

## Appendix

### 1.a.

```
In [14]: # Determine the number of locations.  
num_loc = len(pd.unique(nc['sub_icb_location_name']))  
  
print(" The number of unqie locations is:",  
      (num_loc))
```

The number of unqie locations is: 106

```
In [17]: # Determine the number of service settings.  
num_service_setting = len(pd.unique(nc['service_setting']))  
  
print("Number of service settings is:", num_service_setting)
```

Number of service settings is: 5

```
In [18]: # Determine the number of context types.  
num_context_type = len(pd.unique(nc['context_type']))  
  
print(" The number of unqie locations is:",  
      (num_context_type))
```

The number of unqie locations is: 3

```
In [19]: # Determine the number of national categories.  
num_nat_cat = len(pd.unique(nc['national_category']))  
  
print("Number of national categories is:", num_nat_cat)
```

Number of national categories is: 18

```
In [20]: # Determine the number of appointment status.  
num_app_status = len(pd.unique(ar['appointment_status']))  
  
print("Number of appointment statuses is:", num_app_status)  
ar.appointment_status.value_counts()
```

Number of appointment statuses is: 3

```
Out[20]: Attended    232137  
Unknown     201324  
DNA        163360  
Name: appointment_status, dtype: int64
```

## 2.a.

```
In [309]: # Check the top five locations based on record count from nc  
nc['sub_icb_location_name'].value_counts().head()
```

```
Out[309]: NHS North West London ICB - W2U3Z      13007  
NHS Kent and Medway ICB - 91Q      12637  
NHS Devon ICB - 15N      12526  
NHS Hampshire and Isle Of Wight ICB - D9Y0V      12171  
NHS North East London ICB - A3A8R      11837  
Name: sub_icb_location_name, dtype: int64
```

```
In [1053]: # Create a DataFrame to log the location with the most records  
nth_west_ldn = nc[nc['sub_icb_location_name']  
                 == 'NHS North West London ICB - W2U3Z']  
# View the DataFrame  
nth_west_ldn.head()
```

```
In [337]: # Determine the sum of appointments per each service setting  
nth_west_ldn.groupby(['service_setting'])\  
    .sum('count_of_appointments').reset_index().sort_values(by='count_of_appointments', ascending=False)
```

```
Out[337]:
```

	service_setting	count_of_appointments
1	General Practice	1043225
4	Unmapped	904234
2	Other	343642
3	Primary Care Network	240283
0	Extended Access Provision	222006

```
In [955]: # Determine the count of records for each service setting  
nth_west_ldn['service_setting'].value_counts()
```

```
Out[955]:
```

General Practice	4609
Other	2858
Primary Care Network	2791
Extended Access Provision	2415
Unmapped	334

```
Name: service_setting, dtype: int64
```

## 2.b.

```
In [380]: # Determine the bottom location based on record count  
nc['sub_icb_location_name'].value_counts().tail()
```

```
Out[380]: NHS North East and North Cumbria ICB - 00N      4210  
NHS Lancashire and South Cumbria ICB - 02G      4169  
NHS Cheshire and Merseyside ICB - 01V      3496  
NHS Cheshire and Merseyside ICB - 01T      3242  
NHS Greater Manchester ICB - 00V      2170  
Name: sub_icb_location_name, dtype: int64
```

```
In [382]: # Create a DataFrame to log the location with the most records  
grt_manc_00v = nc[nc['sub_icb_location_name']  
                  == 'NHS Greater Manchester ICB - 00V']  
# View the DataFrame  
grt_manc_00v
```

```
In [383]: # Determine the sum of appointments per each service setting  
grt_manc_00v.groupby(['service_setting'])\  
    .sum('count_of_appointments').reset_index().sort_values(by='count_of_appointments', ascending=False)
```

```
Out[383]:      service_setting  count_of_appointments  
0             Unmapped            430394  
1        General Practice         207797  
2  Extended Access Provision          942  
3   Primary Care Network            74  
4                  Other                 4
```

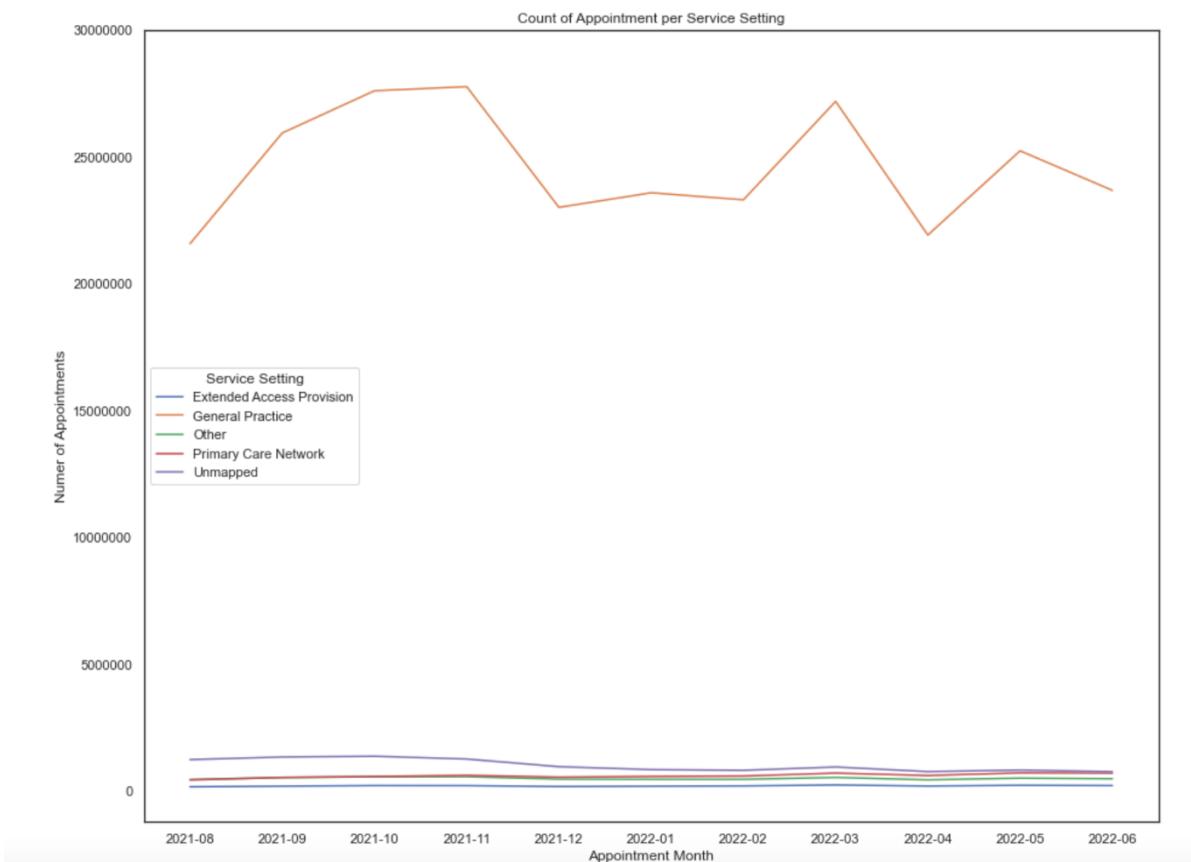
```
In [384]: # Determine the record count of each service setting  
grt_manc_00v['service_setting'].value_counts()
```

```
Out[384]: General Practice      1586  
Extended Access Provision     281  
Unmapped                      279  
Primary Care Network           22  
Other                          2  
Name: service_setting, dtype: int64
```

### 3.a.

```
In [525]: # Aggregate on monthly level and determine the sum of records per month.  
nc_ss = nc.groupby(['appointment_month', 'service_setting'])  
        .sum().reset_index()  
# View output.  
nc_ss.head()
```

```
In [528]: # Plot the appointments over the available date range, and review the service settings for months.  
# Create a lineplot.  
nc_ss_ax = sns.lineplot(x='appointment_month', y='count_of_appointments',  
                        hue='service_setting', data=nc_ss, ci=None)  
  
# Customise and tidy the plot.  
nc_ss_ax.legend(title='Service Setting', loc='center left')  
nc_ss_ax.set_xlabel('Appointment Month')  
nc_ss_ax.set_ylabel('Number of Appointments')  
nc_ss_ax.set_title("Count of Appointment per Service Setting")  
nc_ss_ax.set_yticks([0, 5000000, 10000000, 15000000,  
                    20000000, 25000000, 30000000])  
nc_ss_ax.set_yticks([2500000, 7500000, 12500000, 17500000,  
                    22500000, 27500000], minor=True)  
nc_ss_ax.set_yticklabels(['0', '5000000', '10000000', '15000000', '20000000', '25000000', '30000000'])
```



### 3.b.

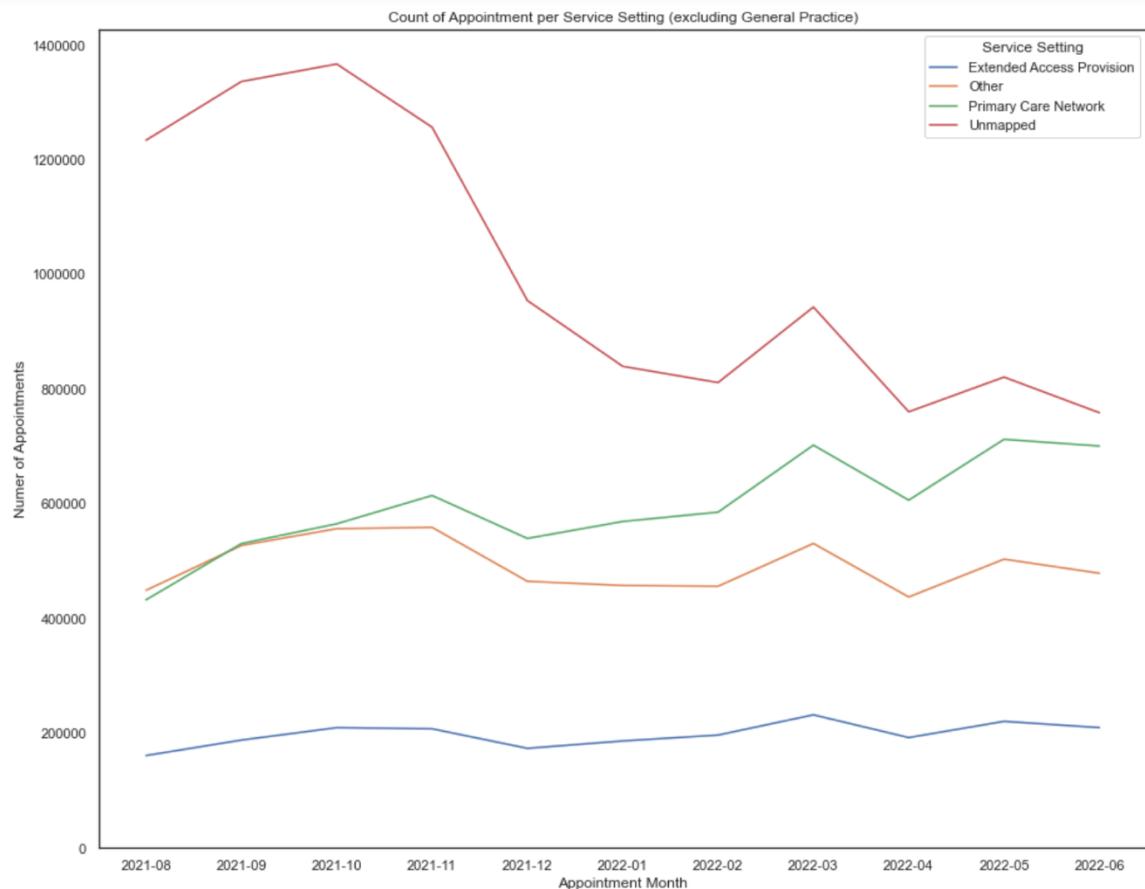
```
In [434]: # Subset the data to remove General Practice
nc_ex_gp = nc_ss[nc_ss['service_setting'] != 'General Practice']
# View DataFrame
nc_ex_gp.head()
```

	appointment_month	service_setting	count_of_appointments
0	2021-08	Extended Access Provision	160927
2	2021-08	Other	449101
3	2021-08	Primary Care Network	432448
4	2021-08	Unmapped	1233843
5	2021-09	Extended Access Provision	187906

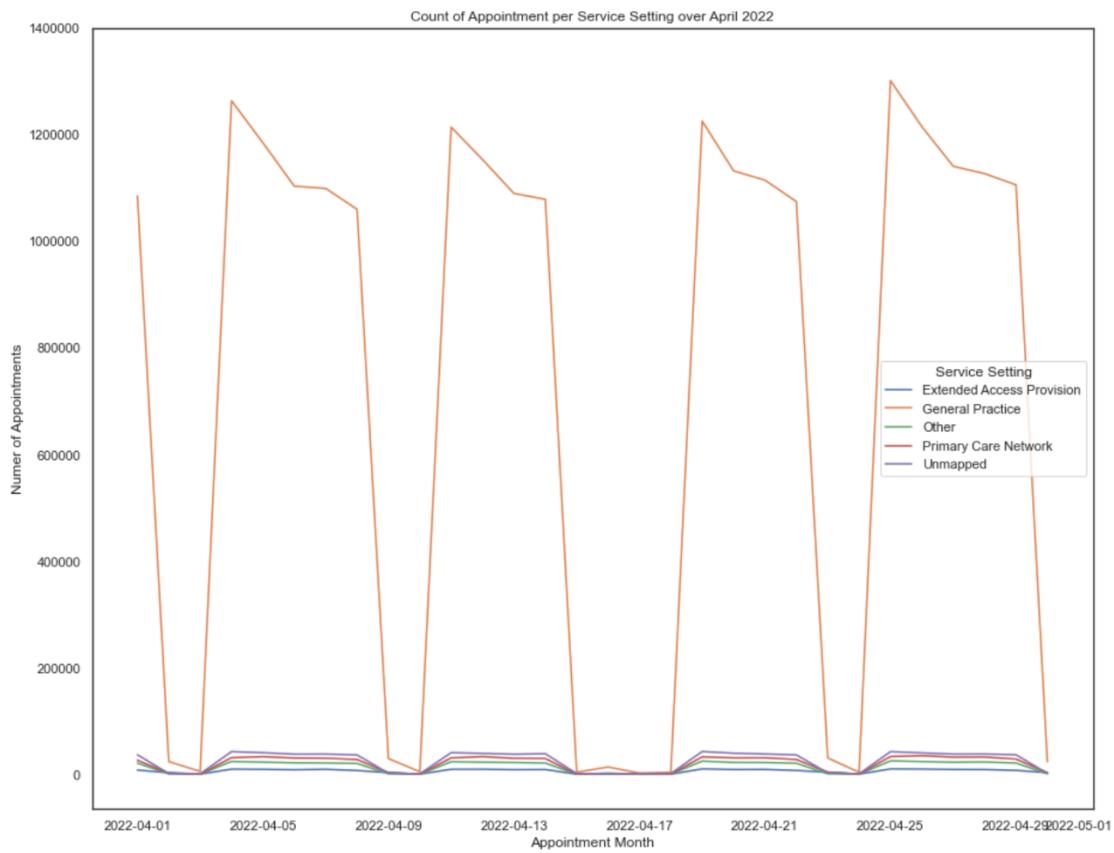
```
In [530]: # Create a lineplot showing service setting excluding General Practice
ex_gp_ax = sns.lineplot(x='appointment_month', y='count_of_appointments',
                        hue='service_setting', data=nc_ex_gp, ci=None)

# Customise and tidy the plot.
ex_gp_ax.set_xlabel('Appointment Month')
ex_gp_ax.set_ylabel('Numer of Appointments')
ex_gp_ax.set_title(
    "Count of Appointment per Service Setting (excluding General Practice)")
ex_gp_ax.legend(title='Service Setting', loc='upper right')
ex_gp_ax.set_yticks([0, 200000, 400000, 600000,
                     800000, 1000000, 1200000, 1400000])
ex_gp_ax.set_yticks([100000, 300000, 500000, 700000,
                     900000, 1100000, 1300000], minor=True)
ex_gp_ax.set_yticklabels(
    ['0', '200000', '400000', '600000',
     '800000', '1000000', '1200000', '1400000'])


```



### 3.c.



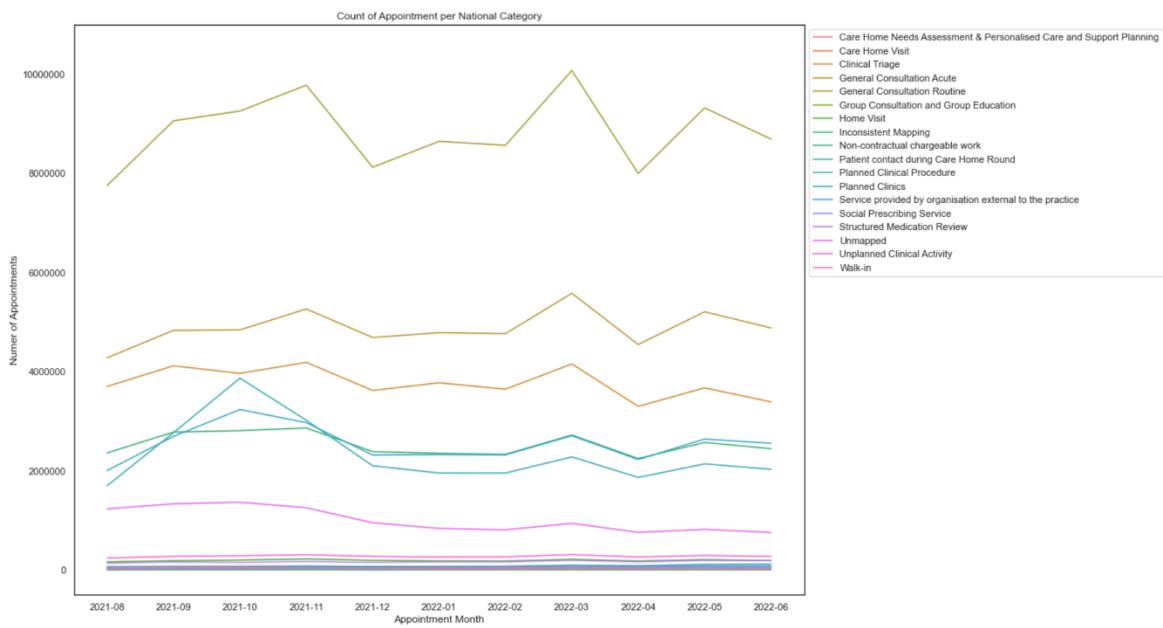
### 4.a.

```
In [68]: # Create a separate data set that can be used in future weeks.
nc_nc = nc.groupby(
    ['appointment_month', 'national_category']).sum().reset_index()
# View output.
nc_nc.head()
```

```
Out[68]: appointment_month          national_category  count_of_appointments
0      2021-08  Care Home Needs Assessment & Personalised Care...        29676
1      2021-08                  Care Home Visit                 47583
2      2021-08                  Clinical Triage                3704207
3      2021-08  General Consultation Acute                4280920
4      2021-08  General Consultation Routine              7756045
```

```
In [581]: # Plot the appointments over the available date range, and review the national categories for months.
# Create a lineplot.
nc_nc_ax = sns.lineplot(x='appointment_month', y='count_of_appointments',
                        hue='national_category', data=nc_nc, ci=None)

# Customise and tidy the plot
nc_nc_ax.legend(bbox_to_anchor=(1, 1))
nc_nc_ax.set_xlabel('Appointment Month')
nc_nc_ax.set_ylabel('Numer of Appointments')
nc_nc_ax.set_title("Count of Appointment per National Category")
nc_nc_ax.set_yticks([0, 2000000, 4000000, 6000000, 8000000, 10000000])
nc_nc_ax.set_yticks([1000000, 3000000, 5000000, 7000000,
                     9000000, 11000000], minor=True)
nc_nc_ax.set_yticklabels(['0', '2000000', '4000000', '6000000',
                         '8000000', '10000000'])
```



## 4.b.

```
In [549]: # Create a new DataFrame featuring the national categories with 10 million + appointments
nc_nc_top = nc_nc[(nc_nc['national_category'].isin(['General Consultation Routine',
                                                    'General Consultation Acute',
                                                    'Clinical Triage', 'Planned Clinics',
                                                    'Inconsistent Mapping', 'Planned Clinical Procedure',
                                                    'Unmapped']))]

# View the DataFrame
nc_nc_top
```

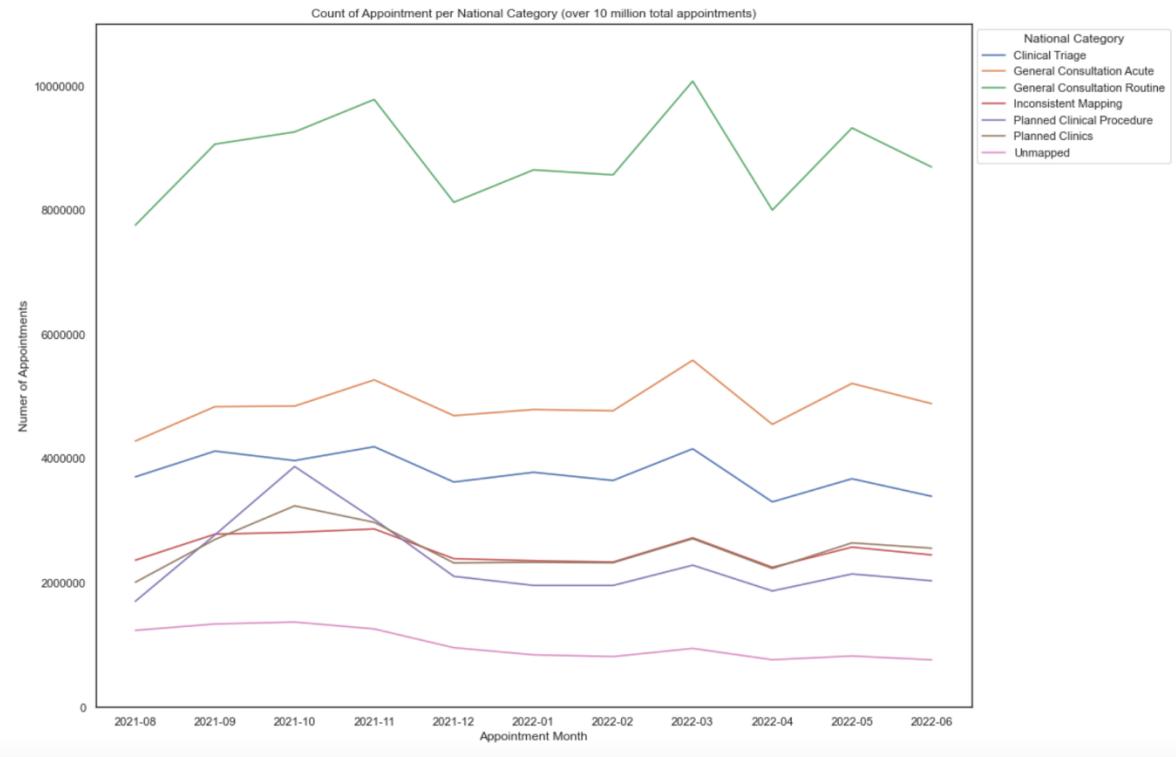
	appointment_month	national_category	count_of_appointments
2	2021-08	Clinical Triage	3704207
3	2021-08	General Consultation Acute	4280920
4	2021-08	General Consultation Routine	7756045
7	2021-08	Inconsistent Mapping	2363093
10	2021-08	Planned Clinical Procedure	1701814
...	...	...	...
184	2022-06	General Consultation Routine	8691384
187	2022-06	Inconsistent Mapping	2447611
190	2022-06	Planned Clinical Procedure	2032073
191	2022-06	Planned Clinics	2556175
195	2022-06	Unmapped	758640

```
In [522]: # Create a new DataFrame to include all national categories which have less than 10 million appointments
nc_nc_bottom = nc_nc[(nc_nc['national_category'].isin(['Unplanned Clinical Activity',
    'Home Visit', 'Structured Medication Review',
    'Service provided by organisation external to the practice',
    'Patient contact during Care Home Round',
    'Care Home Visit', 'Social Prescribing Service',
    'Walk-in',
    'Care Home Needs Assessment & Personalised Care and Support Plan',
    'Non-contractual chargeable work',
    'Group Consultation and Group Education']))]
```

	appointment_month	national_category	count_of_appointments
0	2021-08	Care Home Needs Assessment & Personalised Care and Support Planning	29676
1	2021-08	Care Home Visit	47583
5	2021-08	Group Consultation and Group Education	5161
6	2021-08	Home Visit	165061
8	2021-08	Non-contractual chargeable work	10775
...	...	...	...
192	2022-06	Service provided by organisation external to the practice	81663
193	2022-06	Social Prescribing Service	55066
194	2022-06	Structured Medication Review	187800
196	2022-06	Unplanned Clinical Activity	274491
197	2022-06	Walk-in	35935

## 4.C.

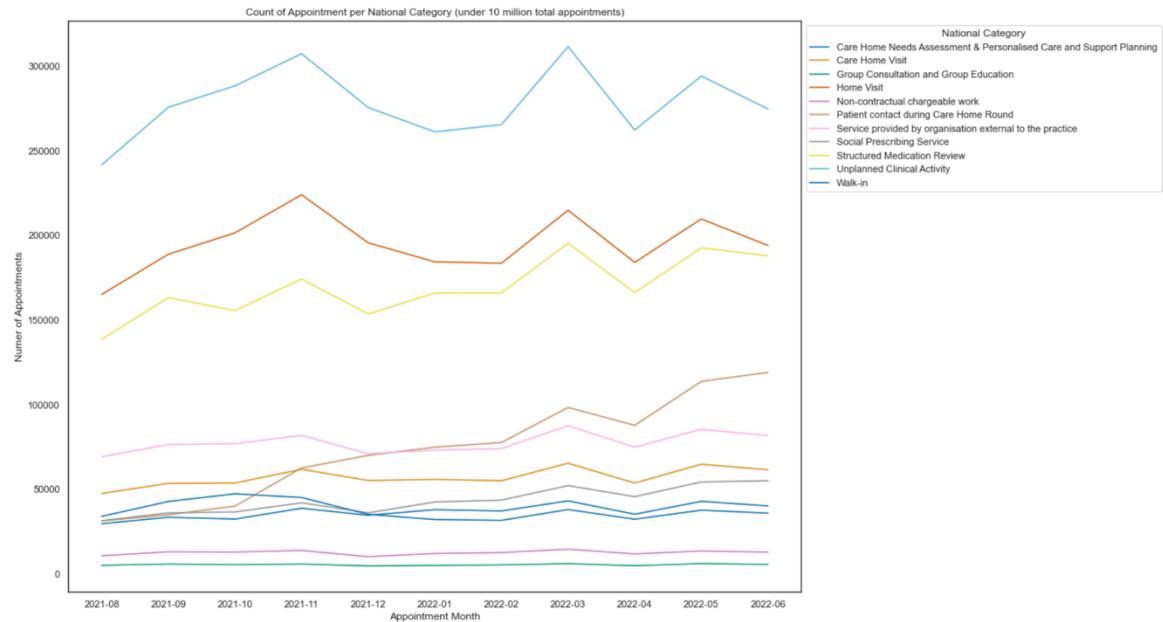
```
In [609]: # Plot the appointments over the available date range, and review the top national categories for months.
# Create a lineplot.
top_ax = sns.lineplot(x='appointment_month', y='count_of_appointments',
                      hue='national_category', data=nc_nc_top, ci=None)
# Customise and tidy the plot
top_ax.legend(title='National Category', bbox_to_anchor=(1, 1))
top_ax.set_xlabel('Appointment Month')
top_ax.set_ylabel('Number of Appointments')
top_ax.set_title("Count of Appointment per National Category (over 10 million total appointments)")
top_ax.set_yticks([0, 2000000, 4000000, 6000000, 8000000, 10000000])
top_ax.set_yticks([1000000, 3000000, 5000000, 7000000,
                  9000000, 11000000], minor=True)
top_ax.set_yticklabels(['0', '2000000', '4000000', '6000000',
                       '8000000', '10000000'])
```



#### 4.d.

```
In [610]: # Plot the appointments over the available date range, and review the bottom national categories for months.
# Create a lineplot.
bottom_ax = sns.lineplot(x='appointment_month', y='count_of_appointments',
                        hue='national_category', palette='colorblind', data=nc_nc_bottom, ci=None)

# Customise and tidy the plot
bottom_ax.legend(title='National Category', bbox_to_anchor=(1, 1))
bottom_ax.set_xlabel('Appointment Month')
bottom_ax.set_ylabel('Numer of Appointments')
bottom_ax.set_title("Count of Appointment per National Category (under 10 million total appointments)")
```



## 5.a.

```
In [140]: # Loop through the messages, and create a list of values containing the # symbol.
tags = []

for y in [x.split(' ') for x in tweets['tweet_full_text'].values]:
    for z in y:
        if '#' in z:
            # Change to lowercase.
            tags.append(z.lower())

tags
```

```
In [141]: # Display the first 30 records.
hashtags = pd.Series(tags).value_counts()
hashtags.iloc[:30]
```

Out[141]:	#healthcare	716
	#health	80
	#medicine	41
	#ai	40
	#job	38
	#medical	35
	#strategy	30
	#pharmaceutical	28
	#digitalhealth	25
	#pharma	25
	#marketing	25
	#medtwitter	24

## 5.b.

```
In [142]: # Convert the series to a DataFrame in preparation for visualisation.  
data = pd.DataFrame(hashtags).reset_index()  
data.rename(columns={data.columns[0]: 'word',  
                    data.columns[1]: 'count'}, inplace=True)  
# Rename the columns.  
data
```

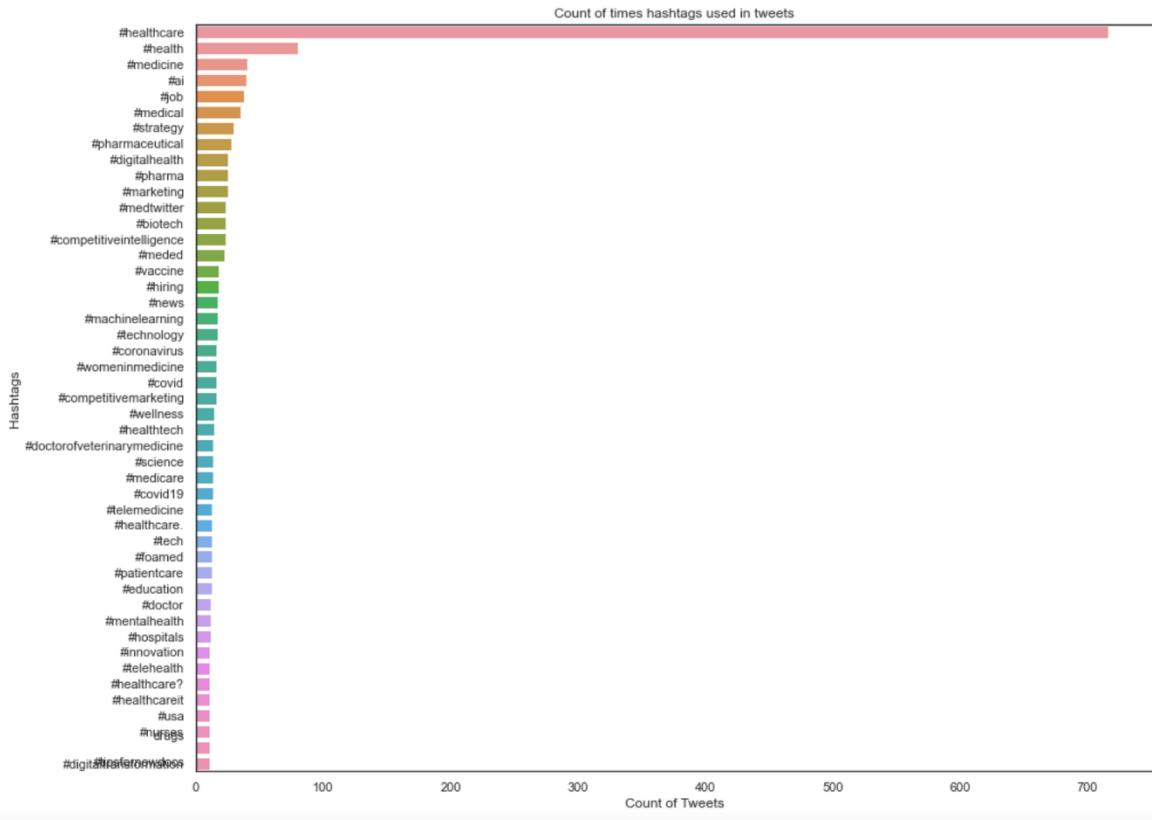
```
Out[142]:
```

	word	count
0	#healthcare	716
1	#health	80
2	#medicine	41
3	#ai	40
4	#job	38
...	...	...
1749	#evestudy	1
1750	#patientdata...	1
1751	#secure	1
1752	#sms	1
1753	\n#csjmu	1

1754 rows × 2 columns

## 5.c.

```
In [664]: # Create a Seaborn barplot indicating records with a count >10 records.  
hashtag_ax = sns.barplot(  
    x='count', y='word', estimator=sum, data=data_above_ten)  
  
# Customise and tidy the plot  
hashtag_ax.set_xlabel('Count of Tweets')  
hashtag_ax.set_ylabel('Hashtags')  
hashtag_ax.set_title(  
    "Count of times hashtags used in tweets")
```



## 5.d.

```
In [148]: # The columns you want to search for outliers in.
cols = ['count']

# Calculate quantiles and IQR.
# Same as np.percentile but maps (0,1) and not (0,100).
Q1 = data_above_ten[cols].quantile(0.25)
Q3 = data_above_ten[cols].quantile(0.75)
IQR = Q3 - Q1
IQR

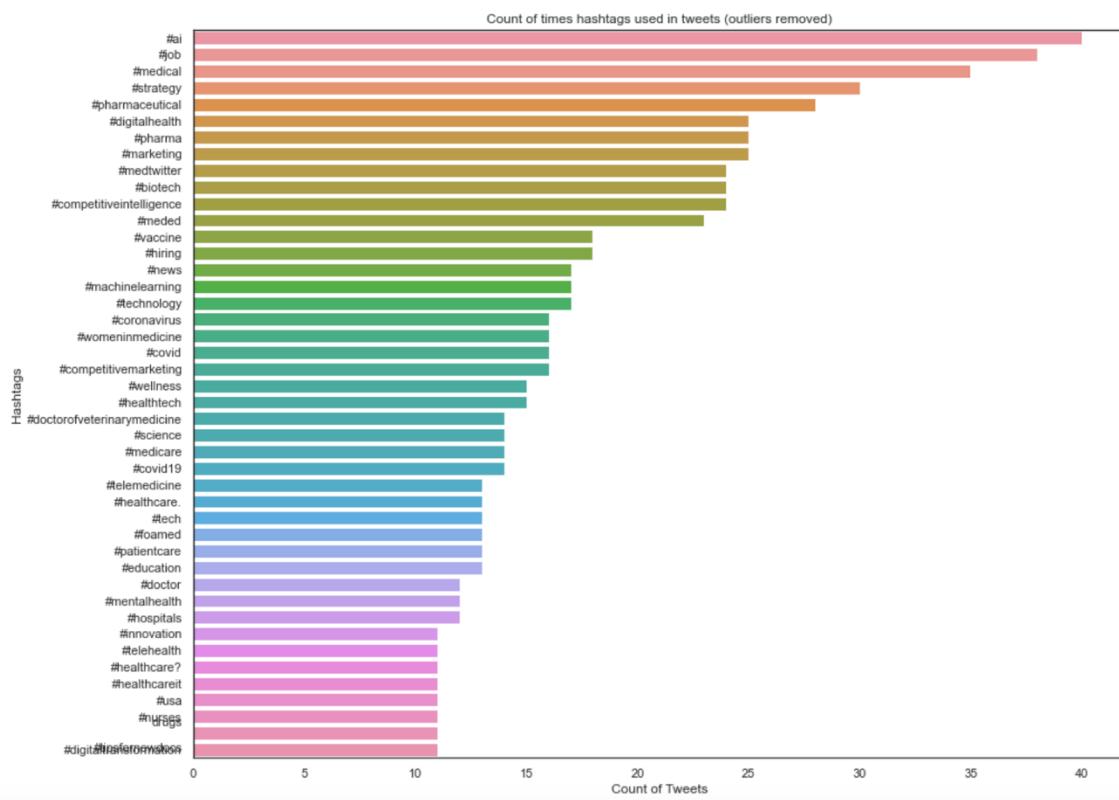
# Return a Boolean array of the rows with (any) non-outlier column values.
condition = ~((data_above_ten[cols] < (Q1 - 1.5 * IQR))
              | (data_above_ten[cols] > (Q3 + 1.5 * IQR))).any(axis=1)

# Filter our DataFrame based on condition.
data_above_ten_non_outlier = data_above_ten[condition]
data_above_ten_non_outlier.shape
```

Out[148]: (44, 2)

## 5.e.

```
In [665]: # Create the plot.  
# View the barplot.  
non_outlier_ax = sns.barplot(x='count', y='word', estimator=sum,  
                             data=data_above_ten_non_outlier)  
  
# Customise and tidy the plot  
non_outlier_ax.set_xlabel('Count of Tweets')  
non_outlier_ax.set_ylabel('Hashtags')  
non_outlier_ax.set_title(  
    "Count of times hashtags used in tweets (outliers removed)")
```



## 6.a.

```
In [1017]: # Filter the data set to only look at data from 2021-08 onwards.
ar_2 = ar[ar['appointment_month'] >= '2021-08']
ar_2
```

	icb_ons_code	appointment_month	appointment_status	hcp_type	appointment_mode	time_between_book_and_appointment	count_of_appointments
3652	E54000034	2021-08	Attended	GP	Face-to-Face	1 Day	6553
3653	E54000034	2021-08	Attended	GP	Face-to-Face	15 to 21 Days	2390
3654	E54000034	2021-08	Attended	GP	Face-to-Face	2 to 7 Days	10547
3655	E54000034	2021-08	Attended	GP	Face-to-Face	22 to 28 Days	937
3656	E54000034	2021-08	Attended	GP	Face-to-Face	8 to 14 Days	4961
...	...	...	...	...	...	...	...
596816	E54000050	2022-06	Unknown	Unknown	Unknown	2 to 7 Days	21
596817	E54000050	2022-06	Unknown	Unknown	Unknown	22 to 28 Days	8
596818	E54000050	2022-06	Unknown	Unknown	Unknown	8 to 14 Days	28
596819	E54000050	2022-06	Unknown	Unknown	Unknown	More than 28 Days	17
596820	E54000050	2022-06	Unknown	Unknown	Unknown	Same Day	10

223418 rows × 7 columns

## 6.b.

```
In [714]: # Create an aggregated data set to review the different features.
ar_agg = ar_2.groupby(['appointment_month', 'hcp_type', 'appointment_status',
                      'appointment_mode', 'time_between_book_and_appointment']).sum().reset_index()

# View the DataFrame.
ar_agg
```

	appointment_month	hcp_type	appointment_status	appointment_mode	time_between_book_and_appointment	count_of_appointments
0	2021-08	GP	Attended	Face-to-Face	1 Day	507835
1	2021-08	GP	Attended	Face-to-Face	15 to 21 Days	194726
2	2021-08	GP	Attended	Face-to-Face	2 to 7 Days	959486
3	2021-08	GP	Attended	Face-to-Face	22 to 28 Days	102111
4	2021-08	GP	Attended	Face-to-Face	8 to 14 Days	398772
...	...	...	...	...	...	...
3749	2022-06	Unknown	Unknown	Unknown	8 to 14 Days	5494
3750	2022-06	Unknown	Unknown	Unknown	More than 28 Days	5115
3751	2022-06	Unknown	Unknown	Unknown	Same Day	1914
3752	2022-06	Unknown	Unknown	Unknown	Unknown / Data Quality	53
3753	2022-06	Unknown	Unknown	Video/Online	2 to 7 Days	1

3754 rows × 6 columns

## 6.c.

```
In [803]: # Determine the total number of appointments per month.
ar_df = ar_agg.groupby(['appointment_month']).sum(
    ['count_of_appointments']).reset_index()
ar_df
# Add a new column to indicate the average utilisation of services.
# Monthly aggregate / 30 to get to a daily value.
ar_df['avg_daily_app'] = (ar_df['count_of_appointments']/30).round(1)
ar_df['utilisation'] = (ar_df['avg_daily_app']/1200000).round(1)
# View the DataFrame

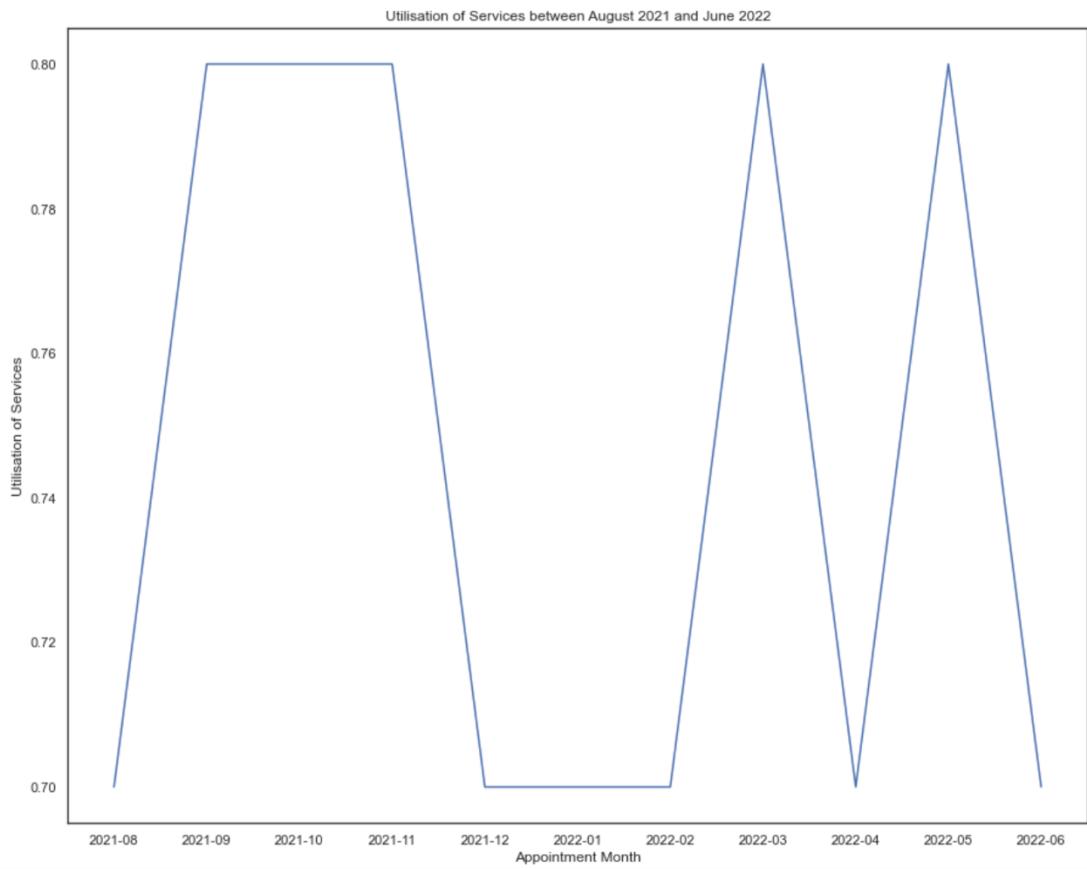
ar_df['utilisation'] = (ar_df['avg_daily_app']/1200000).round(1)
ar_df = ar_df[['appointment_month', 'count_of_appointments',
               'avg_daily_app', 'utilisation']]
ar_df
```

Out[803]:

	appointment_month	count_of_appointments	avg_daily_app	utilisation
0	2021-08	23852171	795072.4	0.7
1	2021-09	28522501	950750.0	0.8
2	2021-10	30303834	1010127.8	0.8
3	2021-11	30405070	1013502.3	0.8
4	2021-12	25140776	838025.9	0.7
5	2022-01	25635474	854515.8	0.7
6	2022-02	25355260	845175.3	0.7
7	2022-03	29595038	986501.3	0.8
8	2022-04	23913060	797102.0	0.7
9	2022-05	27495508	916516.9	0.8
10	2022-06	25828078	860935.9	0.7

## 6.d.

```
In [673]: # Plot monthly capacity utilisation.
# Create a lineplot.
utilisation_ax = sns.lineplot(
    data=ar_df, x='appointment_month', y='utilisation', ci=None)
# Customise and tidy the plot
utilisation_ax.set_xlabel('Appointment Month')
utilisation_ax.set_ylabel('Utilisation of Services')
utilisation_ax.set_title(
    "Utilisation of Services between August 2021 and June 2022")
```



## 7.a.

```
In [806]: # Create a new to isolate the DNA appointment status
dna = ar_agg[ar_agg['appointment_status']=='DNA']

# View the DataFrame
dna.head()
```

	appointment_month	hcp_type	appointment_status	appointment_mode	time_between_book_and_appointment	count_of_appointments
40	2021-08	GP	DNA	Face-to-Face	1 Day	16314
41	2021-08	GP	DNA	Face-to-Face	15 to 21 Days	12960
42	2021-08	GP	DNA	Face-to-Face	2 to 7 Days	43754
43	2021-08	GP	DNA	Face-to-Face	22 to 28 Days	7457
44	2021-08	GP	DNA	Face-to-Face	8 to 14 Days	24346

```
In [946]: # Create a DataFrame to group the values for DNA
dna_group = dna.groupby(['appointment_month']).sum().reset_index()
dna_group
```

```
Out[946]:
```

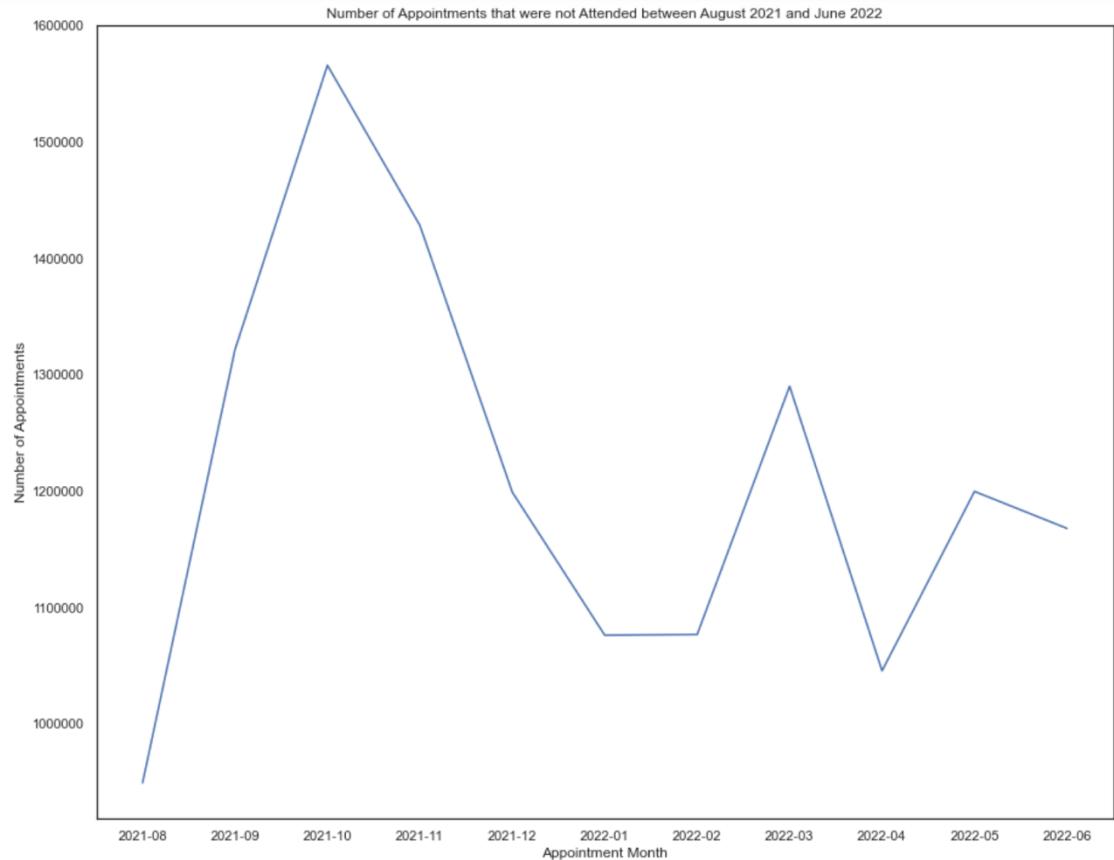
	appointment_month	count_of_appointments
0	2021-08	949137
1	2021-09	1321348
2	2021-10	1565624
3	2021-11	1428087
4	2021-12	1198866
5	2022-01	1076013
6	2022-02	1076658
7	2022-03	1289888
8	2022-04	1045455
9	2022-05	1199518
10	2022-06	1167790

## 7.b.

```
In [950]: # Create a line plot to show number of DNA appointments.
dna_ax = sns.lineplot(x='appointment_month', y='count_of_appointments', data=dna_group, ci=None)

dna_ax.set_xlabel('Appointment Month')
dna_ax.set_ylabel('Number of Appointments')
dna_ax.set_title(
    "Number of Appointments that were not Attended between August 2021 and June 2022")

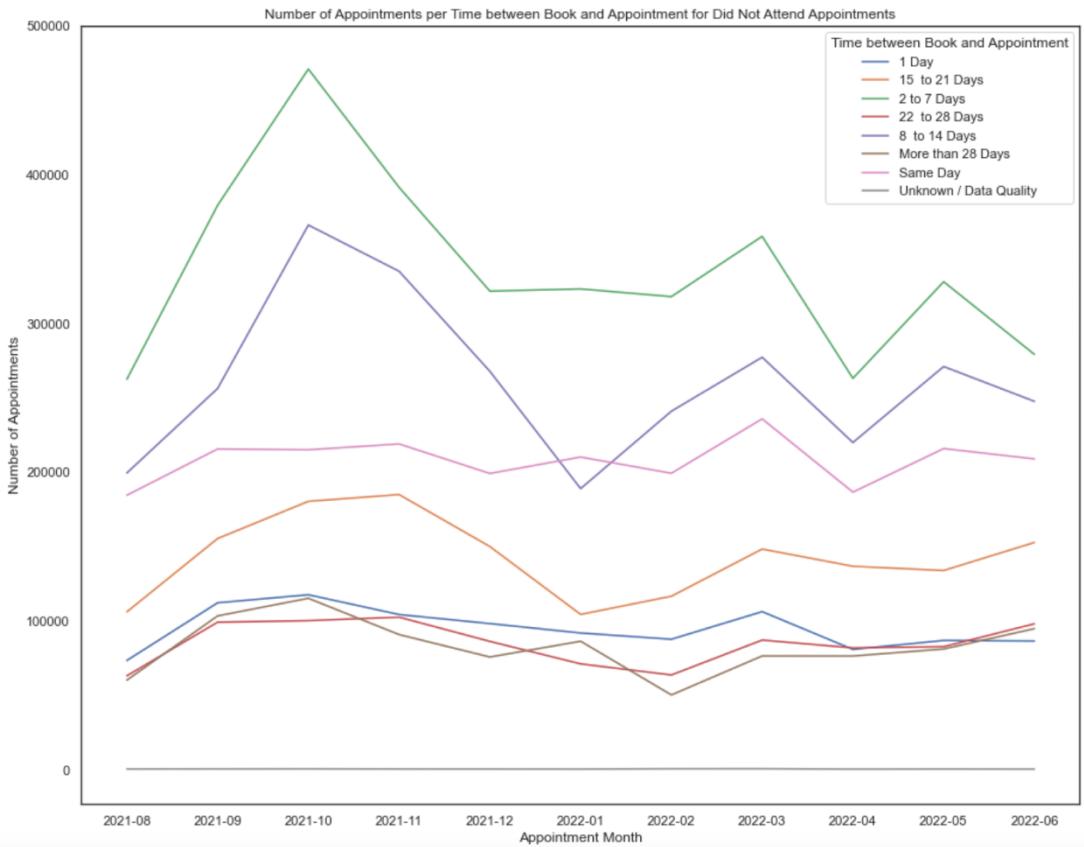
dna_ax.set_yticks([1000000, 1100000, 1200000, 1300000, 1400000, 1500000, 1600000])
dna_ax.set_yticklabels(['1000000', '1100000', '1200000', '1300000', '1400000', '1500000', '1600000'])
```



## 7.C.

```
In [990]: dna_time_ax = sns.lineplot(x='appointment_month', y='count_of_appointments', \
                                   hue='time_between_book_and_appointment', data=dna_time, ci=None)

# Customise and tidy the graph
dna_time_ax.legend(title='Time between Book and Appointment', loc='upper right')
dna_time_ax.set_xlabel('Appointment Month')
dna_time_ax.set_ylabel('Number of Appointments')
dna_time_ax.set_title(
    "Number of Appointments per Time between Book and Appointment for Did Not Attend Appointments")
dna_time_ax.set_yticks([0, 100000, 200000, 300000, 400000, 500000])
dna_time_ax.set_yticks([50000, 150000, 250000, 350000,
                       450000], minor=True)
dna_time_ax.set_yticklabels(['0', '100000', '200000', '300000',
                           '400000', '500000'])
```



## 7.d.

```
In [808]: # Create a new subset on the 8 to 14 Days time between book and appointment
eight_fourteen = ar_agg[ar_agg['time_between_book_and_appointment']
                      == '8 to 14 Days']

# View the output
eight_fourteen.head()
```

	appointment_month	hcp_type	appointment_status	appointment_mode	time_between_book_and_appointment	count_of_appointments
4	2021-08	GP	Attended	Face-to-Face	8 to 14 Days	398772
12	2021-08	GP	Attended	Home Visit	8 to 14 Days	1060
20	2021-08	GP	Attended	Telephone	8 to 14 Days	488939
28	2021-08	GP	Attended	Unknown	8 to 14 Days	7736
36	2021-08	GP	Attended	Video/Online	8 to 14 Days	3561

```
In [761]: # From the new output - groupby hcp_type for 8 to 14 Days time between book and appointment
type_8_14 = eight_fourteen.groupby(['appointment_month', 'hcp_type']).sum().reset_index()

# View the output
type_8_14.head()
```

	appointment_month	hcp_type	count_of_appointments
0	2021-08	GP	975628
1	2021-08	Other Practice staff	1867275
2	2021-08	Unknown	46247
3	2021-09	GP	1253666
4	2021-09	Other Practice staff	2044882

## 7.e.

```
In [810]: # Create a new subset on the 2 to 7 Days time between book and appointment
two_seven = ar_agg[ar_agg['time_between_book_and_appointment']
                  == '2 to 7 Days']

# View the output
two_seven
```

	appointment_month	hcp_type	appointment_status	appointment_mode	time_between_book_and_appointment	count_of_appointments
2	2021-08	GP	Attended	Face-to-Face	2 to 7 Days	959486
10	2021-08	GP	Attended	Home Visit	2 to 7 Days	7012
18	2021-08	GP	Attended	Telephone	2 to 7 Days	975156
26	2021-08	GP	Attended	Unknown	2 to 7 Days	20961
34	2021-08	GP	Attended	Video/Online	2 to 7 Days	13169
...	...	...	...	...	...	...
3723	2022-06	Unknown	Unknown	Face-to-Face	2 to 7 Days	3146
3731	2022-06	Unknown	Unknown	Home Visit	2 to 7 Days	984
3739	2022-06	Unknown	Unknown	Telephone	2 to 7 Days	1000
3747	2022-06	Unknown	Unknown	Unknown	2 to 7 Days	4929
3753	2022-06	Unknown	Unknown	Video/Online	2 to 7 Days	1

484 rows × 6 columns

```
In [811]: # From the new output - groupby hcp_type for 2 to 7 Days time between book and appointment
type_2_7 = two_seven.groupby(['appointment_month', 'hcp_type']).sum().reset_index()

# View the output
type_2_7.head()
```

	appointment_month	hcp_type	count_of_appointments
0	2021-08	GP	2107045
1	2021-08	Other Practice staff	2709478
2	2021-08	Unknown	108437
3	2021-09	GP	2549886
4	2021-09	Other Practice staff	3276358

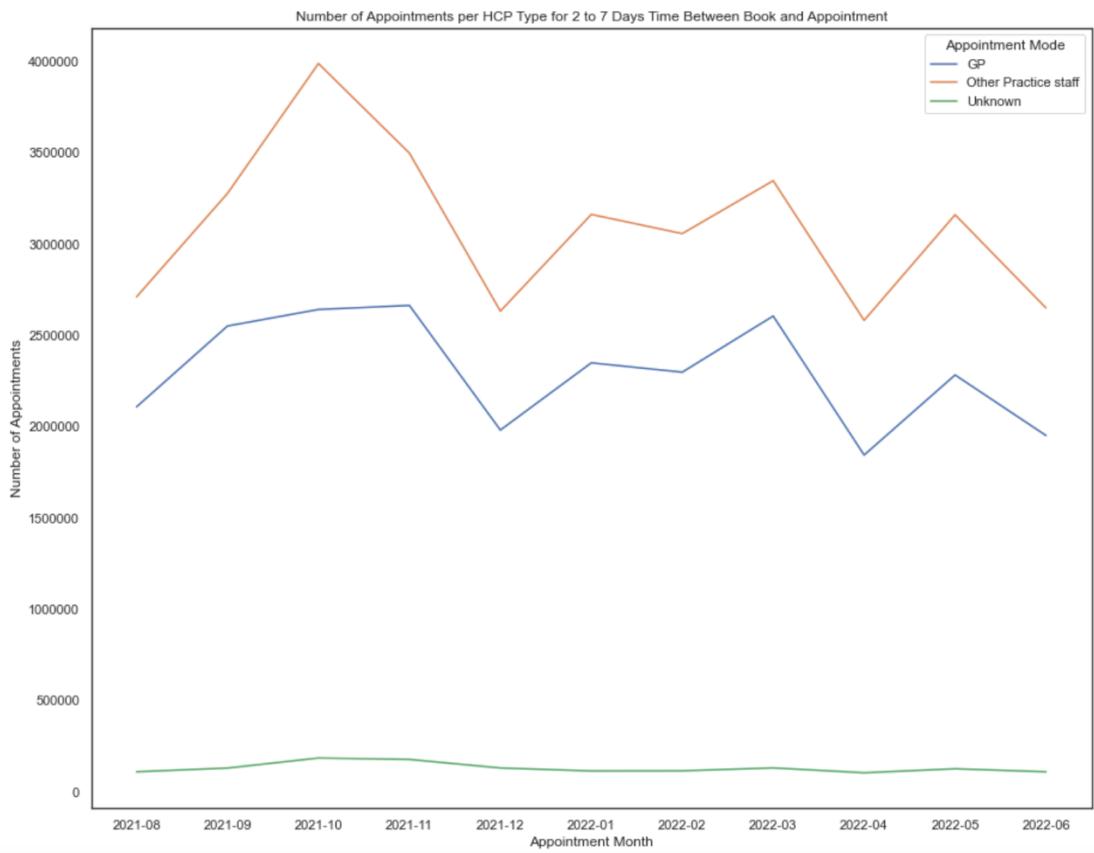
## 7.f.

```
In [812]: # Plot a lineplot to show the relationship of hcp for 8 to 14 days
eight_fourteen_ax = sns.lineplot(x='appointment_month', y='count_of_appointments', \
                                   hue='hcp_type', data=type_8_14, ci=None)

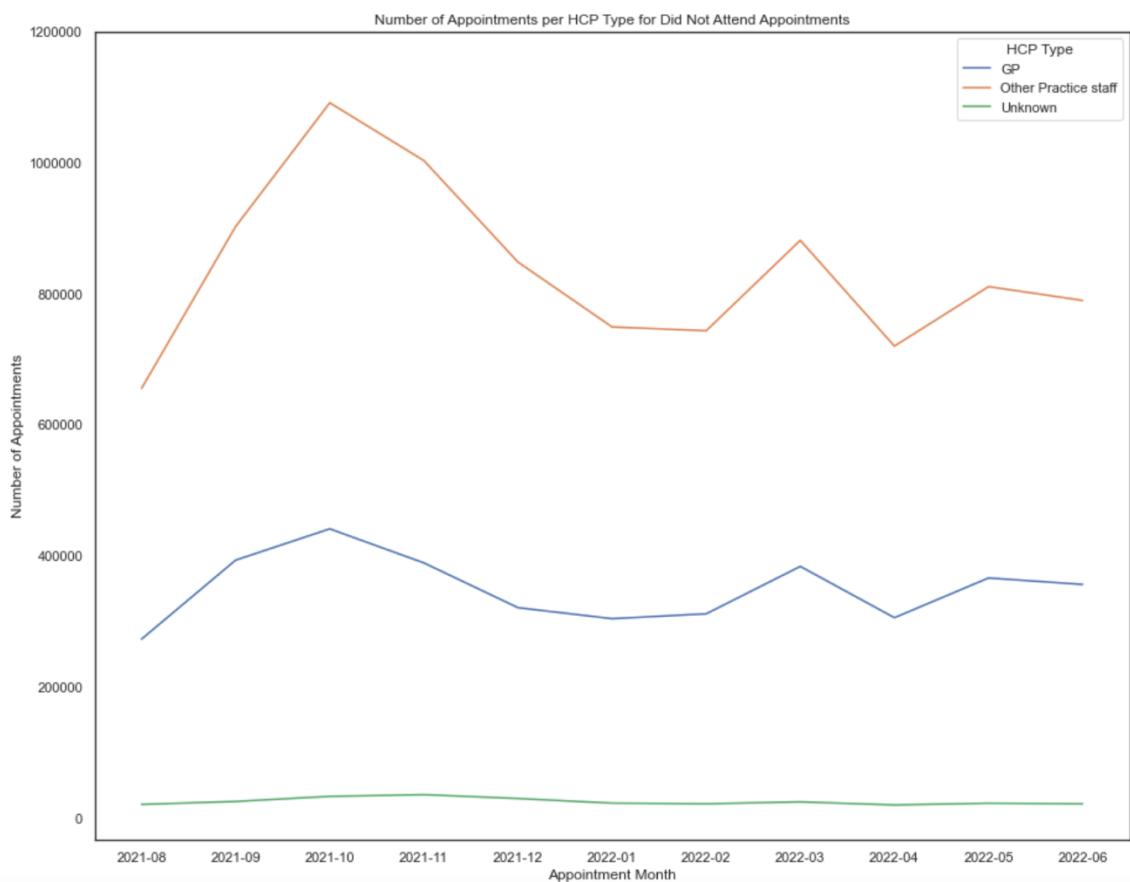
# Customise and tidy the graph
eight_fourteen_ax.legend(title='Appointment Mode', bbox_to_anchor=(1, 1))
eight_fourteen_ax.set_xlabel('Appointment Month')
eight_fourteen_ax.set_ylabel('Number of Appointments')
eight_fourteen_ax.set_title(
    "Number of Appointments per HCP Type for 8 to 14 Days Time Between Book and Appointment")
eight_fourteen_ax.set_yticks([0, 500000, 1000000, 1500000, 2000000, 2500000, 3000000])
eight_fourteen_ax.set_yticks([250000, 750000, 1250000, 1750000,
                            2250000, 2750000], minor=True)
eight_fourteen_ax.set_yticklabels(['0', '500000', '1000000', '1500000', '2000000', '2500000',
                                  '3000000'])
```



7.g.



7.h.



## 7.i.

```
In [983]: # Create a new DataFrame to isolate Other Practice Staff HCP Type
ops_ar = ar_agg[ar_agg['hcp_type'] == 'Other Practice staff']
# Filter new DataFrame further to group appointments for Other Practice Type
ops_ar_group = ops_ar.groupby(['appointment_month', 'hcp_type']).sum().reset_index()
ops_ar_group
```

Out[983]:

	appointment_month	hcp_type	count_of_appointments
0	2021-08	Other Practice staff	10797821
1	2021-09	Other Practice staff	13126731
2	2021-10	Other Practice staff	14942504
3	2021-11	Other Practice staff	14432800
4	2021-12	Other Practice staff	11614470
5	2022-01	Other Practice staff	11705142
6	2022-02	Other Practice staff	11688107
7	2022-03	Other Practice staff	13528583
8	2022-04	Other Practice staff	11217738
9	2022-05	Other Practice staff	12823200
10	2022-06	Other Practice staff	12105677

```
In [985]: # Create a new data frame to focus on Other Practice Staff DNA appointments
ops_dna = dna_type[dna_type['hcp_type'] == 'Other Practice staff']
# Rename column for future merge
ops_dna.rename(columns={'count_of_appointments':'count_dna_appointments'}, inplace=True)
# Merge two df
ops_dna = pd.merge(ops_dna, ops_ar_group[['appointment_month','count_of_appointments']],
                  on='appointment_month', how='left')
# Calculate the percentage appointments that are not attended
ops_dna['percentage_dna'] = ((ops_dna['count_dna_appointments']/ops_dna['count_of_appointments'])*100).round(2)
ops_dna
```

Out[985]:

	appointment_month	hcp_type	appointment_status	count_dna_appointments	count_of_appointments	percentage_dna
0	2021-08	Other Practice staff	DNA	655689	10797821	6.07
1	2021-09	Other Practice staff	DNA	902812	13126731	6.88
2	2021-10	Other Practice staff	DNA	1091576	14942504	7.31
3	2021-11	Other Practice staff	DNA	1003396	14432800	6.95
4	2021-12	Other Practice staff	DNA	848447	11614470	7.31
5	2022-01	Other Practice staff	DNA	749418	11705142	6.40
6	2022-02	Other Practice staff	DNA	743623	11688107	6.36
7	2022-03	Other Practice staff	DNA	881640	13528583	6.52
8	2022-04	Other Practice staff	DNA	720150	11217738	6.42
9	2022-05	Other Practice staff	DNA	810955	12823200	6.32
10	2022-06	Other Practice staff	DNA	789911	12105677	6.53