RESEARCH ARTICLE

Healthcare use by people who use illicit opioids (HUPIO): development of a cohort based on electronic primary care records in England [version 2; peer review: 2 approved]

Dan Lewer1,2*, Prianka Padmanathan3*, Muhammad Qummer ul Arfeen1, Spiros Denaxas1, Harriet Forbes3, Arturo Gonzalez-Izquierdo1, Matt Hickman3,4

1Institute of Health Informatics, University College London, London, NW1 2DA, UK
2Institute of Epidemiology and Healthcare, University College London, London, WC1E 7HB, UK
3Population Health Sciences, Bristol Medical School, University of Bristol, Bristol, BS8 1UD, UK
4National Institute of Health Research Biomedical Research Centre, Bristol NHS Foundation Trust and the University of Bristol, Bristol, UK

* Equal contributors

Abstract

Background: People who use illicit opioids such as heroin have substantial health needs, but there are few longitudinal studies of general health and healthcare in this population. Most research to date has focused on a narrow set of outcomes, including overdoses and HIV or hepatitis infections. We developed and validated a cohort using UK primary care electronic health records (Clinical Practice Research Datalink GOLD and AURUM databases) to facilitate research into healthcare use by people who use illicit opioid use (HUPIO).

Methods: Participants are patients in England with primary care records indicating a history of illicit opioid use. We identified codes including prescriptions of opioid agonist therapies (methadone and buprenorphine) and clinical observations such as ‘heroin dependence’. We constructed a cohort of patients with at least one of these codes and aged 18-64 at cohort entry, with follow-up between January 1997 and March 2020. We validated the cohort by comparing patient characteristics and mortality rates to other cohorts of people who use illicit opioid use (HUPIO).

Results: Up to March 2020, the HUPIO cohort included 138,761 patients with a history of illicit opioid use. Demographic characteristics and all-cause mortality were similar to existing cohorts: 69% were male; the median age at index for patients in CPRD AURUM (the database with more included participants) was 35.3 (interquartile range 29.1-42.6); the average age of new cohort entrants increased...
over time; 76% had records indicating current tobacco smoking; patients disproportionately lived in deprived neighbourhoods; and all-cause mortality risk was 6.6 (95% CI 6.5-6.7) times the general population of England.

Conclusions: Primary care data offer new opportunities to study holistic health outcomes and healthcare of this population. The large sample enables investigation of rare outcomes, whilst the availability of linkage to external datasets allows investigation of hospital use, cancer treatment, and mortality.

Keywords
substance use disorders, drug dependence, opioid agonist therapy, illicit drugs, electronic health records
The main treatment for dependence on illicit opioids is opioid agonist therapy (OAT); a pharmacological treatment of long-acting opioids such as methadone or buprenorphine. In England, OAT is provided by specialist drug and alcohol services or GPs, depending on local commissioning arrangements. A large body of evidence demonstrates the effectiveness of OAT across outcomes including mortality, physical and mental health, and criminal activity\textsuperscript{\textemdash}15.

Despite extensive evidence regarding the risks of illicit opioids and the benefits of OAT, there are important unanswered research questions. Epidemiological research and health interventions have focused on outcomes perceived to be ‘drug-related’, such as overdoses and HIV or hepatitis infections. Meanwhile, there is limited research into engagement with primary care services, healthcare quality, and treatment options for non-communicable diseases and mental health problems\textsuperscript{\textemdash}14. These are important areas of research because the population of people who use illicit opioids in England (as in many other countries) is ageing\textsuperscript{\textemdash}15 and the majority of excess deaths are now caused by non-communicable diseases such as liver disease, chronic obstructive pulmonary disease, and cardiovascular disease\textsuperscript{\textemdash}16.

To facilitate research in these areas, we aimed to develop and validate a phenotype that identifies people with a history of illicit opioid use in longitudinal UK primary care electronic health records (Clinical Practice Research Datalink GOLD and AURUM databases).

Methods
Study design
We developed and validated a cohort of people with a history of illicit opioid use, including cross-sectional analysis of participant characteristics at baseline, and a cohort analysis of all-cause mortality rates.

Data sources
The Clinical Practice Research Datalink (CPRD) GOLD and AURUM are databases of anonymised electronic health records from primary care, including approximately 8% and 13% of the populations in the UK and England respectively\textsuperscript{\textemdash}17\textendash}19. Although the databases include similar clinical information, they differ in terms of data collection software, clinical classification system and geographical coverage. CPRD GOLD includes data from GP practices throughout the UK, while CPRD AURUM initially included England only, and more recently practices in Northern Ireland have been added. To maximise comparability we have restricted the cohort to patients registered in England, though the methods can be used to include patients in other parts of the UK.

Entry and exit dates
We selected patients who were registered at participating GP practices between 1 January 1997 and 31 December 2018 for GOLD, and between 1 January 1997 and 31 March 2020 for AURUM. Cohort entry was defined as the latest of 1 January 1997, the first date when good quality data were available for that patient, and the date of the first code indicating illicit opioid use. Cohort exit was the earliest of the date when the patient stopped being observed (‘last collection date’) or participating in CPRD (the patient transferred out of a
participating GP practice), or death as recorded in their primary care record. In addition to these criteria, we excluded patients who were aged under 18 or 65 or older at cohort entry.

**Selection of patients with a history of illicit opioid use**

We focused on patients with a history of opioid use (rather than specifically current use) due to the typically long duration of opioid use\(^{20-22}\) and the likelihood that patients would not have regularly recorded opioid use. We therefore included patients with illicit opioid use recorded prior to the cohort entry date.

CPRD data include two main types of codes: product codes and clinical codes. Product codes indicate a prescription made in a primary care setting, whilst clinical codes indicate a diagnosis or other clinical observation (sometimes also a prescription). We selected patients by identifying product codes indicating a prescription of OAT and clinical codes indicating a history of illicit opioid use, such as ‘heroin dependence’ (see extended data for a full code list\(^{23}\)). We prioritised specificity over sensitivity, aiming to use codes that are only applied to the target population. Our process for selecting codes is summarised in Figure 1.

**Product codes.** In the UK, treatment for opioid dependence involves the prescription of methadone or buprenorphine\(^{24}\). However, these medications are also licensed for other indications including pain and palliative cough\(^{25,26}\). We therefore developed a method to identify medications that are specific to OAT.

We searched CPRD dictionaries to identify all methadone and buprenorphine product codes (full search terms are available as extended data\(^{23}\)). We found 175 codes in CPRD GOLD and 136 codes in CPRD AURUM. We compared these lists to an existing list of OAT medicines\(^{27}\) and found no additional codes. We then followed a two-step process to identify products specific to OAT. First, we described the age- and sex-distribution of patients at the time of the first prescription. This showed two distinct groups: drugs mainly prescribed to younger men, and drugs mainly prescribed to older women (see extended data\(^{23}\)). Data from specialist drug treatment services shows that the population receiving OAT is three-quarters male and predominantly aged 18–64\(^{28}\). In contrast, the population prescribed opioids for pain relief is mainly older and female\(^{29}\). We therefore excluded medications where more than half of patients were female, the lower quartile of age was younger than 18 years, or the upper quartile of age was older than 64, as these codes are unlikely to relate specifically to OAT. The majority of codes excluded were transdermal buprenorphine patches, which are not indicated for OAT\(^{26}\). Second, a prescribing professional working in a community drug and alcohol service reviewed remaining products to check they are used for OAT.

**Clinical codes.** CPRD GOLD and AURUM differ in the clinical coding system used. CPRD GOLD uses Read codes whilst AURUM uses SNOMED codes. We used keywords to search CPRD dictionaries to find Read and SNOMED clinical codes that may indicate illicit opioid use (methadone; buprenorphine; abus*; addict; dependen*; drug user; heroin; inject; misus*; opiate; opioid; overdose). Our search identified 1,098 Read codes and 1,800 SNOMED codes. Two authors (DL and PP) screened the codes for relevance, with conflicts resolved through discussion.

---

**Figure 1.** Selection of codes that indicate a history of illicit opioid use.
Where codes were likely to indicate illicit opioid use, but did not specifically mention opioids, we classified them as ‘probable’. For example, codes indicating injection of illicit drugs were classified as ‘probable’ because an estimated 94% of people who inject drugs in the UK use heroin\(^9\). These codes have been excluded from our analyses to maximise specificity, but can be included in future research if greater sensitivity is needed.

Some clinical codes described prescriptions, tests or adverse reactions relating to methadone and buprenorphine. We excluded these where the indication was unclear.

After agreeing a list of codes, we checked the age- and sex-distribution of patients with these codes in the same way as we did with the product codes. A small number of codes were either prescribed to a majority of female patients or had an upper quartile of age older than 64. All of these codes represented dependence on medications prescribed for analgesia (for example ‘misuse of Codeine tablets’), which we classified as ‘probable’ and excluded from our analysis.

**External validation**

We validated the HUPIO cohort by comparing it to other samples of people who use opioids. We anticipated the following characteristics: (a) the average age of patients entering the cohort would increase over time, as the cohort of people who use illicit opioids in England is ageing\(^8\); (b) high prevalence of smoking, with a systematic review finding an average of 84% of people enrolled in addiction services currently smoke\(^9\); and 70% of patients starting treatment for opioid dependence in England in 2018 recorded as current tobacco smokers\(^10\). We reported the prevalence of current- and ex-smoking based on existing codelists for smoking histories\(^8\); (c) disproportionate representation of patients living in more deprived areas, as illicit opioid use and opioid-related deaths are consistently associated with deprivation\(^8\); (d) higher mortality rates than the general population, as studies of mortality in this population consistently show very high mortality rates\(^6\). We compared the standardised mortality ratios (SMR) for our cohort to those reported in existing studies of all-cause mortality in this population in England, identified by a brief literature search using Pubmed using the terms (opiate OR opioid OR heroin) AND (mortality OR death) without restrictions on language or publication date.

In addition to these characteristics, we reported the proportion of patients with recorded histories of homelessness, prison, and alcohol dependence, based on existing phenotypes and searches of clinical codes. We expected these experiences to be common among people with a history of illicit opioid use\(^6\). However, we did not know how consistently these experiences would be recorded in primary care, and therefore did not use these variables for validation purposes.

**Internal validation**

We used hospital admissions where the patient had a diagnosis of ‘mental and behavioural disorders due to use of opioids’ (ICD10 code F11) to test the sensitivity of the primary care codelist. Among all patients in CPRD with linked Hospital Episode Statistics data, we requested dates of hospital admissions where F11 was recorded in any diagnostic position. We then reported the proportion of these patients who had a primary care code indicating a history of illicit opioid use based on the HUPIO codelist, and whether the primary care code was before the first hospital admission, in the 30 days after the first admission, or more than 30 days after admission. We compared the timing of these events because information may be recorded in primary care databases following receipt of hospital discharge summaries.

**Statistical analysis (estimation of mortality rates and ratios)**

We calculated mortality rates and standardised ratios for the subset of patients with linked ONS mortality data (further detailed about the linkage process are available in CPRD documentation\(^9,10\)). We requested mortality data for these patients, with a final date of follow-up of 1 May 2019. To minimise bias due to delayed death registration, which is likely to occur disproportionately for people who use illicit opioids due to the involvement of coroners in deaths due to drug poisoning, we stopped follow-up six months before this date (i.e. 30 October 2018). To calculate the standardised mortality ratio, we: (a) calculated the duration of follow-up in the HUPIO cohort, stratified by sex, single-year-of-age, and calendar year. We accounted for aging by expanding follow-up for each participant into days, and summarising the number of days by sex, single-year-of-age and calendar year; (b) applying mortality rates in the general population of England\(^7\) to these strata to calculate a number of expected deaths; (c) dividing the number of observed deaths by expected deaths and calculating 95% confidence intervals using the exact Poisson method.

All data manipulation and analysis was conducted using R version 3.6.2\(^8\).

**Patient and public involvement**

People who use illicit opioids were involved in discussions about the need for research into health and healthcare for this population.

**Results**

**Cohort size and characteristics**

After exclusions, the HUPIO cohort included 30,491 individuals from GOLD and 108,270 individuals from AURUM. The derivation of the cohort is shown in Figure 2. The number of participants under active follow-up in GOLD increased over time to a maximum of 11,384 in March 2010 and then decreased as GP practices stopped contributing data; and in AURUM increased over time to a maximum of 44,935 in October 2018.

The majority of patients in both databases had a relevant clinical code and no OAT prescription, while the majority of patients with an OAT prescription also had a relevant clinical code (Figure 3).

The median age at baseline was 33.5 in GOLD and 35.3 in AURUM. The distribution of age groups was similar to that of
Figure 2. Flow-chart of the number of participants in the cohort. OAT - opioid agonist therapy.

Figure 3. Venn diagram showing the overlap between individuals with a history of opioid agonist prescriptions in primary care and clinical codes indicating non-therapeutic opioid use. Note that there are fewer patients in GOLD and this is not reflected in the relative sizes of the circles.
patients entering treatment for opioid dependence in England (see extended data\(^3\)). We observed a linear increase in the mean age of patients entering our cohort, parallel and three years older than the mean age of patients entering treatment for opioid dependence in England (see extended data\(^3\)). When stratified by date, the mean age of patients entering the cohort was similar for GOLD and AURUM. The older average age in AURUM is therefore explained by patients entering the cohort at later dates than in GOLD. In both GOLD and AURUM, 69\% of patients were male; similar to 72\% of patients in opioid agonist treatment in England in 2018\(^4\), and 72\% of participants in the Unlinked Anonymous Monitoring Survey of People who Inject Drugs\(^6\). There was a clear association between deprivation and a history of opioid use, with over 40\% of patients living in the most deprived quintile of neighbourhoods.

In both databases, approximately three-quarters of patients were current smokers (at the most recent record of smoking) and a further 10\% were ex-smokers. Characteristics of patients are shown in Table 1.

Mortality rates and ratios
In CPRD AURUM, linkage to ONS mortality records was conducted for 75,807/108,270 patients (70\%), and in CPRD GOLD, linkage was conducted for 23,241/30,491 patients (76\%). Mortality rates were similar in the two databases. During a combined 910,567 patient-years of follow-up, there were 12,404 deaths (crude mortality rate of 13.6 deaths per 1,000 person-years). Given age- and sex-specific mortality rates in the general population of England, we expected 1,872 deaths, giving an SMR of 6.6 (95\% CI 6.5-6.7). Table 2 provides a summary of follow-up time, deaths, and mortality rates, stratified by database, age and sex. We identified two studies that also reported SMRs in populations with a history of illicit opioid use in England. The first identified 198,247 opiate users from national drug treatment and criminal justice databases between 2005 and 2009\(^8\) and used linkage to national mortality records to estimate an SMR of 5.7 (95\% CI 5.5-5.9). The second identified 6,683 people entering treatment for heroin dependence in South London between 2006 and 2019\(^9\), again using linkage to national mortality records to calculate an SMR of 6.6 (95\% CI 6.1-7.1). In both studies, as in our cohort, the crude mortality rate was higher for men, while the SMR was higher for women.

Internal validation
Among patients hospitalised with a diagnosis of ‘mental and behavioural disorders due to use of opioids’, 89\% of patients in GOLD and 88\% of patients in AURUM also have a record in primary care data indicating a history of illicit opioid use. 72\% and 67\% of hospitalised patients respectively have the first relevant primary care code prior to hospitalisation, 2\% and 3\% have the primary code in the 30 days after hospitalisation, and 15\% and 18\% have the code more than 30 days after hospitalisation. A table of this information is provided in extended data.

Discussion
We developed and validated an electronic healthcare record phenotype that identified approximately 139,000 patients with a history illicit opioid use registered at primary care practices in England. Patient characteristics (age, sex, smoking history, and deprivation) and mortality rates were comparable to other samples of this population.

Strengths and limitations
To our knowledge, this is the first study to develop a method for identifying people with a history of illicit opioid use within primary care records. Earlier studies have focused specifically on people prescribed opioids\(^{40,41}\), general illicit drug use or dependence\(^{42,43}\), and people prescribed OAT\(^{28,44}\). The latter is a limited subset of this population, particularly given that OAT in England is not always prescribed by GPs. These studies have included patients prescribed any methadone or buprenorphine product and excluded those with doses suggesting indications other than OAT (such as pain or palliative cough). Yet over 70\% of daily doses for these medications are missing from CPRD\(^4\), and therefore require imputation or exclusion. The method developed in this study avoids the need for imputation by using products that are specific to OAT. It is possible that we excluded some medicines that are used for OAT in addition to other indications, though few patients with prescriptions of excluded methadone or buprenorphine products had other codes indicating a history of illicit opioid use (see extended data\(^3\)).

For individuals prescribed OAT in primary care, CPRD includes details of these prescriptions. However, in England, many individuals are prescribed OAT in other settings and information about these prescriptions is not available. The main alternative source of national data on people prescribed OAT in England is the National Drug Treatment Monitoring System, which provides data on people in specialist community drug and alcohol treatment services\(^9\). Although the population using these services is likely to overlap with those in primary care, there may also be important differences. For example, people accessing only drug and alcohol services may have more complex drug treatment needs, whilst those only in primary care may have greater physical comorbidity.

The main strength of CPRD in relation to other research datasets for this population is that it offers unique insights into primary healthcare. It can be linked to external datasets to obtain information on care in hospitals, cancer services, and mental health services, as well as causes of mortality. Algorithms have also been developed to facilitate identification of particular aspects of healthcare such as pregnancy\(^{46}\), or health outcomes such as cardiovascular disease\(^{47}\).

There are also limitations in the data, particularly because all data are derived from routine healthcare records. For example, in CPRD there is no systematic recording of the type and frequency of drug use, and the degree of drug dependence. The data on smoking presented in this article suggests that some characteristics of this population are well-captured by GPs, as fewer than
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>Total</td>
<td></td>
<td>30,491</td>
<td>108,270</td>
</tr>
<tr>
<td>Follow-up, years</td>
<td>Median (IQR)*</td>
<td>3.5 (1.8-7.0)</td>
<td>3.8 (1.9-8.1)</td>
</tr>
<tr>
<td>Age at index</td>
<td>18–29</td>
<td>10,585 (34.7)</td>
<td>30,444 (28.1)</td>
</tr>
<tr>
<td></td>
<td>30–39</td>
<td>11,727 (38.5)</td>
<td>42,251 (39.0)</td>
</tr>
<tr>
<td></td>
<td>40–49</td>
<td>5,866 (19.2)</td>
<td>24,955 (23.0)</td>
</tr>
<tr>
<td></td>
<td>50–64</td>
<td>2,313 (7.6)</td>
<td>10,620 (9.8)</td>
</tr>
<tr>
<td></td>
<td>Median (IQR)</td>
<td>33.5 (27.6-40.7)</td>
<td>35.3 (29.1-42.6)</td>
</tr>
<tr>
<td>Sex</td>
<td>Male</td>
<td>21,141 (69.3)</td>
<td>75,065 (69.3)</td>
</tr>
<tr>
<td></td>
<td>Female</td>
<td>9,350 (30.7)</td>
<td>33,205 (30.7)</td>
</tr>
<tr>
<td>Region</td>
<td>East Midlands</td>
<td>1,705 (5.6)</td>
<td>1,846 (1.7)</td>
</tr>
<tr>
<td></td>
<td>East of England</td>
<td>2,735 (9.0)</td>
<td>2,604 (2.4)</td>
</tr>
<tr>
<td></td>
<td>London</td>
<td>3,016 (9.9)</td>
<td>16,876 (15.6)</td>
</tr>
<tr>
<td></td>
<td>North East</td>
<td>911 (3.0)</td>
<td>5,080 (4.7)</td>
</tr>
<tr>
<td></td>
<td>North West</td>
<td>6,128 (20.1)</td>
<td>23,236 (21.5)</td>
</tr>
<tr>
<td></td>
<td>South East</td>
<td>5,675 (18.6)</td>
<td>15,649 (14.5)</td>
</tr>
<tr>
<td></td>
<td>South West</td>
<td>4,407 (14.5)</td>
<td>20,208 (18.7)</td>
</tr>
<tr>
<td></td>
<td>West Midlands</td>
<td>4,071 (13.4)</td>
<td>18,241 (16.8)</td>
</tr>
<tr>
<td></td>
<td>Yorkshire &amp; The Humber</td>
<td>1,843 (6.0)</td>
<td>4,086 (3.8)</td>
</tr>
<tr>
<td></td>
<td>Missing</td>
<td>0 (0.0)</td>
<td>444 (0.4)</td>
</tr>
<tr>
<td>Smoking</td>
<td>Current</td>
<td>23,647 (77.6)</td>
<td>81,968 (75.7)</td>
</tr>
<tr>
<td></td>
<td>Ex-smoker</td>
<td>2,764 (9.1)</td>
<td>11,016 (10.2)</td>
</tr>
<tr>
<td></td>
<td>Never</td>
<td>1,613 (5.3)</td>
<td>6,689 (6.2)</td>
</tr>
<tr>
<td></td>
<td>No records</td>
<td>2,467 (8.1)</td>
<td>8,597 (7.9)</td>
</tr>
<tr>
<td>Quintile of index of multiple deprivation (patient-level)**</td>
<td>1 (least deprived)</td>
<td>2,023 (6.6%)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>3,357 (11.0%)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>4,846 (15.9%)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>4</td>
<td>7,383 (24.2%)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>5 (most deprived)</td>
<td>12,870 (42.2%)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Missing</td>
<td>12 (&lt;0.1%)</td>
<td></td>
</tr>
<tr>
<td>Homelessness</td>
<td></td>
<td>1,143 (3.7)</td>
<td>6,219 (5.7)</td>
</tr>
<tr>
<td>Prison</td>
<td></td>
<td>2,101 (6.9)</td>
<td>8,397 (7.8)</td>
</tr>
<tr>
<td>Alcohol dependence</td>
<td></td>
<td>4,499 (14.8)</td>
<td>22,150 (20.5)</td>
</tr>
</tbody>
</table>

* Follow-up time in this table refers to time that patients are observed within CPRD and primary care records are available. If external databases such as Hospital Episode Statistics or ONS mortality are use, longer follow-up is available.

** Index of Multiple Deprivation was not available for patients in CPRD AURUM at the time of publication.
Table 2. All-cause mortality rates and standardised mortality ratios for patients in primary care in England with a history of illicit opioid use.

<table>
<thead>
<tr>
<th>Database</th>
<th>Sex</th>
<th>Age group</th>
<th>Number at baseline</th>
<th>Follow-up (years)</th>
<th>Observed deaths</th>
<th>Expected deaths</th>
<th>CMR (95% CI)</th>
<th>SMR (95% CI)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>AURUM</strong></td>
<td>Female</td>
<td>18–29</td>
<td>7,176</td>
<td>28,308</td>
<td>105</td>
<td>8.4</td>
<td>3.71 (3.03-4.49)</td>
<td>12.56 (10.28-15.21)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>30–39</td>
<td>8,729</td>
<td>78,140</td>
<td>613</td>
<td>44.7</td>
<td>7.84 (7.24-8.49)</td>
<td>13.71 (12.65-14.84)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>40–49</td>
<td>4,840</td>
<td>60,461</td>
<td>788</td>
<td>79.4</td>
<td>13.03 (12.14-13.98)</td>
<td>9.92 (9.24-10.64)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>50–64</td>
<td>2,490</td>
<td>31,612</td>
<td>771</td>
<td>116.1</td>
<td>24.39 (22.70-26.17)</td>
<td>6.64 (6.18-7.13)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>65+</td>
<td>-</td>
<td>4,672</td>
<td>176</td>
<td>59.4</td>
<td>3.76 (3.21-4.37)</td>
<td>2.97 (2.54-3.44)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>All ages</td>
<td>23,235</td>
<td>203,193</td>
<td>2,453</td>
<td>308.0</td>
<td>12.07 (11.60-12.56)</td>
<td>7.97 (7.65-8.29)</td>
</tr>
<tr>
<td><strong>GOLD</strong></td>
<td>Female</td>
<td>18–29</td>
<td>2,455</td>
<td>10,709</td>
<td>34</td>
<td>3.2</td>
<td>3.18 (2.20-4.44)</td>
<td>10.71 (7.42-14.97)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>30–39</td>
<td>2,612</td>
<td>255,754</td>
<td>1,552</td>
<td>190.0</td>
<td>8.47 (8.31-9.98)</td>
<td>8.17 (7.77-8.58)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>40–49</td>
<td>17,739</td>
<td>226,507</td>
<td>2,472</td>
<td>349.2</td>
<td>14.89 (13.31-15.49)</td>
<td>7.08 (6.80-7.36)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>50–64</td>
<td>721</td>
<td>11,303</td>
<td>277</td>
<td>119.1</td>
<td>42.92 (38.02-48.29)</td>
<td>2.33 (2.06-2.62)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>65+</td>
<td>-</td>
<td>1,977</td>
<td>59</td>
<td>25.8</td>
<td>29.85 (22.72-38.50)</td>
<td>2.29 (1.74-2.95)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>All ages</td>
<td>7,162</td>
<td>73,334</td>
<td>463</td>
<td>41.3</td>
<td>6.21 (5.66-6.80)</td>
<td>11.20 (10.20-12.27)</td>
</tr>
<tr>
<td><strong>Male</strong></td>
<td>18–29</td>
<td>20,481</td>
<td>74,536</td>
<td>463</td>
<td>41.3</td>
<td>6.21 (5.66-6.80)</td>
<td>11.20 (10.20-12.27)</td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td>40–49</td>
<td>1,374</td>
<td>21,245</td>
<td>278</td>
<td>16.1</td>
<td>13.09 (11.59-14.72)</td>
<td>7.61 (7.35-7.87)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>50–64</td>
<td>721</td>
<td>11,303</td>
<td>453</td>
<td>178.4</td>
<td>40.72 (37.06-44.65)</td>
<td>2.54 (2.31-2.78)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>65+</td>
<td>-</td>
<td>11,125</td>
<td>9,219</td>
<td>1387.5</td>
<td>13.70 (13.42-13.99)</td>
<td>6.64 (6.51-6.78)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>All ages</td>
<td>75,807</td>
<td>672,801</td>
<td>9,219</td>
<td>1387.5</td>
<td>13.70 (13.42-13.99)</td>
<td>6.64 (6.51-6.78)</td>
</tr>
</tbody>
</table>

CMR = crude mortality rate (per 1000 person-years) SMR = standardised mortality ratio (compared to the general population of England)
10% had no records and the prevalence of smoking is comparable to that found in other studies. Other characteristics may be less well-captured, for example fewer-than-expected patients had records of homelessness or prison.

Depending on the research question, selection biases are likely to be important. To be included in the CPRD sample, individuals need to be registered with a GP, attend an appointment, and disclose their drug use. At present, we do not know what proportion of this population is registered with a GP. In one study of homeless people who inject drugs in London; a subgroup likely to have relatively high barriers to GP registration, 70% provided GP details9, suggesting that a large proportion of this population is registered. However, disclosure of drug use is likely to differ. Groups more likely to disclose drug use may include those prescribed OAT (either in primary care or specialist drug and alcohol services), and those who are more unwell and therefore have more GP appointments. This latter factor may lead to an overestimation in differences in morbidity and mortality when comparing people with a history of illicit opioid use to the general population. Qualitative research has found both practice-level and individual-level barriers to disclosing and recording illicit drug use9. In particular, patients and GPs who feel more stigma towards illicit drug use may be less likely to discuss the issue. Our internal validation suggested that approximately 90% of patients with opioid use recorded in hospitals also have illicit opioid use recorded in primary care, and the timings of these records suggest they are independent. This supports good sensitivity of the primary care codelist for patients who are registered at a GP practice. It does not provide evidence of sensitivity at a population level, because some individuals are not registered at a GP practice. Consideration of selection biases is important when interpreting analyses using this data and designing sensitivity analyses. Where possible, triangulation with other sources, such as the National Drug Treatment Monitoring System10, can improve confidence in findings.

The data presented here covers a long period of time (1997–2020). Changes happened in both the population and in health services. For example, investment in opioid agonist therapy has changed; while the average age of the population and the prevalence of long-term conditions have increased. This means that selection biases are likely to change over follow-up. In some cases, it may help to restrict analysis to a shorter time-period, or stratify by time-period.

Implications for future research
To date, research into people who use illicit opioids has focused mainly on a narrow range of outcomes such as blood borne viruses and overdoses. The average age of people with a history of illicit opioid use is increasing, and consequently the importance of chronic health conditions is also increasing11. Key areas for future research include the epidemiology of these health issues, assessing the risks and benefits of existing interventions such as OAT in terms of a broader range of health outcomes, and understanding utilisation and quality of general healthcare for this population.

Conclusion
People with a history of illicit opioid use have substantial unmet health needs. Yet to date, large-scale longitudinal studies of healthcare and holistic health outcomes in this population have been limited. We developed and validated a method of identifying people who have a history of illicit opioid use in primary care data to facilitate further research and support improvements in healthcare.

Further details
Ethics and approvals
The study was approved by the MHRA (UK) Independent Scientific Advisory Committee and 19_142R, under Section 251 (NHS Social Care Act 2006). This study is based in part on data from the Clinical Practice Research Datalink obtained under license from the UK Medicines and Healthcare products Regulatory Agency. The data are provided by patients and collected by the NHS as part of their care and support. The interpretation and conclusions contained in this study are those of the authors alone.

Data availability
Underlying data
Researchers can study the HUPIO cohort by applying to the CPRD Independent Scientific Advisory Committee (ISAC). Approval is required if access to anonymised patient level data is being requested for research purposes.

Details of the application process and conditions of access are provided by CPRD at https://www.cprd.com/Data-access.

Extended data

This data file includes:

1. Search terms used to identify codes that may represent a history of illicit opioid use
2. Codelist for identifying people with a history of illicit opioid use
3. Age- and sex-distribution of patients by product and clinical codes
4. Number of patients currently in the cohort
5. Age of patients at cohort entry
6. Internal validation based on hospital admissions for opioid dependence

The phenotype identifying patients with a history of illicit opioid use can also be found on the CALIBER website at
Acknowledgments
We would like to thank Dinah Logan (Specialist Nurse Prescriber) and Dr Ben Watson (Addiction Psychiatry Consultant) for advising on OAT product codes. MH acknowledges support from the National Institute for Health Research (NIHR) Health Protection Research Unit (HPRU) in Behavioural Science and Evaluation.

This study was carried out as part of the CALIBER © resource (https://www.ucl.ac.uk/health-informatics/caliber and https://www.caliberrsearch.org/). CALIBER, led from the UCL Institute of Health Informatics and Health Data Research UK, is a research resource providing validated electronic health record phenotyping algorithms and tools for national structured data sources.

References

https://portal.caliberrsearch.org/phenotypes/lower-hupio-mzx-e2uzxdyvysabtibhkrk. Researchers who are accessing CPRD via the CALIBER platform can use the phenotype directly, while other researchers can download the codelist for their own use.

Data are available under the terms of the Creative Commons Attribution 4.0 International license (CC-BY 4.0).


Open Peer Review

Current Peer Review Status: ✅ ✅

Version 2

Reviewer Report 05 May 2021

https://doi.org/10.21956/wellcomeopenres.18411.r43677

© 2021 McManus S. This is an open access peer review report distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Sally McManus
1 School of Health Sciences, City, University of London, London, UK
2 Violence and Society Centre, City, University of London, London, UK
3 National Centre for Social Research, London, UK

Excellent revisions, no further comments.

Competing Interests: No competing interests were disclosed.

Reviewer Expertise: Population mental health, surveys, cohort studies, inequalities, social determinants

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Version 1

Reviewer Report 22 February 2021

https://doi.org/10.21956/wellcomeopenres.18081.r42593

© 2021 McManus S. This is an open access peer review report distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

Sally McManus
1 School of Health Sciences, City, University of London, London, UK
2 Violence and Society Centre, City, University of London, London, UK
This important methodological study sets out an approach for identifying a cohort of primary care patients in England with a history of illicit opioid use. Consistent with good Open Science principles, codes and algorithms are provided for transparency, to provide detail for future publications based on the cohort identified, and to enable other researchers to apply the same or an adapted approach in their own work with the CPRD, AURUM or related databases. The combined cohort produces an extremely large resource (over 135,000 people), and the longitudinal nature of the data will allow for progression and change to be tracked at the patient-level.

Key external validation work is presented, confirming the cohort profile is consistent with other profiles and with expectations. It would be useful for future validation work to address what proportion of people with a history of illicit opioid use have this noted in their primary care records, and whether biases exist in whether this is a. reported by patients, and b. recorded by practitioners. For example, recognition may be more likely in those who use primary healthcare more, feel more able to disclose illicit behaviour, and for whom opioid use is having greater health consequences. Practitioners can be reluctant to record drug use, and there is useful qualitative research that could provide useful context on such biases (e.g. Davies-Kershaw).

This will be a powerful and pragmatic tool for examining a wider than previously considered range of health outcomes for a cohort of people with a history of illicit opioid use, and their longer term progression.

Minor:
- Spell out IQR acronym in abstract.
- Table 1 – if possible, would be good to have a column showing the profile for the total population and/or patients without a history of illicit opioid use.
- Table 2 – should the word ‘total’ be removed from rows 6 and 14, column 3?

References
Yes

**Are all the source data underlying the results available to ensure full reproducibility?**
Yes

**Are the conclusions drawn adequately supported by the results?**
Yes

**Competing Interests:** No competing interests were disclosed.

**Reviewer Expertise:** Population mental health, surveys, cohort studies, inequalities, social determinants

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

---

**Author Response 20 Apr 2021**

Dan Lewer, University College London, UK

Many thanks for reviewing our article. We have updated the article following your feedback. In addition, we have some specific responses to your comments and questions:

Comment: It would be useful for future validation work to address what proportion of people with a history of illicit opioid use have this noted in their primary care records, and whether biases exist in whether this is a. reported by patients, and b. recorded by practitioners. For example, recognition may be more likely in those who use primary healthcare more, feel more able to disclose illicit behaviour, and for whom opioid use is having greater health consequences. Practitioners can be reluctant to record drug use, and there is useful qualitative research that could provide useful context on such biases (e.g. Davies-Kershaw).

Response: We agree this is central to understanding the cohort. We have added some discussion about the sensitivity of the codelist (two paragraphs above the subheading 'Implications for future research'), including a reference to the article by Davies-Kershaw (thank you for highlighting this). We have also conducted an additional internal validation exercise. This involved selecting all patients in CPRD who had a diagnosis of ICD-10 F11 ('mental and behavioural disorders due to opioid use') and reporting how many appeared in the HUPIO cohort, which is based on primary care records only. This suggested good sensitivity, with approximately 90% of patients identified in hospital also appearing in HUPIO.

**Competing Interests:** None
Dhanya Nambiar
1 Turning Point, Melbourne, Australia
2 Monash University, Melbourne, Australia

Summary:
This paper described a method to establish a cohort of people with a history of opioid use in England using primary care electronic records from 1997 to 2020. Records were searched based on algorithm of primary care notes on opioid use and OAT. At validation, the cohort were found to be comparable to other cohorts of opioid users in the region by age, gender, tobacco use and mortality. The limitations of identifying opioid users from primary care records included people with more physical conditions, who were more connected to primary care and who more likely to disclose their drug use.

Overall, the manuscript is well written and presented, and demonstrates good scientific methods and discussion. I have a few comments on the content and some questions for consideration:

Abstract:
○ Background: Is there a typo in "people who use illicit opioid use"?

Entry and Exit Dates:
○ Over such a long recruitment period, are participants comparable in terms of case ascertainment ie. do you expect that identifying opioid use has become better or worse over the period? And have there been any changes in the healthcare system such as access to OAT or uptake of primary care over the recruitment period?

○ Clinical codes: "...Read and ENOMED clinical codes that may illicit opioid use..." - is there a word missing?

○ External validation: Are all deaths reliably linked to both GOLD and AURUM records?

○ Results - cohort size and characteristics
In Table 1, considering the length of the study period, do you know why the median follow up period was relatively short at under 4 years? Do patients frequently transfer out of GP practices, or is this specific to this cohort?

○ Are participants uniquely identifiable across different participating GP practices over the study period? In future, it would be interesting to compare the frequency at which they present to primary and other healthcare services. This may speak to some of the current limitations such as 1) how representative the cohort is to other international opioid using populations in terms of health service use, and ii) determine if there are subgroups within the cohort with distinct patterns of primary care use (that may have other distinct characteristics such as drug use patterns, homelessness, deprivation, etc).

**Is the work clearly and accurately presented and does it cite the current literature?**
Is the study design appropriate and is the work technically sound?
Yes

Are sufficient details of methods and analysis provided to allow replication by others?
Yes

If applicable, is the statistical analysis and its interpretation appropriate?
Yes

Are all the source data underlying the results available to ensure full reproducibility?
Yes

Are the conclusions drawn adequately supported by the results?
Yes

**Competing Interests:** No competing interests were disclosed.

**Reviewer Expertise:** Health service utilisation, drug use, opioids, cohort studies, mortality

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

---

**Author Response 20 Apr 2021**

**Dan Lewer**, University College London, UK

Many thanks for reviewing our article. We have updated the article following your feedback. In addition, we have some specific responses to your comments and questions:

**Comment:** Over such a long recruitment period, are participants comparable in terms of case ascertainment i.e. do you expect that identifying opioid use has become better or worse over the period? And have there been any changes in the healthcare system such as access to OAT or uptake of primary care over the recruitment period?

**Response:** Yes, there have been changes in both the population and healthcare system that mean case ascertainment is likely to change over time. Two important examples are (1) decreasing investment in OAT since 2010, which may reduce case ascertainment because patients have less overall contact with health services (though OAT is generally delivered in specialist community settings rather than by GPs); (2) the increasing age of the population greater health needs - which may improve case ascertainment as people have a greater need to see their GP. Restricting analyses to limited time periods (or stratifying by time period) may help - which we have now recommended in the paragraph above 'implications for future research'.

**Comment:** Are all deaths reliably linked to both GOLD and AURUM records?
Response: The first version of this article used death data in primary care records. Since then, we have received linked mortality data from the Office for National Statistics for the HUPIO cohort. The mortality rates and ratios in the revised paper use this data. While there are undoubtedly errors in the linkage process, it is generally considered to be high quality and miss few deaths. The linkage uses NHS numbers, a unique identifier that is assigned to everyone registered at a GP practice, in combination with other identifiers. Further information about the linkage methodology is available here: Padmanabhan S, Carty L, Cameron E, Ghosh RE, Williams R, Strongman H. Approach to record linkage of primary care data from Clinical Practice Research Datalink to other health-related patient data: overview and implications. Eur J Epidemiol. 2019;34(1):91-99. doi:10.1007/s10654-018-0442-4

Comment: In Table 1, considering the length of the study period, do you know why the median follow up period was relatively short at under 4 years? Do patients frequently transfer out of GP practices, or is this specific to this cohort?

Response: There are two reasons for this. First, GP practices are continually joining and exiting CPRD. The pool of practices participating in CPRD GOLD is shrinking, and when a practice leaves the database then follow-up for the patient ends. Likewise, the pool of practices for CPRD AURUM is growing and many practices joined later than 1997. Second, this is a mobile population and patients do frequently transfer out of GP practices. We have added a note to the table to clarify that these values are the follow-up duration for primary care data only, and longer follow-up is available where linked data are used (such as Hospital Episode Statistics or Office for National Statistics mortality records). In later analyses we will used matched cohorts of patients who do not have a history of illicit opioid use, for comparison. In these cohorts, follow-up is typically a bit longer because the general population is less mobile than people who use illicit opioids.

Comment: Are participants uniquely identifiable across different participating GP practices over the study period? In future, it would be interesting to compare the frequency at which they present to primary and other healthcare services. This may speak to some of the current limitations such as I) how representative the cohort is to other international opioid using populations in terms of health service use, and ii) determine if there are subgroups within the cohort with distinct patterns of primary care use (that may have other distinct characteristics such as drug use patterns, homelessness, deprivation, etc).

Response: No, unfortunately participants are not uniquely identifiable across different participating GP practices. It is possible that some patients move from one participating practice to another, and therefore appear in the data twice. This is thought to be sufficiently rare in CPRD that each "patient" (in fact a patient-registration episode) can reasonably be considered unique. We agree that it will be interesting to analyse healthcare frequency. CPRD provides linkage to hospital data, and this will allow comparison with existing studies of hospital admissions in patients recruited via specialist OAT clinics.

**Competing Interests:** None