Can we understand population healthcare needs using electronic medical records?

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INTRODUCTION The identification of population-level healthcare needs using hospital electronic medical records (EMRs) is a promising approach for the evaluation and development of tailored healthcare services. Population segmentation based on healthcare needs may be possible using information on health and social service needs from EMRs. However, it is currently unknown if EMRs from restructured hospitals in Singapore provide information of sufficient quality for this purpose. We compared the inter-rater reliability between a population segment that was assigned prospectively and one that was assigned retrospectively based on EMR review.

METHODS 200 non-critical patients aged ≥ 55 years were prospectively evaluated by clinicians for their healthcare needs in the emergency department at Singapore General Hospital, Singapore. Trained clinician raters with no prior knowledge of these patients subsequently accessed the EMR up to the prospective rating date. A similar healthcare needs evaluation was conducted using the EMR. The inter-rater reliability between the two rating sets was evaluated using Cohen's Kappa and the incidence of missing information was tabulated.

RESULTS The inter-rater reliability for the medical 'global impression' rating was 0.37 for doctors and 0.35 for nurses. The inter-rater reliability for the same variable, retrospectively rated by two doctors, was 0.75. Variables with a higher incidence of missing EMR information such as 'social support in case of need' and 'patient activation' had poorer inter-rater reliability.

CONCLUSION Pre-existing EMR systems may not capture sufficient information for reliable determination of healthcare needs. Thus, we should consider integrating policy-relevant healthcare need variables into EMRs.

Keywords: electronic health records, healthcare evaluation mechanisms, needs assessment, patient-centred care

INTRODUCTION

Singapore is ageing at an unprecedented rate. The proportion of the Singapore population aged 65 years and above will increase from 8.4% in 2005 to 18.7% in 2030.⁽¹⁾ In this era of increasing healthcare system burden, the development of tailored packages of services for distinct segments based on population needs holds significant potential for facilitating cost-effective, value-based and patient-centred care.⁽²⁻⁴⁾ It is important to tailor healthcare services to healthcare needs, given that having insufficient services leads to unmet needs and worse clinical outcomes, while excessive or redundant services likely increase cost without improving health.⁽⁵⁻⁷⁾

There are two possible methods to efficiently obtain information on population-level healthcare needs. The first entails prospective collection of healthcare needs information using meso-level information on patient healthcare needs. This refers to, for example, 'whether patient has a functional deficit', as opposed to micro-level information detailing whether it is a deficit in ambulation, dressing or self-feeding. The Simple Segmentation Tool (SST) is one such instrument that can be used by clinicians to capture meso-level information; when aggregated, a snapshot of population-level healthcare needs is obtained. Research that formed the basis of the SST advocated healthcare needs-based

population segmentation⁽³⁾ and the inclusion of variables that were both predictive of future healthcare utilisation and informative for planning services at transitional points of care, such as physical function⁽⁸⁾ and social support level.⁽⁹⁾ The SST has been validated in terms of inter-rater reliability, as well as convergent and predictive validity in an outpatient setting. At the time of writing, the SST was not yet publicly available.

The second method entails retrospective determination of population healthcare needs information using the pre-existing electronic medical records (EMRs). Compared to the prospective method, the EMR system allows the pooling of large patient datasets in a less resource-intensive way, with a relatively high degree of clinical detail. It is thus a promising resource to inform policy decisions about the health services required and identify potential areas for improving healthcare integration.⁽¹⁰⁾

Nonetheless, EMRs may not always capture information in a reliable or accurate manner. This could be due to variations found in clinician data entry, as well as the design of the EMR system, which has minimal data entry fields in order to avoid unduly burdening clinicians who perform data entry. At present, it is not known if EMRs from restructured hospitals in Singapore contain healthcare needs information of sufficient quality to facilitate meso-level healthcare needs-based segmentation.

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Hence, we aimed to evaluate the reliability of the EMRs by determining the inter-rater reliability of a brief patient healthcare needs identification instrument between clinicians who utilise the instrument in the clinic and those who utilise it based only on the EMRs. Secondarily, we aimed to determine the degree of missing information for selected healthcare need variables in the EMRs. We hypothesised that clinicians can reliably utilise EMRs to retrospectively identify healthcare needs information, and that poor reliability is due to a high degree of missing information.

METHODS

This retrospective study utilised a patient dataset containing SST ratings made prospectively in the emergency department, Singapore General Hospital, Singapore. The SST is a brief clinician-administered instrument developed in Singapore that segments patient populations into mutually exclusive health and health-related social service need segments (Appendix). It is designed for use in an outpatient setting, and clinicians trained in its use are expected to first assess a patient as part of their routine clinical assessment before using the instrument to categorise the patient.

The six medical 'global impressions of patient' of the SST were adapted from the original eight based on the 'Bridges to Health' model⁽³⁾ in order to better suit our evaluation of an elderly population. Patients are classified into one of six health categories that best characterises their most salient clinical needs in the medium to long term (months to years), namely: (a) healthy; (b) chronic conditions and asymptomatic; (c) chronic conditions and stable; (d) long course of decline; (e) limited reserve and serious exacerbations; and (f) short period of decline before dying. All patients were only assigned to one category at any point in time. Although the SST version utilised in this study (Appendix) considers 'Acutely ill but curable' to be a global impression of patient category, it is analysed as a complicating factor, as it is a transient patient feature that can coexist with a patient's baseline state (represented by the global impression of patient rating).

The 'complicating factors' section of the SST was designed to measure the degree of need in nine different healthcare-relevant characteristics: (a) functional assessment; (b) social support in case of need; (c) hospital admissions in the last six months; (d) disruptive behavioural issues; (e) polypharmacy; (f) organisation of care; (g) activation in the patient's own care; (h) skilled nursing-type task needs; and (i) acutely ill but with curable conditions.

While primarily designed to capture policy-relevant information, the SST can also be used as a triaging tool for the identification of potentially complex patients who require more detailed evaluation. Depending on the highest level of complicating factor complexity identified in the patient, a complexity category can then be assigned (Box 1). For example, if the highest level of complexity identified is 1 for polypharmacy, the case is deemed to be moderately complex.

The dataset contained SST ratings from 200 non-critical patients aged ≥ 55 years. All patients were Singaporean citizens or permanent residents who provided consent for access to their EMRs. Patients were recruited in a contiguous manner. The

Box 1. Complexity categories in the Simple Segmentation Tool:

0 = No complexity – no need for new integrated health and social needs plan (beyond current provision)

- 1 = Moderate complexity consider new integrated health and social needs evaluation; implement as appropriate
- 2 = High complexity new integrated health and social needs plan needed; urgent implementation

prospective rater group comprised of doctors and nurses who observed patients' interactions with their care provider in the emergency department clinic and reviewed the respective EMRs, then completed the SST for patients at the point of recruitment into the study.

Two medical doctors and one research nurse who had no knowledge of the patients' prior SST ratings were provided with standardised training on the utilisation method of the SST and test cases to familiarise them with the EMRs for purposes of retrospective SST rating. During training, raters were provided with an algorithm (Fig. 1) and a table (Table I) that facilitated retrospective rating. After familiarisation, raters completed an online quiz that tested their ability to administer the SST for sample cases and participated in a discussion session in which they were allowed to rate and discuss six selected patients from the dataset.

All raters were requested to review patient records only up to the time of the emergency department notes during which the prospective SST ratings were made. This was to reduce the risk of bias in retrospective ratings, and ensure that both retrospective and prospective SST rating were done during a similar time point. There was no limit to the earliest possible record that could be reviewed. Raters were allowed to review all data fields in the EMRs that fit within the stipulated time frame.

Retrospective SST raters independently reviewed the records using Singapore General Hospital's Allscripts Sunrise Clinical Manager EMR system. While raters had access to all features of the EMRs, the discharge summaries and emergency department notes were most relevant for obtaining the required healthcare needs information to rate the SST. If a particular piece of information could not be found, raters were required to rate using their best guess and then mark on the SST instrument that the information was missing. The frequency of missing information was then tabulated for all SST data variables.

A sample size calculation was done based on the primary aim of determining the inter-rater reliability of prospective rating versus retrospective rating of the SST global impression rating. We found that a sample size of 139 subjects had 85% power to detect a true Kappa value of 0.60 with a significance level of 0.05.

Retrospective SST ratings by the medical doctors were compared with prospective ratings by the reference physician using Cohen's Kappa coefficient. The inter-rater reliability of SST global impression ratings between the two retrospectively rating doctors was also determined. Meanwhile, the retrospective ratings by the research nurse were compared with the prospective ratings by other research nurses using Cohen's Kappa. This study was approved by the SingHealth Centralised Institutional Review Board (CIRB/2016/2005).

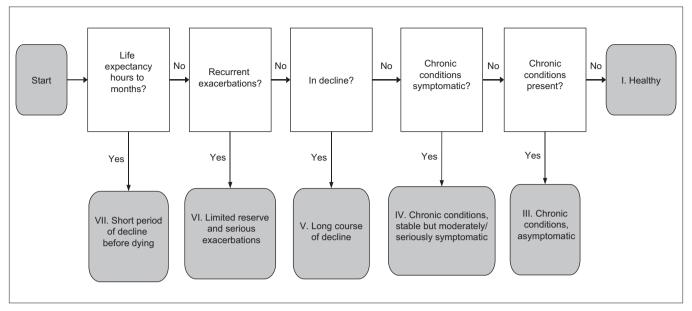


Fig. 1 Flowchart shows algorithm to facilitate retrospective Simple Segmentation Tool global impression rating for all categories except Category II (acutely ill but curable).

Table I. Reference table for rating retrospective complicating factors in the Simple Segmentation Tool.

Complicating factor	EMR software	Keywords	Notes				
Functional assessment	HIDS summary	ADL independent	Usually first 2–3 sentences of HIDS				
ADL/instrumental ADL	DEM record	Community ambulant	summary				
		Wheelchair-bound					
		Bed bound					
Social support in case	HIDS summary	Main caregiver	Decision-making				
of need	DEM record	Main spokesperson	Companionship				
		Living with	Healthcare services				
		Resuscitation status					
		Education given to					
		Family conference					
Hospital admissions in last 6 months	HIDS visit history	-	-				
Disruptive behavioural	HIDS summary	Altered mental status	Time and attention				
issues	DEM record	Aggressive/violent					
		Verbally abusive					
		Psychiatric history					
Polypharmacy	Treatment manager (Sunrise Clinical Manager)	_	To exclude prescribed medications in current visit				
Organisation of care	HIDS correspondence	Referral letter	Look out for suggestion of confusion				
	DEM record	Specialist outpatient clinic letter					
		Agency for Integrated Care					
		involvement					
		Nurse navigator involvement					
Activation in own care	HIDS summary	Presenting chief	Knowledge				
	DEM record	complaints (e.g. omitted	Skill				
		diuretics)	Cooperation issues with adverse medical outcomes				
Skilled nursing-type	HIDS summary	CGT given	E.g. patient with diabetes mellitus				
task needs	DEM record	Types of medication	and hypertension will need				
		Pressure ulcer or other complex chronic wound (e.g. venous ulcer)	monitoring of blood pressure and blood sugar level				

ADL: activities of daily living; CGT: caregiver training; DEM: Department of Emergency Medicine; EMR: electronic medical records; HIDS: hospital inpatient discharge summary

Table II. Distribution of patients according to reference physician's global impression rating.

Parameter	No. (%)						
	Healthy (n = 20)	Chronic, asymptomatic (n = 86)	Chronic, symptomatic (n = 29)	Long course of decline (n = 1)	Limited reserve with serious exacerbations (n = 2)	Short period of decline before dying (n = 2)	
Age*	64 ± 8	67 ± 8	69 ± 9	86	81	72 ± 19	
Female	10 (50)	48 (56)	8 (28)	0 (0)	1 (50)	1 (50)	
Ethnicity							
Chinese	19 (95)	70 (81)	24 (83)	1 (100)	1 (50)	2 (100)	
Indian	1 (5)	6 (7)	1 (3)	0 (0)	0 (0)	0 (0)	
Malay	0 (0)	10 (12)	4 (14)	0 (0)	1 (50)	0 (0)	

^{*}Data presented as mean or mean \pm standard deviation.

Table III. Distribution of patients according to reference physician's complicating factor rating.

Complexity	No. (%)							
	Function*	Social	Past admissions	Disruptive behaviour	Polypharmacy	Care organisation	Patient activation	Skilled nursing task
Level 1	127 (90.1)	109 (76.8)	119 (83.8)	141 (99.3)	83 (58.5)	62 (43.7)	124 (87.3)	136 (95.8)
Level 2	2 (1.4)	23 (16.2)	21 (14.8)	0 (0)	31 (21.8)	80 (56.3)	15 (10.6)	4 (2.8)
Level 3	12 (8.5)	10 (7.0)	2 (1.4)	1 (0.7)	28 (19.7)	0 (0)	3 (2.1)	2 (1.4)

^{*}Percentages are calculated based on n = 141 due to missing data.

RESULTS

Out of 200 patients, 60 patients were reviewed by both retrospective doctors, while 70 were only reviewed by the first retrospective doctor and another 70 were only reviewed by the second retrospective doctor. Thus, retrospective doctors reviewed a total of 130 patient records each, while the research nurse and prospectively rating reference physician reviewed all 200 records. Among the 140 patient records included in inter-rater reliability analysis between prospective and retrospective doctors, 60 records were reviewed by all three retrospective raters. 60 out of the initial 200 case records were excluded from the study, as these patients either decided to withhold consent for access to their EMRs or were utilised as teaching cases by the prospectively rating clinicians and were thus not independently rated. Patients were distributed according to the SST's global impression and complicating factor categories, as rated by the reference physician (i.e. one of the prospectively rating clinicians), whose rating was taken to be the gold standard. Most patients were of low medical severity and were classified in the 'chronic, asymptomatic' category (Table II). There were very few patients in the higher medical severity categories, although this may be attributed to the fact that patients were recruited from the lowest-urgency area of the emergency department.

Table III shows the distribution of patients across three different complexity levels for the various complicating factor variables. With the exception of 'care organisation', the majority of patients were rated to be of Level 1 complexity for all variables. Similar to the global impression ratings (Table II), patients recruited for this study were likely to have lower complexity ratings because they had been pre-triaged to the lowest-urgency area of the emergency department. A 'care organisation' complexity rating of Level 2 suggested that a patient had no main service provider or multiple

non-coordinated providers, which most often seemed to be the case among recruited patients.

Data variables with less missing information such as past admissions, polypharmacy and global impression were found to have higher inter-rater reliability scores (Table IV). The relatively low missing data count for variables such as past admissions and polypharmacy are expected, given that these objective data fields exist within electronic records and are thus routinely captured and available for the raters' reference. On the other hand, global impression ratings can often be inferred through both inpatient and outpatient discharge summaries. The high missing information count (measured in percentages) of the other variables (i.e. function, social, disruptive behaviour, care organisation, patient activation and skilled nursing-type tasks) may have been due to inconsistent capture of such information in structured or unstructured text fields of the EMRs. Furthermore, variations between retrospective doctor and nurse raters in quantifying the incidence of missing information in the EMRs could be due to differences in experience interpreting subjective text fields as well. For example, doctors may have more experience reviewing discharge summaries than nurses, while nurses may be more experienced in reviewing nursing notes due to their respective conventional scopes of responsibilities.

In terms of inter-rater reliability for SST global impression between prospective and retrospective ratings, the Cohen's Kappa score was 0.37 and 0.35 for doctors and nurses, respectively. Although these fell short of a Cohen's Kappa score of 0.6, which was set as the threshold of sufficient reliability in this study, the inter-rater reliability between the two retrospective doctors was significantly better, with a Cohen's Kappa score of 0.75. This suggests that poor comparability between prospective and retrospective ratings is due to missing information within the

Table IV. Missing information and inter-rater reliability results, based on Simple Segmentation Tool data variables.

Rater	No. (%)/Kappa								
	Function	Social	Past admissions	Disruptive behaviour	Polypharmacy	Care organisation	Patient activation	Skilled nursing task	Global impression
Missing information									
Doctor (n = 140)	97 (69)	101 (72)	4 (3)	97 (69)	2 (1)	98 (70)	99 (71)	97 (69)	0 (0)
Nurse (n = 140)	31 (22)	73 (52)	0 (0)	10 (7)	0 (0)	22 (16)	138 (99)	0 (0)	0 (0)
2 doctors+nurse (n = 280)	128 (46)	174 (62)	4 (1)	107 (38)	2 (1)	120 (43)	237 (85)	97 (35)	0 (0)
Inter-rater reliability									
Prospective vs. retrospective doctor* (140 pairs)	0.29	0.05	0.80	0	0.68	-0.09	0.07	-0.05	0.37
Prospective vs. retrospective nurse [†] (137 pairs)	0.16	0.17	0.68	0.26	0.50	0	0.01	0.05	0.35
Retrospective doctor vs. retrospective doctor (40 pairs)	NA	NA	NA	NA	NA	NA	NA	NA	0.75

^{*}Retrospective doctor as reference. †Retrospective nurse as reference. NA: not applicable

EMRs, and not that the process of retrospective rating itself was inherently unreliable.

DISCUSSION

Moving forward, the EMR system has immense potential for patient stratification using real-time big data analytics. It is imperative that important variables are routinely captured as part of a holistic biopsychosocial approach to patient assessment. For instance, physical function,^(12,13) social support⁽¹⁴⁻¹⁸⁾ and patient activation⁽¹⁹⁾ are well-known predictors of re-admission risk and can be used to identify care needs that require intervention. A comprehensive literature review is needed to locate variables with discriminatory and predictive value that can be included in the EMR.

SST variables with poor inter-rater reliability scores are important population healthcare need markers that can facilitate the planning and development of health service interventions. In order to improve the process of information capture by clinicians into the EMRs, one option would simply be to include these variables in the EMR system. This would improve data reliability for population-level health service policy decisions, while potentially reducing the amount of time needed for clinicians to input data into subjective data fields. Specific SST healthcare need variables with poor reliability that could benefit from such an intervention include physical function, social support in case of need, disruptive behavioural issues, care organisation, patient activation, skilled nursing task needs and global impression.

To the best of our knowledge, this is the first study in Singapore that examines the quality of a restructured hospital's EMRs for purposes of retrospective healthcare need identification. The strengths of this study include its evaluation of reliability in retrospective rating for both medical doctors and nurses, as well

as the characterisation of missing healthcare need information for key variables, which could facilitate targeted improvement of the EMR system through the addition of objective data fields.

One possible limitation of this study is that the emergency department EMRs were written by doctors who were prospectively rating patients using the SST. Hence, the quality of their records may be slightly different from the conventional records of doctors who did not use the SST. Nonetheless, our retrospective clinician raters have given feedback that there were no perceptible differences between their patient records and those of other records seen in their usual line of work, and thus the plausible biasing effect of creating records with more complete information is likely to be minimal. If bias had occurred, it would have strengthened the inter-rater reliability between the prospective and retrospective raters, yet strong inter-rater reliability was not observed. Another possible limitation would be that the patients recruited in this study had relatively low medical and social healthcare need complexity. Future studies may benefit from recruitment at sites such as the hospital inpatient department or emergency department areas with higher triage urgency, where patients typically have more healthcare needs.

In conclusion, our results suggest that the clinician's best guess is no substitute for objectively recorded information in the EMRs in terms of identifying healthcare needs. Policymakers may consider integrating important healthcare need data variables into the EMR system as routine data fields to improve data quality and reliability.

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APPENDIX

	Nurse/ Others (Please indicate PGY) Date: _			
Global impression of patient (Circle (〇))	Complicating factors	10	Level	
Healthy	A. Functional assessment, ADL/IADL 0 = no deficit 1 = moderate deficit (1 ADL/IADL) 2 = high deficit (2 or more ADL/IADL)	0	1	2
Acutely ill but curable	B. Social support in case of need 0 = high support (adequate) 1 = moderate support (any) 2 = low support (none)	0	1	2
Chronic condition(s), asymptomatic	C. Hospital admissions in last 6 months 0 = none 1 = 1 to 2 2 = 3 or more	0	1	2
Chronic condition(s), stable but moderately/ seriously symptomatic	D. Disruptive behavioral issues 0 = none 1 = 1 or more, not significantly affecting care 2 = 1 or more, significantly affecting care	0	1	2
Long course of decline	E. Polypharmacy 0 = fewer than 5 prescription medications 1 = 5 to 8 prescription medications 2 = 9 or more prescription medications	0	1	2
Limited reserve and serious exacerbation	F. Organization of care 0 = 1 main provider covering and coordinating for all health issues 1 = no main service provider or multiple non-coordinated providers 2 = patient experiences significant confusion about advice from multiple providers or has no provider at all	0	1	2
Short period of decline before dying	G. Activation in own care 0 = ready, understands and interested in treatment; active cooperation and participative 1 = unsure but willing to cooperate, no major compliance issues 2 = major disconnect, unaware/ no insight, may be defiant with major compliance issues	0	1	2
	H. Skilled nursing type task needs 0 = none 1 = moderate (1 task) 2 = high (2 or more tasks)	0	1	2
o complexity – no need for new integrate cial needs plan (beyond current provisio oderate complexity – consider new integ d social needs evaluation; implement as gh complexity – new integrated health a	n) grated health appropriate			

Guidelines

Circle the category most likely to determine health needs over the next months to years for the most severe active condition. Also, circle category II if patient currently has a non-minor acute condition.

Global impression	Definition		Descriptive examples				
I. Healthy	No more than minimal symptomatic co asymptomatic conditions that increase		Acute URTI, minor injuries, uncomplicated UTI.				
II. Acutely ill but curable	Isolated moderate to major acute conc treatment over a short period (days to assigning Category II, also assign the in category most likely to determine their needs in the medium to long-term (mc	dition that requires months). When dividual to another health service	Hip fracture, joint replacement, pneumonia, DVT, brief psychotic disorder, acute stress disorder). Minor acute conditions (e.g., URTI, minor injuries, uncomplicated UTI) should not be included here.				
III. Chronic conditions, asymptomatic	Chronic conditions (not curable once a persisted >3 months) that are asymptofor increasing preventable risk.	cquired or has	Vascular risk factors such as asymptomatic/uncomplicated diabetes, hypertension, hyperlipidemia; CCF, IHD, CRF, COPD/asthma, depression, bipolar disorder; early asymptomatic visual problems such as glaucoma or AMD; if symptoms that interfere with/restrict normal function are				
IV. Chronic conditions, stable but moderately/ seriously symptomatic	Chronic conditions that are relatively so associated with symptoms that interfeusual function or would generally be sucare seeking (e.g., pain or fatigue).	re with/restrict ufficient to trigger	present (at time of assessment), then assign to Category IV. Stroke, Parkinson's disease, orthogeriatric conditions (e.g., hip fracture, amputation, or joint replacement with residual symptoms, arthritis), stable symptomatic chronic conditions (e.g., stable angina, COPD, diabetic neuropathy), symptomatic psychiatric disorders (e.g., depression, schizophrenia, bipolar disorder), severe visual loss.				
V. Long course of decline	ong course Long (months to years) dwindling cours		A "pure" category V individual would not have a dominant/symptomatic condition and might only manifest with one or more geriatric syndromes (e.g., recurrent falls, incontinence, pressure ulcers, frailty, cognitive impairment). May present concurrently with acute illness (category II).				
VI. Limited reserve and serious exacerbation	erve and exacerbations (3 or more in prior 6 mo ous condition , <u>includes</u> episode at time of		A "pure" category VI individual has recurrent symptoms and frequent complications associated with one dominant chronic condition (e.g., diabetes with recurrent infection/metabolic extremes; frequent flares of COPD, angina, heart failure, CRF that is difficult to control) and is associated with recurrent A&E visits and/or admissions.				
VII. Short period of decline before dying	Severe life limiting conditions for pallia (Life expectancy hours to months).	tive management.	Cancer, end-organ failure (e.g., late CCF, IHD, liver failure, bedridden with severe dementia.)				
A . From this o	Complicating factors	Dania ADI ay Dathian	Level and legends				
(Definiti	nal assessment on: Requiring caregiver assistance to basic or instrumental ADL.)	Basic ADLs: Bathing Instrumental ADLs:	Ability to use telephone Shopping Food preparation Housekeeping Laundry Use of transportation Responsibility for own medication Ability to handle finances				
(Definiti support	Social support (Definition: Family or friends who provide support through decision making, companionship, basic healthcare services.)		Social support refers to aid with decision making, companionship and basic healthcare services should the need arise. If support is available in all 3 domains, this is high.				
 Disruptive behavioural issues (Definition: Disruptive behavioural issues requiring support for caregiver.) 		Behavioral issues are disruptive if they significantly affect care; which is defined as typically requiring time and attention from people in patient's environment or inducing distress. Substance abuse Self-harm Confusion/forgetfulness					
		Depression Dissociative problems Abusiveness					
(Definiti	Polypharmacy (Definition: Large number of prescription medications.)		Prescription medications are defined as medications for which patient has a written prescription. They <u>exclude</u> alternative medications such as traditional Chinese medicine and other herbal supplements.				
F. Organization of care (Definition: Degree of co-ordination amongst providers and absence of multiple doctor medical advice confusion.)		Coordination is defined by documented communication between providers directly (e.g., via letter or email) or indirectly (e.g., via navigator or other dedicate 3 rd party maintaining written records who could be family or other caregiver. Organization is assumed to be functionally uncoordinated if the patient experiences significant confusion.					
(Definiti for man	on in own care on: Knowledge, skill, and confidence aging one's health and healthcare as ooperation with treatment.)	Major compliance/cooperation issues are ones which have led to adverse medica outcomes.					
H. Skilled r (Definiti			Injections Change of feeding tube agement training for patient/caregiver d pressure & blood sugar levels				