The collection of data relating to incidents and near-miss events has become an entrenched and critical component of safety management across high-risk industries worldwide. According to the quality and safety axiom that “every defect should lead to improvement”, reporting systems exist to provide the raw data for continual improvement processes, as well as serving critical functions with respect to the local management of incidents.

In the quest towards enhancing patient safety, health care has also embraced the collection of near-miss and incident data, with all health services in Australia collecting some form of data. Across most health systems, voluntary reporting systems exist for near-miss and minor incidents, with the primary aim of collecting information about vulnerabilities in the health care system, so that remedies can be applied before an actual adverse event takes place. According to the World Health Organization, the primary role of such reporting systems is to enable learning across large health systems.

The fundamental role of patient safety reporting systems is to enhance patient safety by learning from failures of the health care system. We know that most problems are not just a series of random, unconnected one-off events. We know that health-care errors are provoked by weak systems and often have common root causes that can be generalized and corrected. Although each event is unique, there are likely to be similarities and patterns in sources of risk which may otherwise go unnoticed if incidents are not reported and analysed.

Collecting data is only the first step in a process of organisational learning, through which lessons can be drawn from incidents, modifications made to practice, and the risk of adverse events reduced. At the organisational level, aggregate data are available and more sophisticated monitoring and analysis processes can take place, using data from multiple facilities or multiple units within the organisation. This process of using safety-related data has been termed quadruple-loop learning (personal, local, national and international), emphasising the wide-reaching potential for harnessing lessons from incident report data.

Recently, a disparity between the number of reports being made and the rate of meaningful evidence-based change to practice has been identified. This in turn has suggested that the ineffective use of data, and a lack of published learnings from such systems, may be responsible for the deficit of evidence-based change.

In addition to the primary purpose of systemic quality improvement, secondary administrative roles such as local incident management functions, as well as political and organisational reasons for reporting also exist. As reporting systems become embedded in the day-to-day process of health care, these administrative functions (which demand less detailed reports and analysis) may have come to outweigh the primary purpose. Furthermore, the effective use of increasing volumes of data itself becomes increasingly difficult.

Our study explored the utility of incident reporting systems currently used in most Australian health services. Our primary aim was to determine the depth of information available within a typical incident reporting system and whether more sophisticated “human factors” classification schemes relating to incident aetiology and systemic defences can be applied to incident reporting data.

**METHODS**

**Study context**
Our study was part of a project funded by the Australian Commission on Safety and Quality in Health Care to investigate ways to improve learning from patient safety incident data.

**Design and selection of incident reports**
We undertook a retrospective analysis of incident reports relating to patient misidentification, submitted by health professionals to a number of Australian health services.
incident reporting and management systems. Using a text-based search function for “patient identification”, we extracted incidents from the incident reporting system. From these, we selected incidents that met the inclusion criteria of having patient misidentification as the primary incident cause; occurring within a recent 5-year window (2004–2008); and having complete entry of data in the primary database fields.

Measures

The selected incidents were subjected to classification and analysis according to principal natural categories and system safety classification schemes.

Natural categories were first defined in terms of incident types. These categories simply describe what went wrong in the clinical context, and map the error phenotype, which describes the observable manifestation of the error in the context of its occurrence.

To understand the underlying aetiology of each incident, we classified error type using a 10-category scale developed for use in the context of patient safety. The classification describes the genotype of errors made by clinicians and enables an understanding of the reasons behind why an incident occurred from a human factors perspective.

To classify the ways in which the clinical team identified and recovered from an error, we used an error detection mechanisms measure. This classification reflects the current focus on the design of resilient systems rather than simply attempting to eradicate error.

When determining the requirements for systemic analysis of incident data, an important consideration is how large a sample is required to effectively capture the range of natural categories of incident types at an organisational level. Analysis of category saturation was undertaken to describe the rate at which new unique categories were observed in the data, and to determine the point at which no more categories were able to be differentiated.

Analysis

Data were exported from the health services incident reporting systems and imported into a custom-built FileMaker database (FileMaker Inc, Santa Clara, Calif, USA) to preserve the original fields. Classification was undertaken within this database into relevant natural categories. These data were subsequently exported to Microsoft Excel (Microsoft Corp, Redmond, Wash, USA) and PASW Statistics version 18.0 (SPSS Inc, Chicago, Ill, USA) for analysis. Due to the exploratory nature of this study, analyses were primarily restricted to descriptive analyses.

Ethics approval

Our study was approved by the Royal Adelaide Hospital Research Ethics Committee via the Australian Patient Safety Foundation.

RESULTS

Incident types

Of more than 1000 incidents extracted from the health services’ incident reporting systems, 487 incidents met our inclusion criteria. There were 43 unique incident types present within the dataset, giving a ratio of incident types to incidents of about 1:11. However, most of these categories were represented by few instances, with about half the categories (22/43) represented by three or fewer instances of that particular incident type. The five most prevalent categories accounted for 60% of all 487 incidents (Box 1).

The most prevalent incident types involved administering medication to the wrong patient (25.7%) and performing a procedure on the wrong patient (15.2%). Most of these incidents involved phlebotomy, which is a common and relatively low-risk procedure.

A second cluster of prevalent incident types related to mislabeling an order for pathology or medical imaging (7.0%); attributing the results for a diagnostic test (frequently medical imaging) to the wrong patient (6.2%); and labelling a specimen with the wrong patient’s details (6.0%).

The process of category saturation and the rate at which the unique incident types were identified in the dataset are shown in Box 2. The shape of the curve indicates a rapid initial classification of different incident types, with half the total number of categories identified in the first 13.5% of the incidents. All 43 incident types were classified within 76.2% of the dataset.

Aetiology

Of the 487 incidents analysed, 52 (10.7%) had sufficient information to identify and classify a specific error type. Of these, 67.3% (35/52) were skill-based errors (slips or lapses), which is an intuitive finding given the prevalence of errors in automated tasks such as labelling specimens or labelling request forms. Of the remainder, 11.5% (6/52) were deliberative errors, 7.7% (4/52) involved choice of rule, and 5.8% (3/52) involved both matching and perception error types.

Recovery

Two hundred and eighty-eight incident reports (39.1%) had sufficient detailed information to enable classification of the error-detection mechanism. The most prevalent error-detection mechanism was routine check (44.1%, 127/288). Mismatch with the patient’s clinical presentation was the next most prevalent mechanism (15.6%, 45/288), followed by mismatch with the clinician’s expectation (12.5%, 36/288).

DISCUSSION

The principal role of incident reporting systems is to ensure a consistent and coordinated approach to the identification and analysis of incidents so that lessons can be learned and shared across the whole health system. Reporting of near-miss events has been shown to offer numerous benefits compared with retrospective investigation of adverse events. Perhaps the most important of these benefits is the greater frequency of near-miss events, thus allowing quantitative analysis; and the ability to identify and analyse the recovery functions that enabled the accident trajectory to be stopped before an actual adverse event took place.

Efficacy of current incident reporting in Australia

Our study shows that the information contained in the current incident reports is frequently inadequate to allow sophisticated analyses of incident aetiology and, more importantly, the recovery functions involved in averting or ameliorating adverse events. The results suggest that incident reporting may not be meeting the primary objective of enabling continuous improvement of safety and quality.

These results echo a report of the Institute of Medicine in the United States, which suggested that the value of near-miss reporting is required further investigation. Although there is general agreement that near-miss reports contain relevant information about identifiable hazards that cannot be collected by other means, most organisations only take action on serious adverse events, which diminishes the value of reporting large numbers of near misses.

Indeed, the sheer volume of incidents reported means that health care organisational...
Incident reporting systems are also subject to a number of other limitations, including international suggestions of significant underreporting of incidents, and of medical error, as doctors rarely report. Any reporting is highly affected by hindsight bias (based on the degree of harm the patient has suffered, especially considering that doctors tend to disproportionately report the more severe events). Incident reports are thus “a nonrandom sample of identified hazards from a larger unknown universe of hazards”. In light of these limitations, it is critical that we explore alternative ways to learn from incidents in health care.

The category saturation results suggest that a critical threshold exists, beyond which quantitative analyses of existing forms of incident data yield limited new information. For health care incident types related to patient misidentification, this was found to be no more than 400 incidents. In any case, a substantial amount of information was available from only 200 incidents.

Doing more with less — category saturation. Currently, public health systems with established incident reporting systems (such as South Australia, Queensland and New South Wales) receive on average between 30,000 and 100,000 incident reports annually. While there are significant amounts of data available, our study demonstrates that effective learning from incidents requires much less data.

### 1 Principal natural categories of incident type recorded in Australian incident reporting systems, 2004–2008 (n = 487)

<table>
<thead>
<tr>
<th>Incident type</th>
<th>No.</th>
<th>Incident type</th>
<th>No.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Admission</td>
<td></td>
<td>Medication chart for wrong patient</td>
<td>8 (1.6%)</td>
</tr>
<tr>
<td>Admitted under wrong medical record number</td>
<td>6 (1.2%)</td>
<td>Wrong chart in notes</td>
<td>8 (1.6%)</td>
</tr>
<tr>
<td>Armband</td>
<td></td>
<td>Wrong label on chart</td>
<td>25 (5.1%)</td>
</tr>
<tr>
<td>None</td>
<td>8 (1.6%)</td>
<td>Wrong label on consent</td>
<td>2 (0.4%)</td>
</tr>
<tr>
<td>Wrong patient</td>
<td>11 (2.3%)</td>
<td>Wrong labels in notes</td>
<td>10 (2.1%)</td>
</tr>
<tr>
<td>Autologous product</td>
<td></td>
<td>Wrong &quot;not for resuscitation&quot; instruction</td>
<td>1 (0.2%)</td>
</tr>
<tr>
<td>Wrong patient</td>
<td>2 (0.4%)</td>
<td>Wrong notes in file</td>
<td>6 (1.2%)</td>
</tr>
<tr>
<td>Wrong product</td>
<td>2 (0.4%)</td>
<td>Wrong notes with patient</td>
<td>9 (1.8%)</td>
</tr>
<tr>
<td>Booking</td>
<td></td>
<td>Wrong results in file</td>
<td>9 (1.8%)</td>
</tr>
<tr>
<td>Booked under wrong medical record number</td>
<td>1 (0.2%)</td>
<td>Pre-admission</td>
<td></td>
</tr>
<tr>
<td>Calls to family</td>
<td></td>
<td>Letter sent to wrong patient</td>
<td>1 (0.2%)</td>
</tr>
<tr>
<td>Wrong patient’s family called</td>
<td>1 (0.2%)</td>
<td>Procedure</td>
<td></td>
</tr>
<tr>
<td>Clinical information system</td>
<td></td>
<td>Errors in patient details</td>
<td>1 (0.2%)</td>
</tr>
<tr>
<td>Errors in patient details</td>
<td>1 (0.2%)</td>
<td>Incidence</td>
<td></td>
</tr>
<tr>
<td>Discharge</td>
<td></td>
<td>Given wrong summary</td>
<td>74 (15.2%)</td>
</tr>
<tr>
<td>Given wrong summary</td>
<td>2 (0.4%)</td>
<td>Wrong patient</td>
<td></td>
</tr>
<tr>
<td>Given wrong paperwork</td>
<td>1 (0.2%)</td>
<td>Request</td>
<td></td>
</tr>
<tr>
<td>Wrong patient</td>
<td>3 (0.6%)</td>
<td>Wrong details</td>
<td>2 (0.4%)</td>
</tr>
<tr>
<td>Equipment</td>
<td></td>
<td>Wrong label</td>
<td>34 (7.0%)</td>
</tr>
<tr>
<td>Used on two patients</td>
<td>2 (0.4%)</td>
<td>Wrong patient</td>
<td>23 (4.7%)</td>
</tr>
<tr>
<td>Follow-up</td>
<td></td>
<td>Wrong label</td>
<td>30 (6.2%)</td>
</tr>
<tr>
<td>Wrong patient</td>
<td>1 (0.2%)</td>
<td>Wrong patient</td>
<td>11 (2.3%)</td>
</tr>
<tr>
<td>Handover instructions</td>
<td></td>
<td>Wrong patient</td>
<td></td>
</tr>
<tr>
<td>Wrong patient</td>
<td>1 (0.2%)</td>
<td>Script</td>
<td></td>
</tr>
<tr>
<td>Meal</td>
<td></td>
<td>Wrong patient</td>
<td>1 (0.2%)</td>
</tr>
<tr>
<td>Wrong patient</td>
<td>2 (0.4%)</td>
<td>Wrong label</td>
<td>1 (0.2%)</td>
</tr>
<tr>
<td>Medication</td>
<td></td>
<td>Specimen</td>
<td></td>
</tr>
<tr>
<td>Wrong label</td>
<td>7 (1.4%)</td>
<td>Wrong label</td>
<td>29 (6.0%)</td>
</tr>
<tr>
<td>Wrong medication</td>
<td>5 (1.0%)</td>
<td>Transfer</td>
<td></td>
</tr>
<tr>
<td>Wrong patient</td>
<td>125 (25.7%)</td>
<td>Wrong patient</td>
<td>10 (2.1%)</td>
</tr>
<tr>
<td>Multiresistant organism status</td>
<td></td>
<td>Wrong unit</td>
<td>2 (0.4%)</td>
</tr>
<tr>
<td>Wrong status</td>
<td>1 (0.2%)</td>
<td>Transfusion</td>
<td></td>
</tr>
<tr>
<td>Wrong patient</td>
<td>1 (0.2%)</td>
<td>Wrong patient</td>
<td>1 (0.2%)</td>
</tr>
</tbody>
</table>
We suggest that the goal of improving learning from patient safety incident data will be best served by exploring innovative approaches to sampling incidents and eliciting more detailed data for each event. It is apparent that we are lacking quality, not quantity, of incident reports. We should be seeking data that will inform a deeper understanding of the system features that will promptly detect and mitigate the inevitable failures that occur in a complex system such as health care.

Innovative approaches to maximise the efficacy of incident reporting. To facilitate collection of better-quality incident reports, we propose testing a modified incident reporting system in which new incident reports are selected at random and followed up with the individuals involved, to seek more detailed information pertaining to the incident. We envisage that this type of system would complement the existing approaches of incident reporting and root cause analysis, but with better return on investment.

A novel approach to more effective use of incident report data in industries such as health care would first involve risk-based sampling of incidents from traditional incident reporting systems. The aim of this sampling would be the proactive identification of emergent high-risk safety concerns, rather than the analysis of sentinel events after they have occurred. Risk-based sampling would involve introducing risk assessment into the initial data-collection process within the incident reporting system. This approach already exists within other health care safety and quality methods. Health care failure mode and effects analysis, for example, risk assess each failure event in terms of potential severity and probability of occurrence, producing a hazard score that in turn enables prioritisation of safety management activities.26

Second, this new approach would involve eliciting more detailed data on these high-risk incident types. This process could involve targeting specific incident types and using telephone interview or survey techniques to elicit further information. More detailed information on the aetiology of the event could be obtained in this manner, enabling more sophisticated approaches to analysis. Perhaps more importantly, this approach would enable collection of information on the features of near-miss events that relate to the error tolerance or resilience of the system. In this way, the focus of improving learning from patient safety incidents would shift from the negative analytic frame of “what went wrong” to the more positive analytic frame of “what enabled us to maintain the integrity of the system”. This more positive focus on the systemic defences that enable the detection and timely mitigation of incident trajectories could hold an important key for the renewed vigour of quality and safety interventions in health care. Further research is required to examine the resource implications and practicalities of such changes to incident reporting systems.

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We thank the Australian Commission on Safety and Quality in Health Care for funding this study. We also thank the Australian health service for providing access to the data for this study. Ongoing analysis of these incident data are critical to enhancing safety and quality in health care, and we acknowledge the health service’s commitment to safety as a priority over political or other risk associated with providing access to such data. Sarah Michael conducted the search and extracted incidents for analysis. Peter Hibbert helped to initiate the study.

COMPETING INTERESTS

None identified.

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