

## Module Project - PSY6422

### Changes in the number of referrals for autism diagnostic assessment before, during, and after lockdown in England by age group:

#### Research Question:

How did the pandemic affect the number of referrals for autism assessments in NHS practices and did this vary by age?

The pandemic resulted in school closures and disruption in routine for most children in England giving parents unprecedented amounts of time around their school-aged children and also resulting in a period of stress and uncertainty, providing an environment for the exacerbation of symptoms of autism that would be visible to caregivers. This may have created a situation in which parents were well placed to identify potential autism in their children in a way in which they were unable to before. This period of stress and uncertainty also may have exacerbated previously unnoticed symptoms in undiagnosed adults prompting an increase in referrals in older age groups.

## Coronavirus: Autism makes coping with lockdown 'extra hard'

🕒 12 May 2020

<https://www.bbc.co.uk/news/uk-scotland-52618931>

## New special school needed amid pupil autism rise - council

🕒 26 March 2022

<https://www.bbc.co.uk/news/uk-england-tyne-60885874>

### Interview

**Siena Castellon: 'Autistic people are really struggling with how uncertain things are'**

<https://www.theguardian.com/society/2020/nov/11/siena-castellon-autistic-people-are-really-struggling->

with-how-uncertain-things-are

# Autism more common in children in England than previously thought – study

<https://www.theguardian.com/society/2021/mar/29/autism-more-common-in-children-in-england-than-previously-thought-study>

https:

## The Data:

Data was taken from NHS digital who began publishing monthly statistics for the number of autism referrals (amongst other variables such as wait time) for practices in England. The data used here is the result of merging two datasets covering April 2019 - December 2021 and July 2021 - June 2022 and these datasets can be accessed via these links:

<https://digital.nhs.uk/data-and-information/publications/statistical/autism-statistics/april-2019-to-december-2021>

<https://digital.nhs.uk/data-and-information/publications/statistical/autism-statistics/july-2021-to-june-2022>

The data was recorded in this format:

	REPORTING_PERIOD_START	REPORTING_PERIOD_END	STATUS
144294	01/04/2019	30/04/2019	Latest Available Data
144318	01/04/2019	30/04/2019	Latest Available Data
144350	01/04/2019	30/04/2019	Latest Available Data

	BREAKDOWN	PRIMARY_LEVEL
144294	CCG - GP Practice or Residence; Age Group	01G
144318	CCG - GP Practice or Residence; Age Group	01K
144350	CCG - GP Practice or Residence; Age Group	01X

	PRIMARY_LEVEL_DESCRIPTION	SECONDARY_LEVEL	SECONDARY_LEVEL_DESCRIPTION
144294	NHS SALFORD CCG	Under 18	People aged under 18
144318	NHS MORECAMBE BAY CCG	Under 18	People aged under 18
144350	NHS ST HELENS CCG	Under 18	People aged under 18

	METRIC	METRIC_DESCRIPTION
144294	ASD12 Number of new "suspected autism" referrals in the month	
144318	ASD12 Number of new "suspected autism" referrals in the month	
144350	ASD12 Number of new "suspected autism" referrals in the month	

	METRIC_VALUE
144294	10
144318	10
144350	10

The dataset comes from a larger dataset that is also from the NHS: the Mental Health Services Dataset. Data submissions by services are made for each patient whose care is funded fully or partially by the NHS and data submission is optional for services with patients who are not funded by the NHS. This means that individuals seeking private care or assessment are not included in this dataset.

## Data preparation:

These datasets were very large and included metrics that this analysis was uninterested in and also had to be merged and duplications deleted. There were also some formatting changes necessary to make the data more understandable to R such as the conversion of data from character to date class. The final stage of this was to sum the referrals by age across practices as this analysis was not focused on region. The full process of data cleaning and pre-processing is illustrated below.

```
#loading in the data
Apr19_Dec21 <- read.csv(unz(here("data","Zipped_total_data.zip"),"April2019_Dec2021.csv"))
Jul21_Jun22 <- read.csv(unz(here("data","Zipped_total_data.zip"),"July2021_June2022.csv"))

#making a function
fval_tail <- function(df,col_num,value) {
  #find a value in the data frame that matches input 'value', find the element number of the last
#instance of this
  index <- which(df[,col_num] == value)
  output <- tail(index,1)
}

#create a new dataframe for the second dataset with only dates after this
lastdupind <- fval_tail(Jul21_Jun22,1,tail(Apr19_Dec21$REPORTING_PERIOD_START, n=1))
nondup_Jun21_Jul22 <- Jul21_Jun22[-c(1:lastdupind),]

#merge
merged <- rbind(Apr19_Dec21,nondup_Jun21_Jul22)
merged_subset <- subset(merged,merged$METRIC == 'ASD12' &
  merged$SECONDARY_LEVEL != 'NONE' &
  merged$BREAKDOWN != 'CCG - GP Practice or Residence; Age Group')

#format the data classes
classform_ms <- mutate(merged_subset,REPORTING_PERIOD_START =
  as.Date(REPORTING_PERIOD_START, format = "%d/%m/%Y"))
classform_ms <- mutate(classform_ms,REPORTING_PERIOD_END =
  as.Date(REPORTING_PERIOD_END, format = "%d/%m/%Y"))
classform_ms <- mutate(classform_ms,METRIC_VALUE = as.numeric(METRIC_VALUE))

#removing any NA/unknown values
valonly_full <- classform_ms[!is.na(classform_ms$METRIC_VALUE),]
valonly_full <- classform_ms[classform_ms$SECONDARY_LEVEL != "UNKNOWN",]

#making the legend label order in age order
valonly_full$SECONDARY_LEVEL[valonly_full$SECONDARY_LEVEL == 'Under 18'] <- '0 - 17'

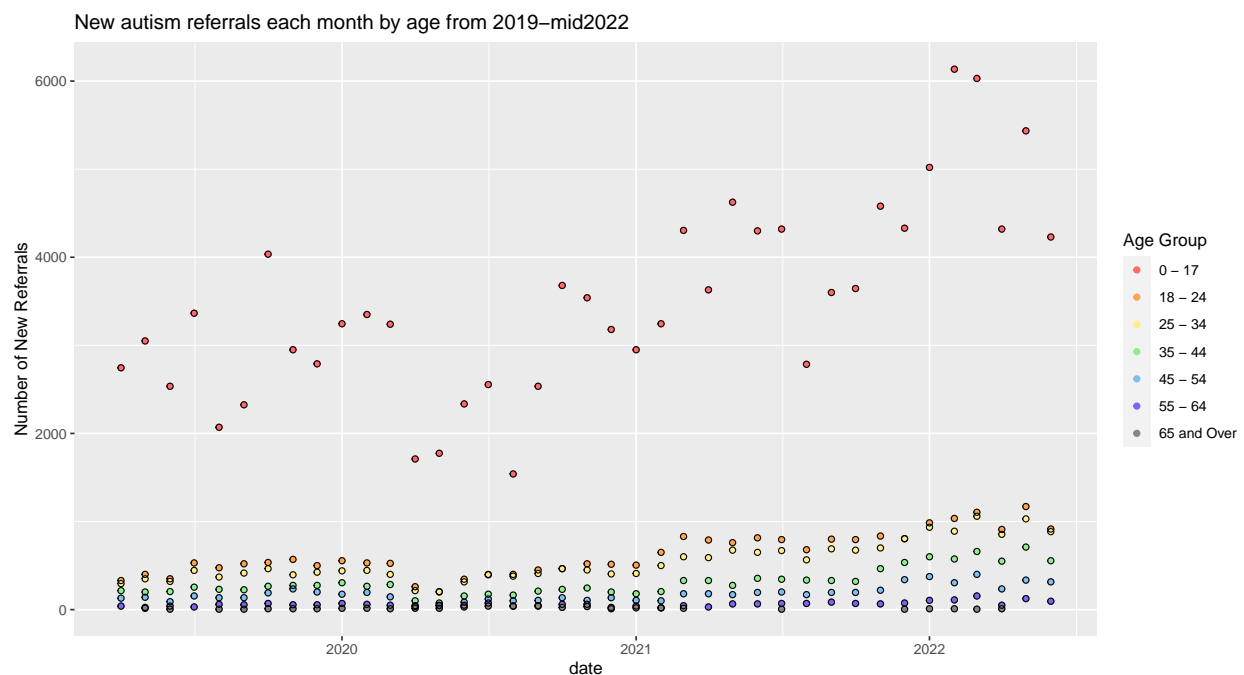
#Sum by age and date
ref_age_date <- aggregate(METRIC_VALUE ~ REPORTING_PERIOD_START + SECONDARY_LEVEL,
  data = valonly_full, FUN = "sum")
ref_age_date <- data.frame(ref_age_date)
```

## Initial visualisation:

I initially visualised the data using a scatter plot but found it difficult to interpret as the points were small and often overlapped, especially for the older age groups with smaller referral numbers.

```
#create an initial visualisation - here, a scatterplot
initialplot <- ggplot(ref_age_date, aes(x = REPORTING_PERIOD_START,
                                         y = METRIC_VALUE, color = SECONDARY_LEVEL)) +
  geom_point(size = 1.5) +
  labs(title = "New autism referrals each month by age from 2019-mid2022",
       x = "date", y = "Number of New Referrals") +
  guides(color=guide_legend("Age Group"))+
  scale_color_manual(values=c("#FF6A6A", "#FFA54F", "#FFEC8B", "#90EE90",
                              "#7EC0EE", "#7A67EE", "#8B8386")) +
  geom_point(shape = 1,size = 1.5,colour = "black")

#plot
plot(initialplot)
```



I next tried visualising the data as a stacked bar chart. This is because I wanted the number of referrals and categories to be more visible and because I wanted to highlight the age group composition and make the proportions more directly comparable. This visualisation method also works well with the data's discrete age group categories.

```
#x values for annotations on barplot
lockdown1 <- fval_tail(ref_age_date,1,"2020-03-01")
lockdown_text_placement <- fval_tail(ref_age_date,1,"2020-02-01")

#designing plot
barplot <- ggplot(ref_age_date, aes(x=REPORTING_PERIOD_START,
                                     y=METRIC_VALUE)) +
  geom_bar(aes(fill=SECONDARY_LEVEL), stat='identity',
```

```

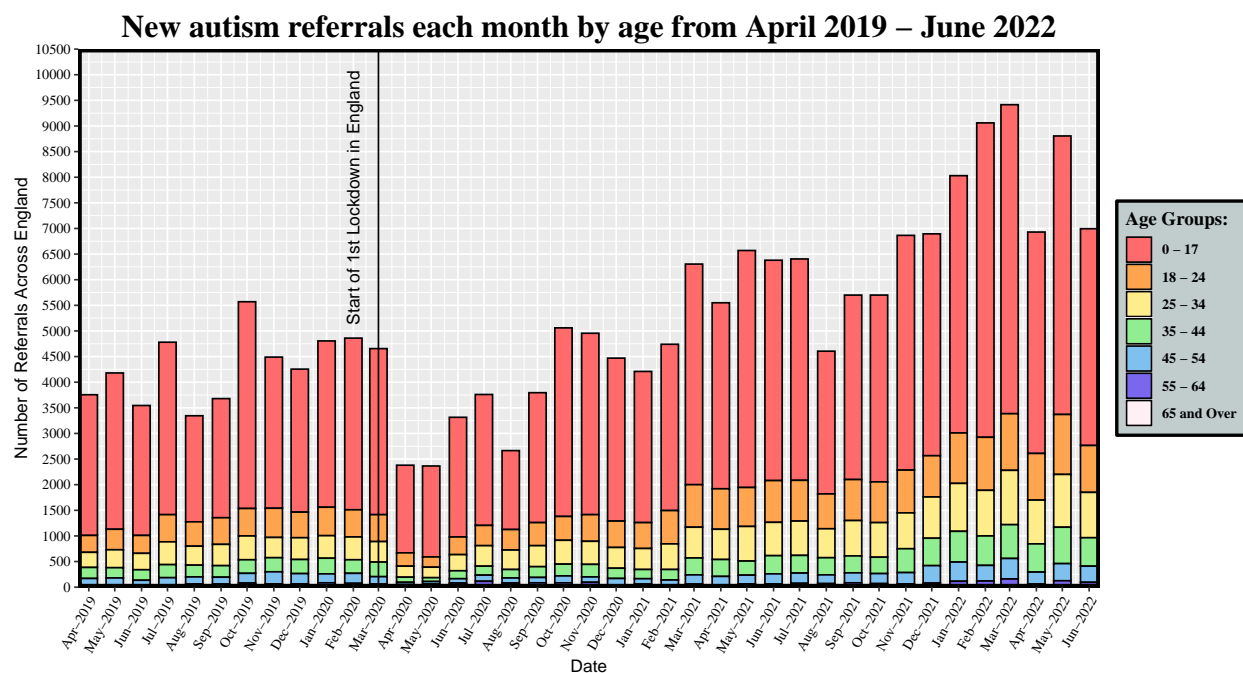
    colour='black', width=20)+
scale_y_continuous(breaks=seq(0,10500,500),expand = c(0,0),
                    limits = c(0, 10500))+
scale_x_date(date_breaks = '1 month', date_labels = "%b-%Y",
              expand = expansion(add = 2))+
theme(axis.text.x=element_text(angle=60, hjust=1),
      plot.title=element_text(family='Times', face="bold",
                               size=20,hjust = 0.5, vjust = 0.5),
      axis.text = element_text(family="Times", colour = "black"),
      legend.text = element_text(family="Times", colour = "black",
                                  face = "bold"),
      legend.background = element_rect(fill = "#C1CDCD",linetype = 1,
                                       colour = "black", linewidth = 1),
      legend.title = element_text(family="Times", colour = "black",
                                   face = "bold", size = 12),
      panel.border = element_rect(colour = "black", fill = NA,
                                   linewidth = 2),)+
labs(title="New autism referrals each month by age from April 2019 - June 2022",
     x="Date", y="Number of Referrals Across England", fill = "Age Groups:")+
scale_fill_manual(values=c("#FF6A6A", "#FFA54F", "#FFEC8B", "#90EE90",
                           "#7EC0EE", "#7A67EE", "#FFF0F5"))+
geom_vline(xintercept = ref_age_date[lockdown1,1])+
annotate("text", x= ref_age_date[lockdown_text_placement,1], y=7600,
         label="Start of 1st Lockdown in England", angle=90)

```

```

#viewing the bar chart
plot(barplot)

```



```

#saving the bar chart
ggsave(here("visualisations", "viz220225546.png"), height = 8, width = 13)

```

Finally, to emphasise the time course component of this data I animated the plot so that the data would be gradually revealed along the x-axis. This draws attention to trends over time in the data alongside differences in age-group proportions.

## Conclusions:

There is a clear decrease in autism referrals after the initial COVID-19 lockdown and a steady increase over the next two years resulting in a rough doubling of referral numbers between the last half of 2019 and the first half of 2022. The magnitude of the increase appears to be relatively similar for all age groups under 45; however, 45 - 54, 55 - 64 appear to be less affected although still seem to display some trend towards increase. 65+ year olds do not have enough referrals to be noticeable on the graph and this does not change over time indicating a relatively negligible contribution to autism referral numbers for NHS England. Therefore, lockdown and the post-lockdown period is associated with an approximate doubling of autism referral numbers for under 45s, a smaller increase in numbers for 45-64 year olds and no visible change in over 65s.

However, this data visualisation is limited in terms of its ability to answer the research question in that it only considers data from NHS CCGs (Clinical Commissioning Groups) and data submission is voluntary for non-NHS funded referrals. This means that data concerning private referrals is largely missing. People may be more likely to seek a private referral if they have struggled to obtain one from their NHS GP and as a result this may mean the referral numbers for adults is underestimated as autism is still traditionally seen as a condition only impacting or being diagnosed in children.

It would be interesting to see if this pattern varied geographically given that the data was recorded for each practice centre and so could be plotted onto a map. It would also be interesting to look at the breakdown of these results by sex as there is a disparity in the number of males vs females diagnosed with autism (generally ranging from 5:1-2:1 depending on the study, Napolitano et al 2022(1)) and it would be interesting to see if this is reflected by a similar trend for referrals or if this is a pattern that emerges at the assessment stage of diagnosis.

## References:

- (1) • Napolitano, A., Schiavi, S., La Rosa, P., Rossi-Espagnet, M. C., Petrillo, S., Bottino, F., Tagliente, E., Longo, D., Lupi, E., Casula, L., Valeri, G., Piemonte, F., Trezza, V., & Vicari, S. (2022). Sex Differences in Autism Spectrum Disorder: Diagnostic, Neurobiological, and Behavioral Features. *Frontiers in Psychiatry*, 13. <https://www.frontiersin.org/articles/10.3389/fpsy.2022.889636>