BMJ Open OpenPrescribing: normalised data and software tool to research trends in **English NHS primary care prescribing** 1998-2016

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ABSTRACT

Objectives We aimed to compile and normalise England's national prescribing data for 1998–2016 to facilitate research on long-term time trends and create an opendata exploration tool for wider use.

Design We compiled data from each individual year's national statistical publications and normalised them by mapping each drug to its current classification within the national formulary where possible. We created a freely accessible, interactive web tool to allow anyone to interact with the processed data.

Setting and participants We downloaded all available annual prescription cost analysis datasets, which include cost and quantity for all prescription items dispensed in the community in England. Medical devices and appliances were excluded.

Primary and secondary outcome measures We measured the extent of normalisation of data and aimed to produce a functioning accessible analysis tool.

Results All data were imported successfully. 87.5% of drugs were matched exactly on name to the current formulary and a further 6.5% to similar drug names. All drugs in core clinical chapters were reconciled to their current location in the data schema, with only 1.26% of drugs not assigned a current chemical code. We created an openly accessible interactive tool to facilitate wider use of these data.

Conclusions Publicly available data can be made accessible through interactive online tools to help researchers and policymakers explore time trends in prescribing.

INTRODUCTION

In 2016. National Health Service in England (NHS) prescribing cost ± 9.20 billion,¹ approximately 9% of the annual NHS budget.² Prescribing behaviour is expected to respond within the dynamic system of evidence-based medicine through changing patterns of disease, innovation in medical treatments and new evidence. Monitoring long-term time trends in prescribing is therefore useful to observe changes in practice, to provide a form of feedback to ensure there are no unexpected or undesirable changes and to facilitate tracking and forecasting of costs.

Strengths and limitations of this study

- We processed publicly available annual data for the whole of England's community dispensing-not a sample.
- We corrected for population size, inflation and (where possible) drugs changing name and/or classification over time.
- We produced a free, openly accessible tool for wider use, displaying trends in items, cost, price per item and quantity per item for each product for 1998-2016, which can be updated annually.
- The tool is limited to product-level data, not individual presentations, and wide-scale correction for dosage was not possible.
- Users can also download our normalised dataset to carry out their own analyses.

Protected by copyright, including for uses related to text and data m NHS Digital publish monthly and annual prescribing datasets from the NHS Business Services Authority, along with static reports on prescribing trends. However, this does ≥ not allow readers to interrogate topics of interest in detail, and the large datasets can be complex to manage. We provide a service a at OpenPrescribing.net that facilitates exploration of outliers and trends for individual general practices in NHS England, which has provided over 250000 analyses to 50000 \overline{a} users over the past year. The detailed dataset that drives this service (running to over 10m rows a month) is only available from 2010 onwards. The annual prescription cost analysis (PCA) data, aggregated nationally (with B no data on individual practice) and by year (with no data on prescribing changes each month) are available back to 1998. These data are freely accessible but consist of individual files for each year of prescribing, which cannot be straightforwardly combined and therefore do not facilitate interrogation of time trends. Additionally, identifiers for individual drugs may change name, or location

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Glossary of prescribing data terminology Box 1

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related

Example presentation: Tra	adorec XL tablets 30)0 mg				
Drug name	BNF chemical name	BNF Section name	BNF Subparagraph name	Items	Quantity	NIC
Tradorec XL_Tab 300mg	Tramadol Hydrochloride	Analgesics	Opioid Analgesics	6374	324167	£152358
The drug name describes the	e full presentation of the d	rug that is the form	lation and strength as well	as the drug's	brand or generic	(product) name

The chemical name is the standard registered name for the active constituent of the medicine. It is not always an individual chemical: examples include 'paracetamol combined preparations' and 'paracetamol and caffeine'.

- Numerical codes representing Chapter, Section, Paragraph and Subparagraph are also supplied. These represent only the first seven characters of each drug's unique 15-character British National Formulary (BNF) code (see box 2).
- Items are functionally equivalent to prescriptions; they do not take into account the quantity (number of boxes/bottles, etc) dispensed to the same person. Items may vary in the quantity prescribed.

Quantity represents the quantity of a drug dispensed, with units of measurement (units/tablets/grams/millilitres, etc) dependent on its formulation.

Net ingredient cost (NIC) represents the basic price of the medicine, that is, the drug tariff price, or, if not listed, the price published by the manufacturer or supplier. NIC may be subject to further charges and/or discounts. Patients who are eligible contribute a fixed fee towards each prescription charge, but this only applies to a minority of items and it is not possible to identify which items in this dataset.

within the British National Formulary (BNF), over time, making simple compilation of the data impossible.

The value of PCA data is indicated by the numerous previous studies using it to assess prescribing trends^{3–5} or to detect changes in response to guidelines or safety alerts.^{6–8} These studies have been focused on data for a small number of drugs, manually aggregated for each bespoke analysis; furthermore, given publishing delays for academic manuscripts, the data are commonly very delayed, and readers cannot easily place the findings in context of current clinical practice or expenditure.

We therefore set out to aggregate all available PCA data into a single data frame for longitudinal analysis of trends, in a service that could be easily updated; to generate an interactive online service where any user can explore and monitor time trends in prescribing using the latest available data and to share all resources for reuse by others as open data.

METHODS

Data sources

Every available PCA annual dataset was downloaded from NHS Digital or National Archives, covering calendar years 1998 to 2016.⁹

Data structure

Protected by copyright, including for use: Each annual PCA dataset includes all items dispensed in England by pharmacy/appliance contractors, dispensing doctors and items personally administered by doctors, whether or not they were prescribed in England or other parts of the UK. Items dispensed in other settings (prisons, 6 hospitals and private prescriptions) are excluded. Prior to 2010, the data were rounded to the nearest 100 and excluded drugs with fewer than 50 items prescribed, accounting for 0.01% of total items.¹⁰ Definitions of excluded drugs with fewer than 50 items prescribed, key terms used in the PCA data (and NHS primary care prescribing data more generally) are given in box 1, and a full glossary of terms is available.¹¹

Every drug presentation (ie, each formulation, dose and product combination) is described by a unique . > drug name and has a unique 15-digit structured British National Formulary (BNF) code, an example of which is given in box 2. The BNF contains an entry for every product available to be prescribed in Britain, including <u>م</u> medicinal products, dietary supplements, complementary therapies and physical appliances such as bandages. The similar technologies hierarchical BNF codes imply a data schema as follows: each presentation of a drug has a product name, which may

British National Formulary code structure. Box 2

Example	presentation	n: Tradorec XI	tablets 300 mg				
Chapter	Section	Paragraph	Subparagraph	Chemical	Product	Presentation	Generic equivalent
04 Central Nervous System	07 Analgesics	02 Opioid Analgesics	0 Opioid Analgesics	40 Tramadol Hydrochloride	BI Tradorec	AC Tradorec XL_Tab 300mg	AM *

*Generic equivalent allows matching with the strength and formulation (presentation) of the generic product (which will always have product code 'AA'). For generic presentations, the product name will match the chemical name (but sometimes with a different abbreviation, for example, 'Tramadol HCI').

be either a brand name or the generic chemical name; as such, each product can be mapped to a chemical. Each chemical is a member of a Paragraph in the BNF (some of which are divided into Subparagraphs, which themselves often approximate to a class of drugs). Each Paragraph belongs to a Section, which is in turn a member of a Chapter (often approximating to a system of the body, such as 'cardiovascular').

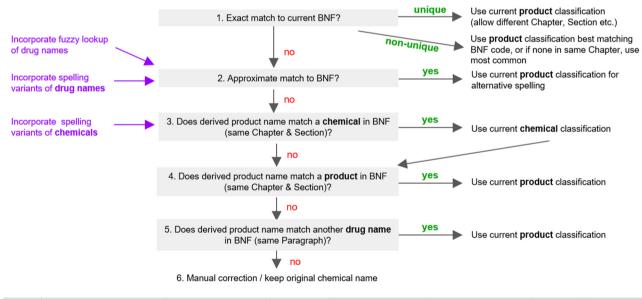
In the PCA data, only the first seven characters of the BNF code for each drug are supplied rather than the full BNF code. Therefore, the 'drug name' is the only source of information on the formulation and dose; however, from this, the BNF code can usually be imputed, but this becomes increasingly difficult for older drugs no longer listed in the BNF. Although each drug's chemical name is also supplied, chemicals are not all unique (eg, 'other preparations'); names may change their spelling over time; and chemicals may move between Paragraphs, Sections and Chapters. Indeed, classifications at any level of the hierarchy can be subject to renaming, spelling change, subdivision, reorganisation and removal.

Data management, aggregation and cleaning

All data were grouped by drug name, combining any available in multiple formulations (despite having identical names). Ultimately, following cleaning, data were grouped to product level. Medical devices/appliances and any other items, Chapter numbers above 15 were excluded.

A key user need was to explore prescribing trends for individual members of a class of drugs over time. This required all data to be normalised, with each individual drug consistently appearing in the correct location in the data schema; that is, all individual presentations of a chemical all mapped under that chemical and all chemicals mapped under the correct Subparagraph/Paragraph (often similar to drug class) of the BNF. To achieve this consistency, we aimed to map each drug to its current position in the latest BNF dictionary up to the level of its 11-character 'product' code through an incremental process. This is summarised below and in figure 1.

Lacking the full BNF code, we attempted to match **ö** each drug name to a current BNF presentation. Those without an exact match (eg, formulation variants no longer available) could sometimes be matched to a similar BNF presentation name, for example, by finding a similar formulation or using the 'fuzzy' lookup add-on for Excel and validated manually.¹² Other drug names could only be matched up to current BNF codes by using their product or chemical names. Matching at each stage was improved by disregarding capitalisation or spacing and spelling changes (eg, Sulphur/Sulfur); these include changes identified within the data and those occurring



Match Type	Drug Name Original	Chemical Original	BNF Code Original	Drug Name Current	Chemical Current	Code Derived	Product Current
1	Aspirin_Pdr Sach 37.5mg	Aspirin	0407010	Aspirin_Pdr Sach 37.5mg	Aspirin	0209000A0AA	Aspirin (Antiplatelet)
2	Frusemide_Tab 20mg	Furosemide	0202020	Furosemide_Tab 20mg	Furosemide	0202020L0AA	Furosemide
2	Amlodipine Besyl_Liq Spec 10mg/5ml	Amlodipine Besylate	0206020	Amlodipine_Oral Soln 5mg/5ml	Amlodipine	0206020A0AA	Amlodipine
3	Felendil XL_Tab 10mg P/R	Felodipine	0206020	(none)	Felodipine	0206020F0	Felendil XL
4	Cozaar-Comp_Tab	Losartan Potassium With Diuretic	0205052	(none)	Losartan Potassium With Diuretic	0205052P0BB	Cozaar-Comp
5	Atorvastatin_Liq Spec 20mg/5ml	Atorvastatin	0212000	Atorvastatin_Oral Soln 20mg/5ml	Atorvastatin	0212000B0AA	Atorvastatin
6	Terbut Sulph_Syr 1.5mg/5ml S/F	Terbutaline Sulphate	0301011	(none)	Terbutaline Sulfate	0301011V0AA	Terbut Sulf

Figure 1 BNF code normalisation process flow chart describing how drug names were matched to the current BNF. Examples of matches at each numerated stage are given in table (bottom), with code/name changes in red. The first example demonstrates a drug which was matched to the current BNF through an exact match by name but had moved from Chapter 4 to Chapter 2. For other types of matching, care was taken to avoid mistakenly matching to similar (but distinctly different) drugs across chapters/sections. BNF, British National Formulary. when many old British spellings (the 'British Approved Name') were replaced with international standard names 'Recommended International Non-Proprietary (the Name').¹³ Remaining drug names in the most-prescribed Chapters (1-6 and 10) were matched to current drug names manually (eg, resolving non-matches due to rearrangement of word order); any others kept their original chemical name and a proxy product name was derived from the drug name field. Full methodology for this matching process is available in our technical documentation online¹⁴ and in online supplementary material.

We measured the extent of normalisation of drug names and classifications and present summary statistics on these.

Normalisation for inflation and population

Prescribing costs were corrected for inflation using the UK's annual consumer price index figures normalised to 2016.¹⁵ Number of items prescribed and costs were divided by the population each year to calculate values per thousand population based on midyear population estimates for England only.¹⁶ We also supply the original number of items and cost in our output.

Interactive analysis tool

Having generated a normalised dataset and a method for updating it, we then set out to implement a free, interactive online data analysis tool where any user can visually explore time trends in prescribing. This was built using Tableau Public, a freely accessible interactive data presentation platform which permits rapid prototyping; however, other front ends onto the same underlying datasets could also be implemented using open-source tools such as Shiny¹⁷ or in Python libraries such as d3¹⁸ with more software engineer resource. Our user needs for the tool were as follows: the ability to display trends in items and cost, normalised for total population change and inflation and also to calculate the average cost per item and quantity per item for each product.

Having delivered the tool, we used it to generate trends data and graphs for a range of clinical areas where prescribing trends have been previously studied and published to demonstrate the ability of our tool to replicate and extend these works.

Data and code

The full compiled and processed PCA datasets are available online via FigShare,¹⁹ SQL code is in online supplementary appendix, and the Trends tool (Tableau workbook) is available to use via https://OpenPrescribing.net/pca.

RESULTS

Data compilation and overall prescribing trends

All data were successfully imported. There were 169100 lines of data in the compiled 1998-2016 dataset (Chapters 1-15) and 169038 in the processed data, the reduction caused by aggregation of a small number of drugs

available in multiple formulations despite having identical names. Total items (14.8 billion), cost (£136.9 billion) and distinct drug names (22 496) remained consistent before and after data processing (table 1, see online supplementary table S1). The inclusion of low volume prescribing in the published datasets from 2010 caused a substantial rise in the number of distinct drugs per vear but not items or cost (table 1). As can be seen from table 1, the inflation-corrected cost in 2016 UK sterling equivalent for all prescribing in NHS England primary care rose from £6.3 billion in 1998 to £10.1 billion in 2004 but then decreased to £8.3 billion in 2016. Items per 1000 population (correcting for crude population growth) has Š grown from 10180 in 1998 to 19196 in 2016, on average increasing by 3.6% per year. The discrepancy between 8 items and cost is caused by variation in the choice of drugs being prescribed and their individual prices, for example, switching from branded drugs to cheaper generic versions including after patent expiry.

Data normalisation

Data were normalised using the methods described above. Of the distinct drug names in the data, 87.5% were matched exactly to a current BNF name, and a further 6.5% matched approximately (table 2). Name changes are particularly prevalent in Chapter 3 (Respiratory) due mainly to the addition in 2004 of a space when a number of doses is given, as is common for inhalers, for example, ç text 'Salbutamol_Inha 100 mcg (200D)' became 'Salbutamol_ Inha 100 mcg (200 D)'.

Of the distinct drug names (23 275, taking into account some drugs having multiple BNF classifications), over 91% could be matched to a current product in the BNF, with no change in code (table 3). Less than 5% could not be matched to a current product and/or chemical code, under 1% of items prescribed. These drugs were assigned ≥ proxy product names (derived from their drug name) so that all data could be presented visually, and for those not matched to a current chemical, the original chemical ğ name was used (mostly 'Other Preparations'). However, normalisation was focused on seven of the most prescribed Chapters with the greatest medical interest (1–6 and 10). <u>0</u> The normalisation of drugs in other Chapters could therefore potentially be improved. In particular, Chapters 9 (Nutrition) and 13 (Skin) have substantial levels of prescribing, but are complex, containing many different drug names and non-drug products such as topical applications and dietary supplements. Other groups with a & particular interest in nutrition or dermatology may wish **g** to expand our work on manual matching: we would be happy to incorporate such amendments into our dataset. Many of the code changes and non-matches have diminished over time, as expected (online supplementary table S2).

Code changes and normalisation outputs are described in table 3. Headers indicate the highest level in the BNF hierarchy at which drugs have been subject to code changes, for example, 'Section' indicates drug names

	Distinct count of drug name	Items	Items per	· 1000	Cost	Inflation- corrected cost	Inflation- corrected cost per 1000	
/ear	n	n	n	Change (%)	£	2016 £	2016 £	Change (%)
998	6338	497.0 million	10180		£4440 million	£6280 million	£128626	
999	6587	513.4 million	10471	+2.9	£5011 million	£6999 million	£142746	+11.0
2000	6613	535.1 million	10868	+3.8	£5284 million	£7318 million	£148648	+4.1
2001	6754	569.2 million	11510	+5.9	£5784 million	£7914 million	£160049	+7.7
2002	6834	598.6 million	12050	+4.7	£6487 million	£8768 million	£176491	+10.3
2003	6893	630.3 million	12625	+4.8	£7113 million	£9488 million	£190035	+7.7
2004	6912	666.0 million	13268	+5.1	£7645 million	£10063 million	£200482	+5.5
2005	6907	698.8 million	13808	+4.1	£7452 million	£9609 million	£189875	-5.3
006	6810	728.4 million	14292	+3.5	£7660 million	£9655 million	£189436	-0.2
007	7056	771.8 million	15022	+5.1	£7810 million	£9614 million	£187112	-1.2
800	7202	816.7 million	15762	+4.9	£7716 million	£9174 million	£177047	-5.4
009	7401	859.2 million	16461	+4.4	£7892 million	£9176 million	£175805	-0.7
010	11703	898.4 million	17065	+3.7	£8162 million	£9193 million	£174636	-0.7
011	11751	931.6 million	17541	+2.8	£8101 million	£8734 million	£164457	-5.8
012	12207	968.9 million	18112	+3.3	£7802 million	£8176 million	£152836	-7.1
013	12318	996.2 million	18494	+2.1	£7846 million	£8022 million	£148920	-2.6
014	12576	1027.0 million	18908	+2.2	£8022 million	£8078 million	£148718	-0.1
2015	12875	1043.5 million	19046	+0.7	£8403 million	£8461 million	£154444	+3.9
2016	13285	1060.9 million	19196	+0.8	£8284 million	£8284 million	£149892	-2.9
otal	22 496	14810.9 million	284680	+3.6	£136914 million	£163006 million	£3 160 254	+1.0
2016 Total 'Drug nar ingredien	13285 22496 me' is the field	1060.9 million 14810.9 million describing the prese (1). 'Change' is the y	19 196 2 84 680 entation of ea	+0.8 +3.6 ich drug, tha	£8284 million £136914 million	£8284 million	£149892 £3 160 254	-2.9 +1.0

which have not changed Chapter but have moved Section. 'No product match' indicates drug names matched to a chemical (9-character BNF code) but with no current matching product (11 characters). 'No chemical match' indicates drug names matched neither to a chemical nor product. The total count of drug names increases when separated by Chapter because four drug names currently exist in two different Chapters.

Interactive data analysis tool

We created a tool which allows anyone to explore the prescribing data, available directly at https://openprescribing.net/pca. Users can search by chemical, Paragraph, Section or Chapter to view time trends in items and costs on stacked charts, where both the overall trends and the relative contribution from each product/chemical can be seen. The cost per item and quantity per item for each product are also shown, which can assist in interpretation of trends in some cases. However, these calculations carry a 'use with caution' note, as items may represent different pack sizes and quantities cannot be reliably summed across preparations because of different strengths and formulations. The page features

The tool can be used to facilitate novel research into time trends and factors associated with changes in practice such as publication of guidelines or evidence land-D marks or changes in price. It can replicate and extend the main findings of previous papers which researched trends for different clinical areas using PCA data. For example, the antipsychotic drug switches which occurred in England following a licence restriction²⁰ can be replicated in the tool and the trends extended to the latest data (figure 2A). This also shows the dramatic reductions in cost that followed the expiry of patents for risperidone and olanzapine. We also replicate antidepressant prescribing trends, previously reported up to 2010⁵ and show that how the overall use of these antidepressants has continued to rise, in particular sertraline (figure 2B). We also replicate findings on the rise of thyroid hormones³ and testosterone,⁴ where we show that prescribing of these drugs continued to rise, with a disproportionate increase in cost (see online supplementary figure S1a,b). We are using this dataset and tool in our academic papers on

 Table 2
 Number and percentage of drug names subject to changes within 1998–2016 PCA data when compared with the current BNF by Chapter

	Name/ change	spelling e	No change		No mat	No match	
Current chapter code (name)	n	%	n	%	n	%	n
1 (Gastrointestinal system)		0.0	1041	99.1	9	0.9	1050
2 (Cardiovascular system)	177	7.2	2278	92.5	8	0.3	2463
3 (Respiratory system)	238	20.8	893	78.0	14	1.2	1145
4 (Central nervous system)	341	8.8	3535	90.7	21	0.5	3897
5 (Infections)	119	9.1	1184	90.7	3	0.2	1306
6 (Endocrine system)	203	13.2	1320	86.0	11	0.7	1534
7 (Obstetrics, gynaecology and urinary tract disorders)	4	0.7	508	90.7	48	8.6	560
8 (Malignant disease and immunosuppression)	2	0.3	538	91.8	46	7.8	586
9 (Nutrition and blood)	210	4.0	4281	81.3	776	14.7	5267
10 (Musculoskeletal and joint diseases)	89	8.5	956	90.8	8	0.8	1053
11 (Eye)	18	3.0%	532	87.2	60	9.8	610
12 (Ear, nose and oropharynx)	16	3.4	373	78.5	86	18.1	475
13 (Skin)	22	1.1	1772	89.5	186	9.4	1980
14 (Immunological products and vaccines)	7	3.0	198	85.0	28	12.0	233
15 (Anaesthesia)	15	5.2	229	78.7	47	16.2	291
Grand total	1461	6.5	19638	87.5	1351	6.0	22450

These name changes include changes in word order, spacing, capitalisation, abbreviation, punctuation (eg, "Califig_(California Syr Of Figs)"/"Califig_California Syr Of Figs"), spelling (eg, "Sulphate"/"Sulfate"), brand name (eg, "Laxoberal_Liq"/"Dulcolax Pico_Liq") and formulation (eg, "Castor Oil"/"Castor Oil_Liq"). The total count of drug names is reduced compared with table 1 because the same drugs can appear over multiple years but only rarely in multiple Chapters.

BNF, British National Formulary; PCA, prescription cost analysis.

trends and variation in NHS prescribing; we encourage others to use our dataset and tool in their own work.

The tool can also be used to complement studies performed in more detailed prescribing data such as the Clinical Practice Research Datalink (CPRD), by giving the full national picture, and giving more longitudinal data that updates with new data releases. For example, several previous publications have reported on patterns of prescribing of smoking cessation medication in The Health Improvement Network database.^{21–23} This included reporting of a possible decline in prescribing despite increased incentives for general practitioners introduced in 2012. We can confirm this decline and show that it continued beyond 2013 (figure 2C). We also show that the slow decline in quinine usage following safety alerts in 2010²⁴ has continued at a similar pace (online supplementary figure S1c). CPRD data contain individual patient records and can therefore be used to assess detailed questions about treatments in specific cohorts of patients. However many labour-intensive CPRD analvses have been conducted to interrogate simple broad prescribing trends which could more straightforwardly be conducted using aggregated and normalised national data, with greater coverage of years and total population. Furthermore, for analyses interrogating national trends and responses to guidelines, in many cases a prescribing change which can only be detected in individual patients'

records, and cannot be detected in national data, may not be relevant in terms of population health or the health service.

Additional tabs in the tool allow discovery of higher-level trends, including Chapter and Section trends, Sections ≥ ranked by items/cost for any selected year, calculation of training, the change in items/cost for each Section between any selected year to the latest year, and the top 20 Paragraphs by items and cost. The Chapter trends page, for example, shows that much of the decline in prescribing costs since the peak in 2004 (table 1) is attributable to a drop in the cost of cardiovascular drugs (figure 3A), and the Section trends page further shows that lipid-regulating drugs lar technologies (Section 2.12) and Drugs for Hypertension and Heart Failure (Section 2.5) experienced the largest cost reductions at that time (figure 3B).

DISCUSSION

Summary

It was possible to aggregate all PCA data from 1998 to 2016 and normalise for most changes in drug names and classifications. Only 87.5% of drug names matched exactly to a current BNF name and 8.7% had undergone some change in classification; however, all drugs in core clinical chapters were reconciled to their current location in the data schema. We generated an interactive online service

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Table 3 Summary of drug code changes within the 1998–2016 prescribing datasets, also separated by (current) chapter. Chapter names can be found in table 2

	BNF cod	e change		No	No				
	Chapter	Section	Paragraph	Sub paragraph	No change	product match	chemical match	Grand total	
Distinct count of drug name	94	52	560	203	21258	815	293	23275	
% of drugs	0.40%	0.22%	2.41%	0.87%	91.33%	3.50%	1.26%	100%	
% of items	0.04%	0.01%	0.84%	0.51%	97.67%	0.84%	0.10%	100%	

Current Chapter number	Chapter	Section	Paragraph	Sub paragraph	No change	No product match	No chemical match	Grand total
1		4	109		1146	6		1265
2	5	3	4	19	2446	7		2484
3	1	6	19		1132	10		1168
4	42	13	173		3831	9		4068
ō	1		161	116	1241	2		1521
6	4	1	2	2	1521	10		1540
7	5		12		528	28		573
3	1		4		579	6	1	591
9	1	7	6	54	4510	507	245	5330
10	27				1041	9		1077
11	1	5	5		571	28	6	616
12					420	44	11	475
13	5	13	28	12	1847	95	30	2030
14			37		195	17		249
15	1				254	37		292
Grand total	94	52	560	203	21262	815	293	23279

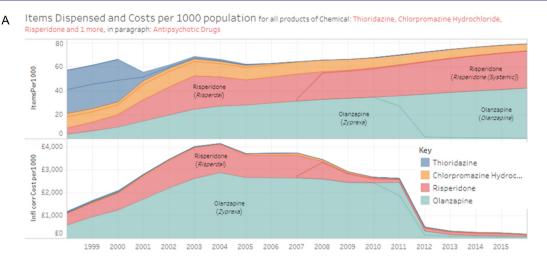
where any user can explore time trends in prescribing broken down by product, chemical, Paragraph, Section and Chapter; this openly accessible interactive data analysis tool provides overviews and insights comparable to previous labour-intensive bespoke data analysis research projects.

Strengths and weaknesses of this study

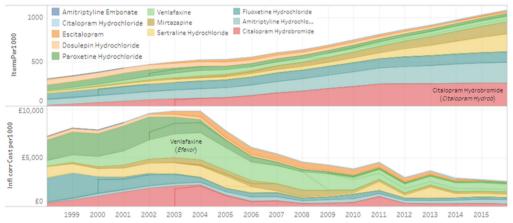
Our tool covers the data for the whole of England's community dispensing, not a sample. We are surprised to note that this is the first project aiming to aggregate long-term trends across the entire prescribing dataset and provide an openly accessible tool for wider use. Many drugs changed name and/or classification over time, but valid chemicals were successfully assigned to all items in Chapters 1-6 and 10, and product names were derived for every drug, allowing maximum consistency in trends analysis. The tool is limited to product-level data due to the wide number of different presentations available.

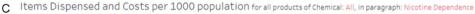
We used items to measure prescribing volume. Quantity is generally more complex for making comparisons as there is wide variation caused by the units, which may

be the number of pills or millilitres, units (eg, inhalers containing multiple doses) or other unit measure. Converting quantities to approximate daily dose sizes (such as defined daily doses) is possible, but the conversion tables available are not sufficiently comprehensive to allow this across the entire dataset, and this would be Ы even more difficult for discontinued drugs. We therefore rejected this option in favour of being able to publish a complete dataset. However, users wishing to analyse data by daily doses can download our BNF-normalised dataset to apply these calculations. Using items also has limitations, as it does not take into account number of packs prescribed per prescription, pack size or dosage. We are launching this tool publicly and will monitor user volume and user feedback: if appropriate we will improve the tool by replicating and expanding it using bespoke software as per our other data analysis tools on OpenPrescribing.net for exploring variation in prescribing at Clinical Commissioning Group (CCG) and individual practice level. We will update the tool annually, dependent on continuing funds for the Open-Prescribing project.



B Items Dispensed and Costs per 1000 population for all products of Chemical: All, in paragraph: Other Antidepressant Drugs, Selective Serotonin Re-Uptake Inhibitors, Tricyclic & Related Antidepressant Drugs





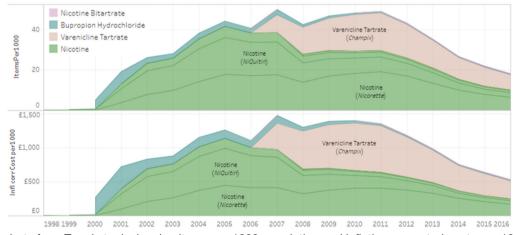


Figure 2 Screen shots from Trends tool, showing items per 1000 population and inflation-corrected costs per 1000 population for selected drugs. Product names are in parentheses. (A) Prescribing trends for four selected antipsychotic chemicals following the safety alert on thioridazine.²⁰ Full dashboard available at https://public.tableau.com/shared/XX7DTWSG2?:display_count=yes. (B) Prescribing trends for selected antidepressant chemicals.⁵ Full dashboard available at https://public.tableau.com/shared/72SJGGP89?:display_count=yes. (C) Prescribing trends for all chemicals within the Paragraph of Nicotine Dependence (smoking cessation medications). Full dashboard available at https://public.tableau.com/shared/6BW9J5RJB?:display_count=yes.

Findings in context of other research

Long-term trends in prescribing have previously been reported on a wide variety of clinical areas, using PCA

data as well as other sources.^{3–5 20 22 23} These are static, not updated and rapidly out of date. Although using CPRD allows a more detailed analysis and investigation of

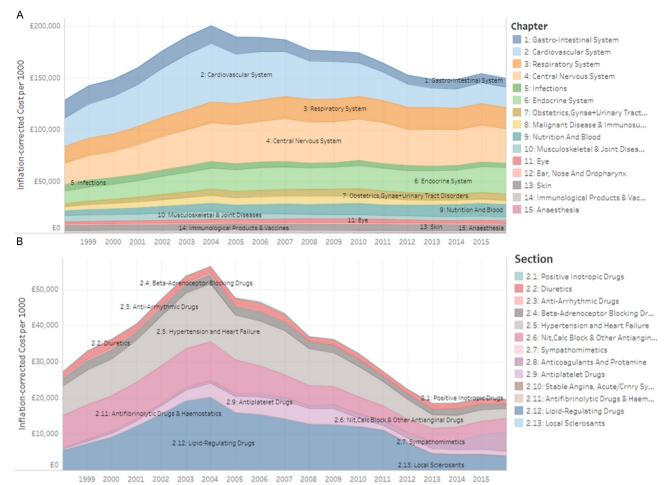


Figure 3 Screenshots from Trends tool, showing inflation-corrected costs per 1000 population (A) by Chapter, and (B) by Section for Chapter 2 (Cardiovascular System).

patient factors associated with prescribing, it takes a great deal of preparation and time to complete. Our tool can replicate some trends found in CPRD and so may provide a useful tool for preliminary investigation of trends. It can also help to confirm whether findings from regional datasets of rich individual patient data (IPD) from electronic health records sources are representative of the national picture, while avoiding repeated work and replication in new IPD datasets. In our related publications on variation and trends in specific disease areas, we report comparisons between trends in PCA data and trends from other more labour-intensive sources such as CPRD in more detail.

6

The UK government produces a 10-year trends document following the annual PCA data release, containing an overall summary of high-level trends and a brief breakdown of six interesting topics with the greatest level or change in prescribed items and cost.²⁵ However, the reported topics are few in number, chosen by NHS Digital, restricted to 10 years of data, do not correct for inflation or population growth, are not easily discoverable by subject specialists and readers are not able to interrogate their own topics of interest in detail. From 2016, the compiled datasets were also made available so users may conduct their own exploration of the data, but, without drug names or categories being normalised, this is little better than the raw data, which we have processed into a normalised longitudinal dataset.

Policy implications and future research

Published papers can provide a useful and detailed insight into prescribing trends,^{3–5} but give a single snapshot which may quickly become out of date. Our tool facilitates ongoing monitoring by researchers and policy-makers to assess prescribing changes in any area of concern or clinical interest they have identified and permits interactive exploration of detailed issues in the technol data, such as individual presentations of chemicals by any interested user. As part of our OpenPrescribing work, we are using prescribing data to investigate adherence to guidelines and changes in practice in various clinical areas to detect anomalous changes in individual practices relative to national trends to send practices alerts and to identify cost-saving opportunities. We have produced various manuscripts using the longitudinal data presented here as part of a range of data sources to describe variation in prescribing in primary care. We are happy to collaborate with other teams of clinicians and academics; we also release our underlying dataset and code as open data for reuse with citation.

CONCLUSIONS

Long-term trends in prescribing are interesting for a number of applications. While previous work on prescribing data has focused on static, manual analysis of a small number of drugs, modern data science approaches make it possible to create interactive services that allow clinicians, healthcare commissioners, policy-makers, academics and any other interested party to interrogate and monitor prescribing trends for any combination of chemicals to identify anomalies or signals of concern and predict spending. We have delivered this using a combination of open-data and freely accessible online tools.

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Contributors BG conceived and supervised the project, HJC designed the methods, conducted the analysis, interpreted the findings, extracted and processed the data in BigQuery, Excel and Tableau with input from BG. HJC and BG wrote the paper. All authors contributed to and approved the final manuscript. BG is the quarantor.

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Patient consent Not required.

Provenance and peer review Not commissioned; externally peer reviewed.

Data sharing statement The full compiled and processed PCA datasets are available online via Figshare (DOI 10.6084/m9.figshare.5447194.v1), SQL code is in supplementary file and the Trends tool (Tableau workbook) is available to use via https://OpenPrescribing.net/pca

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