1	The impact of the COVID-19 pandemic on antimicrobial
2	prescribing at <mark>a</mark> specialist paediatric hospital- an
3	observational study
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15 Abstract

16 Introduction

17 The COVID-19 pandemic has severely impacted healthcare delivery and there are

18 growing concerns that the pandemic will accelerate antimicrobial resistance.

19 Methods

20 Data on patient characteristics and antimicrobial administrations for inpatients

21 treated between 29 April 2019 and Sunday 28 March 2021 were extracted from the

22 electronic health record at a specialist children's hospital in London, UK. Interrupted

23 time series analysis was used to evaluate antibiotic days of therapy (DOT) and the

24 proportion of prescribed antibiotics from the WHO 'Access' classification.

25 Results

A total of 23,292 inpatient admissions were included. Prior to the pandemic there
were an average 262 admissions per week compared to 212 during the pandemic
period. Patient demographics were similar in the two periods but there was a shift
in the specialities that patients had been admitted to. During the pandemic, there

30 was a crude increase in antibiotic DOTs, from 801 weekly DOT before the pandemic

31 to 846. The proportion of Access antibiotics decreased from 44% to 42%. However,

32 after controlling for changes in patient characteristics, there was no evidence for the

33 pandemic having an impact on antibiotic prescribing.

- 35 Conclusion
- 36 The patient population in a specialist children's hospital was affected by the COVID-
- 37 19 pandemic, but after adjusting for such changes there was no evidence that
- 38 antibiotic prescribing was significantly affected by the pandemic. This highlights
- 39 both the value of routine, high-quality EHR data and importance of appropriate
- 40 statistical methods that can adjust for underlying changes to populations when
- 41 evaluating impact of the pandemic on healthcare.

43 Background

44 There is growing concern that the COVID-19 pandemic will accelerate antimicrobial 45 resistance (AMR) - an existing global health threat. High rates of antibiotic use in 46 COVID-19 patients have been reported despite low rates of bacterial co-infections.¹ 47 But perhaps more relevant to children who are generally mildly affected by the 48 disease, are behavioural and structural changes in society and in healthcare settings 49 that might impact how antibiotics are being used. A survey of hospitals and 50 healthcare networks from June 2020 found that 65% of respondents thought that 51 the pandemic had had a negative impact on antimicrobial stewardship (AMS) 52 activities.² Factors such as increased pressure on healthcare workers, less 53 opportunity for isolation of infectious patients and increased rates of empirical 54 antimicrobial use for patients with respiratory symptoms could lead to increased 55 antibiotic use; however, increased focus on hand hygiene in hospitals could lead to 56 reductions in the spread of AMR and social distancing in society might lead to 57 reductions in patients presenting at hospital with respiratory illnesses.³ 58 Understanding the impact of the pandemic on antimicrobial use can inform AMS 59 policies and the response to future pandemics. 60 This study aimed to evaluate the impact of the COVID-19 pandemic on antibiotic 61 prescribing in a tertiary paediatric hospital in London, UK. The changes to the 62 patient population and to the use of all types of antimicrobials were described and

63 multivariable regression models were used to estimate the effect on antibiotic.

64 Methods

65 Setting

66 The UK implemented restrictions to limit the spread of COVID-19. A first nationwide

67 lockdown was implemented on 23 March 2020 and schools had moved online on 20

68 March 2020. This was followed by a month-long second national lockdown in

69 November 2020 and a third lockdown in January 2021.⁴

70 Great Ormond Street Hospital (GOSH) is a paediatric tertiary care hospital in 71 London with an established AMS team.⁵ The AMS team comprises an antimicrobial 72 pharmacist, an infectious disease consultant and a microbiology consultant and 73 their work include a weekly hand over and ward rounds on four days of the week. 74 AMS activities continued at the same level compared to pre-pandemic, however, the 75 face-to-face stewardship rounds transitioned to a virtual format using the 76 comprehensive electronic patient record. As part of a systems response to the 77 pandemic, most complex paediatric inpatients in North Central London CCG were 78 cared for at GOSH from April 2020, instead of their local hospital. Working patterns 79 were also affected with more staff working remotely, being off sick or being 80 deployed to other hospitals.

81 Data

This study used routinely collected deidentified hospital data from inpatients at GOSH between 29 April 2019 and 28 March 2021 and who spent at least one night in hospital (ethical approval (17/LO/0008)). Admissions data was linked to data on treatment speciality, surgical encounters and medication prescribing data. Patients 86 older than 25 years of age when admitted were excluded from the study (<1% of

87 admissions) but no other exclusion criteria were applied.

88 Descriptive statistics of patient characteristics were derived from information

89 recorded at admission (see Table S1 for definitions).

Administration of any antimicrobial on a calendar day, regardless of the number of
administrations, represented one day of therapy (DOT). The number of patient days,
including the day of discharge, was used as the denominator to calculate DOTs per
1000 patient days. Antibiotics administered were then grouped into Access, Watch
and Reserve groups as developed by the WHO⁶ and the proportion of Access
antibiotics was calculated. All analysis was carried out using R version 4.0.3.⁷

96 Interrupted time series model

97 Interrupted time series models were used to compare counts of weekly antibiotic 98 DOTs and the percent of Access antibiotics before the pandemic with the first year 99 of the pandemic. The hypothesis was that the pandemic would cause an immediate 100 and constant shift in antimicrobial consumption, commonly referred to as a level 101 change with no lag. A negative binomial model with the number of patients days 102 (logged) included as an offset was used for antibiotic DOTs and a binomial model 103 was used to model the percent of Access antibiotics. See Table S5 for full list of 104 variables tested for inclusion in the model. Model residuals were checked for signs 105 of autocorrelation and tested for using the Breusch-Godfrey test and the final model 106 was selected using the Akaike Information Criterion.

107 **Results**

108 There were 23,292 inpatient admissions (14,449 individual patients). There were

109 46 weeks included in the pre-COVID-19 period and 54 weeks in the COVID-19

110 period with each week contributing a minimum of 1,450 patient days. During the

111 pre-COVID-19 period 44% of antibiotic DOTs were from the Access group compared

112 to 42% during the pandemic.

113 There was no meaningful difference in the median age between patients admitted

114 before and during the pandemic but those admitted in the COVID period more likely

to get at least one antibiotic, antiviral or antifungal during their stay during the

116 pandemic (Table 1). A positive COVID-19 test was found for 134 admissions. Median

117 weekly patient days by speciality before and during the pandemic and variation

118 over time can be found in Table S3 and Figure S1.

There was an increase in crude antibiotic and antiviral DOTs between the period
before and during the pandemic (Table S4). Antibiotic DOTs by AWaRe group can be
found in Table S5.

122 There was considerable variation in antibiotic DOTs per 1000 patient days between

123 specialities but no speciality experienced a substantial change during the pandemic

124 period (Figure 1a). There was substantial variation in the proportion of Access

125 DOTs between specialities and the two specialities with highest antibiotic DOTs saw

126 a decrease in the percent of Access DOTs (Figure 1b).

No statistically significant difference in antibiotic consumption could be detected for
either antibiotic DOTs (incidence rate ratio 1.01 (95% confidence interval: 0.95,

129 1.08) or percent of Access antibiotics (odds ratio 0.83, (95% confidence interval:

130 0.04, 16.1))(Table S6). There was no evidence of autocorrelation - residual plots and

131 autocorrelation tests can be found in supplementary materials (Figures S1 and S2).

132 Table S2 shows the variables included in the final model for both outcomes.

133 Discussion

134 We found an increase in crude antibiotic DOTs per 1000 patient days but after

135 adjusting for changes to the patient population using statistical modelling, there was

136 no evidence of significant changes to antibiotic use during the first year of the

137 pandemic.

138 The variation in changes in patient bed days between specialities explains most of

the crude increase in antibiotic DOTs as there is substantial variation in antibiotic

140 DOTs between specialities (Figure 1a). The large increases in intensive care patients

141 during the COVID-19 pandemic are likely the result of transfers from other

142 hospitals, as is the increase in number of cancer patients. These are patient groups

143 with intrinsically greater use of antimicrobials. For specialities such as paediatric

144 respiratory medicine, the reduction in bed days is likely a consequence of a decrease

145 in demand due to behavioural changes during lockdown. The increase in the

146 proportion of emergency admissions and the decrease in surgeries will have also

147 accounted for some of the crude differences.

Antimicrobial use in patients with COVID-19 have been widely reported but far less
is known about the wider impact on antibiotic prescribing for all patients.⁸ Multiple
studies have examined the pandemic impact on antibiotic prescribing in primary

151 care in England and report a decrease in GP prescribing but an increase in dental 152 prescribing.⁹⁻¹¹ There is more limited data regarding secondary care but there was a 153 4.8% increase in total prescribing rate between 2019 and 2020 in England but the 154 patient population was vastly different to previous years¹². This study provides new 155 insights on the pandemic impact on inpatient antimicrobial use in children whilst 156 considering the complex changes to patient population. It also highlights the value 157 of a dedicated AMS team.

All data in this study was routinely collected and digitally extracted from a database.This study demonstrates how hospital EHR data can be used to evaluate important

160 system changes and antimicrobial use monitoring. The value of EHR data featured

161 heavily in the UK's five year National Action Plan to tackle antimicrobial

162 resistance.¹³ Despite this, a recent systematic review found that few antimicrobial

163 use studies used solely digitally extracted data.¹⁴

164 A study strength is the use of a large routinely collected, comprehensive patient-165 level dataset. We used two different antibiotic use metrics which captured both 166 volume changes as well as the antibiotic type. The rich data and metric choice 167 provide a more accurate overview of the changes to antimicrobial use and their 168 appropriateness. Interrupted time series models were used which allowed 169 controlling for the substantial 1. Hughes, S., Troise, O., Donaldson, H., Mughal, N. & 170 Moore, L. S. P. Bacterial and fungal coinfection among hospitalized patients with 171 COVID-19: a retrospective cohort study in a UK secondary-care setting. Clin. 172 Microbiol. Infect. 26, 1395-1399 (2020).

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Characteristic	Pre- COVID- 19, N = 26,531	COVID- 19, N = 22,545	p-value ^a
Age (years)	5.3 (1.8, 10.7)	5.2 (1.4, 11.2)	0.14
Male	6,449 (55%)	6,375 (56%)	0.3
Any theatre encounter	5,577 (48%)	5,023 (44%)	<0.001
Admission type			<0.001
Elective	9,802 (84%)	8,404 (74%)	
Emergency	690 (5.9%)	1,185 (10%)	
Other	27 (0.2%)	46 (0.4%)	
Transfer	1,203 (10%)	1,798 (16%)	
Antibiotics during stay	5,925 (51%)	6,324 (55%)	<0.001
Antifungals during stay	945 (8.1%)	1,153 (10%)	<0.001
Antivirals during stay	475 (4.1%)	605 (5.3%)	<0.001
Antiprotozoal during stay	99 (0.8%)	117 (1.0%)	0.2
Immunosuppressants during stay	413 (3.5%)	445 (3.9%)	0.14
¹ Median (IQR); n (%) ^a Wilcoxon rank sum test	t; Pearson's (Chi-squared t	est

215 Table 1: Patient level characteristics before and during the COVID period

Figure 1

 $^{\rm a}{\rm Antibiotic}$ DOT per 1000 bed days by speciality before and during the pandemic





