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School of Computing  
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**Assessed Coursework  
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**To be completed by the student submitting work**  
(please make sure that you complete ALL fields)

Module Code

C O M P E I 2 2 M

Module Title

Data Science (33862)

Coursework  
Number

Coursework 2

Name of Lecturer  
Marking Work

Professor Roy Ruddle

**\*\*You must fill this in\*\***

Deadline Date

3rd December 2021

Date Handed In

3rd December 2021

My signature here confirms acceptance of the  
declaration of academic integrity printed  
overleaf.

Student  
Name:

Himanshu Jaiswal

Signature:

*[Handwritten Signature]*

SID  
Number:

2 0 1 5 7 7 1 6 2

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Feedback section to be completed by the marker

Section	Marks Available	Marks Awarded	Comments on sections
<b>Total</b>			
Late Penalty (marks to be deducted)			

General Comments on Work

Other members of the group:


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*\*For the submission of group work*

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# Coursework 2

## Group Analysis Case Study

### Introduction

This report is submitted as coursework 2 for the Data Science module taught as a part of the MSc Data Science and Analytics programme at the University Of Leeds.

In this group analysis case study,

#### We aim to:

- Analyse the data containing NHS Dental Statistics for England for the period **April 2018 to March 2019**, at the level of Clinical Commissioning Groups (CCGs).
- Using the data and analysis, answer the following questions:
  - ❖ What patterns are there in the number/age of patients treated?
  - ❖ What is the effect of deprivation?
  - ❖ A person aged 55 in 2019 will reach the State Pension age in 2031. What types of CCGs face the greatest shortage of dentists in 2031?

#### The data:

#### *We used the following datasets for our analysis:*

- **The CCG and Patient Part:**

1. nhs-dent-stat-eng-jan-jun-18-anx3-ps-prac.csv ([Link to dataset](#))
2. nhs-dent-stat-eng-jul-dec-18-anx3-ps-prac.csv ([Link To dataset](#))
3. nhs-dent-stat-eng-jan-jun-19-anx3-ps-prac.csv ([Link to dataset](#))

- **The Dentist Part:**

1. NHS Dental Statistics for England 2019-20 Annex 3\_Workforce csv.csv ([Link to dataset](#))

- **The Deprivation Part:**

1. File\_7\_-\_All\_IoD2019\_Scores\_\_Ranks\_\_Deciles\_and\_Population\_Denominators\_3.csv ([Link to dataset](#))

The datasets are **publicly available** on the UK government's website.

## Data Characterisation:

### Understanding the data:

We merged the above mentioned data files into one single dataset using the join method in python.

There are 11 columns in our final file:

1. **PSEEN\_END\_DATE:** Last date of the month in which the patient was seen by a dentist.
2. **CCG\_CODE:** CCG code of the CCG commissioning health care.
3. **CCG\_ONS\_CODE:** The code of the CCG with the ONS. ONS stands for Office for National Statistics.
4. **CCG\_NAME:** The name of the Clinical Commissioning Group.
5. **PATIENT\_TYPE:** Whether a patient is a child or an adult.
6. **AGE\_BAND:** Determines the age of the patient.
7. **PATIENTS\_SEEN:** Number of patients seen in a month within a given CCG.
8. **POPULATION:** Population of the CCG.
9. **DENTIST\_AGE\_GROUP:** Categorisation of the dentist's age.
10. **NUMBER\_OF\_DENTISTS:** number of dentists in that particular CCG.
11. **AVG\_IMD\_SCORE:** IMD stands for Index of Multiple Deprivation. It is to classify the relative deprivation of small areas.

## Data Preparation

### Profiling the data:

#### Characterisation Tasks

The following table explains the variable type of each column of the data along with the minimum and maximum values by giving examples from each column. The empty cells represent that there is no minimum or maximum value for that particular column.

Columns	Data Type	Min.	Max.	Data Sample
<b>PSEEN_END_DATE</b>	Numeric	30/04/2018	30/06/2019	28/02/2019
<b>CCG_CODE</b>	Nominal	00C	99Q	99K, 07J
<b>CCG_ONS_CODE</b>	Nominal	E38000001	E38000230	E38000100
<b>CCG_NAME</b>	Nominal	----	----	NHS Darlington CCG
<b>PATIENT_TYPE</b>	Categorical	----	----	Adult,Child
<b>AGE_BAND</b>	Numeric	0	18	11,12,13..
<b>PATIENTS_SEEN</b>	Numeric	3	499,995	42, 398, 1022, 927
<b>POPULATION</b>	Numeric	0	962,934	1269, 1328
<b>DENTIST_AGE_GROUP</b>	Ordinal	Under 35	55+	35-44,45-54
<b>NUMBER_OF_DENTISTS</b>	Numeric	4	258	11,13,18..
<b>AVG_IMD_SCORE</b>	Numeric	7.18	52.139	9.179, 10.4



Columns	Distinct Values	Average Value	Length
<b>PSEEN_END_DATE</b>	12	----	10
<b>CCG_CODE</b>	198	----	3-11
<b>CCG_ONS_CODE</b>	192	----	9-11
<b>CCG_NAME</b>	192	----	11-57
<b>PATIENT_TYPE</b>	2	Child	5
<b>AGE_BAND</b>	19	9	1-2
<b>PATIENTS_SEEN</b>	8491	7903	3-8
<b>POPULATION</b>	2664	15164	4-9
<b>DENTIST_AGE_GROUP</b>	4	----	2-3+
<b>NUMBER_OF_DENTISTS</b>	127	43	1-4
<b>AVG_IMD_SCORE</b>	191	22	5-6

## Data Quality:

There were several data quality issues within the data:

### 1. Missing/Spurious Values:

- 1.1. Columns **CCG\_CODE**, **CCG\_ONS\_CODE** and **CCG\_NAME** have a few spurious values which can be observed as a case of missing values. Few rows are having “UNALLOCATED” as their value, which can be assumed to

be a case of missing values but since they are just a few, we can ignore them for our analysis.

- 1.2. Also, **Dentist\_Age\_Groups** and **Number Of Dentists** have missing values in some of the rows which is again a missing data issue.

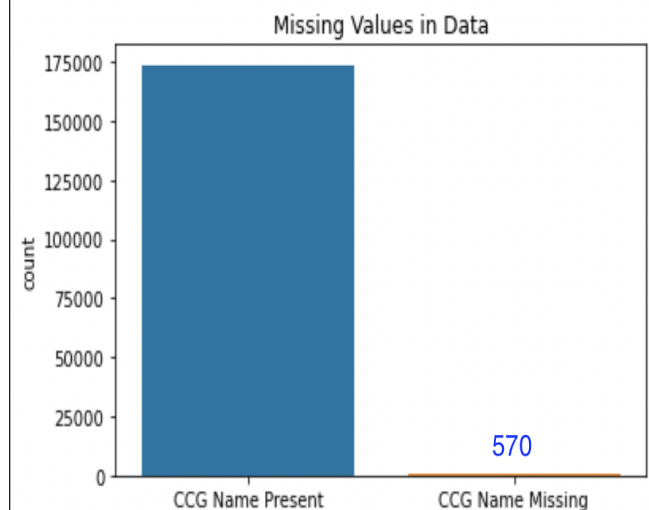
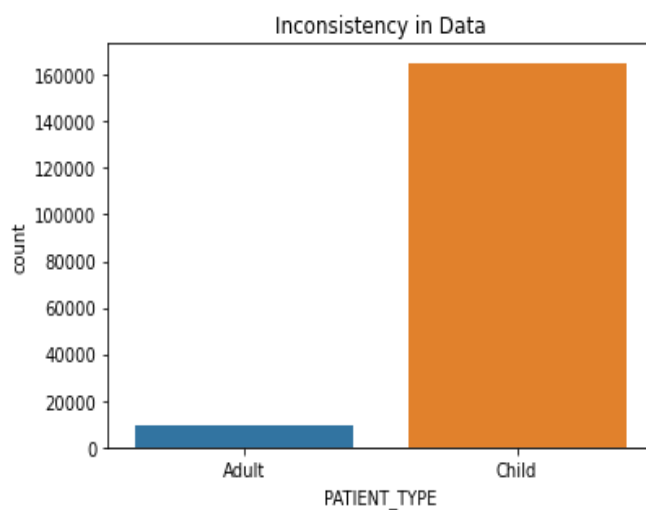
## 2. Duplicate Records:

- 2.1. While merging the datasets, we observed that the number of Dentist's data for every CCG is not available and thus few duplicate records were formed in the dataset.

So, we performed **DATA STANDARDISATION** and cleaned the data to minimize duplicacy of records.

## 3. Inconsistency:

- 3.1. The number of records in the patients data shows inconsistency as for adults we have cumulative data for the past 24 months but for children, it is just for the past 12 months. So, we have to be mindful of this fact while performing analysis.



# Detailed Analysis:

## a). Analysis of patterns seen in number/age of patients treated :

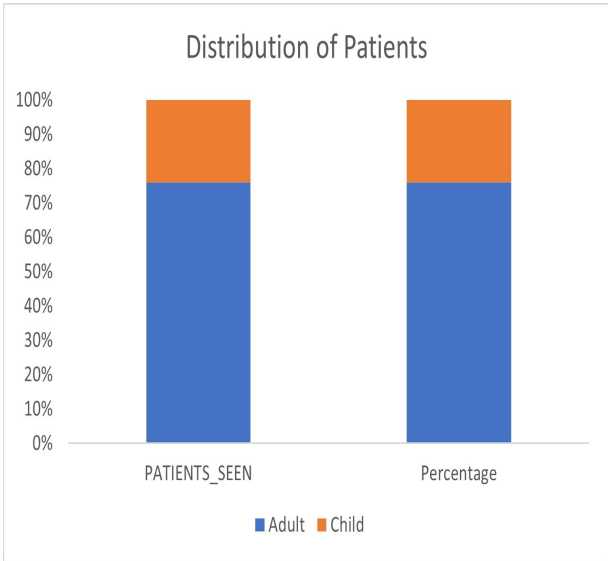


Fig. 1(a)

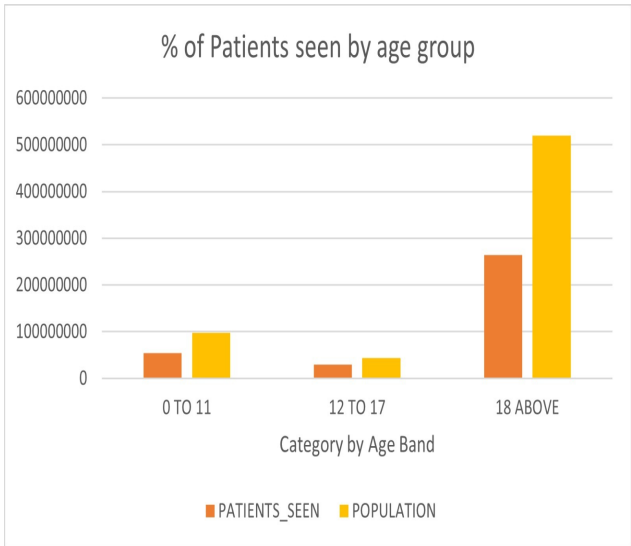


Fig. 1(b)

The bar chart Fig. 1(b) depicts the ratio of the population to the patients treated in different age bands. The population of the adults being the maximum, it makes sense from the chart that we have the maximum number of patients treated in that age group. We have the least number of patients who are children considering the small population of them. Furthermore, we could see how we have more patients in the age band of 0 to 11 than 12 to 17. It again makes sense considering the fact that babies need more care & regular health checkups in their age. Fig. 1(a) shows the distribution of patients based on the patients seen and the percentage of patients seen in adults and children. As we can see there are more adult(76%) patients than children(24%).

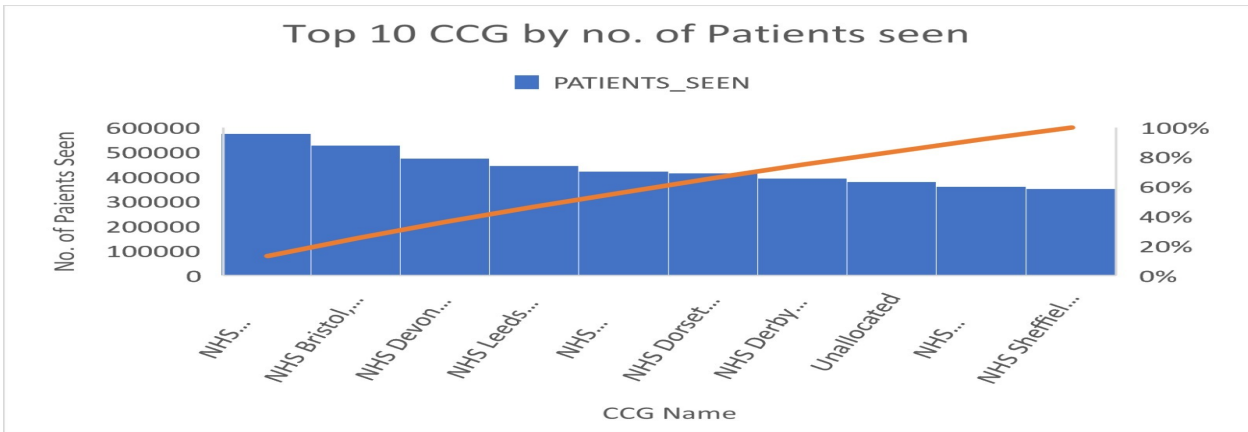


Fig. 1(c)

The graph above shows the top 10 CCGs by the number of patients seen which depicts where the most number of patients are seen in CCG's.

## b). Analysis on effects of deprivation:

The average IMD Score explains how deprived the CCGs are, the score is directly proportional to deprivation in the CCGs. The following scatter plot shows different ranges of avg IMD scores of all CCGs which has been broken down into five categories. Category 1: 40 & above, Category 2: 30 to 39, Category 3: 20 to 29, Category 4: 10 to 19, Category 5: Below 10.

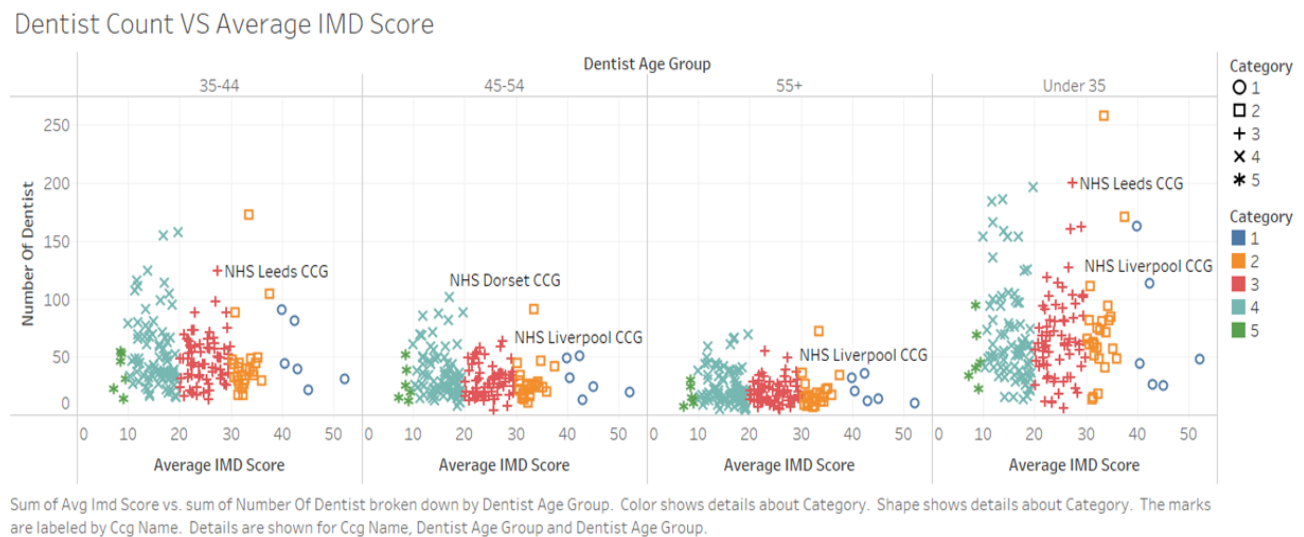


Fig. 2(a)

Less the age-group of the dentist is, the more scattered the data points(CCGs) are in our scatter plot. The age group under 35 showed the highest scatter which explains how high the dentist count is with some variance & there are least number of dentists above the age of 55. We have many young dentists serving across most of the CCGs having avg. IMD score ranging from 19 to 30. Though we could analyse some of the patterns as explained above from our plot, this isn't enough to conclude that there is any dependency between the dentist count and avg IMD score and so the deprivation.

From the below graph we can infer that as the Percentage of the Patients\_seen decreases the AVG\_IMD\_SCORE also decreases across the categories 1-5 of CCGs. Here we can also observe that as the age of the dentist increases, the dentist count decreases in each of these categories. So we can conclude that the more the CCG is deprived, the more patients have been seen in those CCGs.



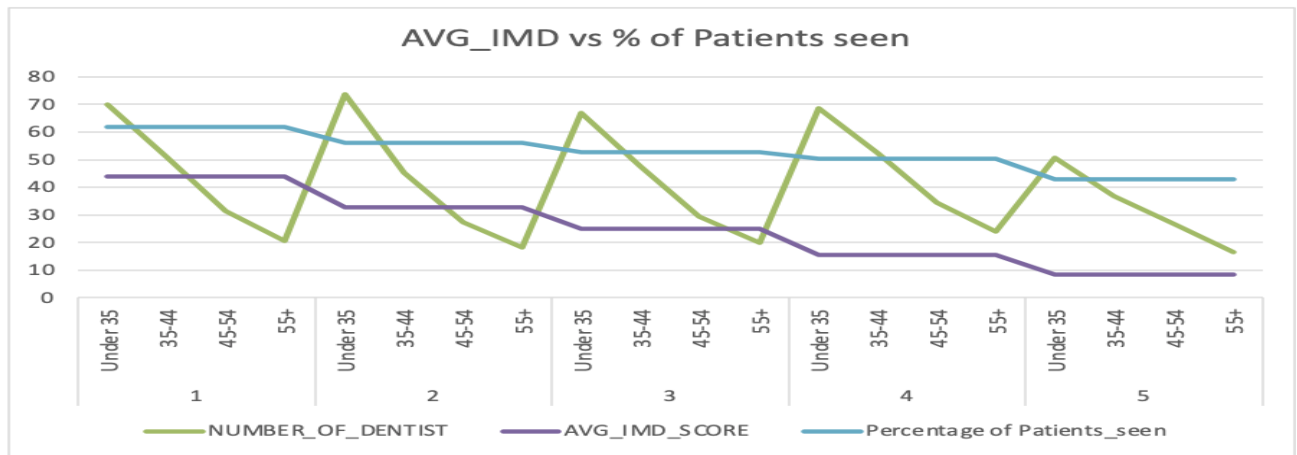


Fig. 2(b)

c). A person aged 55 in 2019 will reach the State Pension age in 2031. What types of CCGs face the greatest shortage of dentists in 2031?

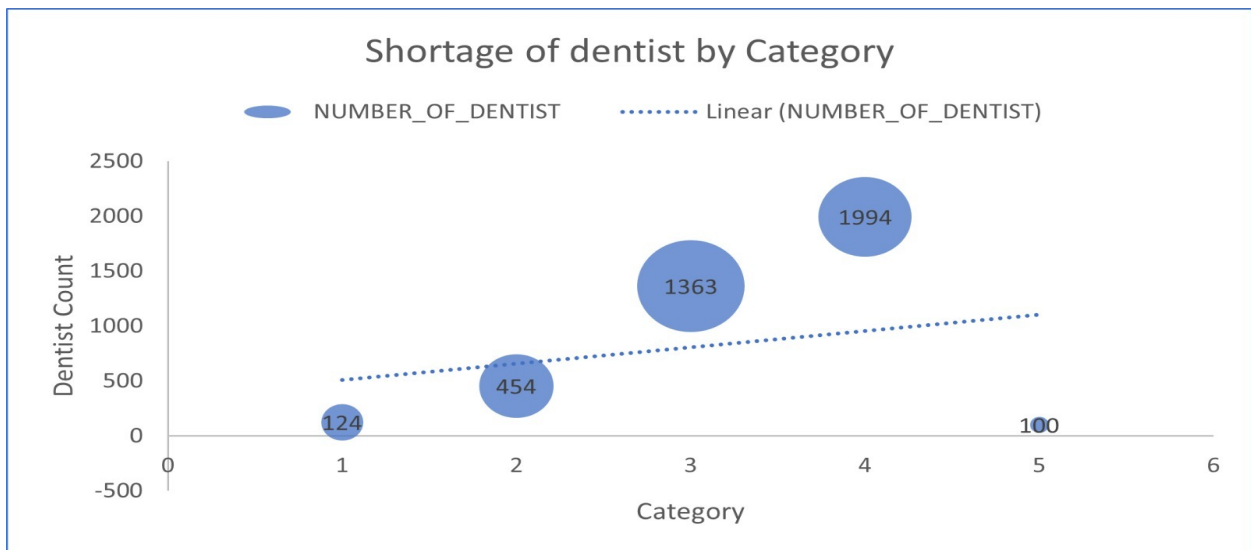


Fig. 3(a)

The graph plots the number of dentists above the age of 55 for all the 5 categories of CCGs from which we can infer that the one in which there will be the most shortage of dentists would be category 4 (avg IMD ranging from 10-19) followed by 3 (avg IMD ranging from 20-29) in 2031. We can also say that the least shortage of dentists would be in category 5 (avg IMD ranging below 10) and 1 (avg IMD ranging above 40) as it is clearly evident from the graph above. In the graph below Fig. c(2) we can see that % of patients who are 55+ are mostly 30% of total dentist count which is very significant. Therefore, when these 30% of dentists will retire there will be a huge shortage of dentists in 2031.

Orange color percentage represents 55+ aged dentists and blue color percentage depicts below 55 aged dentists in 2019.

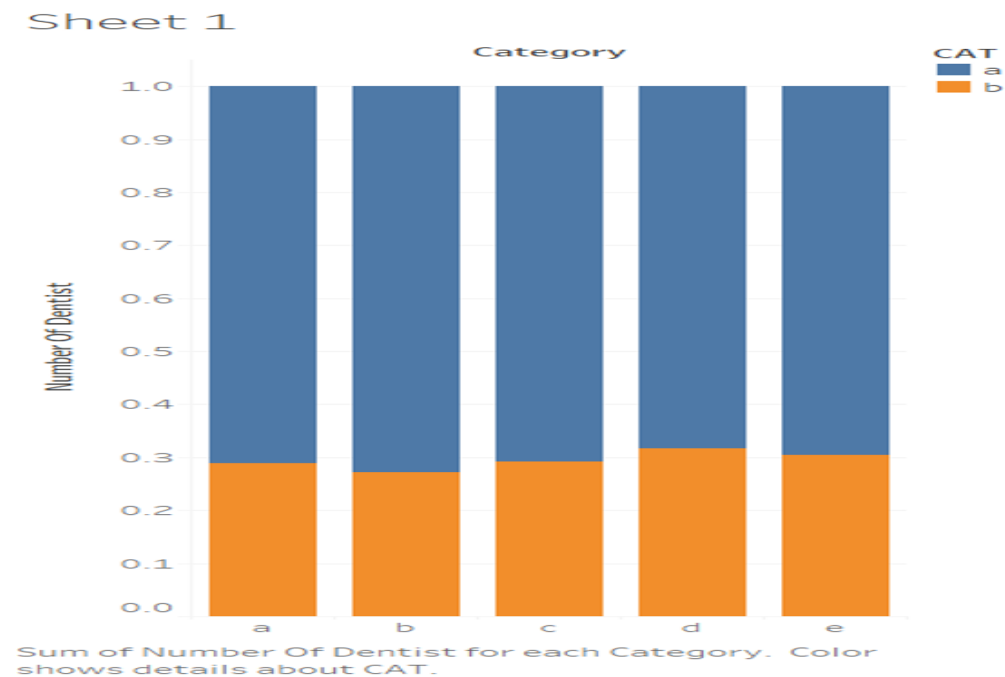


Fig. c(2)

## Conclusion:

- We have inferred by performing the analysis on our data that there are maximum patients aged above 18 and least from the age group 12-17.
- We have analysed that the percentage of patients seen is directly proportional to average IMD score.
- Among the patients that have been seen, 76% are adults and 24% are children.
- We have analysed that the number of dentists are maximum for the age-group under 35 and as the age increases, the number of dentists decreases so the minimum is for the age-group above 55.
- The CCG which has average IMD score between 20 - 40 will face the most shortage of dentists in 2031.
- This data and the analysis will help NHS in future to allocate funds and dentists to ensure improvement in deprived areas, and will also be used as evidence so that the government and organization can study in which area they should provide economical and healthcare support.