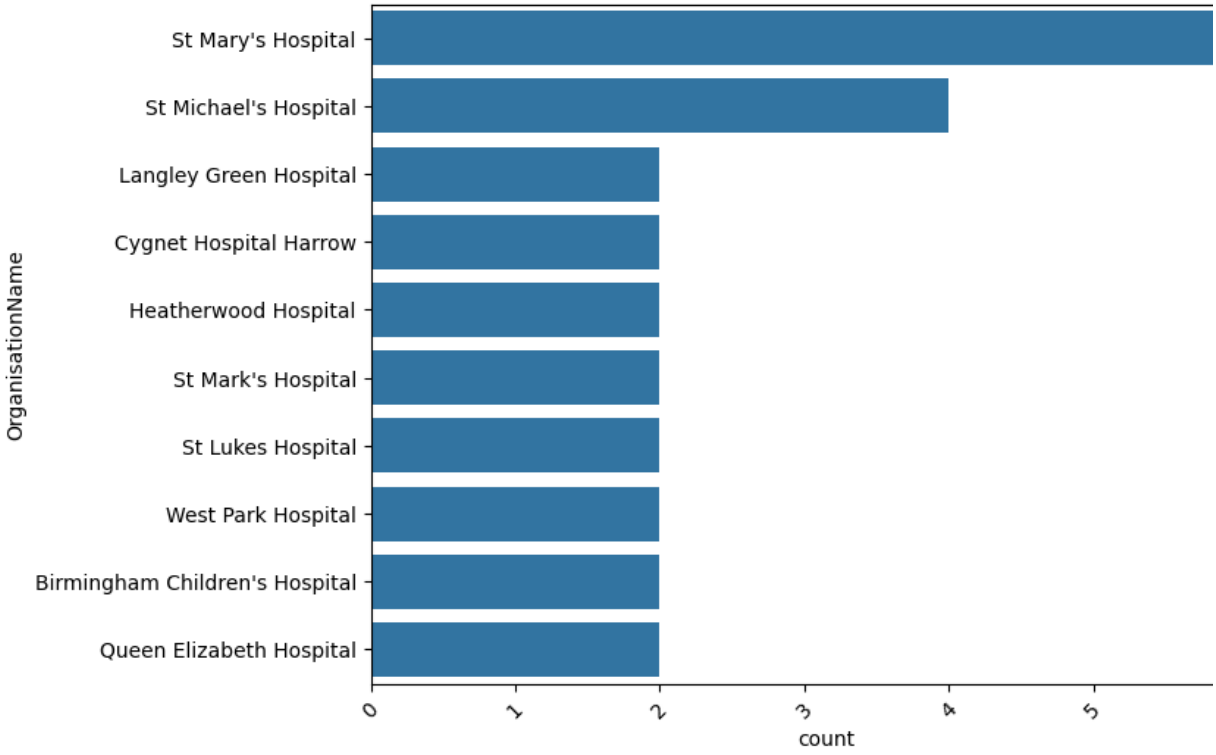

NDA07
RY20N
RXPCG
RXP09
RXN10
RXM56
RWX85
RWX84
RWYC6
RWY11
OrganisationCode
0.0
0.2
0.4
0.6
0.8
count

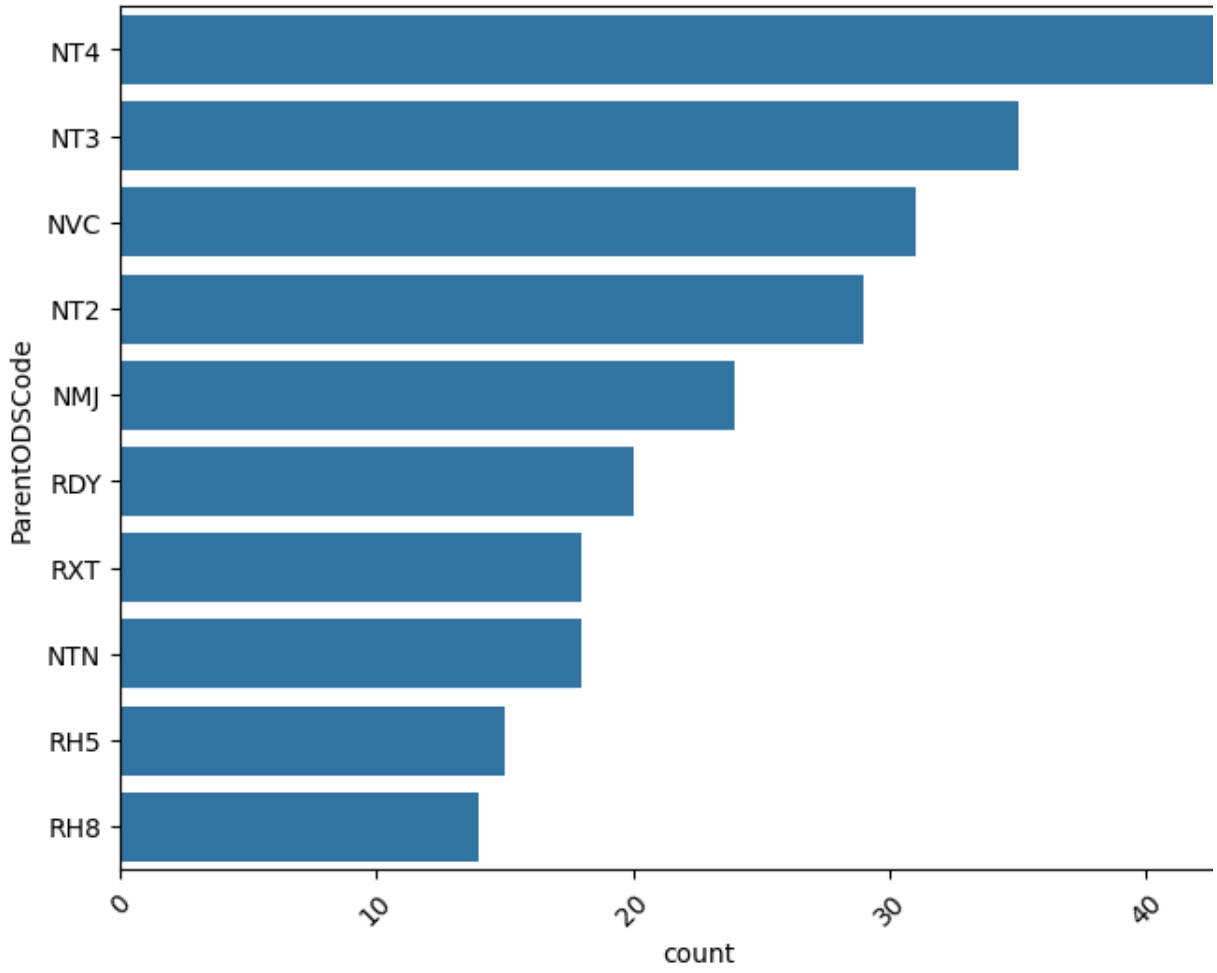

Hospital
OrganisationType
0
200
400
600
800
1000
count

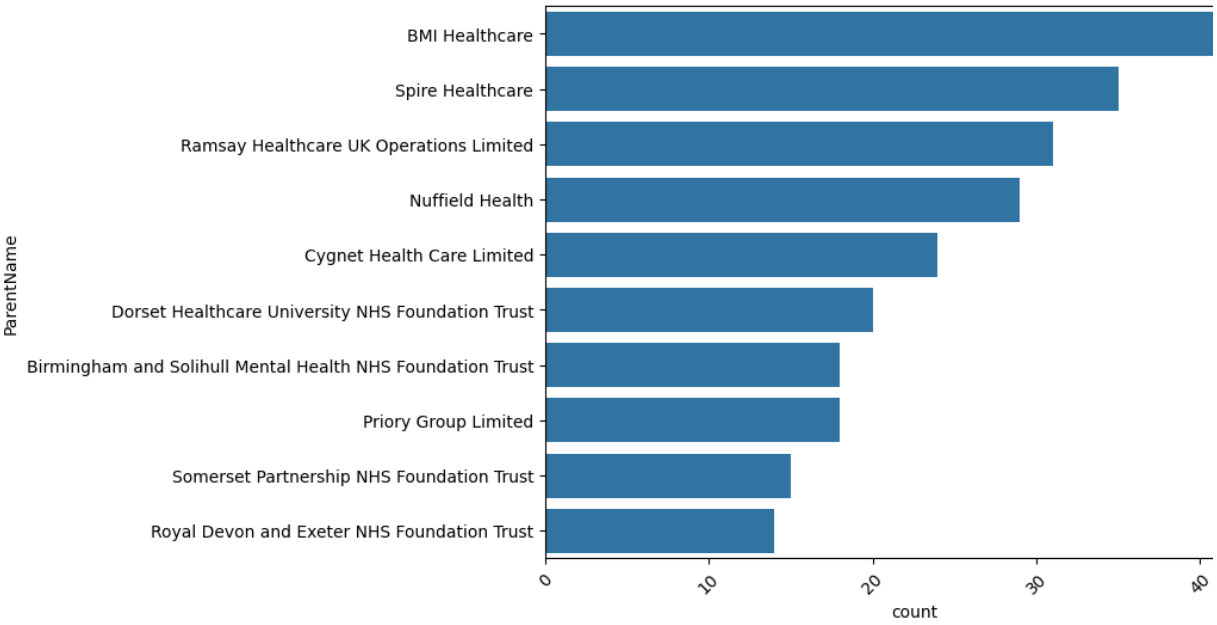

Hospital
UNKNOWN
Mental Health Hospital
SubType
0
200
400
600
800
count


NHS Sector
Independent Sector
Sector
0
100
200
300
400
500
600
700
count


OrganisationStatus
Visible
0
200
400
600
800
1000
count

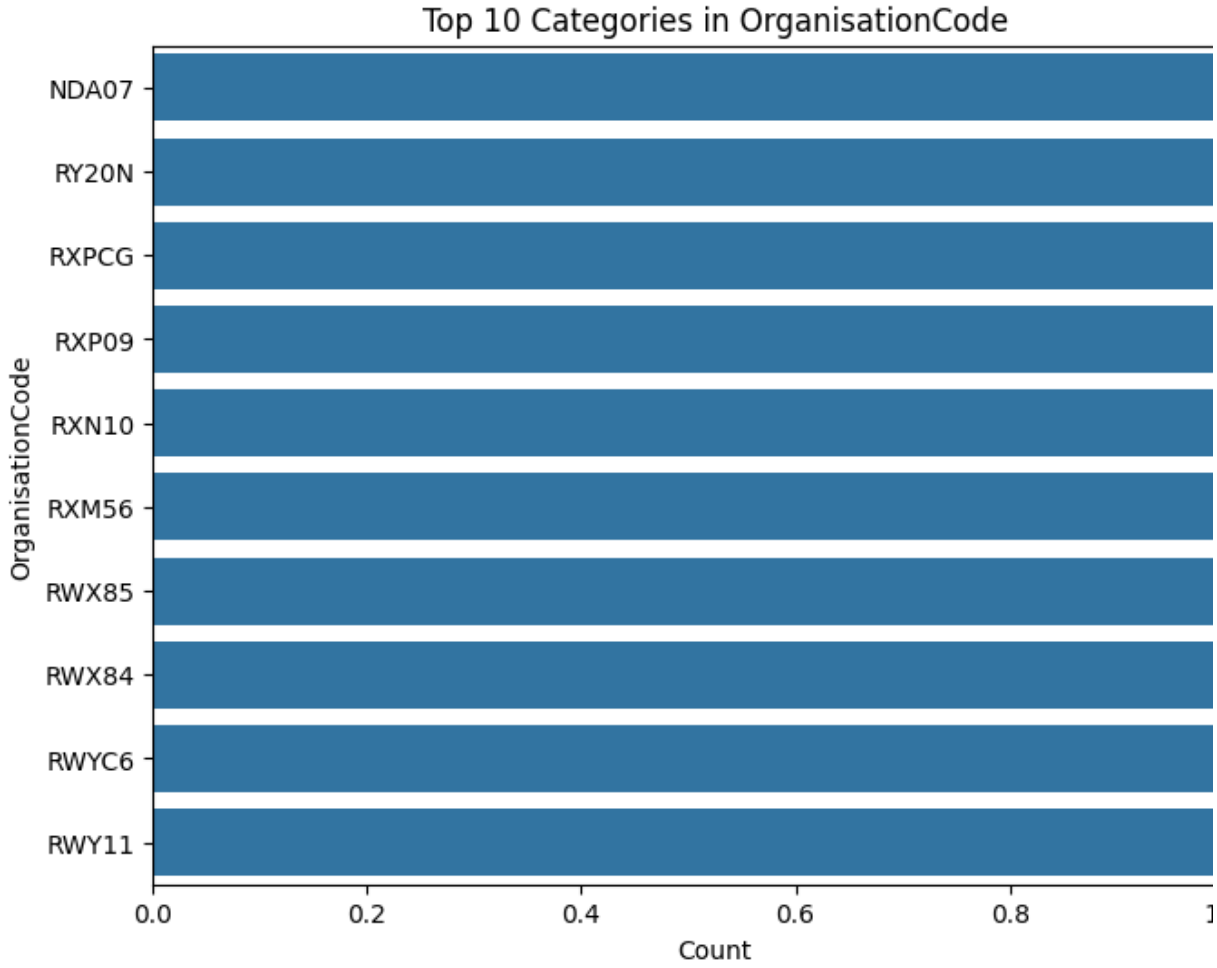

OrganisationName
St Mary's Hospital
St Michael's Hospital
Langley Green Hospital
Cygnet Hospital Harrow
Heatherwood Hospital
St Mark's Hospital
St Lukes Hospital
West Park Hospital
Birmingham Children's Hospital
Queen Elizabeth Hospital
0
1
2
3
4
5
count

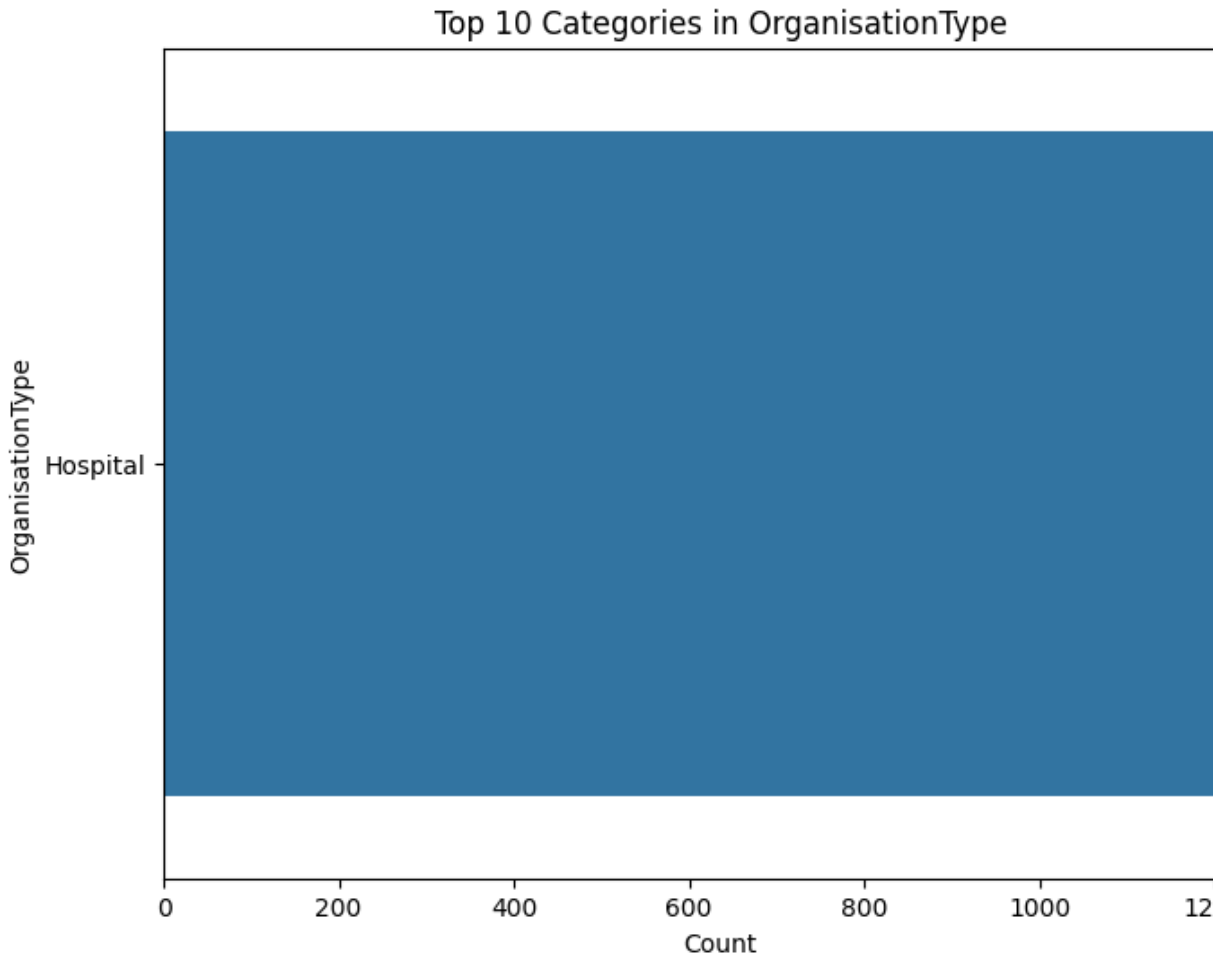


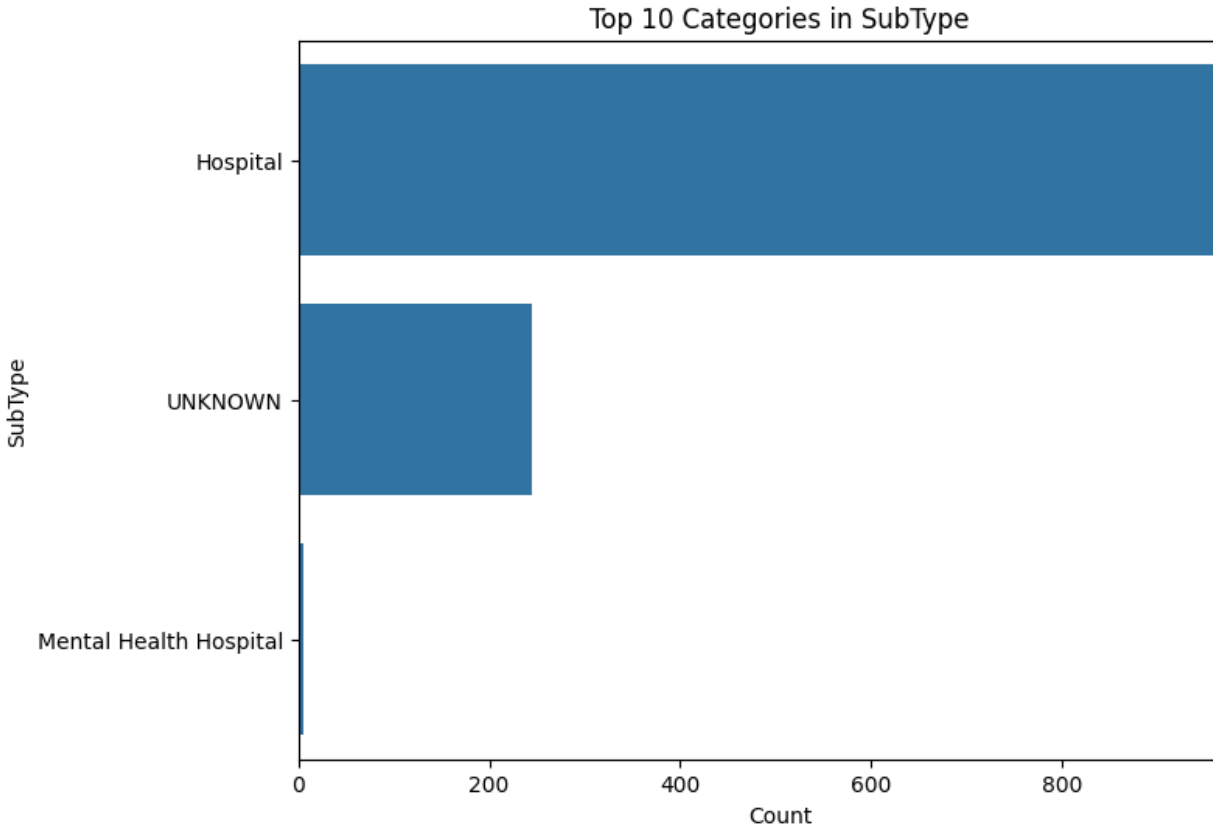


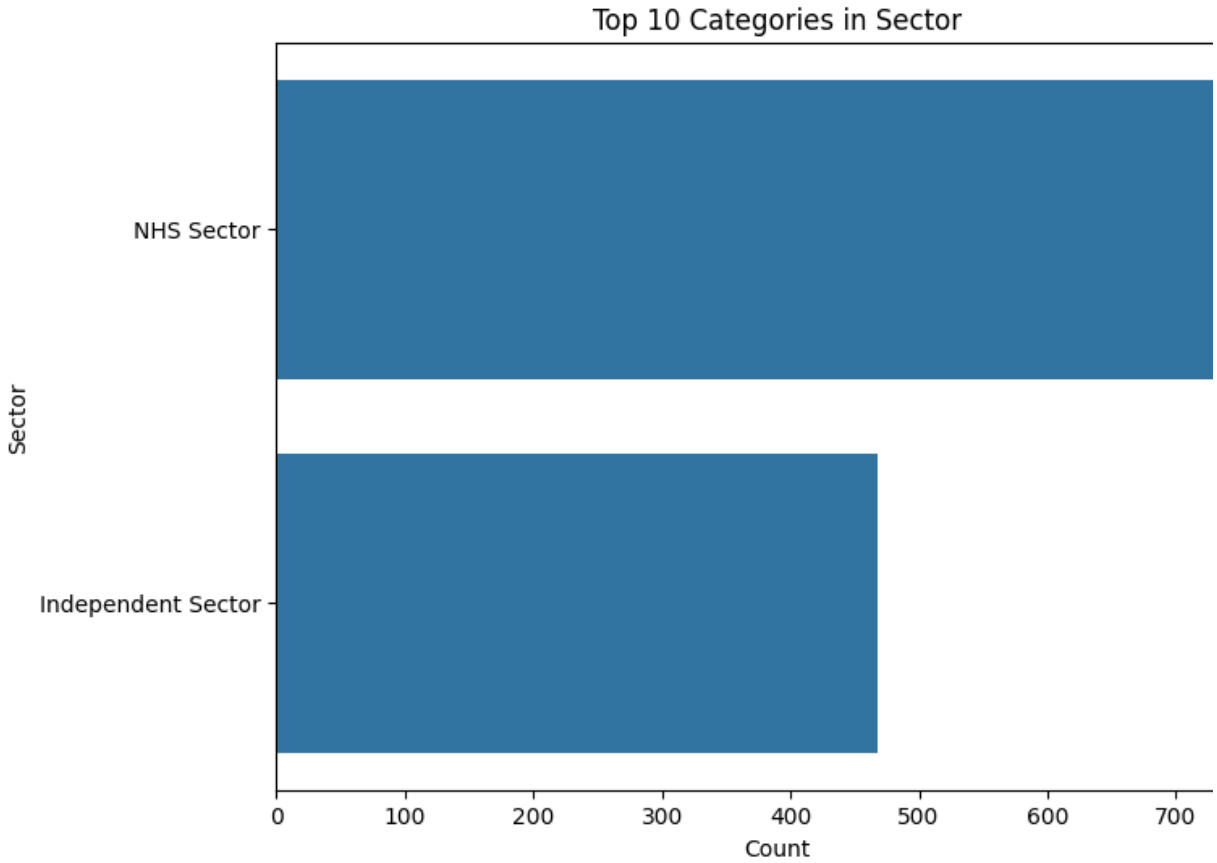
Object-Type BarPlot

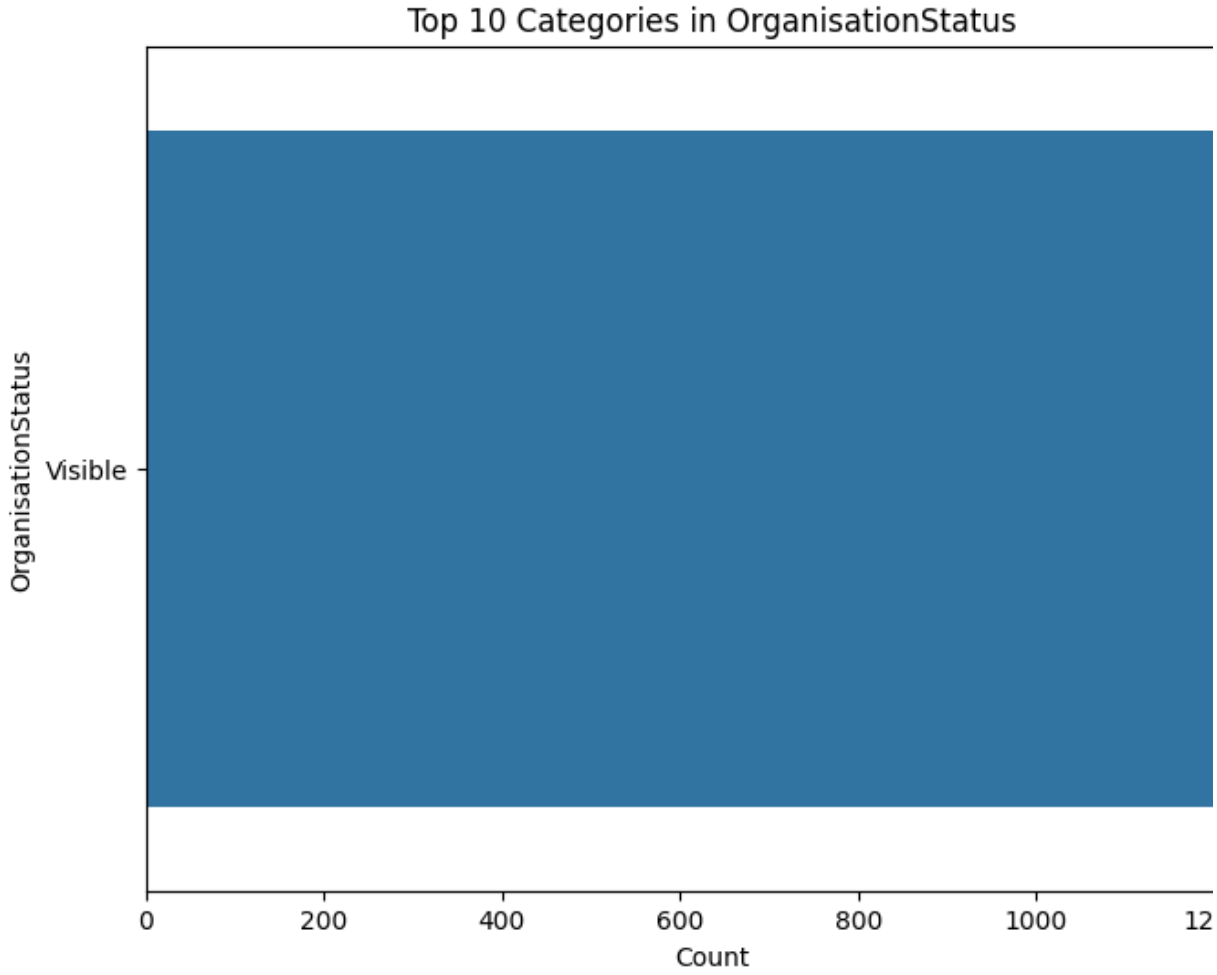
In [ ]:
```
for col in object_columns:
    plt.figure(figsize=(8, 6))
    top_values = data[col].value_counts().head(10)  # Get the top 10 most
    sns.barplot(y=top_values.index, x=top_values.values)
    plt.title(f'Top 10 Categories in {col}')
    plt.xlabel('Count')
    plt.ylabel(col)
    plt.show()
```

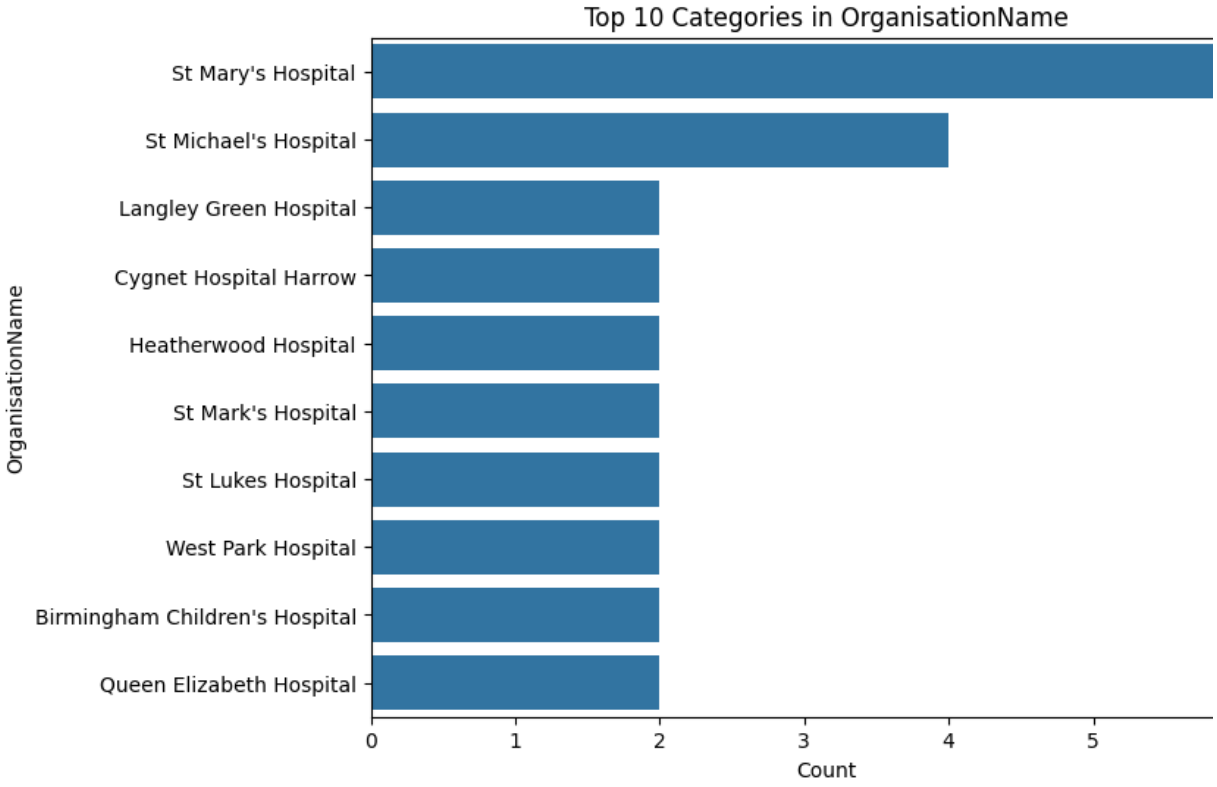


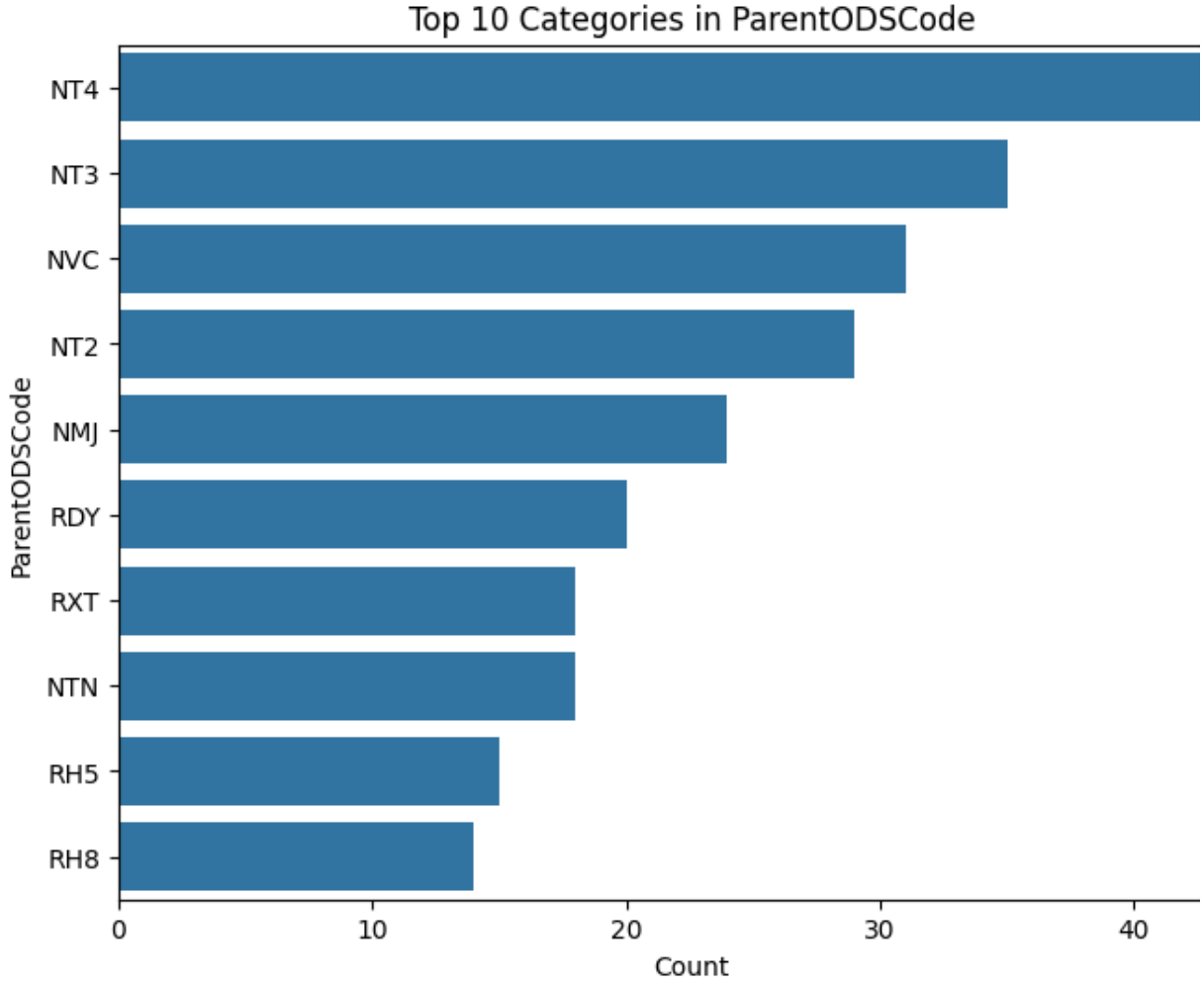

Top 10 Categories in OrganisationType
OrganisationType
Hospital
0
200
400
600
800
1000
12
Count

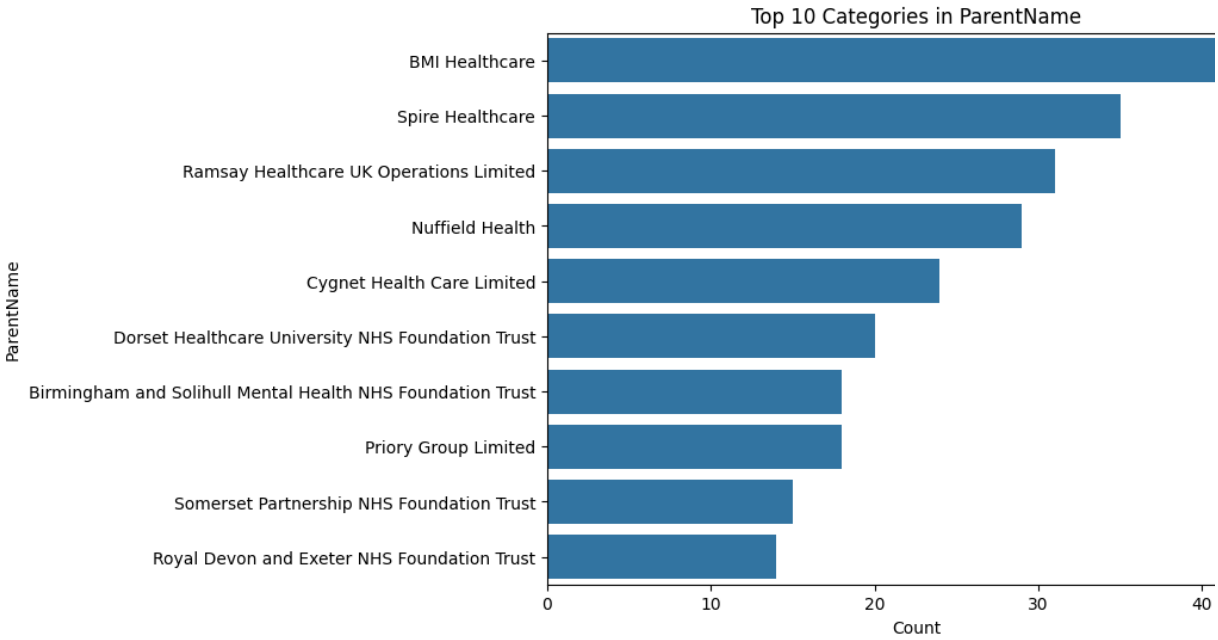

Top 10 Categories in SubType
SubType
Hospital
UNKNOWN
Mental Health Hospital
0
200
400
600
800
Count


Top 10 Categories in Sector
NHS Sector
Independent Sector
Sector
0
100
200
300
400
500
600
700
Count


Top 10 Categories in OrganisationStatus
Visible
OrganisationStatus
0
200
400
600
800
1000
12
Count

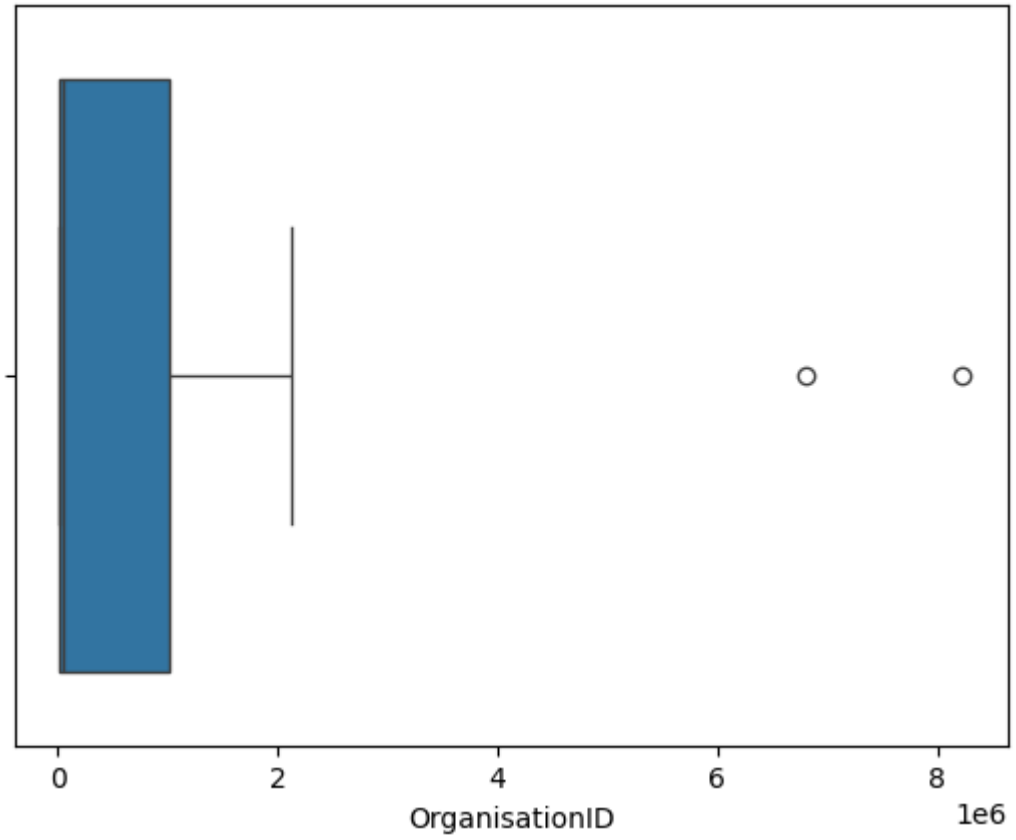

Top 10 Categories in OrganisationName
St Mary's Hospital
St Michael's Hospital
Langley Green Hospital
Cygnet Hospital Harrow
Heatherwood Hospital
St Mark's Hospital
St Lukes Hospital
West Park Hospital
Birmingham Children's Hospital
Queen Elizabeth Hospital
OrganisationName
0
1
2
3
4
5
Count

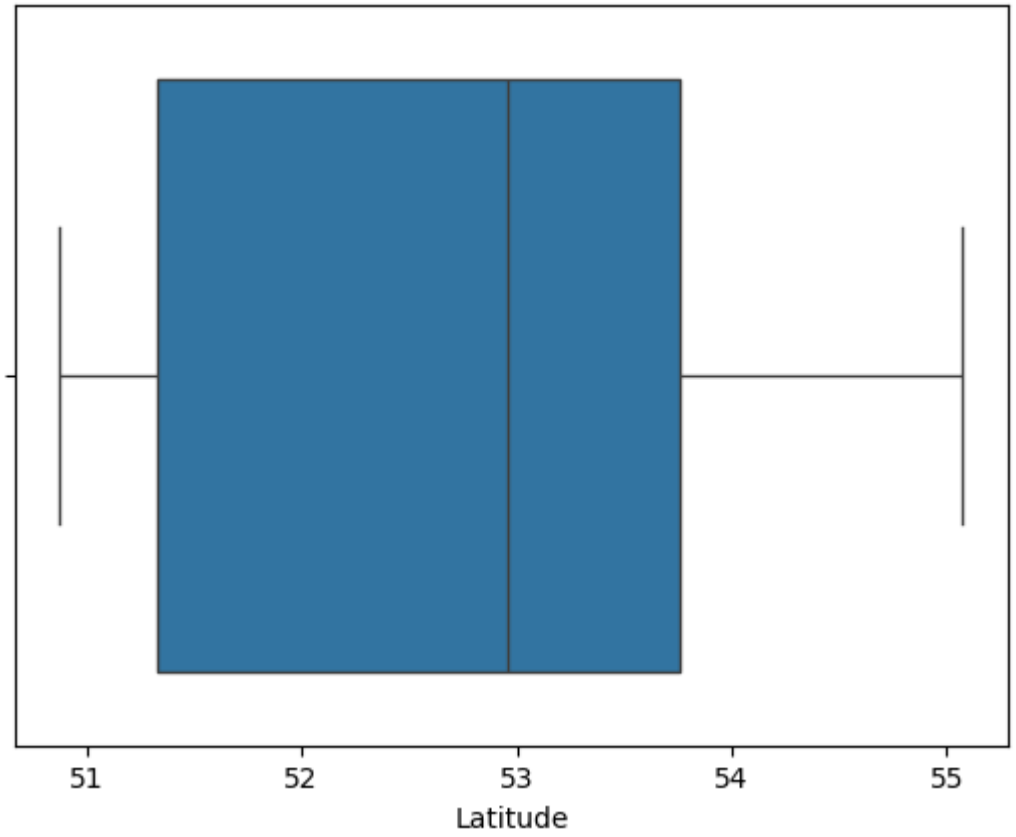

Top 10 Categories in ParentODSCode
NT4
NT3
NVC
NT2
NMJ
RDY
RXT
NTN
RH5
RH8
ParentODSCode
0
10
20
30
40
Count

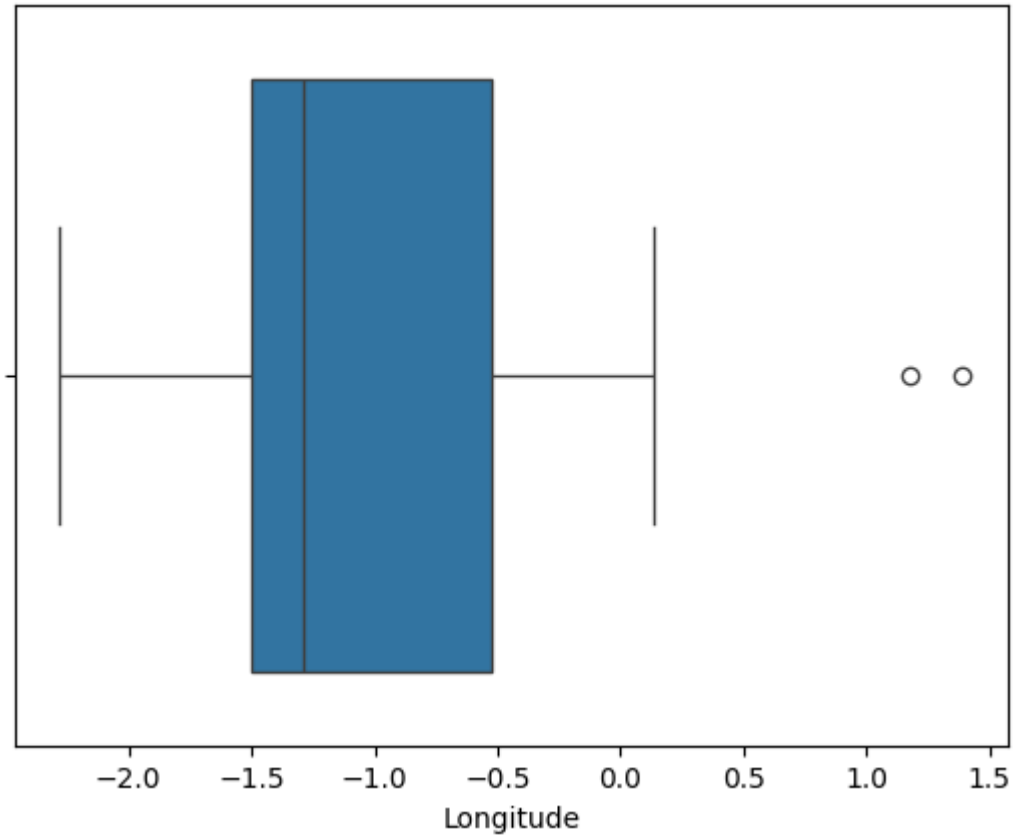


Numeric Columns Boxplot

In [ ]:
```
for col in numerical_columns:
    sns.boxplot(x=data[col])
    plt.show()
```

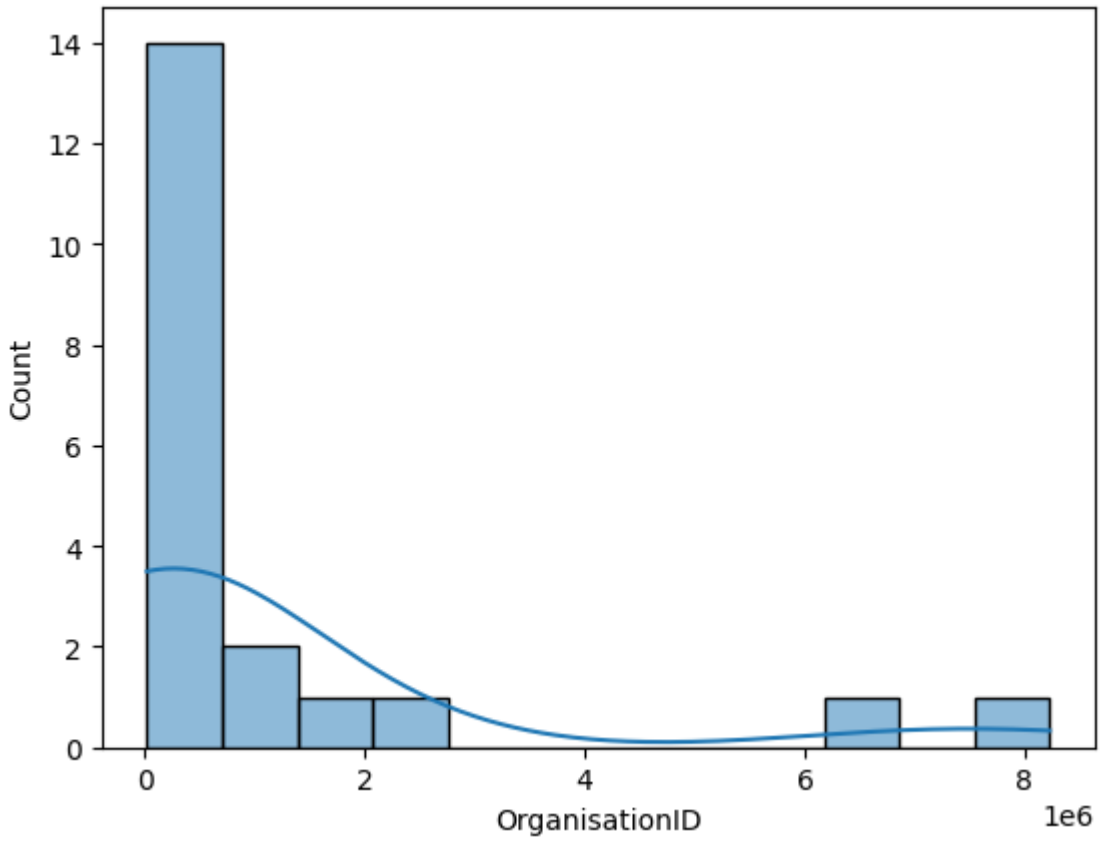


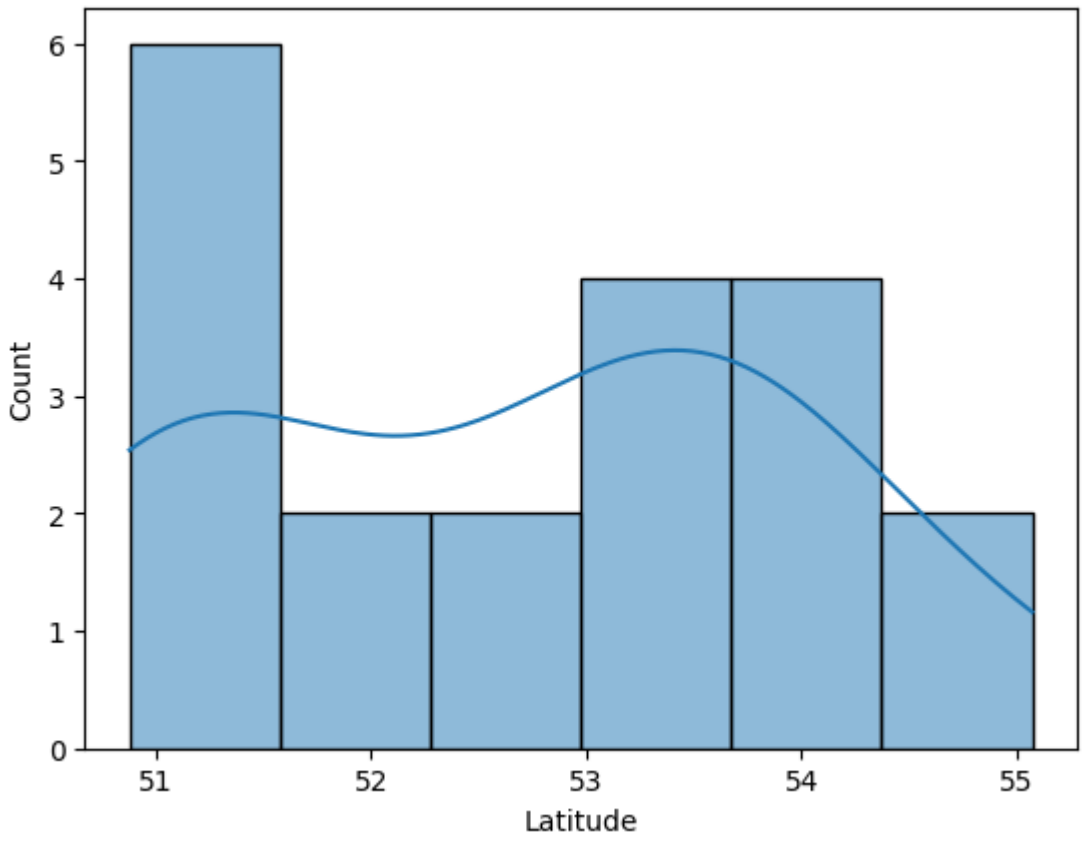


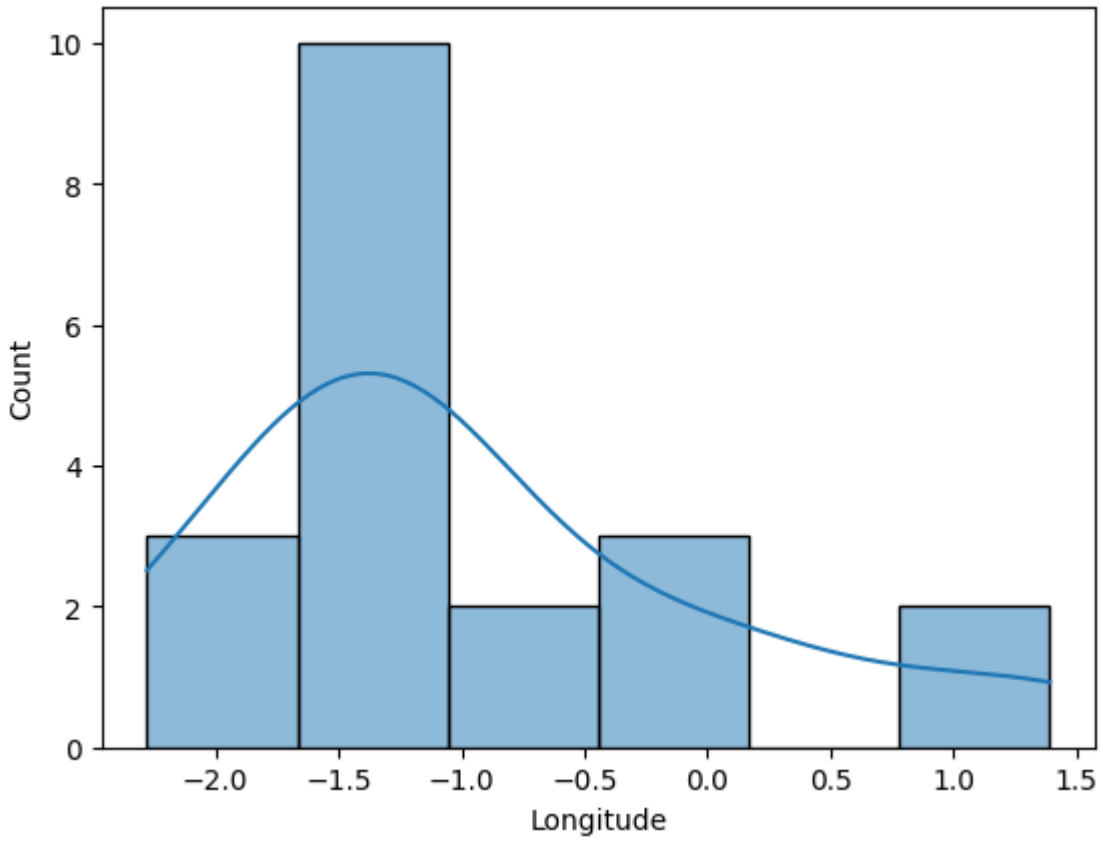


Histograms with a Kernel Density Estimate (KD

In [ ]:
```
for col in numerical_columns:
    sns.histplot(data[col], kde=True)
    plt.show()
```

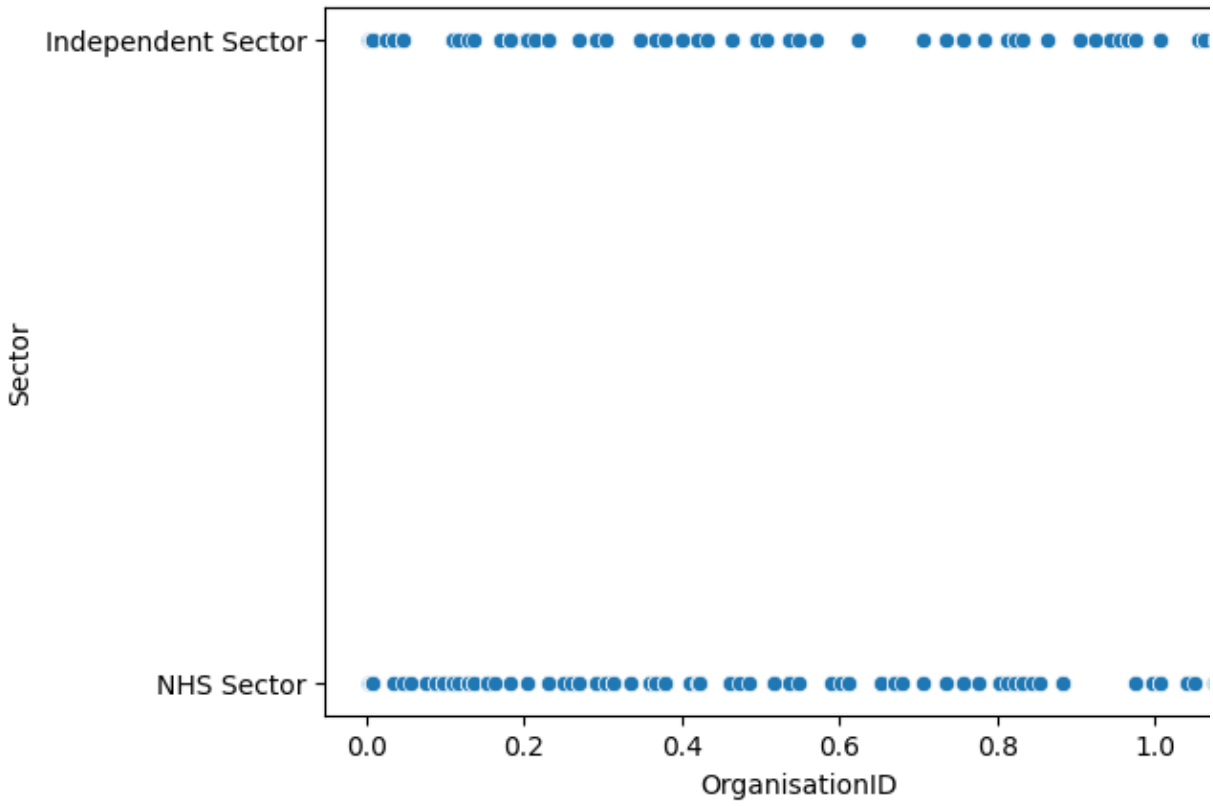






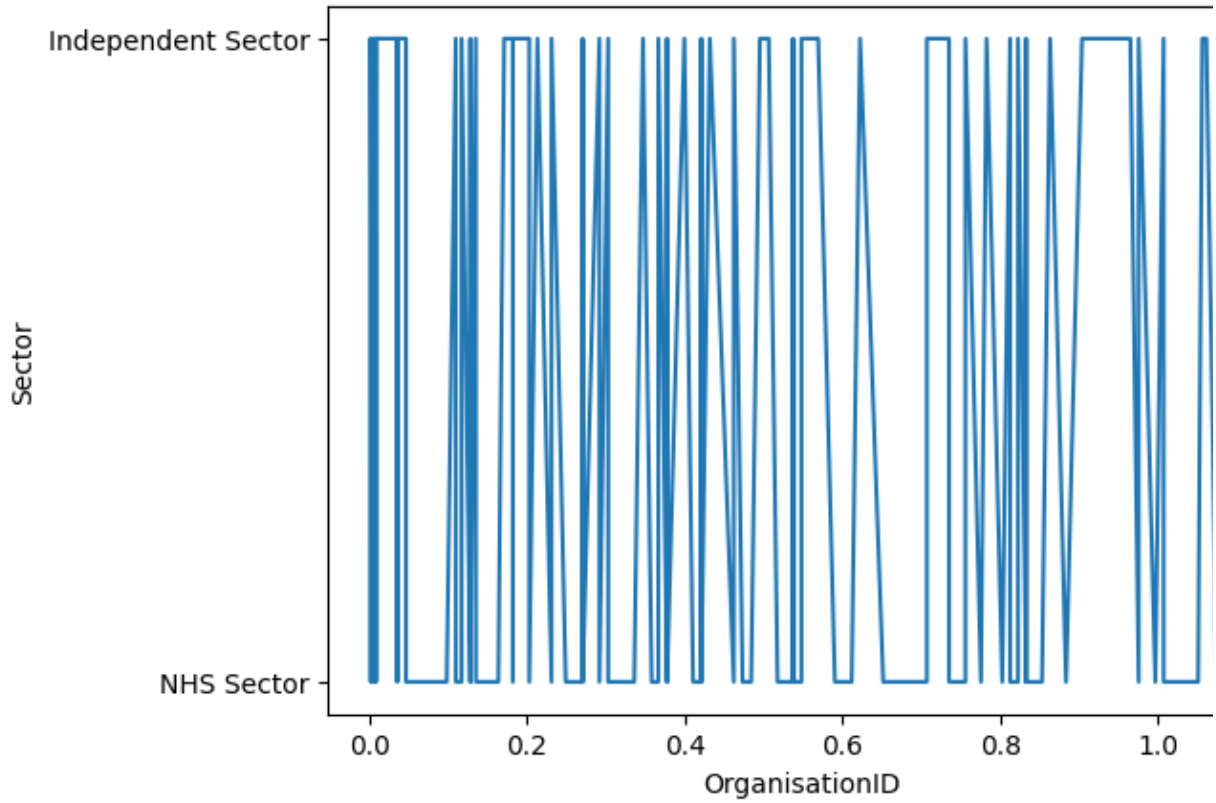
In [ ]:
```
sns.scatterplot(data=data, x='OrganisationID', y='Sector')
```

Out[ ]:
```
<Axes: xlabel='OrganisationID', ylabel='Sector'>
```



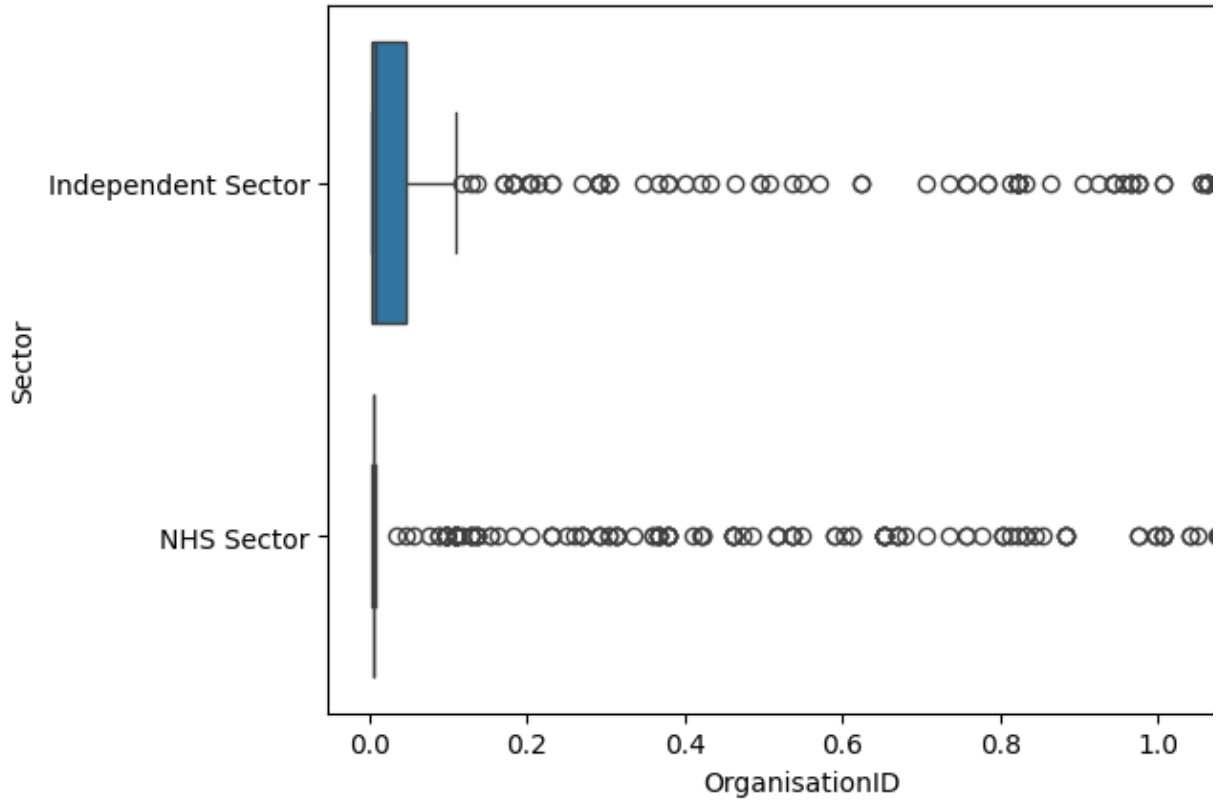
In [ ]:
```
sns.lineplot(data=data, x='OrganisationID', y='Sector')
```

Out[ ]:
```
<Axes: xlabel='OrganisationID', ylabel='Sector'>
```



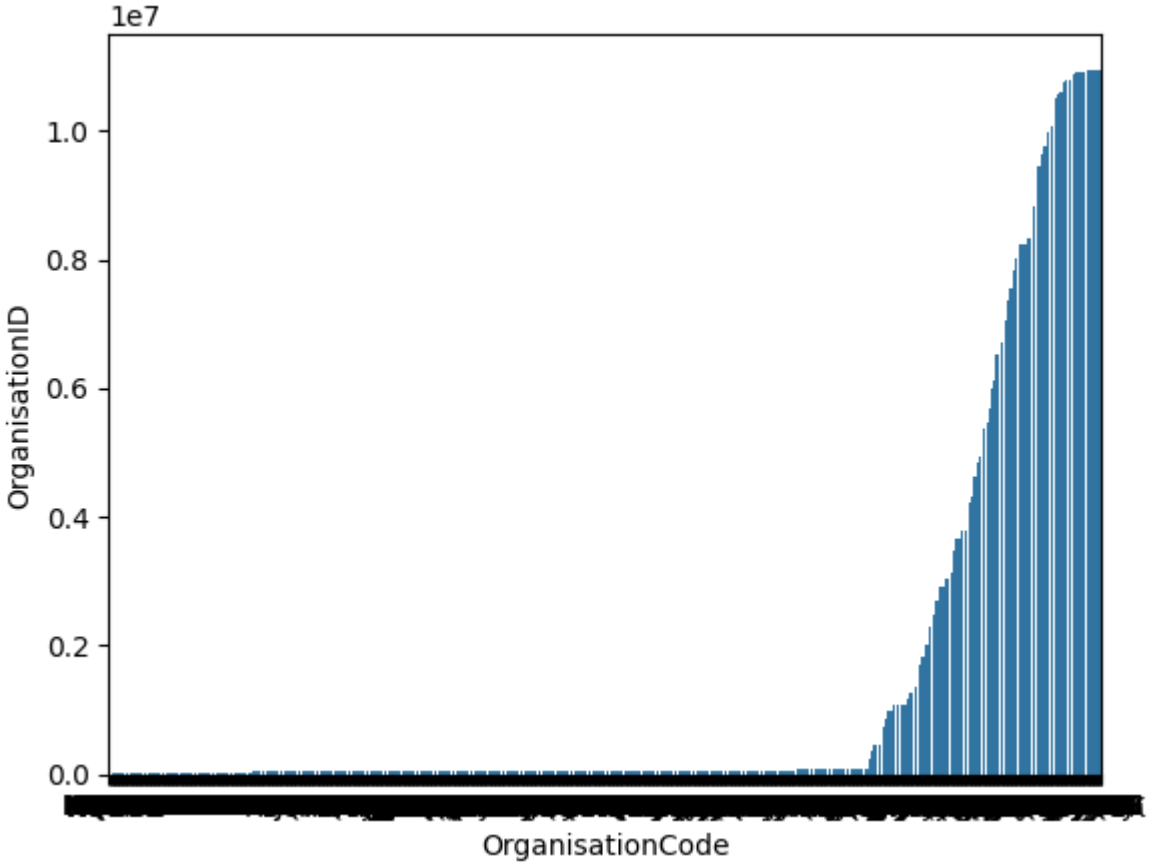
In [ ]:
```
sns.boxplot(data=data, x='OrganisationID', y='Sector')
```

Out[ ]:
```
<Axes: xlabel='OrganisationID', ylabel='Sector'>
```



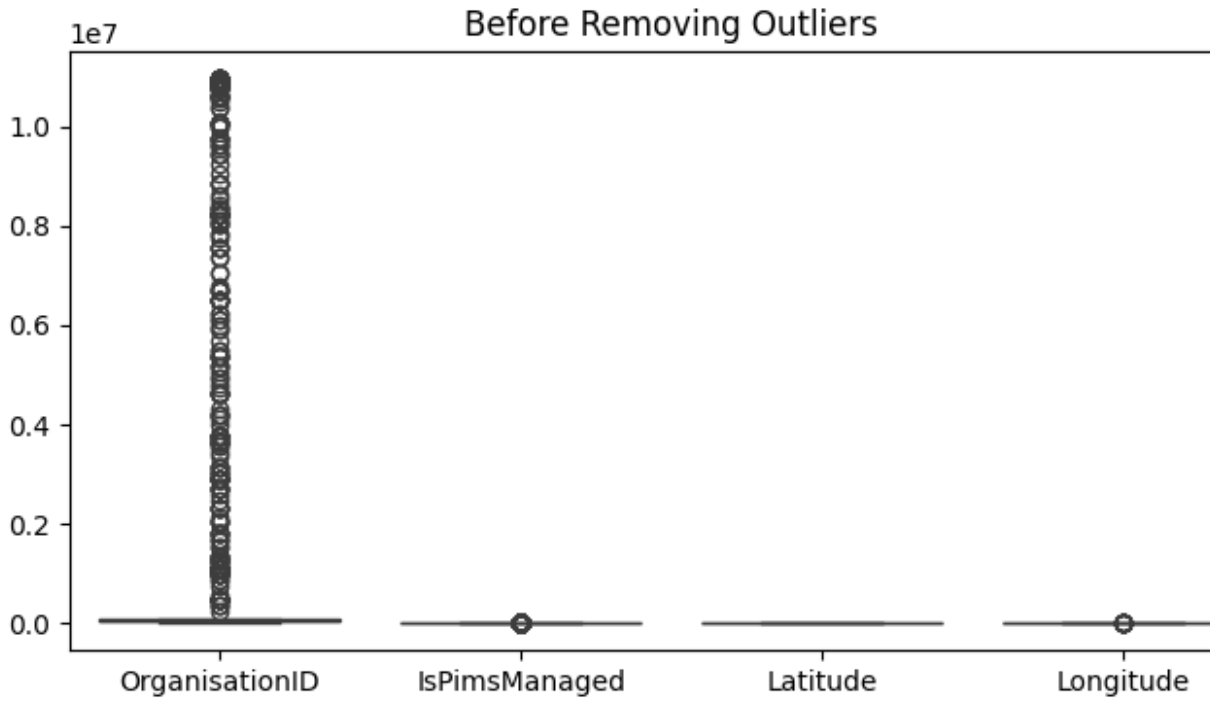
In [ ]:
```
sns.barplot(data=data, x='OrganisationCode', y='OrganisationID')
```

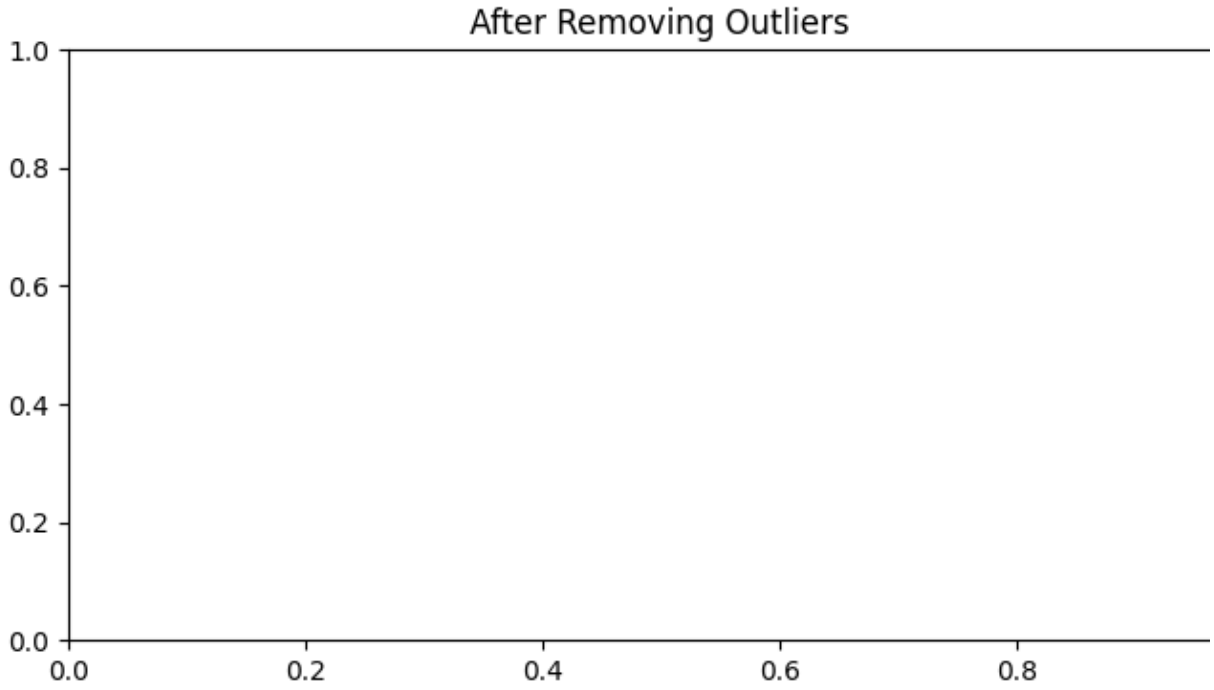
Out[ ]:
```
<Axes: xlabel='OrganisationCode', ylabel='OrganisationID'>
```



Removing Outliers

In [4]:
```
import matplotlib.pyplot as plt
import seaborn as sns

# Boxplot before removing outliers
plt.figure(figsize=(8, 4))
sns.boxplot(data=data)
plt.title("Before Removing Outliers")
plt.show()

# Boxplot after removing outliers
plt.figure(figsize=(8, 4))
sns.boxplot(data=data_no_outliers)
plt.title("After Removing Outliers")
plt.show()
```


Before Removing Outliers
1e7
1.0
0.8
0.6
0.4
0.2
0.0
OrganisationID
IsPimsManaged
Latitude
Longitude


After Removing Outliers
1.0
0.8
0.6
0.4
0.2
0.0
0.0
0.2
0.4
0.6
0.8