

# 1 Utilising a data capture tool to populate a cardiac rehabilitation registry: a feasibility study

## 2 Abstract

3 **Background:** Clinical registries are effective for monitoring clinical practice, yet manual data  
4 collection can limit their implementation and sustainability. The objective of this study was to assess  
5 the feasibility of using a data capture tool to collect cardiac rehabilitation (CR) minimum variables  
6 from electronic hospital administration databases to populate a new CR registry in Australia.

7 **Methods:** Two CR facilities located in Melbourne, Australia participated, providing data on 42  
8 variables including: patient socio-demographics, risk factors and co-morbidities, CR program  
9 information (e.g. number of CR sessions), process indicators (e.g. wait time) and patient outcomes  
10 (e.g. change in exercise capacity). A pre-programmed, automated data capture tool (GRHANITE™)  
11 was installed at the sites to extract data available in an electronic format from hospital sites.  
12 Additionally, clinicians entered data on CR patients into a purpose-built web-based tool (REDCap).  
13 Formative evaluation including staff feedback was collected.

14 **Results:** The GRHANITE™ tool was successfully installed at the two CR sites and data from 176  
15 patients (median age=67 years, 76% male) were securely extracted between September – December  
16 2017. Data pulled electronically from hospital databases was limited to seven of the 42 requested  
17 variables. This is due to CR sites only capturing basic patient information (e.g. socio-demographics,  
18 CR appointment bookings) in hospital administrative databases. The remaining clinical information  
19 required for the CR registry were collected in formats (e.g. paper-based, scanned or Excel  
20 spreadsheet) deemed unusable for electronic data capture. Manually entered data into the web-tool  
21 enabled data collection on all remaining variables. Compared to historical methods of data  
22 collection, CR staff reported that the REDCap tool reduced data entry time.

23 **Conclusions:** The key benefits of a scalable, automated data capture tool like GRHANITE™ cannot be  
24 fully realised in settings with under-developed electronic health infrastructure. While this approach

25 remains promising for creating and maintaining a registry that monitors the quality of CR provided to  
26 patients, further investment is required in the digital platforms underpinning this approach.

27 **Key words:** cardiac rehabilitation; registry; data scraping

## 28 **Introduction**

29 The ability to quantify healthcare quality relies on the implementation of appropriate systems  
30 that can accurately capture how care is being delivered [1]. In a recent scientific statement, the  
31 American Heart Association called for the systematic redesign of cardiovascular care to enable a  
32 'learning healthcare system' which uses information technology and data infrastructures to enhance  
33 optimal healthcare delivery [2]. In Australia, the Commission on Safety and Quality of Health Care  
34 (the Commission) promotes the use of clinical registries to systematically monitor healthcare,  
35 highlight variations in outcomes, and inform quality improvement efforts [3]. Ischaemic heart  
36 disease ranks as the highest priority area identified by the Commission that would benefit from  
37 registry development due to the high burden of disease, serious consequences associated with poor  
38 quality care and strong clinical support [4]. This follows the success of cardiac registries  
39 internationally including the Global Registry of Acute Coronary Events (GRACE)[5] and effective  
40 system-wide changes seen by countries such as Sweden which has established more than 100 health  
41 registries including some that have been maintained for more than 25 years [6].

42 A key component of secondary prevention of heart disease is cardiac rehabilitation (CR).  
43 Although CR is extremely effective in preventing cardiovascular recurrent events and complications  
44 [7] and recommended in clinical guidelines [8, 9], there is variability in program delivery and quality  
45 [10] some of which stems from a lack of uniform data collection and monitoring systems. The need  
46 to develop quality indicators and implement systems that collect standardised CR outcome data is  
47 recognised by several national associations internationally [11-13] including the Australian  
48 Cardiovascular Health and Rehabilitation Association's (ACRA; the Australian association of CR  
49 professionals) [14]. Specifically, ACRA recommend that all CR services collect a minimum set of data

50 and report on key performance indicators to promote continuous quality improvement of services  
51 and benchmarking[14]. Despite these calls, quality indicator data from CR sites are, for the most  
52 part, not systematically collected or collated. One jurisdiction in Australia, Queensland, has recently  
53 established the Queensland Cardiac Outcomes Registry (QCOR) which includes the collection of CR  
54 quality indicator variables as part of the registry and will be the first state in Australia to  
55 systematically collect CR data [15]. In the state of Victoria, the Victorian Cardiac Outcomes Registry  
56 (VCOR) [16] collects data on cardiac patients across 35 hospitals on three modules (percutaneous  
57 coronary intervention, heart failure and the early treatment of acute myocardial infarction).  
58 However, CR data are not included within VCOR.

59 Globally, custodians of CR registries have noted challenges, common to any registry, such as  
60 site investment or 'buy-in', privacy and security considerations, as well as limited resources for  
61 contributing data [17]. Indeed, sites are often required to manually enter data, which is time-  
62 consuming for clinical staff and increases the risk of data errors [18]. Ideally, data collection should  
63 be automated and linked to administrative databases or electronic medical records (EMRs). With  
64 advances in technology, this is becoming more feasible. Automated data capture techniques using  
65 specially-designed software can be used to extract routinely-collected data. Such software can also  
66 incorporate automated safeguards built-in to the data entry systems to ensure privacy protection.  
67 This has been previously demonstrated within primary care and other settings in Australia [19] using  
68 the GRHANITE™ (GeneRic Health Network Information for the Enterprise [20]):  
69 <https://www.grhanite.com/>) tool.

70 The aim of this manuscript was to assess the feasibility of extracting routinely-collected  
71 minimum data (as defined by the NSW division of ACRA [21]) from CR sites and hospital  
72 administration databases using the GRHANITE™ automated data capture tool in order to populate a  
73 Victorian CR Registry (VCRR).

## 74 **Methods**

75 *Overarching design of VCRR*

76 This feasibility study consisted of a 3-month (September-December 2017) data collection  
77 period involving quantitative data capture from two pilot sites and formative evaluation of the  
78 process including feedback from CR clinicians. The design of the registry was guided by technical  
79 standards outlined by the Commission [3], as illustrated in a logic model (Figure 1).

80 **\*FIGURE 1\***

81 **Figure 1** Clinical Quality Registries Information Model [16]. Reproduced with permission from *Logical Design*  
82 *for Australian Clinical Quality Registries*, developed by the Australian Commission on Safety and Quality in  
83 Health Care (ACSQHC), for use exclusively in Australia. ACSQHC: Sydney. 2012.

84 Acronyms: CQR: clinical quality registry; MBS: Medicare Benefits Schedule; PBS: Pharmaceutical Benefits  
85 Scheme

86

87 *Selection of the minimum variables for the VCRR*

88 The registry comprised a minimum set of variables selected from the New South Wales  
89 (NSW) ACRA association quality indicators and data dictionary which was based on expert  
90 consensus [21]. The 42 selected data elements consisted of: demographic information (e.g. sex, age),  
91 disease/condition (e.g. principal referral diagnosis) risk factors and co-morbidities (e.g. diabetes  
92 status, smoking status), intervention (e.g. number of CR sessions), process indicators (e.g. CR wait  
93 time) and individual patient outcomes (e.g. change in pre-post exercise capacity) (Table 1).

94 **\*TABLE 1 \***

95 **Table 1.** Victorian Cardiac Rehabilitation Registry minimum variables

96 *Setting and recruitment*

97 In the state of Victoria in South East Australia, there are 136 CR programs, delivered across  
98 publicly and privately-funded hospitals and community health settings. The national association of  
99 CR professionals (ACRA) has a State-level directory of all CR facilities which was used to identify one

100 public and one private site to invite to participate in the study. These sites were purposively selected  
101 to ensure sample representation of: funding sources (public and private), settings (acute hospital,  
102 rehabilitation hospital), and location (metropolitan and suburban). Site 1 was a large publicly-funded  
103 program, which runs a six-week CR program for approximately 40 outpatients per week. Site 2 was a  
104 private facility primarily funded through health insurance funds and the Department of Veteran  
105 Affairs, which runs a 12-week program for approximately 15 outpatients per week. Participating  
106 sites were offered a stipend of AU\$6,000 (USD\$4700) to cover cost related to staff time for the set-  
107 up of automated data collection. Both CR sites agreed to participate.

#### 108 *Ethics approval*

109 The study was approved by the Human Research Ethics Committee (HREC) at the University  
110 of Melbourne (HREC number: 1748609) and included a waiver of consent for individual patient data  
111 (which was de-identified). Site-specific research ethics approval was also obtained. Staff who  
112 participated in qualitative interviews provided informed consent.

#### 113 *Automated collection procedure (GRHANITE™)*

114 The team at the University of Melbourne's Health and Biomedical Informatics Centre  
115 Research Information Technology Unit (led by DB) assisted in the development of the data extraction  
116 protocol and worked with the sites' Information Technology (IT) teams to create an interface regime.  
117 This required the development of a "mapping" document which linked the variables requested from  
118 the research team with the variables collected and available electronically at the sites. The overview  
119 of the study methods can be seen in Figure 2.

#### 120 **\*FIGURE 2\***

121 **Figure 2** Overview of the study methods

122

#### 123 *Manual web-based data collection (REDCap)*

124 To capture variables that were not available electronically at the sites, a secure web-based  
125 data collection form was designed using the REDCap (Research Electronic Data Capture:  
126 <https://www.project-redcap.org/>) software. The web-based form included three sections (Section 1:  
127 identifiable patient information; Section 2: pre-CR data; Section 3: post-CR data) and was trialled for  
128 two weeks at both sites, with feedback from the CR sites informing refinement of the data entry  
129 template. Once finalised, clinicians entered data for patients who were enrolled in the CR programs  
130 during the data collection period. The REDCap data collection forms contained mandatory fields to  
131 reduce missing data and in-built logic checks to increase the accuracy of data. Authorised staff were  
132 provided with a secure log-in which enabled access to the REDCap template; data access restrictions  
133 ensured clinicians could only view data from their site. Additional detail on REDCap is provided in  
134 Supplementary File 2.

#### 135 *Data extraction and linkage*

136 CR data were extracted from the sites via the University of Melbourne's GRHANITE™  
137 research data acquisition system. The GRHANITE™ interface was installed at both sites and  
138 scheduled to extract pre-determined variables on patients who participated in the CR program  
139 during the data collection period. GRHANITE™ enabled data to be extracted in a de-identified  
140 manner by incorporating advanced privacy-preserving hashing techniques to generate unique  
141 'signatures'. These data were then securely transmitted to the VCRR database based on the  
142 University of Melbourne's server, with data stored in Microsoft SQL. Further details regarding data  
143 security and storage can be found in Supplementary File 1 and 2.

#### 144 *Data quality*

145 The system highlighted any GRHANITE™ data extraction failures or omissions and IT  
146 representative at each site reviewed the data to ensure it was coherent before it was forwarded to  
147 the central registry. The REDCap data collection forms contained mandatory fields to reduce missing  
148 data (data must be entered before being able to move to the next section) and in-built logic checks  
149 to increase the accuracy of data. Missing patient records were assessed by comparing the number

150 of patients booked CR appointments in the electronic administrative database (total numbers) with  
151 number of patients manually entered into REDCap.

### 152 *Formative evaluation*

153           Semi-structured interviews were conducted within one week of the completed data  
154 collection period (December 2017) to ascertain any barriers or enablers to implementation of the CR  
155 registry. Individual interviews were held with clinical staff members involved with clinical data  
156 collection at the two pilot sites (N=3). The interviews were conducted by a member of the research  
157 team (ET). They were audio-recorded and then transcribed verbatim except to preserve anonymity.

158           The interview guide consisted of three parts: (i) historical approaches to data collection, (ii)  
159 barriers to measuring and collecting variables and (iii) recommendations for future registry  
160 implementation. Feedback provided by the clinicians was synthesised under the same three  
161 headings and identified barriers were coded in themes and sub-categories using content  
162 analysis[22].

## 163 **Results**

### 164 *Characteristics of patients included in VCRR*

165           The combined electronic and manual data revealed that across the two sites, 176 patients had a  
166 booked CR appointment, 115 patients (65.34%) completed the initial CR appointment and 48  
167 patients (27.27%) completed the CR program (achieved patient goals and/or attended an agreed  
168 number of exercise and education sessions) within the data extraction period. The study sample was  
169 predominantly male (76%) with a mean age of 67 years and 83% spoke English as their preferred  
170 language (Table 2). The participant's sociodemographic characteristics differed across the two sites,  
171 with participants at Site 2 being 10 years older on average (74 years vs. 65 years) and having a lower  
172 baseline exercise capacity (95m less on the six-minute walk test) (Table 2).

### 173 **\*TABLE 2\***

174 **Table 2.** Characteristics of patients included in VCRR

175 Variables available from the electronic hospital administrative databases were limited to  
176 seven (age, sex, postcode, Aboriginal and Torres Strait Islander status, preferred language, CR  
177 booking, referral date) for each of the patients. This is due to hospital administrative databases at  
178 the sites only collecting basic information on patient sociodemographic characteristics and CR  
179 appointment bookings. **Data extracted from the manual entry component (REDCap) enabled  
180 collection of all 42 variables in the minimum data set, supplementing the electronic data.**

### 181 *CR Quality*

182 The minimum variables extracted were useful in informing assessment of CR site quality in  
183 many instances (Table 3). There were site-specific differences in process indicators of care,  
184 suggesting the minimum variables are sensitive. For example, participants in Site 1 experienced a  
185 longer wait time to receive CR (44 days vs. 19 days) and were less likely to be screened for  
186 depression (54% vs. 92%). None of the identified smokers (across either site) were reported to have  
187 been referred for smoking cessation.

188 There was a large amount of missing and unknown data from the manual-entry source.  
189 Discrepancies existed between the number of patients booked CR appointments in the hospital  
190 administrative database (n=176) and those who attended the initial assessment and were entered  
191 into REDCap (n=115). Reasons for non-attendance to the initial session were not routinely collected  
192 and therefore unable to be ascertained for all cases. Further, many values in the post-CR  
193 assessment were reported as unknown (e.g. CR medication status was unknown for 44% of patient  
194 who completed a post-CR assessment).

### 195 **\*TABLE 3\***

196 **Table 3.** CR process indicators

197

### 198 *CR Staff Perceptions of Data Capture Processes*

199 Feedback from the two sites revealed that the manual entry component was straight-forward,  
200 easy to use, and quicker than traditional forms of data collection (i.e., clinician-selected variables



201 entered into an Excel spreadsheet; Table 4). The training provided was perceived as sufficient and  
202 staff felt in-built features such as mandatory fields enabled them to feel more confident about the  
203 data quality. Staff expressed desire to have the capacity to search more easily for entered patient  
204 data (a feature that is available in REDCap but was not highlighted during the training session) and  
205 additional information about the rationale/evidence for some of the selected minimum variables. All  
206 interviewees wanted to continue using REDCap and preferred this approach over traditional  
207 methods; as described by the CR co-ordinator at Site 2 *“I just can see that REDCap is the bright new  
208 future that we can start to get the cardiac rehab product out there with consistency between  
209 programs... Because at the moment we can all say that we are doing cardiac rehab and we can all be  
210 members of ACRA but I don’t know what you’re providing and you don’t know what I am doing  
211 unless you are there”.*

212 Five main barriers were identified regarding historic methods of measuring and entering  
213 variables (see Supplementary file 3): i) workload and competing responsibilities (e.g. time  
214 constraints), ii) environmental context and resources (e.g. information technology issues, and not  
215 having access to a quiet and secure space to enter data); iii) patient factors (e.g. patient  
216 needs/concerns conflicting with data collection requirements); iv) care delivery processes and co-  
217 ordination (e.g. referrals getting lost because sent via post/fax ) and v) outcome expectations (e.g.  
218 reduced confidence in data because of measurement errors).

219 **\*TABLE 4 \***

220 **Table 4.** Feedback from sites on web-based data entry

221

222 **Discussion**

223 To our knowledge, this was the first study to assess the feasibility of utilising a data capture  
224 tool to automatically extract minimum CR registry variables within public and private facilities. While  
225 CR sites collected large amounts of clinical data, the majority of these data (i.e., 83% of the 42

226 variables) were not readily-available in an appropriate electronic format rendering automated data  
227 extraction unfeasible. Until such time that the current infrastructure in public and private CR settings  
228 in Australia develops, the key benefits of scalable, automated data capture tools like GRHANITE™  
229 will remain unrealised. While this approach remains promising for creating and maintaining a  
230 registry that monitors the quality of CR provided to patients, further investment is required in the  
231 digital platforms underpinning this approach including ensuring electronic platforms are i) accessible  
232 to CR sites, ii) fit for purpose and, iii) capturing high quality data. In the interim, a web-based data  
233 collection tool housed on the REDCap system can enable standardised data to be collated from  
234 various CR sites with known limitations associated with manual data entry. These key findings are  
235 discussed further below.

236 Greater emphasis must be placed on ensuring CR staff have access to EMRs[9]. In general, allied  
237 health and community-based settings have had low-levels of adoption of electronic health  
238 infrastructure compared to acute settings and primary care [23]. To ensure more timely access,  
239 national associations such as ACRA, the Cardiac Society of Australia and New Zealand (CSANZ) and  
240 the National Heart Foundation (NHF) need to facilitate advocacy efforts at the local, state and  
241 national-level for improved electronic infrastructure within the CR setting. For example, ACRA could  
242 provide guidance to CR co-ordinators and managers to push the agenda within local settings;  
243 enhanced CR representation on state-based cardiac clinical networks could drive the issue at a state-  
244 level; and the development of a national strategic plan and committee could be established with the  
245 aim of improving monitoring of CR and enhancing national efforts.

246 Future digital health investments will be driven by specific business needs and the identification  
247 and demonstration of local and system-wide benefits[24]. Consequently, a clear business case for  
248 enhanced monitoring of CR is required which details the digital requirements necessary to fulfil the  
249 current gap. Additionally, the workplace will likely need to up-skill to ensure adequate digital  
250 capability. Well-developed and robust change management is a crucial factor in deploying new

251 systems and clinicians must be involved in the process and actively champion health technology  
252 activities [24].

253 Ideally, as EMR uptake increases, all CR minimum variables would be available electronically, and  
254 a registry could be pre-filled. In other countries CR registries have begun to simultaneously link with  
255 administrative electronic databases to enable auto-filling of data (e.g. the Danish registry and  
256 Canadian registry) [17, 25]. In states where different EMR systems are being implemented, flexible  
257 tools like GRHANITE™ will be crucial in enabling interoperability of data across various systems  
258 (including public and private) whilst adhering to privacy and security concerns.

259 Ultimately, the success of data capture through EMRs will depend on multiple factors, including  
260 minimum variables being: i) clearly defined, ii) entered consistently across sites, iii) of sufficient  
261 reliability/validity, and iv) extractable. The CR field can begin to prepare for this now by ensuring  
262 quality indicators are clearly defined and comparable across states.

263 In the interim, CR data collection can be improved via the use of a standardised web-based  
264 tool housed on platforms such as the REDCap system. REDCap had multiple advantages including: i)  
265 ease of implementation without any need for the sites' IT departments, ii) usable at both public and  
266 private CR sites, iii) secure and password-protected access, iv) straight-forward and quick data  
267 entry, v) in-built functions (e.g. mandatory fields, character limits, drop down options, automated  
268 reports) to enhance data quality and completeness, vi) available for use at no costs for affiliated  
269 research institutes. Further, REDCap was supported by those entering the data who expressed an  
270 interest in continuing beyond the study period.

271 Use of the web-based tool, however, could be enhanced. For example, future studies should  
272 incorporate data quality checks early in the data collection period that include a comparison of  
273 enrolled and entered patient data to ensure such data match and reasons for missing data are  
274 ascertained. In Australia CR sites often refer patients to more convenient programs (e.g. closer to  
275 home); such information needs to be captured on all patients so that reasons for non-attendance

276 can be more accurately documented. Additionally, unknown data requires additional clarification.  
277 For example, post-CR medication status had larger amounts of unknown responses than other  
278 variables and is potentially not being checked at post-CR interviews. Automated alerts could be in-  
279 built for this variable to clarify the reason for the unknown information.

## 280 **Study limitations**

281 We acknowledge that this study has limitations. Due to the small sample size and Victorian  
282 setting, results from this feasibility study may not be generalizable to other settings and saturation  
283 of themes in the staff interviews were not realised. Additionally, the ‘snap-shot’ method of data  
284 collection meant that many patients had not completed CR at the time of data extraction. Further,  
285 enhanced methods are required to ensure all who enrolled into the CR programs were captured  
286 even if they did not attend the initial assessment session to reduce reporting bias towards CR  
287 attenders.

## 288 **Implications and future recommendations**

289 The transition to digital health systems holds great potential for enhancing clinical care within  
290 the CR setting. However, many jurisdictions have been slow to adopt e-health infrastructure limiting  
291 the application of tools like GRHANITE™. Key organisations need to advocate for EMRs in CR  
292 programs so that automated data-capture technologies can increase the viability of CR registries in  
293 the future. Efforts must also focus on preparing the field for the digital transition and preparing a  
294 clear business case delineating the local- and system-wide benefits and the digital requirements so  
295 systems are built in a way that is fit for purpose.

296 In the interim, a web-based data entry tool shows promise as an approach that should be  
297 explored further and could enable the monitoring of CR quality across the private and public sector.

298

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30 | patients, further investment is required in the digital platforms underpinning this approach.

31 | **Key words:** cardiac rehabilitation; registry; data scraping

## 32 | Introduction

33 | The ability to quantify healthcare quality relies on the implementation of appropriate systems  
34 | that can accurately capture how care is being delivered [1]. In a recent scientific statement, the  
35 | American Heart Association called for the ~~The need to~~ systematically redesign of cardiovascular care  
36 | to ~~be a~~ enable a 'learning healthcare system' which uses information technology and data  
37 | infrastructures to enhance optimal healthcare delivery ~~has recently been highlighted in a Scientific~~  
38 | ~~Statement~~[2]. In Australia, the Commission on Safety and Quality of Health Care (the Commission)  
39 | promotes the use of clinical registries to systematically monitor healthcare, highlight variations in  
40 | outcomes, and inform quality improvement efforts [3]. Ischaemic heart disease ranks as the highest  
41 | priority area identified by the Commission that would benefit from registry development due to the  
42 | high burden of disease, serious consequences associated with poor quality care and strong clinical  
43 | support ~~and the existence of a current national registry (Australian Cardiac Outcome Registry) that~~  
44 | ~~could be expanded in the future to include non-surgical interventions~~[4]. This follows the success of  
45 | cardiac registries internationally including the Global Registry of Acute Coronary Events (GRACE)[5]  
46 | and effective system-wide changes seen by countries such as Sweden which has established more  
47 | than 100 health registries including some that have been maintained for more than 25 years [6].

48 | A key component of secondary prevention of heart disease is cardiac rehabilitation (CR).  
49 | Although CR is extremely effective in preventing cardiovascular recurrent events and complications

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[7] and recommended in clinical guidelines [8, 9], there is variability in program delivery and quality [10] some of which stems from a lack of uniform data collection and monitoring systems. The [need to develop quality indicators and implement systems that collect standardised CR outcome data is recognised by several national associations internationally](#) [11-13]. ~~National Heart Foundation of Australia recognises the need to “develop national key performance indicators for secondary prevention services and implement systems to collect standardised outcome data” [11, 12] including . Moreover, evaluation and quality improvement has been identified as a core component in the delivery of comprehensive CR programs by~~ the Australian Cardiovascular Health and Rehabilitation Association’s (ACRA; the [national Australian](#) association of CR professionals) [14]. Specifically, ACRA recommend that all CR services collect a minimum set of data and report on key performance indicators to promote continuous quality improvement of services and benchmarking [14]. Despite these calls, quality indicator data from CR sites are, for the most part, not systematically collected or collated. [One jurisdiction in Australia, Queensland, has recently established the](#) ~~The~~ Queensland Cardiac Outcomes Registry (QCOR) [has recently expanded to which](#) includes the collection of CR quality indicator variables as part of the registry and will be the first state in Australia to systematically collect CR data [15]. In [the state of](#) Victoria, the Victorian Cardiac Outcomes Registry (VCOR) [16] collects data on cardiac patients across 35 hospitals on three modules (percutaneous coronary intervention, heart failure and the early treatment of acute myocardial infarction). However, CR data are not included within VCOR.

[Globally,](#) ~~c~~ Custodians of CR registries have noted challenges, common to any registry, such as site investment or ‘buy-in’, privacy and security considerations, as well as limited resources for contributing data [17]. Indeed, sites are often required to manually enter data, which is time-consuming for clinical staff and increases the risk of data errors [18]. Ideally, data collection should be automated and linked to administrative databases or electronic medical records (EMRs). With advances in technology, this is becoming more feasible. Automated data capture techniques using specially-designed software can be used to extract routinely-collected data. Such software can also

76 incorporate automated safeguards built-in to the data entry systems to ensure privacy protection.  
77 This has been previously demonstrated within primary care and other settings in Australia [19] using  
78 the GRHANITE™ (GeneRic Health Network Information for the Enterprise [20]:  
79 <https://www.grhanite.com/> [15]) tool.

80 ~~Accordingly,~~The aim of this manuscript was to ~~we~~ assessed the feasibility of extracting  
81 routinely-collected minimum data (as defined by the NSW division of ACRA [21]) from CR sites and  
82 hospital administration databases using the GRHANITE™ automated data capture tool in order to  
83 populate a Victorian CR Registry (VCRR).

## 84 **Methods**

### 85 Overarching design of VCRR

86 This feasibility study consisted of a 3-month (September-December 2017) data collection  
87 period involving quantitative data capture from two pilot sites and formative evaluation of the  
88 process including feedback from CR clinicians. The design of the registry was guided by technical  
89 standards outlined by the Commission [3], as illustrated in a logic model (Figure 1).

### 90 **\*FIGURE 1\***

91  
92 ~~Figure 1~~ Clinical Quality Registries Information Model [16]. Reproduced with permission from Logical Design  
93 for Australian Clinical Quality Registries, developed by the Australian Commission on Safety and Quality in  
94 Health Care (ACSQHC), for use exclusively in Australia. ACSQHC: Sydney. 2012.  
95 ~~Australian Commission on Safety and Quality in Health Care's (2012) [16] Clinical Quality Registries Information~~  
96 ~~Model~~

97 Acronyms: CQR: clinical quality registry; MBS: Medicare Benefits Schedule; PBS: Pharmaceutical Benefits  
98 Scheme

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100 Selection of the minimum variables for the VCRR

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101 The registry comprised a minimum set of variables selected from the New South Wales  
102 (NSW) ACRA association quality indicators and data dictionary which was based on expert  
103 consensus [21]. The 42 selected data elements consisted of: demographic information (e.g. sex, age),  
104 disease/condition (e.g. principal referral diagnosis) risk factors and co-morbidities (e.g. diabetes  
105 status, smoking status), intervention (e.g. number of CR sessions), process indicators (e.g. CR wait  
106 time) and individual patient outcomes (e.g. change in pre-post exercise capacity) (Table 1).

107 **\*TABLE 1 \***

108 **Table 1.** Victorian Cardiac Rehabilitation Registry minimum variables

109 Setting and recruitment

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110 In the state of Victoria in South East Australia, there are 136 CR programs, delivered across  
111 publicly and privately-funded hospitals and community health settings. TACRA - the national  
112 association of CR professionals (ACRA) has a State-level directory of all CR facilities which was used  
113 to identify one public and one private site to invite to participate in the study. These sites were  
114 purposively selected to ensure sample representation of: funding sources (public and private),  
115 settings (acute hospital, rehabilitation hospital), and location (metropolitan and suburban). Site 1  
116 was a large publicly-funded program, which runs a six-week CR program for approximately 40  
117 outpatients per week. Site 2 was a private facility primarily funded through health insurance funds  
118 and the Department of Veteran Affairs, which runs a 12-week program for approximately 15  
119 outpatients per week. Participating sites were offered a stipend of AU\$6,000 (USD\$4700) to cover  
120 cost related to staff time for the set-up of automated data collection. Both CR sites agreed to  
121 participate.

122 Ethics approval

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123 The study was approved by the Human Research Ethics Committee (HREC) at the University  
124 of Melbourne (HREC number: 1748609) and included a waiver of consent for individual patient data

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125 (which was de-identified). Site-specific research ethics approval was also obtained. Staff who  
126 participated in qualitative interviews provided informed consent.

127 Automated Data collection procedure (GRHANITE™)

128 The team at the University of Melbourne’s Health and Biomedical Informatics Centre  
129 Research Information Technology Unit (led by DB) assisted in the development of the data extraction  
130 protocol and worked with the sites’ Information Technology (IT) teams to create an interface regime.  
131 This required the development of a “mapping” document which linked the variables requested from  
132 the research team with the variables collected and available electronically at the sites. The overview  
133 of the study methods can be seen in Figure 2.

134 **\*FIGURE 2\***

135 Figure 2 Overview of the study methods

136  
137 Manual Amendment to the study protocol to add a manual data entry component web-based data  
138 collection (REDCap)

139 In order to To capture variables that were not available electronically at the sites, a secure  
140 web-based data collection form was designed using the REDCap (Research Electronic Data Capture:  
141 <https://www.project-redcap.org/>) software. The amendment was approved by the University of  
142 Melbourne’s HREC in July 2017. The web-based form included three sections (Section 1: identifiable  
143 patient information; Section 2: pre-CR data; Section 3: post-CR data) and was trialled for two weeks  
144 at both sites, with feedback from the CR sites informing refinement of the data entry template. Once  
145 finalised, clinicians entered data for patients who were enrolled in the CR programs during the data  
146 collection period. The REDCap data collection forms contained mandatory fields to reduce missing  
147 data and in-built logic checks to increase the accuracy of data. Authorised staff were provided with a  
148 secure log-in which enabled access to the REDCap template; d. Data access restrictions ensured

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149 [clinicians could only view data from their site. Additional detail on REDCap is provided in](#)

150 [Supplementary File 2.](#)

151

152 *Data extraction and linkage*

153 CR data were extracted from the sites via the University of Melbourne’s GRHANITE™  
154 research data acquisition system. The GRHANITE™ interface was installed at both sites and  
155 scheduled to extract pre-determined variables on patients who participated in the CR program  
156 during the data collection period. GRHANITE™ enabled data to be extracted in a de-identified  
157 manner by incorporating advanced privacy-preserving hashing techniques to generate unique  
158 ‘signatures’. These data were then securely transmitted to the VCRR database based on the  
159 University of Melbourne’s server, with data stored in Microsoft SQL. Further details regarding data  
160 security and storage can be found in Supplementary File 1 [and 2](#).

161 *Data quality*

162 The system highlighted any GRHANITE™ data extraction failures or omissions and IT  
163 representative at each site reviewed the data to ensure it was coherent before it was forwarded to  
164 the central registry. [The REDCap data collection forms contained mandatory fields to reduce missing  
165 data \(data must be entered before being able to move to the next section\) and in-built logic checks  
166 to increase the accuracy of data. Missing patient records were assessed by comparing the number  
167 of patients booked CR appointments in the electronic administrative database \(total numbers\) with  
168 number of patients manually entered into REDCap.](#)

169 *Formative evaluation*

170 Semi-structured interviews were conducted within one week of the completed data  
171 collection period (December 2017) to ascertain any barriers or enablers to implementation of the CR  
172 registry. Individual interviews were held with clinical staff members involved with clinical data  
173 collection at the two pilot sites (N=3). The interviews were conducted by a member of the research  
174 team (ET). They were audio-recorded and then transcribed verbatim except to preserve anonymity.

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175 | The interview guide consisted of three parts: (i) historical approaches to data collection, (ii)  
176 | barriers to measuring and collecting variables and (iii) recommendations for future registry  
177 | implementation. Feedback provided by the clinicians was synthesised under the same three  
178 | headings and identified barriers were coded in themes and sub-categories using content  
179 | analysis[22].

## 180 | **Results**

### 181 | Characteristics of patients included in VCRR

182 | The combined electronic and manual data revealed that across the two sites, 176 patients had a  
183 | booked CR appointment, 115 patients (65.34%) completed the initial CR appointment and 48  
184 | patients (27.27%) completed the CR program (achieved patient goals and/or attended an agreed  
185 | number of exercise and education sessions) within the data extraction period. The study sample was  
186 | predominantly male (76%) with a mean age of 67 years and 83% spoke English as their preferred  
187 | language (Table 2). The participant's sociodemographic characteristics differed across the two sites,  
188 | with participants at Site 2 being 10 years older on average (74 years vs. 65 years) and having a lower  
189 | baseline exercise capacity (95m less on the six-minute walk test) (Table 2).

### 190 | \*TABLE 2\*

#### 191 | Table 2. Characteristics of patients included in VCRR

192 | Variables available from the electronic hospital administrative databases were limited to  
193 | seven (age, sex, postcode, Aboriginal and Torres Strait Islander status, preferred language, CR  
194 | booking, referral date) for each of the patients. This is due to hospital administrative databases at  
195 | the sites only collecting basic information on patient sociodemographic characteristics and CR  
196 | appointment bookings. ~~The remaining clinical information selected for the CR registry minimum data~~  
197 | ~~set were collected on paper based records and manually transferred by clinicians onto an Excel~~  
198 | ~~spreadsheet or scanned into patient records and deemed unusable for electronic data capture.~~  
199 | ~~Amendment to the study protocol to add a manual data entry component~~

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200 ~~In order to capture variables that were not available electronically at the sites, a secure web-~~  
201 ~~based data collection form was designed using the REDCap (Research Electronic Data Capture;~~  
202 ~~<https://www.project-redcap.org/>) software. The amendment was approved by the University of~~  
203 ~~Melbourne's HREC in July 2017. The web-based form included three sections (Section 1: identifiable~~  
204 ~~patient information; Section 2: pre-CR data; Section 3: post-CR data) and was trialed for two weeks~~  
205 ~~at both sites, with feedback from the CR sites informing refinement of the data entry template. Once~~  
206 ~~finalised, clinicians entered data for patients who were enrolled in the CR programs during the data~~  
207 ~~collection period. The REDCap data collection forms contained mandatory fields to reduce missing~~  
208 ~~data and in-built logic checks to increase the accuracy of data. Authorised staff were provided with a~~  
209 ~~secure log in which enabled access to the REDCap template. Data access restrictions ensured~~  
210 ~~clinicians could only view data from their site. Additional detail on REDCap is provided in~~  
211 ~~Supplementary File 2.~~

212 ~~Combining electronic data and REDCap data extracts via GRHANITE™~~  
213 ~~The GRHANITE™ data capture software was configured to extract data from both the electronic data~~  
214 ~~(from hospital administrative databases) and manually entered clinical data (from REDCap) into the~~  
215 ~~study database hosted on the University of Melbourne's server and secured within the University's~~  
216 ~~IT infrastructure. The unique 'signatures' generated by GRHANITE™ enabled anonymous record~~  
217 ~~linkage between the electronic and manually entered data. Data extracted from the manual entry~~  
218 ~~component (REDCap) enabled collection of all 42 variables in the minimum data set, supplementing~~  
219 ~~the electronic data. The overview of the amended study methods can be seen in Figure 2.~~

220 **\*FIGURE 2\***

221 **Figure 2 Overview of amended study methods**

222 **Characteristics of patients included in VCRB**

223  
224 The combined electronic and manual data revealed that across the two sites, 176 patients had a  
225 booked CR appointment, 115 patients (65.34%) completed the initial CR appointment and 48

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226 patients (27.27%) completed the CR program (achieved patient goals and/or attended an agreed  
227 number of exercise and education sessions) within the data extraction period. The study sample was  
228 predominantly male (76%) with a mean age of 67 years and 83% spoke English as their preferred  
229 language (Table 2). The participant's sociodemographic characteristics differed across the two sites,  
230 with participants at Site 2 being 10 years older on average (74 years vs. 65 years) and having a lower  
231 baseline exercise capacity (95m less on the six-minute walk test) (Table 2).

232 **\*TABLE 2\***

233 **Table 2.** Characteristics of patients included in VCRP

234  
235 **CR Quality**

236 The minimum variables extracted were useful in informing assessment of CR site quality in  
237 many instances (Table 3). There were site-specific differences in process indicators of care,  
238 suggesting the minimum variables are sensitive. For example, participants in Site 1 experienced a  
239 longer wait time to receive CR (44 days vs. 19 days) and were less likely to be screened for  
240 depression (54% vs. 92%). None of the identified smokers (across either site) were reported to have  
241 been referred for smoking cessation.

242 There was a large amount of missing and unknown data from the manual-entry source.  
243 Discrepancies existed between the number of patients booked CR appointments in the hospital  
244 administrative database (n=176) and those who attended the initial assessment and were entered  
245 into REDCap (n=115). Reasons for non-attendance to the initial session were not routinely collected  
246 and therefore unable to be ascertained for all cases. Further, many values in the post-CR  
247 assessment were reported as unknown (e.g. CR medication status was unknown for 44% of patient  
248 who completed a post-CR assessment).

249  
250 **\*TABLE 3\***

251 **Table 3.** CR process indicators

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252

253 CR Staff Perceptions of Data Capture Processes

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254 Feedback from the two sites revealed that the manual entry component was straight-forward,  
255 easy to use, and quicker than traditional forms of data collection (i.e., clinician-selected variables  
256 entered into an Excel spreadsheet; Table 4). The training provided was perceived as sufficient and  
257 staff felt in-built features such as mandatory fields enabled them to feel more confident about the  
258 data quality. Staff expressed desire to have the capacity to search more easily for entered patient  
259 data (a feature that is available in REDCap but was not highlighted during the training session) and  
260 additional information about the rationale/evidence for some of the selected minimum variables. All  
261 interviewees wanted to continue using REDCap and preferred this approach over traditional  
262 methods; a-s described by the CR co-ordinator at Site 2. "I just can see that REDCap is the bright new  
263 future that we can start to get the cardiac rehab product out there with consistency between  
264 programs... Because at the moment we can all say that we are doing cardiac rehab and we can all be  
265 members of ACRA but I don't know what you're providing and you don't know what I am doing  
266 unless you are there".

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267 Five main barriers were identified regarding historic methods of measuring and entering  
268 variables (see Supplementary file 3): i) workload and competing responsibilities (e.g. time  
269 constraints), ii) environmental context and resources (e.g. information technology issues, and not  
270 having access to a quiet and secure space to enter data); iii) patient factors (e.g. patient  
271 needs/concerns conflicting with data collection requirements); iv) care delivery processes and co-  
272 ordination (e.g. referrals getting lost because sent via post/fax ) and v) outcome expectations (e.g.  
273 reduced confidence in data because of measurement errors).

274 **\*TABLE 4 \***

275 **Table 4.** Feedback from sites on web-based data entry

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276

277 **Discussion**

278 To our knowledge, this was the first study to assess the feasibility of utilising a data capture  
279 tool to automatically extract minimum CR registry variables within public and private facilities ~~in~~  
280 ~~Australia~~. While CR sites collected large amounts of clinical data, the majority of these data (i.e., 83%  
281 of the 42 variables) were not readily-available in an appropriate electronic format rendering  
282 automated data extraction unfeasible. Until such time that the current infrastructure in public and  
283 private CR settings in Australia develops, the key benefits of scalable, automated data capture tools  
284 like GRHANITE™ will remain unrealised. While this approach remains promising for creating and  
285 maintaining a registry that monitors the quality of CR provided to patients, further investment is  
286 required in the digital platforms underpinning this approach including ensuring electronic platforms  
287 are i) accessible to CR sites, ii) fit for purpose and, iii) capturing high quality data. In the interim, a  
288 web-based data collection tool housed on the REDCap system can enable standardised data to be  
289 collated from various CR sites with known limitations associated with manual data entry. These key  
290 findings are discussed further below.

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#### 291 Enhancing access and use of EMRs

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292 Greater emphasis must be placed on ensuring CR staff have access to EMRs[9]. In general, allied  
293 health and community-based settings have had low-levels of adoption of electronic health  
294 infrastructure compared to acute settings and primary care [23]. To ensure more timely access,  
295 national associations such as ACRA, the Cardiac Society of Australia and New Zealand (CSANZ) and  
296 the National Heart Foundation (NHF) need to facilitate advocacy efforts at the local, state and  
297 national-level for improved electronic infrastructure within the CR setting. For example, ACRA could  
298 provide guidance to CR co-ordinators and managers to push the agenda within local settings;  
299 enhanced CR representation on state-based cardiac clinical networks could drive the issue at a state-  
300 level; and the development of a national strategic plan and committee could be established with the  
301 aim of improving monitoring of CR and enhancing national efforts.

302 ~~Within Victoria (and likely other states) Future,~~ digital health investments will be driven by  
303 specific business needs and the identification and demonstration of local and system-wide  
304 benefits[24]. Consequently, a clear business case for enhanced monitoring of CR is required which  
305 details the digital requirements necessary to fulfil the current gap. Additionally, the workplace will  
306 likely need to up-skill to ensure adequate digital capability. Well-developed and robust change  
307 management is a crucial factor in deploying new systems and clinicians must be involved in the  
308 process and actively champion health technology activities\_[24].

### 309 ~~Ensuring EMRs are fit for purpose~~

310 Ideally, as EMR ~~uptake increases~~ ~~develop in Australia~~, all CR minimum variables would be  
311 available electronically, and a registry could be pre-filled. In other countries CR registries have begun  
312 to simultaneously link with administrative electronic databases to enable auto-filling of data (e.g. the  
313 Danish registry and Canadian registry)\_[17, 25]. In states where different EMR systems are being  
314 implemented, flexible tools like GRHANITE™ will be crucial in enabling interoperability of data across  
315 various systems (including public and private) whilst adhering to privacy and security concerns.

316 Ultimately, the success of data capture through EMRs will depend on multiple factors, including  
317 minimum variables being: i) clearly defined, ii) entered consistently across sites, iii) of sufficient  
318 reliability/validity, and iv) extractable. The CR field can begin to prepare for this now by ensuring  
319 quality indicators are clearly defined and comparable across states.

### 320 ~~Monitoring CR in settings with under-matured electronic platforms~~

321 ~~Many states are a long way from having fully integrated electronic health systems. Between~~  
322 ~~2004-2013 Victoria invested over \$300 million to reform the IT ecosystem with the HealthSMART~~  
323 ~~initiative which was eventually abandoned due to a 'one-size-fit-all approach' being~~  
324 ~~unsuccessful[21]. Consequently, the responsibility of developing digital solutions was placed back on~~  
325 ~~health services providers resulting in a wide range of clinical information systems implemented to~~

326 ~~varying degrees across hospitals and health centres[22]. Many CR sites have no access to EMRs and~~  
327 ~~as demonstrated in this feasibility study are relying on paper-based data collection methods.~~

328 In the interim, CR data collection can be improved via the use of a standardised web-based  
329 tool housed on platforms such as the REDCap system. REDCap had multiple advantages including: i)  
330 ease of implementation without any need for the sites' IT departments, ii) usable at both public and  
331 private CR sites, iii) secure and password-protected access, iv) straight-forward and quick data  
332 entry, v) in-built functions (e.g. mandatory fields, character limits, drop down options, automated  
333 reports) to enhance data quality and completeness, vi) available for use at no costs for affiliated  
334 research institutes. Further, REDCap was supported by those entering the data who expressed an  
335 interest in continuing beyond the study period.

336 Use of the web-based tool, however, could be enhanced. For example, future studies should  
337 incorporate data quality checks early in the data collection period that include a comparison of  
338 enrolled and entered patient data to ensure such data match and reasons for missing data are  
339 ascertained. In Australia CR sites often refer patients to more convenient programs (e.g. closer to  
340 home); such information needs to be captured on all patients so that reasons for non-attendance  
341 can be more accurately documented. Additionally, unknown data requires additional clarification.  
342 For example, post-CR medication status had larger amounts of unknown responses than other  
343 variables and is potentially not being checked at post-CR interviews. Automated alerts could be in-  
344 built for this variable to clarify the reason for the unknown information.

#### 345 **Study limitations**

346 We acknowledge that this study has limitations. Due to the small sample size and Victorian  
347 setting, results from this feasibility study ~~may not be~~ ~~are not~~ generalizable to other settings and  
348 saturation of themes in the staff interviews were not realised. Additionally, the 'snap-shot' method  
349 of data collection meant that many patients had not completed CR at the time of data extraction.  
350 Further, enhanced methods are required to ensure all who enrolled into the CR programs were

351 captured even if they did not attend the initial assessment session to reduce reporting bias towards  
352 CR attenders.

### 353 **Implications and future recommendations**

354 The transition to digital health systems holds great potential for enhancing clinical care within  
355 the CR setting. However, many jurisdictions have been slow to adopt e-health infrastructure limiting  
356 the application of tools like GRHANITE™. Key organisations need to advocate for EMRs in CR  
357 programs so that automated data-capture technologies can increase the viability of CR registries in  
358 the future. Efforts must also focus on preparing the field for the digital transition and preparing a  
359 clear business case delineating the local- and system-wide benefits and the digital requirements so  
360 systems are built in a way that is fit for purpose.

361 In the interim, a web-based data entry tool shows promise as an approach that should be  
362 explored further and could enable the monitoring of CR quality across the private and public sector.

363

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**Table 1.** VCRR minimum variables

CORE DATA		
Person identifying information	1.	Name
	2.	Medicare number
	3.	Patient Unit Record number
	4.	Date of birth
	5.	Sex
	6.	Postcode
	7.	Culturally and linguistically diverse (CALD)
	8.	Aboriginal and Torres Strait Islander status
Provider organization	9.	Service provider name
CQR SPECIFIC DATA		
Disease/condition	10	Principal CR referral diagnosis
Risk factors and co-morbidities (for risk adjustment)	11	Interventions/complications (e.g. PCI, CABG)
	12	Diabetes diagnosis
	13	Smoking status
	14-18	Prescribed medications (i. oral antiplatelet, ii. Beta-blockers, iii. ACE-I/ARB, iv. lipid-lowering, v. sublingual nitrate)
	19	Waist circumference
	20	Exercise capacity
Intervention	21	CR program model
	22	CR referral date
	23	CR commencement date
	24	Number of CR sessions attended
	25	CR completion status



	26	Reason for CR withdrawal (if applicable)
Process indicators of evidence based care	27	CR wait time (CR commencement date – CR referral date)
	28	Screened for depression
	29	Positive cases for depression referred for management
	30	Current/recent smokers referred or provided with smoking cessation advice
	31-35	Prescribed medications (i. oral antiplatelet, ii. Beta-blockers, iii. ACE-I/ARB, iv. lipid-lowering, v. sublingual nitrate)
	36	Provided a symptom-management plan
	37-40	Referred for ongoing care (i. General Practitioner, ii. specialist/Cardiologist, iii. CR follow-up, iv. Phase 3 CR or equivalent)
Individual patient	41	Pre-post change in exercise capacity
outcome measures	42	Pre-post change in waist circumference

Acronyms: ACE-1: angiotensin-converting enzyme; ARB: angiotensin receptor blockers; CR: cardiac rehabilitation

**Table 2.** Characteristics of patients included in the VCRR

	SITE 1	SITE 2	Total	Missing %
	Freq (%);	Freq (%);	Freq (%);	
	Mean [SD]	Mean [SD]	Mean [SD]	
	n=131	n=45	n=176	
Male	99 (75.57)	35 (77.78)	134 (76.14)	0
Age (years)	64.96 [11.82]	74.11 [9.21]	67.30 [11.88]	0
Aboriginal or Torres				
Strait Islander	1 (0.76)	0 (0)	1 (0.57)	0
English not preferred				
language	30 (22.90)	0 (0)	30 (17.04)	0
Referral indication				34.65*
STEMI	20 (15.26)	4 (8.89)	24 (13.64)	
NSTEMI	14 (10.68)	1 (2.22)	15 (8.52)	
CT surgery	37 (28.24)	9 (20.00)	46 (26.14)	
Interventions				34.65*
Non-elective PCI	19 (14.50)	4 (8.89)	23 (13.07)	
Elective PCI	30 (22.90)	6 (13.33)	36 (20.45)	
CT surgery	37 (28.24)	4 (8.89)	41 (23.29)	
Diabetic	25 (19.01)	7 (15.55)	32 (18.18)	34.65*
Smoker	8 (6.11)	1 (2.22)	9 (5.11)	34.65*
Exercise capacity†	480.50 [93.22]	383.91 [126.89]	456.61 [110.11]	47.16*

Acronyms: CT: cardiothoracic; Freq: frequency; NSTEMI: non-ST elevated myocardial infarction; PCI: percutaneous coronary intervention; SD: standard deviation; STEMI: ST-elevated myocardial infarction.

\*Manually entered data had missing variables; † six-minute walk test

**Table 3.** Process indicators of evidence based care

Process indicator	SITE 1	SITE 2	Total	Unknown/ missing*
	Freq (%); Mean [SD] n=89 <sup>†</sup>	Freq (%); Mean [SD] n=26 <sup>†</sup>	Freq (%); Mean [SD] n=115 <sup>†</sup>	Freq (%) n=115 <sup>†</sup>
CR wait time (days)	44.26 [22.53]	19.21 [19.46]	38.94 [24.13]	2 (1.74)
Screened for depression	48 (53.93)	24 (92.31)	72 (62.61)	43 (37.39)
Positive case for depression referred	1 (2.38) <sup>‡</sup>	2 (22.22) <sup>‡</sup>	3 (5.88) <sup>‡</sup>	43 (84.31) <sup>‡</sup>
No. of smokers	8 (8.99)	1 (3.85)	9 (7.83)	2 (1.74)
Smokers referred for cessation	0 (0) <sup>§</sup>	0 (0) <sup>§</sup>	0 (0) <sup>§</sup>	3 (33.33) <sup>§</sup>
Post-CR medications				
Antiplatelet	21 (23.60)	20 (76.92)	41 (35.65)	74 (64.35)
Beta-blockers	18 (20.22)	12 (46.15)	30 (26.09)	75 (65.22)
ACE-I/ARB	14 (15.73)	10 (38.46)	24 (20.87)	75 (65.22)
Lipid-lowering	21 (23.60)	13 (50.00)	34 (29.57)	75 (65.22)
Sublingual nitrate	11 (12.36)	5 (19.23)	16 (13.91)	75 (65.22)
Provided a symptom - management plan	48 (53.93)	22 (84.61)	70 (60.87)	45 (39.13)
Referred for ongoing care	44 (49.44)	24 (92.31)	68 (59.13)	47 (40.87)

Acronyms: ACE-1: angiotensin-converting enzyme; ARB: angiotensin receptor blockers; CR: cardiac rehabilitation; Freq: frequency; SD: standard deviation.

\*These data were part of a prospective 3-month snap-shot, as such not all data were known at the time of data extraction highlighting issues using these data to compare sites; <sup>†</sup>Denominator = number of patient records entered into REDCap; <sup>‡</sup>Denominator = number of patients screened positive for depression; <sup>§</sup>Denominator = number of identified smokers

**Table 4.** Feedback from sites on use of web-based data entry

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*How sites were traditionally collecting clinical information about CR participants*

- paper-based medical notes or hard copy worksheets
- data manually transferred into an Excel spreadsheet when time allowed
- collected variables were determined individually by the sites and relied on clinician knowledge of CR 'best practice' and influenced by management requirements

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*Identified issues with traditional methods of data collection*

- time consuming
- unnecessary data collected (i.e. not used in analysis or reporting)
- analysis of data in Excel was challenging
- unable to compare data across sites
- collected data was influenced by patient needs, time constraints and perceived importance of the clinical information

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*Experience using the REDCap web-based standardized templates*

- straight-forward and easy
- data entry was quick
- training was sufficient
- appreciated quick responses if any questions arose
- reports more professional compared to Excel

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*Future use of the REDCap web-based templates*

- potential to improve the consistency between CR programs
  - expressed desire to continue using REDCap
  - staff wanted to be able to search more easily for previously entered patients
  - additional evidence/rationale behind why certain variables were selected as the minimum data is required
  - would like available data to be automatically imported from hospital databases
  - would like to enter data during the patient assessment (e.g. via an I-Pad)
-

**Figure 1**  
[Click here to download high resolution image](#)

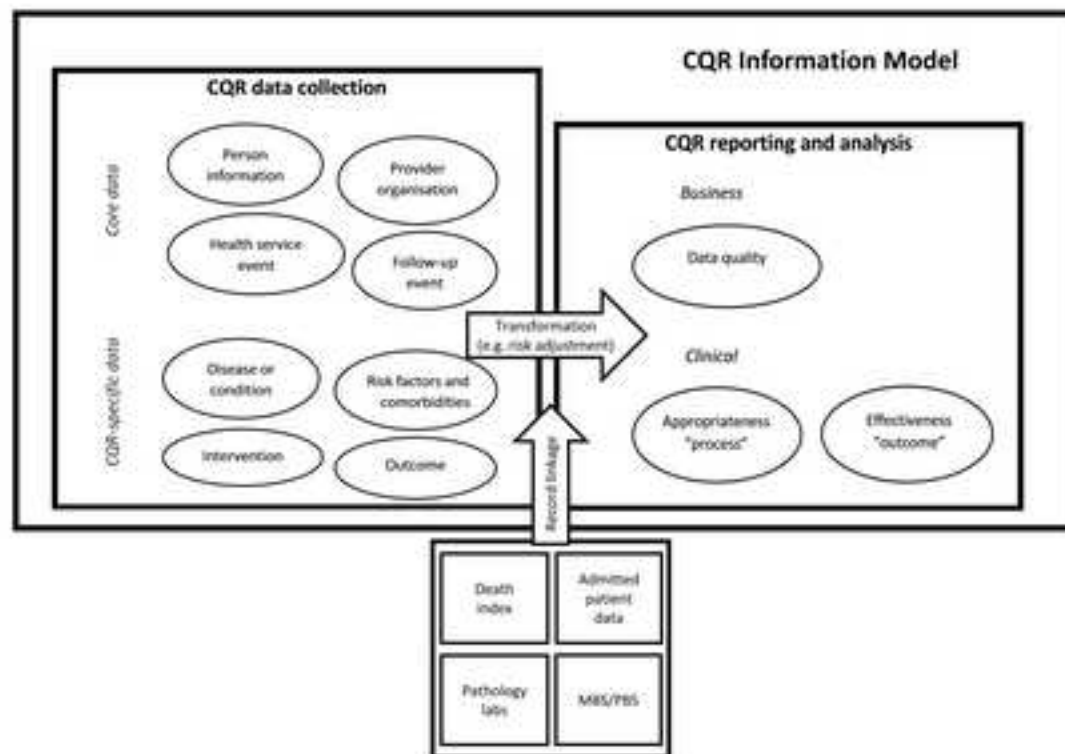
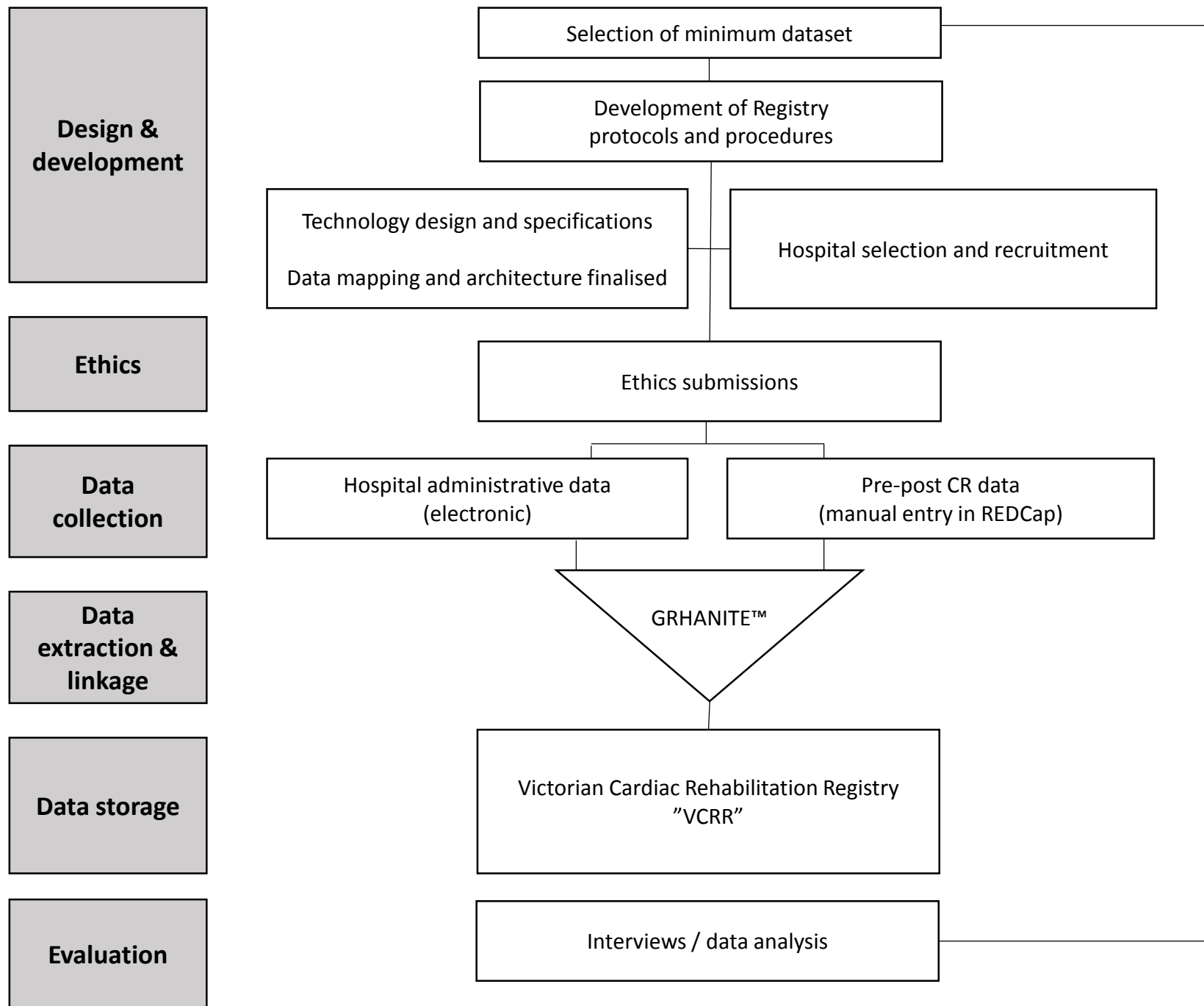


Figure 2



**Supplementary Material**

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