Electronic health record system risk assessment: a case study from the MINET

Khin Than Win, Hai Phung, Lis Young, Mai Tran, Carole Alcock, Ken Hillman

Abstract

This article discusses the risk assessment of a health information system. A case study was conducted at the South Western Sydney Area Health Service to examine the potential risks of the Maternal and Infant Network (MINET) health information system using Failure Mode Effect Analysis (FMEA). FMEA was conducted by utilising safety attributes identified by the authors. Potential failure modes of the system were identified by the study. From this study, it can be concluded that FMEA is an appropriate risk-assessment method for MINET.

Keywords: Health information; risk-assessment; safety.

Introduction

A thorough literature search indicates that there has been little study to date on risk-assessment methods for electronic health record systems. Although there have been assessments of project management and security risk to systems, there has been no safety risk assessment. The main objective of this study was to identify possible risks in a health information system (the Maternal and Infant Network: MINET) using Failure Mode Effect Analysis (FMEA). In particular, this article describes a risk-assessment case study conducted at the Simpson Centre for Health Services Research. Identifying possible failures from the system could mitigate or prevent its potential failure, while also enhancing the safety of the system.

Background

The report To err is human: building a safer health system (Kohn et al. 2000) emphasised the importance of safety in healthcare. There have been many reports of medical misadventure; for example, 98,000 Americans die each year as a result of preventable medical errors (Kohn et al. 2000). The Institute of Medicine estimates the number of lives lost to preventable medication errors alone represents over 7000 deaths annually, which is more than the number of injuries in US work places (Institute of Medicine 2000). The 1998 National Survey of New Zealand has documented that 4.5% of all hospital admissions were associated with highly preventable adverse events (Davis et al. 2001). In Australia, more than 55,000 patients become disabled and as many as 18,000 unnecessary deaths occur each year due to medical errors (Weingart et al. 2000).

As electronic health record systems are part of the overall healthcare system, it is important to ensure that electronic health data are secure and dependable in order to reduce the risk of occurrence of medical errors. The identification of safety requirements of electronic health record systems would also help to reduce errors (Win et al. 2002). Exploring undesirable events that can occur from electronic health record systems would assist in identifying risk.

There are specific risk-assessment methods available for different systems. With proper risk assessment, potential risks can be identified and avoided, resulting in a safer health record system, and ultimately in safer healthcare. Awareness of risk and safety requirements is important, as it will assist in re-engineering of the appropriate electronic health record systems for healthcare organisations.

Safety and dependability

Drawing on the relationship framework for dependability and data quality and the literature review under-
taken, we have identified safety attributes for Electronic Health Records (EHRs). However, as EHRs can have different purposes for different information management and systems, safety attributes may also differ (Schloeffel & Jeselon 2002; Shiffman et al. 1999). Attributes of dependability include availability, reliability, security and safety (Sommerville 2001).

Data quality

Data quality has been defined as ‘the totality of features and characteristics of a data set that bears on its ability to satisfy the needs that result from the intended use of the data’ (Arts et al. 2002). Wherever possible, data quality should not be compromised, because low quality health data will have a significant impact on decision-making processes on information management.

Data quality and dependability

Box 2 presents characteristics involved in healthcare data quality. It shows how data quality is related to the dependability and lists appropriate measures needed to ensure data quality.

Electronic Health Record Systems (EHRs)

A Health Information Network for Australia identified the electronic health record as:

... an electronic longitudinal collection of personal health information, usually based on the individual or family, entered or accepted by health care professionals, which can be distributed over a number of sites or aggregated at a particular source, including a handheld device. The information is organised primarily to support continuing, efficient and quality health care (A Health Information Network for Australia, 2000).

Humphreys has defined health records which are used in health services research for monitoring public health and outcomes as ‘population health records’ (Humphreys 2000). However, these records are also used for data acquisition, record keeping, communication, integration, surveillance, information storage and retrieval, and data analysis. These attributes also apply to EHRs, as defined by Perreault and Wiederhold (1990); therefore, MINET can be categorised as one of the EHRs.

MINET case study

The case study was conducted at the Simpson Centre for Health Services Research, from the Maternal and
Infant Network (MINET) database. MINET was selected as a case study for this research as it involves different electronic health data from different sources. The MINET database contains health data on infants and children in the South Western Sydney Area Health Service (SWSAHS) from the prenatal period to school age (0-5 years). MINET involves community-based data from the Ingleburn Baby Information System (IBIS Database) and obstetric and gynaecological data (OBSTET) from the hospital data system. MINET caters for all persons living in the SWSAHS. These data are important for public health and health service research, because the prenatal and infant and early childhood periods are critical for the promotion of good health and the development of personal characteristics for adolescence and adulthood (Halldorsson et al. 1999). It is also important that the databases are accurate for the purposes of health service research. As part of MINET, OBSTET data are downloaded to the Simpson Centre for Health Data Research only and Simpson Centre does not have any control over how the OBSTET data are collected and processed.

Currently, IBIS Version 4 is being used in the SWSAHS. IBIS uses Optical Mark Recognition (OMR) to capture data. IBIS is part of a Local Area Network, which enables sharing of information with other service points for mothers and their babies. There are two types of data for IBIS: baseline and follow-up data. The IBIS baseline form is used for the first visit at the baby clinic and the IBIS follow-up form is used for subsequent visits.

**Methodology**

Different methods have been explored in order to identify one that is appropriate for risk assessment of EHRs. Failure Mode Effect Analysis (FMEA) was proposed because risk assessment conducted through this method involves identifying the possible failure modes of the system before failure can occur (Win, Cooper & Alcock 2004). An alternative method of risk assessment is root cause analysis, for example fault-tree analysis; however, this method identifies the source of error after the event. Fault-tree analysis is more suitable for retrospective studies of systems in which adverse events or errors have already occurred, or to track back to the root-cause conditions. With FMEA, failure mode can be predicted and action taken to prevent the condition from occurring in the first place (Win 2004). It is clearly important to identify any possible risks first to ensure the system’s safety, so FMEA is a more suitable approach compared with root-cause analysis in this case.

**Failure Mode Effect Analysis (FMEA)**

Failure Mode Effect Analysis is a structured approach to the prediction and identification of the consequences of failures in a system. To conduct an FMEA, the processes involved in the system can be subdivided into sub-processes and possible failure modes of these processes. Upon identification, their potential effects can be estimated and analysed to prevent the possible failures.

**Results and discussion**

Possible failure modes of MINET can be predicted as illustrated in Box 7.
Potential effects, severity, probability and hazard score of each potential failure mode are described in Box 8.

These identified hazards were notified to the authorised personnel responsible for the system and necessary actions were identified.

The following section presents the analysis of possible failure modes identified in Box 7. These include Forms not marked appropriately and Missing forms. Forms that are not marked appropriately could be ruled out at the time of the scanning process. This can be due to human error, and the Simpson Centre needs to trace them back to the Community Health Centre. Patients’ data are filled in manually at the Community Health Centres and compiled to scan.

Duplication of MRNs and Same MRN for different patients are also possible failures in the source of the data, as identified in Box 7. It is noted that different medical record numbers (MRNs) are used in different services. In addition, maternal and infant MRNs can differ, leading to the possibility of incorrect association of maternal and infant MRNs; for example, parents may be unmarried, or a mother might not necessarily change her surname upon marriage, in which case the mother’s and infant’s surnames will be different. In some cases, a mother’s surname could be changed from the previous childbirth history as a result of subsequent marriage or divorce. Duplication of MRNs for the same person at different services or at the same service is always a possibility, and several problems could result from wrong linkage of the data. Data may be linked to the wrong patient at the time of analysis and there can be errors in predicting health indicators.

Possible failure modes during scanning and downloads have been identified in Box 7. Scanning is done in batch processing, and some documents could be misplaced and lost during this process. To avoid this problem, it is advisable to have computerised data entry at the point of care.

Files cannot be linked if the system is unavailable due to such factors as power loss or application failure. The centre has not yet experienced application failure, and, as it is not a ‘real time’ system, power loss for a limited period is acceptable. There would also be loss of data should the data file be corrupted, and the Simpson Centre regularly backs up data so that this can be prevented. There is also a high probability that data from different databases cannot be matched properly, as different versions of IBIS have different data units. These problems need to be addressed at the time of form download and during statistical analysis.

Wrong data linkage can lead to incomplete and inaccurate data, and in the case of clinical data this could result in a significant and immediate impact on the patient. The Simpson Centre uses aggregated data for statistical analysis, and records not matched perfectly are excluded from the analysis. Disruption may occur in the scanning process due to mechanical problems with the machine, power failure, or inexperienced scanning operators. When detected during the scanning process, the problem can be rectified to minimise impact on the system; the only impact would then be on the scanning task the person was performing at the time.
Conclusion
The Simpson Centre uses data mainly for health services research, an important role, as it could affect public health research in, for example, healthcare processes, disease patterns, disease surveillance, prevention of disease and health promotion. Data from MINET are used for health services research, and conducting a risk assessment study has had a positive effect, as inaccurate or incomplete information can have an impact on health outcome indicators. It is very important that electronic health data from MINET are complete and accurate, as MINET is used for data collection, analysis and interpretation of data for early intervention, planning, prevention and evaluation.

Although MINET data are health data, it can be seen that not all safety attributes identified are appropriate for this system. As MINET is aggregated data and it is used for de-identified data, the requirement that a patient’s name and identification be displayed on every screen is not applicable to MINET. However, it can be concluded that unique patient identification is very important for MINET, as there could be different potential effects (as described above) if a patient could not be uniquely identified. Alerts and reminders could assist in follow-up and referral for community health, but they do not have a significant impact on the health service research data for MINET. They could, however, be appropriate for disease surveillance health research in different aggregated data sets. Data regarding medication and dosage could be important for maternal and child health data in community care; for example, errors in a child’s medication dosage could result in a serious outcome. Including information regarding these data will add value to the system.

As for all health informatics systems, issues of privacy, confidentiality and security are important in MINET. Those involved with data entry and processing are given a clear explanation of privacy policy and are therefore aware of confidentiality concerns. Access level is decided by the user group, and the administrator needs to set the level accordingly to ensure the privacy of data.

As data from a number of sources are used for research, common standards are important for different databases. Completeness of data is essential for MINET databases, as incomplete data will result in errors in statistical analysis, which will in turn have an impact on healthcare indicators. FMEA has identified possible failures of the system and is therefore an appropriate risk-assessment method for the MINET.

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8: Failure mode effect analysis of different processes

<table>
<thead>
<tr>
<th>Potential failure mode:</th>
<th>Potential effect</th>
<th>Severity</th>
<th>Probability</th>
<th>Hazard score</th>
</tr>
</thead>
<tbody>
<tr>
<td>1.1. Forms are not marked appropriately</td>
<td>Incomplete data</td>
<td>High</td>
<td>Medium</td>
<td>6</td>
</tr>
<tr>
<td>1.2. Forms missing</td>
<td>Incomplete data</td>
<td>High</td>
<td>Medium</td>
<td>6</td>
</tr>
<tr>
<td>1.3. Different MRN for the same person</td>
<td>Data unavailable or misleading data for the research purpose</td>
<td>Medium</td>
<td>Medium</td>
<td>4</td>
</tr>
<tr>
<td>1.4. Same MRN for different patient</td>
<td>Misleading data for the research purpose</td>
<td>High</td>
<td>Medium</td>
<td>6</td>
</tr>
<tr>
<td>2.1. Forms are printed malaligned</td>
<td>Documents unscannable</td>
<td>Low</td>
<td>Low</td>
<td>1</td>
</tr>
<tr>
<td>2.2. Overwriting existing file</td>
<td>Data loss</td>
<td>High</td>
<td>Low</td>
<td>3</td>
</tr>
<tr>
<td>2.3. Disruption of the scanning process</td>
<td>Incomplete data entry</td>
<td>Low</td>
<td>Medium</td>
<td>2</td>
</tr>
<tr>
<td>3.1. MRN cannot be matched</td>
<td>Mother and infant linked wrongly</td>
<td>High</td>
<td>Medium</td>
<td>6</td>
</tr>
<tr>
<td>3.2. Files cannot be linked</td>
<td>Data unavailable for research purpose</td>
<td>High</td>
<td>Low</td>
<td>3</td>
</tr>
<tr>
<td>3.3. File corrupted</td>
<td>Data unavailable</td>
<td>High</td>
<td>Low</td>
<td>3</td>
</tr>
<tr>
<td>3.4. Data cannot be matched</td>
<td>Impact in analysis of data</td>
<td>High</td>
<td>High</td>
<td>9</td>
</tr>
<tr>
<td>4.1. The system is attacked by the intruder</td>
<td>Breach of confidentiality</td>
<td>High</td>
<td>Low</td>
<td>3</td>
</tr>
<tr>
<td>4.2. Staff member stolen the patient information</td>
<td>Breach of confidentiality</td>
<td>High</td>
<td>Low</td>
<td>3</td>
</tr>
</tbody>
</table>
References


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