

BMJ Open The effect of primary care availability on antibiotic consumption in Hungary: a population based panel study using unfilled general practices

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ABSTRACT

Objective We analyse the effect of primary care availability on antibiotic consumption and on the quality of antibiotic prescribing behaviour.

Design Retrospective panel design, secondary analysis of settlement-level administrative panel data (n=2320 settlements, T=72 months).

Participants and setting We analyse antibiotic consumption of the population of villages in Hungary, over years 2010 to 2015. We exploit the geographical and time variation in unfilled (mainly single-handed) general practices as a source of exogenous variation in the availability of primary care. We control for socioeconomic characteristics and settlement fixed effects in a panel regression framework.

Outcome measures Antibiotic expenditures and days of therapy (DOT); consumption of narrow-spectrum and broad-spectrum antibiotics; consumption of Access, Watch and Reserve antibiotics according to the AWaRe categorisation; number of visits to the general practitioner (GP).

Results If the general practice of a village becomes unfilled, the number of GP visits decreases on average by 9.3% (95% CI 5.6% to 12.9%), antibiotics DOT decrease on average by 3.2% (95% CI 1.0% to 5.4%) and expenditures on antibiotics decrease on average by 2.5% (95% CI 0.3% to 4.7%). The negative effect on antibiotic consumption is stronger in settlements where secondary care is less available, and where antibiotics were previously overprescribed. The quality of prescribing behaviour measured by the relative changes in the narrow-spectrum vs broad-spectrum as well as the Access versus Watch and Reserve antibiotics deteriorates significantly as a consequence of worse primary care availability.

Conclusions Limited availability of primary care reduces the consumption of antibiotics and at the same time impairs the quality of prescriptions through a decrease of the number of doctor-patient encounters.

INTRODUCTION

'Antibiotic resistance is one of the most significant threats to patients' safety in Europe. It is driven by overusing antibiotics and prescribing them inappropriately.'¹ Global consumption of antibiotic drugs increased

Strengths and limitations of this study

- To estimate the effects of access to primary care on the consumption of antibiotics, we use geographical and time variation in unfilled general practices in Hungary.
- Our results are based on administrative records of antibiotic use of the entire population of Hungary living in villages between years 2010 to 2015.
- We analyse the quantity and quality of antibiotic consumption, by separating the narrow-spectrum versus broad-spectrum as well as the Access versus Watch-Reserve antibiotic prescriptions according to the AWaRe classification.
- The external validity of the findings is restricted to healthcare systems with mostly single-handed general practices, which serve as the first point of contact to the system.

by 35% between 2000 and 2010.² Antibiotics can be life-saving, however, overuse of them can increase the susceptibility to infections and disease.³ The WHO identifies antimicrobial resistance as one of the top 10 threats to global health in 2019.⁴

We analyse the effect of primary care availability on antibiotic consumption and on the quality of antibiotic prescribing behaviour. In European countries, with the exception of Turkey, the average age of general practitioners (GPs) is between 45 and 55 years⁵ so lack of supply in primary care is either already a problem, or will be a problem in the near future, in many places. For instance, in Britain, around 10% of GP positions were unfilled in 2015, and the situation is worsening daily.⁶

Little is known about the relationship between primary care availability and antibiotic consumption. Limited access to primary care is likely to lead to fewer consultations, hence lower access to prescriptions. However, it can also have opposite effects since if GPs spend more time with their patients then the

prescription of antibiotics can be reduced.⁷ Results from Switzerland and also from 17 European countries show that the density of physicians is positively related to outpatient antibiotic consumption.^{8,9} Based on regional data from Hungary, a previous study did not find evidence for a relationship between antibiotic use and the density of pharmacies or the number of enrolled patients per GP.¹⁰ In this paper, we use more detailed longitudinal administrative data from a later period from Hungary and exploit the process of general practices becoming unfilled as a natural experiment to analyse the question and reach different conclusions.

Institutional background in Hungary

The following summary of primary care in Hungary is based on Gaál *et al*, 2011, Wilm *et al*, 2015.^{11,12} In Hungary, the municipalities are responsible for the provision of primary care. The majority of primary care services are provided by general practitioners (family physicians or family paediatricians). Local governments designate primary care districts to the GPs, who are not allowed to refuse patients from their district. On the other hand, patients have the right to choose their GP. The overwhelming majority of GPs work in single-handed practices, group practices are very rare.¹² To work in a primary care district with a territorial supply obligation, family doctors must purchase a licence. Thus practices can be bought and sold, while local governments still keep a control over who serves as a GP. GP care is provided for free, and at least in principle, GPs act as gatekeepers. However, several specialist care services can be accessed without the referral of a GP (eg, dermatology, ear, nose and throat diseases, gynaecology, traumatology). GPs' financing is mostly based on capitation, with some supplements depending for example on the age composition of the patients and the type of settlements covered. In 2015, there were 6277 practising GPs in Hungary, the average number of patients per GP was 1566. Meanwhile, there are 3155 settlements in Hungary, and the total population is 9.8 million.¹³

A severe problem of the Hungarian primary care system is that around 5% of the – overwhelmingly single-handed – GP practices are unfilled. The high ratio of unfilled GP positions in Hungary is despite the fact that the government offers financial subsidies if a GP takes up a practice that has been unfilled for at least 6 months. The unfilled practices are mostly located in remote, less developed areas of the country. If a practice becomes unfilled it will be partly covered by the GPs of neighbouring primary care districts. Hence, on the one hand, deputy GPs in the remote areas have to travel a lot, so as to provide some care at the unfilled practices. On the other hand, patients in unfilled districts cannot access a GP locally every day of the week. If they need to see a GP when the deputy GP is not around, they have to travel to another district (typically to another settlement).

Consumption of antibiotics in Hungary

Hungary had the ninth lowest per capita consumption of antibiotics for systemic use out of 30 European Union

(EU)/European Economic Area (EEA) countries in 2016 (the defined daily dose (DDD) per 1000 inhabitant-days was 15.4 in Hungary, 21.9 in the EU/EEA).¹⁴ However, quality indicators of antibiotic prescribing behaviour do not show such a bright picture. For instance Juhasz *et al*,¹⁵ report some inappropriateness of antibiotic treatment in acute cystitis and Marton *et al*¹⁶ already showed alarming evidence of high incidence of antibiotic resistance in Hungary almost three decades ago. Regarding the quality indicators in international comparison, the ratio of broad-spectrum to narrow-spectrum antibiotics in community consumption is the second largest in the EU/EEA¹⁷ (Broad-spectrum antibiotics are known to have a high potential to cause antimicrobial resistance.^{18,19}). Out of 12 quality indicators such as the seasonal variation of antibiotic prescriptions or the use of certain broad-spectrum antibiotics, Hungary is in the worst quartile of EU/EEA countries in five cases, in the second worst quartile in four cases and in the better half in only three cases.²⁰

Furthermore, the recent WHO Essential Medicines List²¹ classifies antibiotics into Access, Watch, Reserve and Other categories to ensure appropriate prescribing behaviour (AWaRe classification). In Hungary, compared with the median value in the European region, a smaller share of antibiotic consumption belongs to the Access group (48% vs 56%) and a larger share to the Watch group (38% vs 29%) as well as to the Reserve group (0.5% vs 0.2%), putting the country into the worst quartile in terms of these quality measures as well.¹⁸ Responding to these results, the prescription rate of antibiotics has become part of the primary care performance indicator system in Hungary since 2011.²²

METHODS

Data

We combine three data sources to create settlement-level monthly administrative panel data (n=2320 settlements, T=72 months over years 2010 to 2015).

General practices

We obtained the list of filled and unfilled general practices from the National Health Insurance Fund Administration (NHIFA) for the time between 2004 to 2017. (These data were provided to us by the NHIFA free of charge in the framework of secondary use of public data.) We know the zip code of the site of the practice, the list of settlements the practice caters for and the start and end date (month) of the practice having been unfilled. In those settlements where there is only one general practice, we exactly know whether it is unfilled or not in a given month. However, our information is less precise in settlements (mainly towns and cities) where more than one general practice is available. We know if some of them are unfilled, but cannot identify the exact part of the settlement that is not covered by GP care. To alleviate the consequences of this data limitation, we focus our analysis on villages, of which 77% are covered by only one general practice.

Meanwhile, the lack of GP care availability is typically an issue in villages: although only 30% of the population live there, 81% of unfilled general practices are located there.

Our main explanatory variable on the settlement level is the ratio of the unfilled practices relative to the total number of practices (filled plus unfilled) covering the given settlement. If a village is covered by one unfilled practice then the ratio of unfilled practices equals one there. (Most of the villages (72.5%) are covered by practices that cater for adults and children as well; 18.3% of the practices covering villages cater for adults only and 9.2% for children only).

Settlement-level demographic, socioeconomic and GP care data

We also use the T-STAR municipal statistical system of the Central Statistical Office of Hungary. This contains wide-range annual settlement-level demographic data (such as the gender and age composition of the population) and socioeconomic data (such as the unemployment rate).

Furthermore, the database contains, by settlement, the annual number of visits to all GPs based in the given settlement. Thus if a general practice caters for several settlements then all the visits are attached to the basis settlement. Due to this limitation, we take the following steps to generate the database of the number of GP visits. First, using the register of general practices, for each settlement we identify the settlement that provides the GP care (this is important for small villages, which do not host a health centre). Second, we calculate the population of the catchment area of the general practices of a given settlement, using the population statistics from T-STAR. Finally, we divide the number of GP visits with the estimated population to construct the per capita indicator on the catchment area level.

Consumption of antibiotics

Data on the use of antibiotics, covering the entire population of Hungary, come from the National Healthcare Services Centre (NHSC) through an agreement between the NHSC and the Institute of Economics, Centre for Economic and Regional Studies of the Hungarian Academy of Sciences. They contain information on pharmaceuticals in the Anatomical Therapeutic Chemical (ATC) group 'J01' (antibiotics with systemic use) that were purchased through pharmacies, hence only relate to the ambulatory setting and exclude hospital care. Only prescription drugs are recorded (over-the-counter sales are excluded) but this is not a limitation because of the prescription-only status of antibiotics in Hungary (Some non-prescription sales do occur but they constitute only around 2% of all sales.²³). Our data cover prescriptions both in primary care and specialist outpatient care, hence we measure the overall effect on the use of antibiotics.

The NHSC data set covers years 2010 to 2015. It provides information on the name, identifier code and detailed ATC code of the antibiotic medication, the date of the purchase, the amount purchased and the associated expenditure (the sum of social security and out-of-pocket

payments). It also includes the zip code of the patient's address, the patient's year of birth and gender and the International Statistical Classification of Diseases code of the diagnosis for which the prescription was given. Using the zip codes, we construct monthly indicators of antibiotic use (days of therapy (DOT) and expenditure of all antibiotics as well as of various subcategories) on the settlement level.

We define various subcategories so as to analyse the quality of antibiotic prescription. First, following Cormican *et al.*,¹⁷ we distinguish the consumption of narrow-spectrum penicillins, cephalosporins, macrolides (ATC codes J01(CA+CE+CF+DB+FA01)) and of broad-spectrum penicillins, cephalosporins, macrolides and fluoroquinolones (J01(CR+DC+DD+(F-FA01)+MA)). Then, we define a quality index as the ratio of the consumption of the narrow-spectrum and the broad-spectrum categories. (This is the reciprocal of the index used by Juhasz *et al.*¹⁵ Since there are a lot of zero values in the consumption of narrow-spectrum antibiotics on the settlement-month level, the original index is not defined in many cases, hence we use its reciprocal). To lower the risk of antibiotic resistance, a higher value of this quality indicator is desirable.

Second, we use the AWaRe (Access – Watch – Reserve) classification of the list of essential medicines²¹ to assess the quality of consumption. The Access category includes first and second choice antibiotics that should be widely available in healthcare settings, while antibiotics in the Watch and Reserve categories have a higher potential for resistance to develop and their use as first and second choice treatment should be limited.¹⁸ Since the Reserve category comprises only a small fraction (0.5%) of antibiotic consumption in Hungary, we analyse the consumption of Watch and Reserve categories jointly.

Finally, to generate descriptive statistics and to relate the consumption of antibiotics to total pharmaceutical consumption, we also use publicly available county-level statistics on pharmaceutical sales from the NHIFA. (Since they come from the same administrative sources, the NHSC and NHIFA antibiotic consumption data are very similar both in their levels and in their county-level variation (the correlation is above 0.95)).²⁴ Similarly to the NHSC data above, only prescribed pharmaceutical purchases in public pharmacies are included in these statistics.

Statistical analysis

Regressions for the quantity and quality of antibiotic consumption

In our baseline analysis, we estimate the effect of unfilled general practices on antibiotic consumption using the settlement-month panel of antibiotic DOT and expenditures, which was created from the NHSC data. We restrict the analysis to villages and estimate the following panel model:

$$antibiotics_{s,t} = \gamma_0 + \gamma_u unfilled_{s,t} + \gamma_d D_t^{date} + \gamma_h h_{s,t} + \nu_s + \varepsilon_{s,t} \quad (1)$$

Table 1 Average antibiotic days of therapy per 1000 inhabitant-days by gender and age

Age group	Males			Females		
	Mean	SE	Median	Mean	SE	Median
0–5	18.43	0.03	16.64	17.10	0.03	15.18
6–17	17.25	0.03	14.75	17.79	0.03	15.50
18–39	10.41	0.01	9.78	15.39	0.02	14.85
40–59	9.12	0.01	8.75	13.83	0.02	13.47
60–	13.91	0.02	13.35	13.54	0.02	13.19
Total	11.97	0.01	10.68	14.77	0.01	14.03

Statistics weighted by the size of each settlement-gender-age group. SE is the standard error of the mean. NHSC data, years 2010 to 2015.

where s indicates settlements and t the time measured in months. In various model specifications the outcome variables ($antibiotics_{s,t}$) are the level and logarithm of antibiotic DOT per 1000 inhabitant-days as well as the logarithm of social security expenditures per 1000 inhabitants. (The monthly settlement-level DOT or expenditures are zero only for a tiny fraction of observations, hence the logarithms are missing only in very few cases). Also, as quality indicators, we examine the DOT of the subcategories (narrow-spectrum, broad-spectrum, Access and Watch-Reserve) as well as the quality index (ratio of narrow-spectrum and broad-spectrum antibiotics) defined above.

Our main explanatory variable is the ratio of unfilled general practices among all practices covering settlement s in month t ($unfilled_{s,t}$). Further explanatory variables include time fixed effects (D_t^{date}) to control for time trends and various settlement-level indicators ($h_{s,t}$) from the T-STAR database to control for health needs and healthcare supply. In particular, we proxy health needs, that is, the demand-side determinants of antibiotic consumption, with the age-gender distribution and the socioeconomic characteristics of the settlement (ratio of females, ratio of residents aged 0 to 17 years and aged 60+ years, local unemployment rate, prescription exemption rate). According to table 1, the use of antibiotics differs by gender and age, hence the age-gender distribution can partly capture the settlement-specific demand for antibiotics. Socioeconomic characteristics are expected to affect the demand via income and health effects. Finally, we measure the supply of healthcare beyond primary care with the local (micro-regional) specialist outpatient capacities (annual specialist outpatient hours per 1000 inhabitants).

We also include settlement-level fixed effects in equation (1), (v_s), which capture all time-invariant characteristics of the settlements (such as region, distance to the nearest town or location of the GP practice). Hence

this estimation method yields unbiased estimates for the parameters even if we allow that settlements with unfilled and filled practices may systematically differ from each other across time-invariant dimensions that we cannot measure (Wooldridge, 2010; section Fixed Effects Methods).²⁵ In other words, the inclusion of settlement fixed effects allows us to identify the impact of primary care availability on antibiotic consumption from the within-settlement time variation of filled/unfilled status, not just from the cross-sectional geographical variation.

The dependent variables in equation (1) are calculated on per capita basis and hence the models are inherently heteroscedastic, with larger variances observed in smaller settlements (as also shown in online supplementary figure A1 in appendix A1). Hence we estimate the equation with weighted fixed effects least squares, using the (time-averaged) population of the settlements as weights. According to online supplementary figure A2 in appendix A1, the weighted residuals are much closer to homoscedasticity. We cluster SEs on the micro-regional level to take into account potential interdependencies of shocks in the local healthcare systems (in line with Cameron *et al*²⁶; the clustering formula is provided by Cameron and Trivedi, 2009²⁷ (section Cluster-robust standard errors)).

For the sake of comparison, we re-estimate equation (1) with a random effects specification for v_s as well. This specification assumes that the settlement-level time-invariant unobserved factors are uncorrelated with the explanatory variables (Wooldridge, 2010;²⁵ section Random Effects Methods). We choose between the fixed effects and the random effects specifications with a Hausman-type test (Wooldridge, 2010;²⁵ section The Hausman Test Comparing the RE and FE Estimators), implemented with the *xtoverid* function²⁸ of the Stata statistical software.²⁹ All other calculations are performed in Stata as well.

Heterogeneity analyses

We also investigate effect heterogeneities. First, we extend equation (1) and interact the ratio of unfilled general practices ($unfilled_{s,t}$) with the micro-regional specialist outpatient hours per 1000 inhabitants (population-weighted mean=1.22, SD=0.80), with the binary indicator of the presence of a pharmacy in the settlement (weighted proportion=0.775) and, to capture the socioeconomic conditions, with settlement-level unemployment rate (weighted mean=0.105, SD=0.071) and with the prescription exemption rate (weighted mean=0.039, SD=0.025).

Second, we analyse effect heterogeneities according to the initial consumption level of antibiotics. Our aim is to understand if the negative estimated effect of unfilled practices originates from settlements where the use of antibiotics was relatively high, conditional on health needs and healthcare supply. To do so, first we estimate a cross-sectional regression of the annual antibiotic use in year 2010, and generate the residuals:

$$antibiotics_{s,2010} = \gamma_0^{2010} + \gamma_u^{2010} unfilled_{s,2010} + \gamma_h^{2010} h_{s,2010} + \varepsilon_{s,2010} \quad (2)$$

The outcome variable in equation (2) is the logarithm of the annual antibiotic DOT in 2010, and settlement-level controls (h_s) are the same as in equation (1). The generated residuals have a SD of 0.35 (with zero mean). Then, we re-estimate equation (1), with adding the interaction term between the ratio of unfilled general practices and the year 2010 residual of equation (2). Here we use data for years 2013 to 2015, so that we can interpret the interaction term as the heterogenous effect according to the baseline use of antibiotics.

Regressions for GP care use

Finally, to investigate the mechanisms of how primary care availability affects the use of antibiotics, we estimate the impact of the ratio of unfilled general practices on the annual number of GP visits, and on antibiotic prescriptions per GP visit. We use the settlement-level (more precisely, GP catchment area level) GP care use statistics as described in section Settlement-level demographic, socio-economic and GP care data and restrict the analysis to villages. We estimate the following fixed effects model:

$$\ln y_{s,t} = \beta_0 + \beta_u unfilled_{s,t} + \beta_d D_t^{date} + \beta_h h_{s,t} + \nu_s^{GP} + \varepsilon_{s,t}^{GP} \quad (3)$$

The notations and the explanatory variables are essentially the same as in equation (1). Since the data are on the annual level, the time dummies are year dummies now. The dependent variables ($\ln y_{s,t}$) are the logarithm of the per capita annual number of GP visits and the logarithm of the ratio of antibiotic prescriptions to GP visits.

Patient and public involvement

No patients were involved in the research, as it was based on the retrospective analysis of administrative data.

RESULTS

Descriptive analysis

Unfilled general practices

The number and the share of unfilled general practices increased from 150 (2.3%) in 2008 to 350 (5.5%) in 2016. Based on the (censored) data covering years 2004 to 2017, the observed mean duration of a practice being unfilled is 4.9 years, the median is 3.3 years.

Figure 1 shows the spatial distribution of settlements with unfilled general practices as of December 2015. While unfilled practices are scattered all over Hungary, their distribution is more dense in the Eastern part of the country, especially in the poorer North-Eastern part. Between years 2010 to 2015 (in the period on which our main antibiotic analysis is based), about two-third of unfilled general practices were in villages, and the share of unfilled practices increased from 5.8% to the substantial 10.9% in this settlement category. As the analysis presented in online supplementary appendix A2 shows, unfilled general practices are on average more common

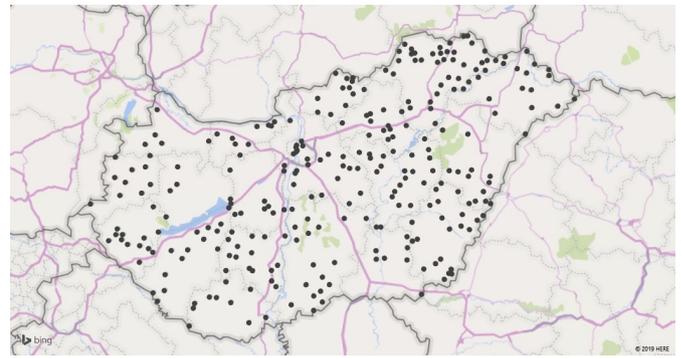


Figure 1 Spatial distribution of settlements with unfilled general practices, December 2015 (data source: National Health Insurance Fund Administration).

in the less affluent villages, for example, where the unemployment rate is higher.

Within villages over years 2010 to 2015, the annual per capita number of GP visits was 8.0.

Consumption of antibiotics

Based on the NHIFA data, throughout years 2010 to 2015, the average per capita annual expenditures on antibiotics (social security plus out-of-pocket payments) were 4.5 EUR (1350 HUF), which accounted for 3.0% of the total per capita expenditures on prescribed pharmaceuticals (148 EUR (44 300 HUF)). Thirty per cent of antibiotic expenditures were paid by the social security. The DOT of antibiotics was 14.2 per 1000 inhabitant-days throughout years 2010 to 2015, without major time trend (15.1 in 2010, 14.8 in 2015).

Antibiotic consumption is strongly seasonal: compared with the summer months, expenditures more than double in the winter (online supplementary figure A3 in appendix A3). According to table 1, antibiotic DOT per 1000 inhabitant-days is higher for women than for men (by 23%), and higher for children than for adults, but the age differences are more pronounced for men than for women. Gender differences are partly due to the common use of antibiotics for the treatment of urinary tract infections among women. There is a non-negligible variation in per capita antibiotic consumption across counties in Hungary. In line with Matuz *et al.*,¹⁰ the use of antibiotics tends to be higher in the poorer areas (Eastern, North-Eastern and Southern Hungary – online supplementary figure A4 in appendix A3).

Effects on the quantity of antibiotic consumption

The estimation results of equation (1) are reported in table 2. The fixed effects logarithmic models show that when the single general practice covering a village becomes unfilled, then antibiotic DOT decreases by 0.033 on the logarithmic scale, that is, by 3.2% (95% CI 1.0% to 5.4%) and antibiotic expenditures by 0.026 on the logarithmic scale, that is, by 2.5% (95% CI 0.3% to 4.7%). (If γ is an effect parameter in the logarithmic model then $100 \cdot (\exp(\gamma) - 1)$ is the percentage effect on the original scale, which is roughly $100 \cdot \gamma$ if γ is close to zero). In

Table 2 Effect of the ratio of unfilled general practices on antibiotic use

	(1): DOT per 1000 inhabitant-days		(2): Log of per capita DOT		(3): Log of per capita expenditures	
	FE	RE	FE	RE	FE	RE
Ratio of unfilled practices	-0.561*** (0.185)	-0.537*** (0.182)	-0.0327*** (0.0115)	-0.0314*** (0.0115)	-0.0257** (0.0114)	-0.0248** (0.0113)
Ratio of females	22.46*** (2.802)	22.03*** (2.706)	1.445*** (0.190)	1.429*** (0.179)	1.569*** (0.205)	1.546*** (0.189)
Age 0-17 ratio	-7.166 (4.644)	-0.391 (3.677)	-0.235 (0.265)	0.0650 (0.214)	-0.387 (0.281)	-0.0507 (0.220)
Age 60 plus ratio	-0.575 (3.484)	-5.976* (3.126)	-0.0751 (0.247)	-0.321 (0.208)	-0.399 (0.259)	-0.647*** (0.216)
Unemployment rate	-3.402*** (1.048)	-1.633* (0.964)	-0.135** (0.0637)	-0.0389 (0.0609)	-0.164** (0.0700)	-0.0675 (0.0656)
Prescription exemption rate	1.988 (1.976)	4.157** (1.942)	0.233** (0.118)	0.355*** (0.116)	0.273** (0.118)	0.405*** (0.119)
Specialist outpatient capacities	-0.00453 (0.0851)	-0.00140 (0.0852)	-0.000401 (0.00458)	-0.000241 (0.00453)	0.00424 (0.00426)	0.00394 (0.00433)
Monthly date effects	yes	yes	yes	yes	yes	yes
Settlement effects	yes	yes	yes	yes	yes	yes

Cluster-robust SEs in brackets with clustering on the micro-regional level, *p<0.1, **p<0.05, ***p<0.01. Settlement-month level NHSC data, weighted by the population of settlements, years 2010 to 2015. Number of observations: 162 720 in model (1), 162 311 in models (2) and (3) (number of settlements: 2316). DOT, days of therapy; FE, fixed effects; NHSC, National Healthcare Services Centre; RE, random effects.

absolute terms, the reducing effect is 0.6 DOT per 1000 inhabitant-days if the ratio of unfilled practices increases from zero to one.

Table 2 also shows that the random effects models yield almost identical estimates for the impact of the unfilled practices, so the above results are robust to this specification change. Meanwhile, the parameters of some control variables differ substantially across the fixed/random effects specifications. As a consequence, the Hausman-type overidentification test rejects the random effects model in favour of the fixed effects specification even at the 0.1% significance level for all three dependent variables. In the following we will display the fixed effects results.

Further specification checks were also performed. The estimation results change only slightly if the main explanatory variable – the continuous indicator of the ratio of unfilled practices – is replaced with a categorical variable that takes three distinct values: it equals 0 if the ratio of unfilled practices is 0, equals 1 if the ratio of unfilled practices is between 0 and 1 and equals 2 if the ratio of unfilled practices is 1. In other words, here we distinguish the settlements where no practices are unfilled, where all practices are unfilled and where some practices are filled and some are unfilled (middle category). Additionally, to check for non-linearities, we include the squared terms of all control variables in the model.

According to the parameter estimates in this specification, if all GP practices (typically the single available GP practice) become unfilled, that implies a 3.7% lower per capita antibiotic DOT (95% CI: 1.5% to 5.8%) compared with the ‘none unfilled’ category, which is very close to the estimate of the linear specification presented in table 2 (ie, 3.2%). Meanwhile, if there are more than one GP practices and some but not all of them become unfilled, that has no significant effect on antibiotic consumption (effect size is -0.5%, 95% CI -2.5% to +1.4%).

Finally, specification checks (not shown here in detail) reveal that above the contemporaneous effect, past and future (ie, previous and next year’s) unfilled status of the general practice have no significant effect on antibiotic use. Thus, we find no evidence for adjustment in antibiotic use after the immediate negative effect of a practice becoming unfilled. We also do not see evidence that people would adjust their antibiotic use in anticipation of a practice becoming unfilled.

Heterogeneity analyses

The heterogeneity analyses are reported in table 3. According to the first column, a poorer supply of health-care services (as captured by the availability of specialist care and pharmacies) strengthens the negative effect of unfilled general practices on antibiotic consumption. Lower supply of specialist care implies less opportunity for substitution of primary care, leading to the stronger effect of the availability of primary care on antibiotic use. Meanwhile, there is no effect heterogeneity according to

Table 3 Heterogeneous effect of the ratio of unfilled general practices

	Dependent variable: log of per capita DOT	
	(1): Heterogeneity by settlement characteristics	(2): Heterogeneity by baseline antibiotic use
Ratio of unfilled practices (RUGP)	-0.0746*** (0.0231)	-0.0304** (0.0151)
RUGP × pharmacy present in settlement	0.0611*** (0.0185)	
RUGP × specialist outpatient hours in micro-region/1000 inhabitants	0.0147* (0.00815)	
RUGP × unemployment rate in settlement	-0.0688 (0.112)	
RUGP × prescription exemption rate in settlement	-0.187 (0.239)	
RUGP × residual of antibiotic use in 2010		-0.112* (0.0588)
Ratio of females	1.454*** (0.189)	1.173** (0.460)
Age 0–17 ratio	-0.229 (0.266)	-0.383 (0.505)
Age 60 plus ratio	-0.0844 (0.246)	-0.0123 (0.455)
Unemployment rate	-0.130** (0.0645)	-0.143 (0.0896)
Prescription exemption rate	0.247** (0.122)	0.239 (0.219)
Specialist outpatient capacities	-0.00214 (0.00509)	-0.0102 (0.0290)
Monthly date effects	yes	yes
Settlement fixed effects	yes	yes

Cluster-robust SEs in brackets with clustering on the micro-regional level, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Settlement-month level NHSC data, weighted by the population of settlements, years: 2010 to 2015 in model (1), 2013 to 2015 in model (2). Number of observations: 162 311 in model (1), 79.432 in model (2) (number of settlements: 2316).

DOT, days of therapy; NHSC, National Healthcare Services Centre; RUGP, ratio of unfilled general practices.

socioeconomic conditions measured by the local unemployment rate and the prescription exemption rate.

The results reported in the second column of [table 3](#) suggest that a major fraction of the negative effect of unfilled general practices originates from settlements where the baseline use of antibiotics was comparatively high. In settlements where the initial use of antibiotics was higher by one standard deviation (by 0.35 on logarithmic scale), the negative effect on the logarithm of antibiotic consumption is stronger by 0.039, comparable with the magnitude of the average effect.

Effects on quality indicators

We investigate some further indicators of antibiotic use that can provide evidence whether the quality of antibiotic prescriptions is affected by the unfilled status of general practices.

First, we look at the effect of the ratio of unfilled general practices on the use of narrow-spectrum versus broad-spectrum antibiotics. According to fixed effects panel regressions ([table 4](#)), this quality index decreases substantially by 0.023 (from an average level of 0.18) if a practice becomes unfilled, suggesting quality deterioration. The numerator (consumption of narrow-spectrum antibiotics) decreases by 7.5% (ie, by 0.078 on the logarithmic scale), while the denominator (consumption of broad-spectrum antibiotics) decreases by a statistically insignificant 1.4%.

Second, [table 4](#) also shows that the consumption of Access antibiotics decreases statistically significantly by 3.1% (ie, by 0.032 on the logarithmic scale) if a practice becomes

unfilled, while the consumption of Watch and Reserve antibiotics decreases by a statistically insignificant 1.9%. These results again suggest a quality worsening effect.

GP care use and antibiotics: understanding the mechanism

We also examine how the effect of unfilled general practices on the number of GP visits interferes with the effect on antibiotic use. Our regression results show that living in a village where the single general practice is unfilled implies on average 9.3% fewer GP visits (95% CI 5.6% to 12.9%; -0.098 on the logarithmic scale), holding settlement-level controls and time invariant characteristics fixed, and netting out time effects ([table 5](#), first column). Due to the measurement error in the GP care use indicator (measured at the centre of a catchment area, as explained in section Settlement-level demographic, socio-economic and GP care data), the true effect could be even higher. Nevertheless, the results indicate a substantial negative effect of unfilled practices on GP care use.

It is also known from our results that per capita antibiotic DOT decreases on average only by 3.2% ([table 2](#)). The smaller decrease in antibiotic use than in GP visits suggests that the limited availability of GP care is partly compensated by a higher likelihood of antibiotic prescription per GP visit. Indeed, as shown in the second column of [table 5](#), a general practice becoming unfilled increases antibiotic prescriptions per GP visits by 7.5% (ie, by 0.073 on the logarithmic scale). Hence the overall negative effect of unfilled practices on antibiotic use stems from

Table 4 Effect of the ratio of unfilled general practices on the use of selected antibiotic categories

	Log of per capita DOT				
	Quality index	Broad spectrum	Narrow spectrum	Access	Watch & Reserve
Ratio of unfilled practices	-0.0233** (0.0101)	-0.0139 (0.0129)	-0.0781** (0.0387)	-0.0319** (0.0137)	-0.0187 (0.0176)
Ratio of females	-0.134 (0.122)	1.625*** (0.212)	1.500*** (0.519)	1.466*** (0.246)	1.376*** (0.290)
Age 0–17 ratio	0.154 (0.1860)	-0.373 (0.279)	-1.087 (0.766)	-0.0951 (0.329)	-0.411 (0.404)
Age 60 plus ratio	0.257* (0.154)	-0.268 (0.273)	0.0341 (0.704)	-0.162 (0.302)	-0.101 (0.369)
Unemployment rate	-0.0393 (0.0423)	-0.0904 (0.0637)	-0.591*** (0.194)	-0.224** (0.0934)	-0.0415 (0.0788)
Prescription exemption rate	-0.0371 (0.0819)	0.209* (0.109)	0.276 (0.407)	0.230 (0.153)	0.407*** (0.150)
Specialist outpatient capacities	-0.00293 (0.00381)	-0.00119 (0.00463)	0.00252 (0.0211)	0.00290 (0.00795)	-0.00168 (0.00864)
Monthly date effects	yes	yes	yes	yes	yes
Settlement fixed effects	yes	yes	yes	yes	yes
Observations	162 256	162 256	141 208	161 866	161 261

Cluster-robust SEs in brackets with clustering on the micro-regional level, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. Settlement-month level NHSC data, weighted by the population of settlements, years: 2010 to 2015. Number of settlements: 2316. Quality index: ratio of DOT of narrow-spectrum vs broad-spectrum antibiotics.

DOT, days of therapy; NHSC, National Healthcare Services Centre.

the limited access to GP care, and not from the deputy GPs prescribing fewer antibiotics per visit.

DISCUSSION

Using administrative data from Hungary, we estimated the effect of primary care availability on antibiotic use. We exploited the geographical and time variation in unfilled general practices as a source of exogenous variation in primary care supply. We estimated a negative

3.2% average effect on per capita antibiotic use if the ratio of unfilled general practices increase from 0 to 1 in a settlement (eg, if the only general practice becomes unfilled in a settlement). Larger effects are observed for settlements with limited availability of specialist care and pharmacies. While the use of antibiotics decreases as a consequence of the supply-side effects, we also find evidence that the quality of antibiotic prescriptions deteriorates.

Table 5 Effect of the ratio of unfilled general practices on annual GP visits and antibiotics prescription per GP visit

	Log of annual GP visits per capita	
	Log of annual GP visits per capita	Log of antibiotic DOT per GP visit
Ratio of unfilled practices	-0.0980*** (0.0207)	0.0726*** (0.0230)
Ratio of females	1.440*** (0.333)	-0.252 (0.388)
Age 0–17 ratio	-1.126*** (0.371)	1.042** (0.437)
Age 60 plus ratio	0.592 (0.392)	-0.632 (0.475)
Unemployment rate	-0.0202 (0.0847)	-0.117 (0.0999)
Prescription exemption rate	0.498*** (0.180)	-0.339* (0.195)
Specialist outpatient capacities	-0.00425 (0.00596)	0.000564 (0.00585)
Year effects	yes	Yes
GP catchment area fixed effects	yes	Yes

Cluster-robust SEs in brackets with clustering on the micro-regional level, * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. GP catchment area-year level T-STAR data, weighted by the population of GP catchment areas, years: 2010 to 2015. Number of observations: 8505 (number of GP catchment areas: 1438).

DOT, days of therapy; GP, general practitioner.

In our main specification, we estimated panel regressions with settlement-level fixed effects. This strengthens the identification of the parameters of our interest because we could control for any time-invariant unmeasured characteristics of the settlements that could interfere with the probability of a general practice being unfilled. We also demonstrated that the use of settlement-level random effects, or a non-linear specification in the explanatory variables give almost identical estimates for the impact of unfilled practices on antibiotic indicators, so our results are robust to these specification changes.

Our results come from a healthcare system where general practices are overwhelmingly single-handed and group practices are extremely rare. Therefore, our main explanatory variable, the ratio of unfilled practices in a village, captures the availability of primary care reasonably well. We note, however, that using a more conventional indicator such as the number of GPs per 1000 inhabitants gives similar results: higher density of GPs is associated with higher use of antibiotics. Specifically, if the number of available GPs per 1000 inhabitants in a settlement increases by one (the average of this measure is 1.7 in our sample) then per capita antibiotic use increases by 2.2% (95% CI 1.3% to 3.2%), according to our calculations.

Our results are in line with the literature showing that geographical distance to healthcare facilities negatively affects healthcare use (Haynes *et al*, Hyndman *et al*, Elek *et al*^{30–32}, among others). Also, we contribute to the literature on the determinants of antibiotic prescribing behaviour of GPs. There is evidence that pay-for-performance programmes can reduce broad-spectrum antibiotic prescribing.^{33 34} Our results indicate that reduced availability of primary care worsens the quality of antibiotic prescribing; the share of broad-spectrum antibiotics increases.

The study is subject to some limitations. The external validity is restricted to similar healthcare systems, that is to systems with mostly single-handed general practices, which serve as the first point of contact to the system. Also, while the consumption of pharmaceuticals other than antibiotics is also likely to be affected by the availability of primary care, it remains for future research to investigate these effects. Finally, despite the richness of the administrative data used, we face some data limitations. There could still remain unobserved time-varying confounders that are related both to antibiotic use and the availability of GP care (such as changing population health status and health behaviour). We do not have information on who exactly – among the other GPs – provide service as deputy GP and how exactly the opening hours change if a general practice becomes unfilled. Also, we do not know to what extent patients switch from an unfilled general practice to another, filled one. These details could provide further insights into the mechanisms how primary care availability affects antibiotic use.

Overall, the results suggest that given the socioeconomic status of a settlement, and given the supply of secondary care, the availability of primary care has a

non-negligible effect on the quantity and quality of antibiotic consumption. The main message is that antibiotic use can be influenced by supply-side factors, the usage is sensitive to the access to primary care. The overuse of antibiotics might be limited if the number of avoidable visits to primary care physicians is decreased. However, it is important to keep in mind that for avoiding antibiotic resistance, it is desirable not only to reduce the overall use of antibiotics, but also to shift consumption towards narrow-spectrum antibiotics. According to our results, such a quality improvement cannot be achieved with limiting the availability of primary care.

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