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Examining opioid prescribing trends for non-cancer pain using an estimated oral morphine equivalence measure: a retrospective cohort study between 2005 and 2015

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Abstract

Background: Over the past 20 years prescription of opioid medicines has markedly increased in the UK, despite a lack of supporting evidence for use in commonly occurring, painful conditions. Prescribing is often monitored by counting numbers of prescriptions dispensed, but this may not provide an accurate picture of clinical practice.

Aim: To use an estimated oral morphine equivalent (OMEQ_e) dose to describe trends in opioid prescribing in non-cancer pain, and explore if opioid burden differed by deprivation status.

Design & setting: A retrospective cohort study using cross-sectional and longitudinal trend analyses of opioid prescribing data from Welsh Primary Care General Practices (PCGP) took place. Data were used from the Secure Anonymised Information Linkage (SAIL) databank.

Method: An OMEQ_e measure was developed and used to describe trends in opioid burden over the study period. $OMEQ_e$ burden was stratified by eight drug groups, which was based on usage and deprivation.

Results: An estimated 643 436 843 milligrams (mg) $OMEQ_e$ was issued during the study. Annual number of prescriptions increased 44% between 2005 and 2015, while total daily $OMEQ_e$ per 1000 population increased by 95%. The most deprived areas of Wales had 100 711 696 mg more $OMEQ_e$ prescribed than the least deprived over the study period.

Conclusion: Over the study period, $OMEQ_e$ burden nearly doubled, with disproportionate $OMEQ_e$ prescribed in the most deprived communities. Using $OMEQ_e$ provides an alternative measure of prescribing and allows easier comparison of the contribution different drugs make to the overall opioid burden.

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How this fits in

It is known that opioid prescribing has increased in the UK over the past 20 years. Measures of prescribing vary and are not always reflective of what is seen in practice, nor do they allow easy identification of populations or individuals most at risk. This study used an OMEQ_e to standardise prescribing data. It demonstrated anomalies in prescription numbers and opioid burden. The use of OMEQ provides more easily comparable data across a range of opioid medicines and warrants consideration as a standard measure of prescribing.

Introduction

The number of prescriptions for opioid medicines issued in the UK has increased substantially over the past 20 years.¹⁻⁶ In particular, prescriptions for 'strong' opioids, such as morphine, oxycodone, and fentanyl, have seen greater increases than those classed as 'weak', such as codeine and dihydrocodeine.^{1,2,6} Prescribing continued to increase even when evidence to support using these medicines for people living with non-cancer pain is largely absent.⁷⁻¹¹

National and international concerns have focused on strong opioids.¹²⁻¹⁴ However, dose and duration of use are more likely indicators of harm or potential for dependence than the choice of drug itself.^{11,15-20} It has been estimated that adverse events occur in as many as 78% of people using opioids over extended periods of time.¹¹⁻¹³ Higher doses¹⁴⁻¹⁷ have been associated with depression and anxiety,¹⁸⁻²⁰ and an increased risk of dependence and misuse.²¹⁻²⁴ It has been proposed the burden or risk of opioids would be more accurately discussed in mg doses or dose equivalents, rather than number of prescriptions alone.^{2,25}

An accurate estimation of opioid burden and risk is especially important in areas of high socioeconomic deprivation, which are associated with poorer health outcomes, higher incidence of chronic pain,^{2,26,27} and mental health disorders compared with the general population.²⁸ Deprivation is associated with higher prescribing of potentially dependence-forming medicines, including opioids, especially for chronic, non-cancer pain in the UK²⁹ and internationally.³⁰ Furthermore, concomitant use of other medicines, such as benzodiazepines and antidepressants with opioids, have also been disproportionately reported in more deprived areas and confer additional risk of harm to the user.³¹⁻³³

Wales has historically high levels of deprivation.³⁴ In 2016, 23% of the Welsh population lived in poverty, more than in England (22%), Scotland (19%), or Northern Ireland (20%).³⁵ The south of the country contains the majority of the most deprived areas in Wales,³⁶ and also has the highest opioid-related death rates in England and Wales.²⁹ However, only one comprehensive analysis of Welsh opioid prescribing has been undertaken.⁴

The aim of this study was to examine opioid prescribing trends in Wales between 2005 and 2015 using an estimated measure of daily OMEQ dose to standardise data. Analysis of OMEQ_e by deprivation quintile determined if opioid burden varied in distinct areas of socioeconomic deprivation.

Method

Data source

The study used individuals' anonymised data held in the SAIL databank, which is part of the national e-health records research infrastructure for Wales.^{37,38}

Each individual was allocated a unique anonymised linkage field (ALF) number. The ALF allowed cross-linking between different existing datasets, providing a record of all healthcare interactions for each individual whose data is available to SAIL. A dataset was produced by cross-linking individuals' anonymised records from the PCGP and Welsh Index of Multiple Deprivation (WIMD) 2011 datasets, based on the local super output areas (LSOAs) contained within the PCGP dataset.

At the time of this study, the databank contained complete data from 1 January 2005–31 December 2015 and so 11 years of available data were examined.

Opioid prescriptions

Prescriptions are automatically assigned Read codes on the electronic patient record, when issued in primary care, providing consistent identification of data.^{1,3,37,38} Read codes are a thesaurus of clinical terms used to record interactions, diagnoses, and interventions in primary care settings in Wales. A

list of Read codes was compiled for all prescribable oral and transdermal opioid medicines used for analgesia, including combination products, for example, paracetamol and codeine (co-codamol), using the NHS Information Authority's clinical terminology browser. Products licensed for the management of misuse and injectable opioids, which are reserved for palliative care, were excluded.

Only data for people aged ≥18 years between 2005 and 2015 without a recorded cancer diagnosis (identified using Read codes for cancer diagnoses or treatment) at any time between 2004 and 2015 were included in the analysis.

All data were subjected to repeated cross-sectional sampling to determine prescribing trends over the study period.

Estimated oral morphine equivalent dose

At the time of this study, dispensing data were not included within SAIL datasets. The prescribed drug product, including strength, was available from PCGP data, but not administration directions and quantity of each opioid product prescribed. Therefore, actual oral morphine equivalent dose for each individual could not be calculated.

An OMEQ_e measure was developed using data available from SAIL (**Table 1**). For each product, the recommended daily dose per day was taken from the *British National Formulary*³⁹ and electronic medicines compendium (emc).⁴⁰ The daily dose was converted to a daily OMEQ_e value, based on available conversion tables.^{8,39} Daily OMEQ_e for each product was multiplied by the number of prescriptions issued each year to determine annual totals (**Table 1**). Results were stratified by drug, with less frequently prescribed medicines (oral diamorphine, dipipanone, hydromorphone, meptazinol, methadone tablets, pentazocine, pethidine, and tapentadol) grouped as 'other' opioids.

Measuring utilisation

The number of prescriptions and number of patients per year were calculated per drug in repeat cross-sections for each year and further stratified by deprivation quintile. Data were standardised to annual population size for the SAIL databank, using data from the Office for National Statistics (ONS)⁴¹ and StatsWales.⁴² Deprivation data were adjusted by each quintile's annual population.⁴²

Data analysis

Data were extracted from the study tables within SAIL using Structured Query Language (SQL) code. Percentage change rate of number of prescriptions issued and number of people receiving prescriptions over the study period were also noted. Data were stratified into eight drug groups.

Shapiro-Wilk calculations showed data were non-parametric. Therefore, Kruskal-Wallis tests were used to examine differences in mean prescribing over the study period in the different drug groups and deprivation quintiles. Statistical analysis was conducted using IBM SPSS Statistics software (version 25.0) and figures drawn using Excel (version 16.30; retrieved from https://office.microsoft.com/excel).

Deprivation scores

The WIMD is the official measure used by Welsh Government to determine relative deprivation of areas within Wales.³⁶ The WIMD is a weighted total score of deprivation based on income (23.5%), employment (23.5%), health (14%), education (14%), geographical access of services (10%), community safety (5%), physical environment (5%), and housing (5%). Scores are not linear, so areas in group two are not twice as deprived as those in group four. Indices are published every 3 years.⁴³ The 2011 index was recommended by SAIL for use in this study, as representative of the full 11-year period. There were no significant changes in LSOA or WIMD areas in that time. Data are presented in quintiles, with WIMD1 being the most deprived areas and WIMD5 the least deprived.

Results

Prescribing data were extracted from 345 PCGPs across Wales. A total of 22 641 424 prescriptions for opioids were included in the analysis. Between 2005 and 2015, opioid prescriptions increased by 44% from 692 to 994 prescriptions per 1000 population annually. The total daily $OMEQ_{e}$, issued from all included practices in Wales, more than doubled in the 11 years examined, from 37 662 651

Table 1 Example of calculations for $OMEQ_{e}$ (mg) using 2005 data for female subjects

	Units used for calculating annualised $OMEQ_{\scriptscriptstyle\!\mathrm{e}}$				
Drug product	Recommended daily dose ^{a,b}	Oral morphine equivalent of daily dose (mg) ^{c,d}	Annual number of prescriptions	Annualised total OMEQ _e burden (mg)	
Buprenorphine					
10 mcg per hour	1 patch per week	24	28	672	
52.5 mcg per hour	1 patch twice a week	126	354	44 604	
Codeine					
Co-codamol 8/500	2 tablets 4 times a day	6.4	17 952	114 893	
Codeine phosphate 30 mg	2 tablets 4 times a day	24	16 293	391 032	
Zapain capsules (30/500)	2 tablets 4 times a day	24	112	2688	
Dihydrocodeine					
Co-dydramol 10/500	2 tablets 4 times a day	8	153 047	1 224 376	
DHC Continus 90 mg MR tablet	1 tablet twice a day	18	1009	18 612	
Remedeine tablet	2 tablets 4 times a day	16	1295	20 720	
Fentanyl					
Durogesic 100 mcg per hour patch	1 patch every 3 days	360	131	47 160	
Fentanyl 200 mcg SL lozenge	1 lozenge 4 times a day	120	40	4800	
Fentanyl 25 mcg per hour patch	1 patch every 3 days	90	3429	308 610	
Morphine					
Morphgesic SR 10 mg m/r tablet	1 tablet twice a day	20	73	730	
MXL 60 mg m/r capsule	1 capsule once a day	60	23	1380	
Oramorph 10 mg/5 ml liquid 100 ml	5 mL every 2 hours	120	573	68 760	
Sevredol 20 mg tablet	1 tablet every 6 hours	120	299	35 880	
Oxycodone					
Longtec 20 mg m/r tablets	1 tablet twice a day	80	1	80	
Oxycodone HCl 20 mg capsule	1 capsule every 4 hours	240	250	60 000	
OxyContin 80 mg m/r tablet	1 capsule twice a day	320	262	83 840	
Tramadol					
Dromadol XL 200 mg m/r tablet	1 tablet once daily	20	11	220	
Tramadol 50 mg capsule	2 capsules 4 times a day	40	93 918	3 756 720	
Tramacet 325 mg/37.5 mg	2 tablets 4 times a day	30	4450	133 500	
Other					
Co-proxamol 32.5 mg/325 mg tablet	2 tablets 4 times a day	26	82 015	2 132 390	
Hydromorphone HCl 1.3 mg capsule	1 capsule every 4 hours	58.5	6	351	
Pethidine HCl 50 mg tablet	1 tablet every 4 hours	30	2381	71 430	

Annualised total = estimated oral morphine equivalent of daily dose x annual number of prescriptions. Process repeated for each drug product and totalled for each year. $a^{39 b40 c8} d^{64} OMEQ_e$ = estimated oral morphine equivalent.

		Oral or trans	dermal opioids	
	Total daily OMEQe (mg) dose prescribed	Annual total daily OMEQe dose (mg) per 1000 population	OMEQe dose (mg) per prescription issued	Annual number of prescriptions issued per 1000 population
Buprenorphine	23 641 528			
2005	977 464	422	98	4
2015	2 756 458	1142	37	31
Rate change, %	182	170.5	-61.7	606.3
Codeine	223 817 156			
2005	13 743 115	5916	17	357
2015	25 593 382	10 581	19	549
Rate change, %	86.2	78.8	16.2	53.9
Dihydrocodeine	44 600 874			
2005	4 368 806	1887	12	154
2015	3 471 460	1438	13	109
Rate change, %	-20.5	-23.8	7.7	-29.2
Fentanyl	64 138 905			
2005	2 695 290	1164	186	6
2015	6 496 270	2691	147	18
Rate change, %	141.0	131.2	-21.2	193.2
Morphine	91 132 530			
2005	3 293 220	1422	86	17
2015	17 047 800	7063	68	104
Rate change, %	417.7	396.6	-20.6	525.6
Oxycodone	45 120 680			
2005	1 316 480	569	105	5
2015	6 165 400	2554	100	26
Rate change, %	368.3	349.3	-4.9	372.4
Tramadol	144 173 635			
2005	7 865 695	3397	36	95
2015	14 252 335	5905	38	156
Rate change, %	81.2	73.8	5.7	64.4
Other	8 888 696			
2005	3 446 735	1719	27	56
2015	699 711	347	58	5
Rate change, %	-79.7	-79.8	117.4	-91.0

Table 2 Daily OMEQ $_{\!\!e}$ (mg) issued on prescription, given as annual totals and adjusted to population, stratified by drug

Results are rounded to nearest whole number. Rate change (%) calculated using original, unrounded data. Original data are available from the authors on request.





Figure 1 Comparison of the percentage contribution of each opioid prescribed by total prescriptions issued and total daily OMEQ_e dose (mg) in Wales between 2005 and 2015

mg to 76 428 768 mg. When adjusted to population, annualised daily OMEQ_e per 1000 population increased by 95% (from 16 266 mg to 31 665 mg) over the study period (**Table 1**).

Total estimated oral morphine equivalent prescribed

Codeine was the most commonly prescribed opioid (**Table 2**), with just under 12.5 million prescriptions issued and the highest annual total OMEQ_e prescribed for the study duration (*Figure 1*). Codeine OMEQ_e per 1000 population increased by 79%, from 5916 mg to 10 581 mg. Tramadol was the second most commonly prescribed opioid in Wales with a 74% increase, from 3397 mg to 5905 mg OMEQ_e per 1000 population, although annual total OMEQ_e started to reduce from 2014 (*Figure 2*).

Large increases were noted in 'strong' opioids (morphine, oxycodone, fentanyl, and buprenorphine) during the study (*Figure 2*). Morphine $OMEQ_e$ increased by 397%, from 1422 mg to 7063 mg per 1000 population (*Table 2*). By 2015, morphine was prescribed at three times the equivalent dose of either oxycodone (increased 349%, from 569 mg to 2554 mg per 1000 population) or fentanyl (increased 131%, from 1164 mg to 2691 mg per 1000 population).

Overall, 71% of the total opioid burden in the areas of Wales covered by the SAIL databank was accounted for by three drugs: codeine (35%), tramadol (22%), and morphine (14%). Statistically significant differences were found between the 11-year total OMEQ_e when each drug group was compared with the others (P<0.001, H = 73.5, η^2 = 0.8).



Figure 2 Trends in opioid prescribing across Wales, 2005–2015. Annual daily OMEQ. in mg per 1000 population, stratified by drug

Opioid prescribing trends by deprivation

Figure 3 illustrates the trends in annualised daily OMEQ_e of all oral and transdermal opioids stratified by the WIMD (2011). Over the study, people in the most deprived quintiles (WIMD1) were prescribed an estimated 100 711 696 mg more OMEQ_e than in the least deprived (WIMD5) (*Table 3*).

Between 2005 and 2015, OMEQ_e doubled in all but the least deprived (WIMD5) areas (**Table 3**). Twenty-eight per cent (176 824 265 mg of 622 969 068 mg) of total OMEQ_e was issued in the most deprived areas of Wales. In contrast, 12% (76 112 569 mg) were prescribed in the least deprived areas. Throughout the study, OMEQ_e prescribed in WIMD1 areas remained more than twice those noted in WIMD5 areas (**Table 3**) for both total OMEQ_e (mg) and OMEQ_e per 1000 population. Despite large percentage increases in all quintiles, the difference between total OMEQ_e prescribed per quintile were statistically significant (*P*<0.001, H = 34.5, $\eta^2 = 0.61$).

Discussion

Summary

This study identified trends in opioid prescribing in Wales, similar to those previously reported in other parts of the UK.^{1,3,6,26,27,44} A marked increase in opioid burden in Wales between 2005 and 2015 was noted. Using the OMEQ_e measure described, opioid burden in the study population nearly doubled in 11 years. Increasing deprivation was associated with higher OMEQ_e and, consequently, a higher burden per person, despite rises in percentage terms being similar in all WIMD 2011 quintiles.





Figure 3 Trends in opioid prescribing across Wales, 2005–2015. Annual daily OMQE_e (mg) per 1000 population, stratified by deprivation. Welsh Index of Multiple Deprivation 2011 (WIMD 2011), where WIMD1 = most deprived, WIMD5 = least deprived.

Strengths and limitations

Large sets of prescribing and diagnostic data have been validated as an accurate means for conducting healthcare population research,^{45,46} as they reduce recall bias and regional variation. In this study, anyone registered with included practices and prescribed an opioid medicine were included in the analysis, avoiding selection bias. This is the first study of Welsh data to utilise OMEQ_e to better understand the burden of opioid prescribing on the population. Using linkage systems within SAIL datasets, data from people with a recorded cancer diagnosis could be excluded from analysis. The data confidently reflects prescribing for non-cancer pain, unlike other recent studies that assumed the majority of prescribing was attributable to persisting, non-cancer pain based on longevity of prescribing and dose forms used.^{2,6}

Other studies have suggested large increases in prescribing are attributable to a range of drugs.^{2,6,47} The current study showed that three drugs were responsible for the majority of prescribing. This may, in part, be owing to the effective use of National Prescribing Indicators, which, in particular, have encouraged morphine to be used as first-line 'strong' opioid.⁴⁸

Prescribing data provide an indication of intention to treat but does not confirm consumption. It also does not indicate the diagnosis or how long an individual might have been using the medication. Moreover, data presented here did not identify people receiving more than one opioid medicine and, so, would have higher individual OMEQ_e burdens.

Table 3 Trends in $OMEQ_e$ (mg) prescribing stratified by deprivation

	Oral or transdermal opioids		
Deprivation quintile	Total daily OMEQ _e dose (mg) prescribed	Total daily OMEQ _e dose (mg) per 1000 population	
WIMD1			
2005	10 319 636	21 757	
2015	21 167 919	43 176	
Rate change, %	105.1	98.4	
Total prescribed ^a	176 824 265		
WIMD2			
2005	8 590 375	18 203	
2015	17 399 026	35 475	
Rate change, %	102.5	94.9	
Total prescribed ^a	146 459 878		
WIMD3			
2005	7 684 060	17 108	
2015	15 342 942	32 564	
Rate change, %	99.7	90.3	
Total prescribed ^a	129 880 669		
WIMD4			
2005	5 374 595	12 242	
2015	10 878 897	23 534	
Rate change, %	102.4	92.2	
Total prescribed ^a	93 691 687		
WIMD5			
2005	4 486 035	9 381	
2015	8 721 170	17 557	
Rate change, %	94.4	87.2	
Total prescribed ^a	76 112 569		

^aTotal prescribed 2005–2015. Annual OMEQ_e calculated as per method and stratified by deprivation quintile (Welsh Index of Multiple Deprivation [WIMD2011], where WIMD1 = most deprived, WIMD5 = least deprived). OMEQ_e = estimated oral morphine equivalent.

It was not possible to access dispensing data, which provides details required to accurately calculate OMEQ. The authors' estimated measure (OMEQ_e) required assumptions to be made in regard of daily dose prescribed. Also, quantity could not be verified in order to calculate duration of use. However, the trends are similar to those reported elsewhere in the UK.^{1–3,6,27,44}

Further analysis is required to determine an individual's daily intake, where multiple opioids and strengths of products are prescribed. While prescription numbers have started to stabilise or reduce since the end of the study,^{6,48,49} concerns remain about the number of people receiving supramaximal opioid doses and lengthy durations of use.^{11,32}

In the study, opioid medicines were identified by Read codes and accuracy of data extraction depended on the inclusivity of the coding used. Similar rationales for deciding which opioid products to include in analysis of primary care prescribing have been adopted by other UK-based authors.^{1,3,6,27} However, incomplete coding lists could result in an under-representation of prescribing.

Comparison with existing literature

Examining trends by prescription numbers alone is likely to underestimate the opioid burden within a population. Using English data, Curtis *et al* demonstrated a 34% growth in prescription numbers equated to a 127% increase in OMEQ burden between 1998 and 2016.⁶ In the present study, a 44% increase in prescription numbers in Wales, translated into a 95% increase in opioid burden using the OMEQ_e measure described.

Another measure of prescribing is defined daily doses (DDD), devised by the World Health Organization:⁵⁰ DDDs are 'the assumed average maintenance dose per day for a drug used for its main indication in adults.' However, DDDs vary for each drug and between formulations of the same drug.⁵⁰ When OMEQ was used to compare prescribing in four Nordic countries, it demonstrated noteworthy differences in patterns of opioid consumption compared with those seen with DDDs.⁴³ 'Weak' opioids, such as codeine, carry higher DDD values than 'strong' opioids like morphine. Countries where codeine predominated, appeared to have high overall opioid prescribing, which was reversed when OMEQ was used and the contribution of 'strong' opioids accounted for.⁴³

Prescribers' understanding of OMEQ is poor.⁵¹⁻⁵³ Use of OMEQ as a measure of prescribing might improve comprehension of opioid equivalence and lead to safer prescribing.

Substantial increases in opioid prescribing, with higher levels in more deprived populations, were also reported in other parts of the UK^{2,27} and internationally.^{54–57} Increased levels of prescribing in areas of high socioeconomic deprivation has been linked to greater reported pain intensity.²⁶ However, limited evidence supports the notion that opioids are effective at reducing pain, particularly in the longer term.^{8,58,59} High-dose opioids (above 120 mg OMEQ) have been associated with increased levels of pain.^{60,61} In the context of this and previous studies,^{2,6,26,27} the implications of increased opioid prescribing in more deprived areas are concerning. It exposes the most vulnerable people to higher levels of medicines, which may be ineffective at best, and could cause additional health and wellbeing complications.¹¹

Implications for practice

OMEQ is a useful measure of opioid utilisation in the general population and an individual basis.²⁵ This study has demonstrated differences between assumed burden of opioid prescribing using OMEQ_e and prescriptions issued, which might have important clinical implications. Evaluating opioid prescribing using OMEQ would provide easily comparable data that better reflects clinical practice. Reasons for disparities in opioid burden between areas of deprivation need further investigation. Lack of availability and acceptability of non-pharmacological management and services have been suggested among reasons why prescribing is favoured.^{62,63} Use of OMEQ as a measure of opioid burden should be considered as a means of identifying 'at risk' populations and individuals, as prescription numbers reduce.

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Ethical approval

This research was approved by the Information Governance Review Panel (IGRP) of the SAIL databank, based in Swansea University (SAIL identification number: 0507)

Provenance

Freely submitted; externally peer reviewed.

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References

- Zin CS, Chen LC, Knaggs RD. Changes in trends and pattern of strong opioid prescribing in primary care. Eur J Pain 2014; 18(9): 1343–1351. DOI: https://doi.org/10.1002/j.1532-2149.2014.496.x
- Mordecai L, Reynolds C, Donaldson LJ, de C Williams AC. Patterns of regional variation of opioid prescribing in primary care in England: a retrospective observational study. Br J Gen Pract 2018; 68(668): e225–e233. DOI: https://doi.org/10.3399/bjgp18X695057
- Ruscitto A, Smith BH, Guthrie B. Changes in opioid and other analgesic use 1995-2010: repeated cross-sectional analysis of dispensed prescribing for a large geographical population in Scotland. *Eur J Pain* 2015; 19(1): 59–66. DOI: https://doi.org/10.1002/ejp.520
- Davies E, Phillips C, Rance J, Sewell B. Examining patterns in opioid prescribing for non-cancer-related pain in Wales: preliminary data from a retrospective cross-sectional study using large datasets. *Br J Pain* 2019; **13(3)**: 145–158. DOI: https://doi.org/10.1177/2049463718800737
- Bedson J, Chen Y, Hayward RA, et al. Trends in long-term opioid prescribing in primary care patients with musculoskeletal conditions: an observational database study. Pain 2016; 157(7): 1525–1531. DOI: https://doi.org/ 10.1097/j.pain.00000000000557
- Curtis HJ, Croker R, Walker AJ, et al. Opioid prescribing trends and geographical variation in England, 1998-2018: a retrospective database study. Lancet Psychiatry 2019; 6(2): 140–150. DOI: https://doi.org/10.1016/S2215-0366(18)30471-1
- Stannard C. Where now for opioids in chronic pain? Drug Ther Bull 2018; 56(10): 118–122. DOI: https://doi.org/10. 1136/dtb.2018.10.000007
- Faculty of Pain Management, Royal College of Anaesthetists. Opioids aware: a resource for patients and healthcare
 professionals to support prescribing of opioid medicines for pain; https://www.fpm.ac.uk/opioids-aware (accessed
 26 Nov 2020).
- Rosenblum A, Marsch LA, Joseph H, Portenoy RK. Opioids and the treatment of chronic pain: controversies, current status, and future directions. *Exp Clin Psychopharmacol* 2008; 16(5): 405–416. DOI: https://doi.org/10. 1037/a0013628
- Ballantyne JC, Shin NS. Efficacy of opioids for chronic pain. Clin J Pain 2008; 24(6): 469–478. DOI: https://doi.org/ 10.1097/AJP.0b013e31816b2f26
- Els C, Jackson TD, Kunyk D, et al. Adverse events associated with medium- and long-term use of opioids for chronic non-cancer pain: an overview of Cochrane reviews. *Cochrane Database Syst Rev* 2017; **10(10)**: CD012509. DOI: https://doi.org/10.1002/14651858.CD012509.pub2
- Lin JC, Chu LF, Stringer EA, et al. One month of oral morphine decreases gray matter volume in the right amygdala of individuals with low back pain: confirmation of previously reported magnetic resonance imaging results. *Pain Med* 2016; **17**(8): 1497–1504. DOI: https://doi.org/10.1093/pm/pnv047
- Scherrer JF, Salas J, Sullivan MD, et al. The influence of prescription opioid use duration and dose on development of treatment resistant depression. Prev Med 2016; 91: 110–116. DOI: https://doi.org/10.1016/j.ypmed.2016.08.003
- 14. Gomes T, Mamdani MM, Dhalla IA, *et al*. Opioid dose and drug-related mortality in patients with nonmalignant pain. *Arch Intern Med* 2011; **171(7)**: 686–691. DOI: https://doi.org/10.1001/archinternmed.2011.117
- Merrill JO, Von Korff M, Banta-Green CJ, et al. Prescribed opioid difficulties, depression and opioid dose among chronic opioid therapy patients. *Gen Hosp Psychiatry* 2012; **34(6)**: 581–587. DOI: https://doi.org/10.1016/j. genhosppsych.2012.06.018
- 16. Scherrer JF, Salas J, Copeland LA, et al. Prescription opioid duration, dose, and increased risk of depression in 3 large patient populations. Ann Fam Med 2016; **14(1)**: 54–62. DOI: https://doi.org/10.1370/afm.1885
- Morasco BJ, Yarborough BJ, Smith NX, et al. Higher prescription opioid dose is associated with worse patientreported pain outcomes and more health care utilization. J Pain 2017; 18(4): 437–445. DOI: https://doi.org/10. 1016/j.jpain.2016.12.004
- Salas J, Scherrer JF, Schneider FD, et al. New-onset depression following stable, slow, and rapid rate of prescription opioid dose escalation. Pain 2017; 158(2): 306–312. DOI: https://doi.org/10.1097/j.pain. 000000000000763
- Fischer B, Murphy Y, Kurdyak P, Goldner EM. Depression a major but neglected consequence contributing to the health toll from prescription opioids? *Psychiatry Res* 2016; **243**: 331–334. DOI: https://doi.org/10.1016/j. psychres.2016.06.053
- Mazereeuw G, Sullivan MD, Juurlink DN. Depression in chronic pain: might opioids be responsible? Pain 2018; 159(11): 2142–2145. DOI: https://doi.org/10.1097/j.pain.00000000001305
- Ballantyne JC, LaForge KS. Opioid dependence and addiction during opioid treatment of chronic pain. Pain 2007; 129(3): 235–255. DOI: https://doi.org/10.1016/j.pain.2007.03.028
- Minozzi S, Amato L, Davoli M. Development of dependence following treatment with opioid analgesics for pain relief: a systematic review. Addiction 2013; 108(4): 688–698. DOI: https://doi.org/10.1111/j.1360-0443.2012. 04005.x
- Campbell G, Nielsen S, Larance B, et al. Pharmaceutical opioid use and dependence among people living with chronic pain: associations observed within the pain and opioids in treatment (point) cohort. Pain Med 2015; 16(9): 1745–1758. DOI: https://doi.org/10.1111/pme.12773
- Degenhardt L, Lintzeris N, Campbell G, et al. Experience of adjunctive cannabis use for chronic non-cancer pain: findings from the pain and opioids in treatment (point) study. Drug Alcohol Depend 2015; 147: 144–150. DOI: https://doi.org/10.1016/j.drugalcdep.2014.11.031

- Nielsen S, Gisev N, Bruno R, et al. Defined daily doses (DDD) do not accurately reflect opioid doses used in contemporary chronic pain treatment. *Pharmacoepidemiol Drug Saf* 2017; 26(5): 587–591. DOI: https://doi.org/10. 1002/pds.4168
- Todd A, Akhter N, Cairns J-M, et al. The pain divide: a cross-sectional analysis of chronic pain prevalence, pain intensity and opioid utilisation in England. BMJ Open 2018; 8(7): e023391. DOI: https://doi.org/10.1136/bmjopen-2018-023391
- Chen T-C, Chen L-C, Kerry M, Knaggs RD. Prescription opioids: Regional variation and socioeconomic status evidence from primary care in England. Int J Drug Policy 2019; 64: 87–94. DOI: https://doi.org/10.1016/j.drugpo. 2018.10.013
- Marmot M, Bell R. Fair society, healthy lives. Public Health 2012; 126(Supp 1): S4–S10. DOI: https://doi.org/10. 1016/j.puhe.2012.05.014
- Office for National Statistics. Deaths related to drug poisoning in England and Wales: 2018 registrations. 2019; https://www.ons.gov.uk/peoplepopulationandcommunity/birthsdeathsandmarriages/deaths/bulletins/deathsrelate dtodrugpoisoninginenglandandwales/2018registrations (accessed 26 Nov 2020).
- Keyes KM, Cerdá M, Brady JE, et al. Understanding the rural-urban differences in nonmedical prescription opioid use and abuse in the United States. Am J Public Health 2014; 104(2): e52–e59. DOI: https://doi.org/10.2105/AJPH. 2013.301709
- Stannard CF. Pain and pain prescribing: what is in a number? Br J Anaesth 2018; 120(6): 1147–1149. DOI: https:// doi.org/10.1016/j.bja.2018.03.002
- Taylor S, Annand F, Burkinshaw P, et al. Dependence and withdrawal associated with some prescribed medicines: an evidence review. 2019; https://assets.publishing.service.gov.uk/government/uploads/system/uploads/ attachment_data/file/829777/PHE_PMR_report.pdf (accessed 26 Nov 2020).
- Torrance N, Veluchamy A, Zhou Y, et al. Trends in gabapentinoid prescribing, co-prescribing of opioids and benzodiazepines, and associated deaths in Scotland. Br J Anaesth 2020; 125(2): 159–167. DOI: https://doi.org/10. 1016/j.bja.2020.05.017
- Edwards RT, Charles JM, Thomas S, et al. A national programme budgeting and marginal analysis (PBMA) of health improvement spending across Wales: disinvestment and reinvestment across the life course. BMC Public Health 2014; 14(1): 837. DOI: https://doi.org/10.1186/1471-2458-14-837
- 35. Barnard H. Poverty in Wales 2018. 2018; https://www.jrf.org.uk/report/poverty-wales-2018 (accessed 26 Nov 2020).
- Welsh Government. Welsh index of multiple deprivation: 2011. 2011; https://gov.wales/welsh-index-multipledeprivation-full-index-update-ranks-2011 (accessed 26 Nov 2020).
- Lyons RA, Jones KH, John G, et al. The Sail databank: linking multiple health and social care datasets. BMC Med Inform Decis Mak 2009; 9(1): 3. DOI: https://doi.org/10.1186/1472-6947-9-3
- Ford DV, Jones KH, Verplancke J-P, et al. The Sail Databank: building a national architecture for e-health research and evaluation. BMC Health Serv Res 2009; 9: 157. DOI: https://doi.org/10.1186/1472-6963-9-157
- Joint Formulary Committee. British National Formulary. 2020; https://about.medicinescomplete.com/publication/ british-national-formulary (accessed 26 Nov 2020).
- 40. Datapharm. Electronic medicines compendium. Up to date, approved, regulated prescribing and patient information for licensed medicines. 2020; https://www.medicines.org.uk/emc (accessed 26 Nov 2020).
- 41. Office for National Statistics. 2011 census analysis comparing rural and urban areas of England and Wales. London: ONS; 2013.
- StatsWales. Population estimates by local authority and year. 2020; https://statswales.gov.wales/Catalogue/ Population-and-Migration/Population/Estimates/Local-Authority/populationestimates-by-localauthority-year (accessed 26 Nov 2020).
- Svendsen K, Borchgrevink P, Fredheim O, et al. Choosing the unit of measurement counts: the use of oral morphine equivalents in studies of opioid consumption is a useful addition to defined daily doses. Palliat Med 2011; 25(7): 725–732. DOI: https://doi.org/10.1177/0269216311398300
- Foy R, Leaman B, McCrorie C, et al. Prescribed opioids in primary care: cross-sectional and longitudinal analyses of influence of patient and practice characteristics. BMJ Open 2016; 6(5): e010276. DOI: https://doi.org/10.1136/ bmjopen-2015-010276
- 45. Williams T, van Staa T, Puri S, Eaton S. Recent advances in the utility and use of the general practice research database as an example of a UK primary care data resource. *Ther Adv Drug Saf* 2012; **3**(2): 89–99. DOI: https://doi. org/10.1177/2042098611435911
- 46. Jones KH, Ford DV, Jones C, et al. A case study of the Secure Anonymous Information Linkage (SAIL) gateway: a privacy-protecting remote access system for health-related research and evaluation. J Biomed Inform 2014; 50: 196–204. DOI: https://doi.org/10.1016/j.jbi.2014.01.003
- Karanges EA, Blanch B, Buckley NA, Pearson S-A. Twenty-Five years of prescription opioid use in Australia: a whole-of-population analysis using pharmaceutical claims. Br J Clin Pharmacol 2016; 82(1): 255–267. DOI: https:// doi.org/10.1111/bcp.12937
- All Wales Medicines Strategy Group. National prescribing indicators 2019–2020. 2019; https://awmsg.nhs.wales/ files/national-prescribing-indicators/national-prescribing-indicators-2019-2020 (accessed 26 Nov 2020).
- 49. OpenPrescribing.net. Opioid analgesics. 2020; https://openprescribing.net/bnf/040702 (accessed 26 Nov 2020).
- World Health Organization Collaborating Centre for Drug Statistics Methodology. Guidelines for ATC classification and DDD assignment: 2020. 2019; https://www.whocc.no/filearchive/publications/2020_guidelines_web.pdf (accessed 26 Nov 2020).
- Rennick A, Atkinson T, Cimino NM, et al. Variability in opioid equivalence calculations. Pain Med 2016; 17(5): 892–898. DOI: https://doi.org/10.1111/pme.12920

- 52. Shaheen PE, Walsh D, Lasheen W, *et al.* Opioid equianalgesic tables: are they all equally dangerous? *J Pain* Symptom Manage 2009; **38**(**3**): 409–417. DOI: https://doi.org/10.1016/j.jpainsymman.2009.06.004
- 53. Institute for Safe Medication Practices Canada. Sink or swim? Helping patients and practitioners to understand opioid potencies and overdose risk. *ISMP Canada Safety Bulletin* 2017; **17**(8): 1–6.
- 54. Joynt M, Train MK, Robbins BW, et al. The impact of neighborhood socioeconomic status and race on the prescribing of opioids in emergency departments throughout the United States. J Gen Intern Med 2013; 28(12): 1604–1610. DOI: https://doi.org/10.1007/s11606-013-2516-z
- Wagemaakers FN, Hollingworth SA, Kreijkamp-Kaspers S, et al. Opioid analgesic use in Australia and the Netherlands: a cross-country comparison. Int J Clin Pharm 2017; 39(4): 874–880. DOI: https://doi.org/10.1007/ s11096-017-0492-9
- 56. Kapoor S, Thorn BE. Healthcare use and prescription of opioids in rural residents with pain. *Rural Remote Health* 2014; **14**(**3**): 2879.
- 57. Prunuske JP, St Hill CA, Hager KD, et al. Opioid prescribing patterns for non-malignant chronic pain for rural versus non-rural US adults: a population-based study using 2010 NAMCS data. BMC Health Serv Res 2014; 14(1): 563. DOI: https://doi.org/10.1186/s12913-014-0563-8
- Furlan AD, Sandoval JA, Mailis-Gagnon A, Tunks E. Opioids for chronic noncancer pain: a meta-analysis of effectiveness and side effects. CMAJ 2006; 174(11): 1589–1594. DOI: https://doi.org/10.1503/cmaj.051528
- Chaparro LE, Furlan AD, Deshpande A, et al. Opioids compared with placebo or other treatments for chronic low back pain: an update of the Cochrane review. Spine 2014; 39(7): 556–563. DOI: https://doi.org/10.1097/BRS. 00000000000249
- Cohen SP, Christo PJ, Wang S, et al. The effect of opioid dose and treatment duration on the perception of a painful standardized clinical stimulus. Reg Anesth Pain Med 2008; 33(3): 199–206. DOI: https://doi.org/10.1097/ 00115550-200805000-00002
- Tompkins DA, Campbell CM. Opioid-induced hyperalgesia: clinically relevant or extraneous research phenomenon? Curr Pain Headache Rep 2011; 15(2): 129–136. DOI: https://doi.org/10.1007/s11916-010-0171-1
- 62. Finestone HM, Juurlink DN, Power B, et al. Opioid prescribing is a surrogate for inadequate pain management resources. Can Fam Physician 2016; **62(6)**: 465–468.
- McCrorie C, Closs SJ, House A, et al. Understanding long-term opioid prescribing for non-cancer pain in primary care: a qualitative study. BMC Fam Pract 2015; 16(1): 121. DOI: https://doi.org/10.1186/s12875-015-0335-5
- 64. Pain Management Centre, Oxford University Hospitals NHS Foundation Trust. Opioid calculator for calculation of oral morphine equivalent daily dose (MED) in mg/day. 2020; https://www.ouh.nhs.uk/services/referrals/pain/ documents/opioid-calculator.xlsx (accessed 26 Nov 2020).