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# Age And Ageing

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#### Comparing the predictive ability of the Revised Minimum Dataset Mortality Risk Index (MMRI-R) with nurses' predictions of mortality among frail older people: a cohort study.

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Keywords:	End of life care, Older people, Community nurse, Mortality, Advance care planning
Keypoints:	Identifying patients who are approaching the end of their lives can facilitate advance care planning, Prognostication is challenging amongst frail older patients with multiple progressive and fluctuating long-term conditions, The MMRI-R is comparable to community nurses' intuition at identifying patients approaching end of life, The MMRI-R and nurses predictions performed better in combination than alone, Case- management nurses such as UK community matrons should be trusted to initiate the early implementation of anticipatory care



> Comparing the predictive ability of the Revised Minimum Dataset Mortality Risk Index (MMRI-R) with nurses' predictions of mortality among frail older people: a cohort study.

### Abstract

#### Objectives

To establish the accuracy of community nurses' predictions of mortality among older people with multiple long-term conditions, to compare these with a mortality rating index and to assess the incremental value of nurses' predictions to the prognostic tool.

#### Design

A prospective cohort study using questionnaires to gather clinical information about patients case managed by community nurses. Nurses estimated likelihood of mortality for each patient on a 5point rating scale. The dataset was randomly split into derivation and validation cohorts. Cox proportional hazard models were used to estimate risk equations for the MMRI-R and nurses' predictions of mortality individually and combined. Measures of discrimination and calibration were calculated and compared within the validation cohort.

#### Setting

Two NHS Trusts in England providing case-management services by nurses for frail older people with multiple long-term conditions.

#### Participants

867 patients on the caseload of 35 case-management nurses. 433 and 434 patients were assigned to the derivation and validation cohorts respectively. Patients were followed up for 12 months.

#### Results

249 patients died (28.72%). In the validation cohort MMRI-R demonstrated good discrimination (Harrell's c-index 0.71) and nurses' predictions similar discrimination (Harrell's c-index 0.70). There was no evidence of superiority in performance of either method individually (p=0.83) but the MMRI-R and nurses' predictions together were superior to nurses' predictions alone (p=0.01).

#### Conclusions

Patient mortality is associated with higher MMRI-R scores and nurses' predictions of 12-month mortality. The MMRI-R enhanced nurses' predictions and may improve nurses' confidence in initiating anticipatory care interventions.

# Introduction

The identification of people requiring palliative and supportive care underpins national and international strategies to improve the quality of end of life care [1-5]. Identifying patients who are approaching the end of their lives can facilitate advance care planning to align future care to patients' wishes and preferences [6]. Identifying patients in need of supportive or palliative care is challenging, particularly for frail older patients with multiple progressive and fluctuating long-term conditions due to the complex interplay between pathology, clinical findings, therapeutic response, co-morbidities, psychosocial factors, and rate of decline [7]. The evidence suggests that doctors tend to overestimate survival and feel inadequately prepared to make predictions [8].

A number of clinician-administered identification tools, such as the Prognostic Indicator Guide [9], have been developed to support primary care teams to recognise patients who may require palliative care but few have been validated in primary care settings [10]. The Palliative Prognostic Score [11] includes clinicians' prediction of survival as a significant predictor of mortality, although the validity of this has recently been challenged [12]. A small exploratory randomised controlled trial of general practitioners in Australia demonstrated that the use of the Supportive and Palliative Care Indicators Tool as a screening tool was better at predicting actual death than unguided clinical intuition but incurred a higher false positive rate [13]. Most mortality risk indices for community dwelling older adults have limited clinical value as they have been developed on samples with low mortality rates or do not assess short-term mortality [14].

Less attention has been afforded to the role of nurses in identifying patients in need of supportive and palliative care [15, 16] despite the increasing number of primary care nurses who case manage older adults with frailty at high risk of clinical deterioration, hospitalisation and death. In the UK, case-management nurses, known as community matrons, were introduced in 2005 to coordinate and manage the care of older adults in their own homes or care homes [17, 18]. Supporting patient choice around end of life care through anticipatory care planning is considered a key responsibility of community matrons [18] but it is unclear whether this role has been adopted in practice.

With an increasing emphasis on identifying older people with advanced frailty for whom a supportive or palliative care approach to case management may be beneficial we assessed the ability of community nurses to recognise patients who may be approaching the end of life and compared this to an existing mortality risk index. The Minimum Dataset Mortality Index (MMRI) [19] was developed using routinely collected data from a large sample of nursing home residents in the United States of whom 23% died within six months. The index encompasses demographic data, comorbidities, symptoms, cognition and functional status. It demonstrated very good discrimination and calibration across a wide range of mortality risk levels and has subsequently been simplified [20] and validated within the UK [21] without significant reduction in performance.

#### Methods

#### Objectives, study design and source of data

A prospective cohort study was undertaken to establish the accuracy of community nurses' predictions of mortality among older people with multiple long-term conditions, to compare these with the MMRI-R and to assess the incremental value of nurses' predictions to the MMRI-R. All

community nurses employed by two neighbouring NHS Trusts in the East Midlands, who provided a case-management service to older people living in the community, were invited to participate.

#### Predictor variables

The nurses completed an electronic questionnaire for each patient on their caseload to provide demographic information (age and gender) and clinical information relating to comorbidities (cancer, renal failure and heart failure), symptoms (shortness of breath, dehydration, loss of appetite and cognition), functional dependency and recent admission to a care home. Nurses were also asked to estimate the likelihood of mortality within 12 months for each patient using a 5-point Likert scale (highly unlikely, unlikely, unsure, likely, highly likely). Data was pseudonymised in a secure NHS data warehouse prior to analysis.

#### Outcome:

The outcome of the study was all-cause mortality determined through cross-reference with a national database of deaths derived from information contained within the electronic patient record. Each patient was followed up for a minimum of 12 months.

#### Sample size:

In the two participating NHS Trusts, 45 case-management nurses had an estimated total caseload of 1400 patients. We estimated non-response from 20% of nurses and that 15% of patients would die over a 12-month period based on retrospective analysis of caseload data. The dataset was stratified according to 12-month survival and randomly split into derivation and validation cohorts of equal size.

#### Model derivation:

MMRI-R scores were calculated from the published bedside scoring tool (Supplementary table 1 available at Age and Ageing online) [21]. The maximum score possible was 85 with increasing scores indicating greater mortality risk. Nurses' predictions of death were rated on a five-point scale with five indicating that the patient was highly likely to be in the last year of life. Cox proportional hazard models were used to define three separate risk equations. Coefficients were estimated for the MMRI-R and the nurses' predictions of survival separately and then in combination. Risk groups were created to facilitate the comparison of observed survival probabilities. Cut-points were 25<sup>th</sup> and 75<sup>th</sup> centile of each prognostic index in the derivation cohort indicating low, intermediate and high mortality risk. Kaplan-Meier survival estimates were calculated across risk groups for each model.

#### Model validation:

Discrimination, the ability to separate those at high risk from those at low risk, was assessed using two measures. Royston-Sauerbrei  $R^2_D$  is based on prognostic separation and is a measure of explained variation on the log relative hazard scale [22]. Harrell's *c*-index measures concordance by quantifying the rank correlation between predicted risk and observed survival times [23]. Values of 0.5 indicate no predictive discrimination whilst a value of 1 indicates perfect separation of patients with different outcomes [24]. Differences in Harrell's c-indices were calculated in order to compare predictive performance of each risk equation [25].

Calibration, or how accurately estimated survival reflected actual survival, was assessed according to methods outlined by Royston [26]. The predicted mean survival curves were compared with the

Kaplan-Meier survival curves for each risk group within the validation dataset. Stata Version 14.2 was used for all analyses.

#### Ethical Approval and Funding

Ethical approval was granted by Nottingham 2 Medical Research Ethics Committee (12/EM/0322 09/11/2012). Site specific approval and research and development approval was obtained from both participating NHS Trusts (DHRD/2012/071). This study was supported by a Learning Beyond Registration Doctoral Scholarship awarded to AC by the East Midlands Professional Deanery and was sponsored by the University of Nottingham.

# Results

#### Participants

A total of 45 nurses case-managing 1330 predominantly older patients with multiple long-term conditions were invited to complete electronic questionnaires for each of their patients. Of these, 10 nurses declined or were unable to participate leaving 35 community case-management nurses (77%) submitting individual questionnaires for a total of 883 patients (mean 25 patients, SD 14.4). Eight questionnaires were duplicated and were eliminated from analysis. A further eight patients with incomplete identification details were lost to follow-up and were also excluded from analysis. A total of 867 patients were randomly assigned to the derivation cohort (n=433) or validation cohort (n=434) (Supplementary figure 1 available at Age and Ageing online). Questionnaires were submitted between July and September 2013. Survival was determined on 1<sup>st</sup> October 2014, one year after receipt of the last questionnaire.

The characteristics of the patients assigned to the derivation and validation cohorts are reported in Table 1. MMRI-R scores ranged from 2 to 62 (mean 22.5, SD 10.6). Characteristics of the matrons are reported in Supplementary Table 2 available at Age and Ageing online). Almost two thirds of the matrons had worked previously as a district nurse.

#### Outcomes

A total of 249 deaths were recorded giving an overall mortality of 29%. Among those who died deaths occurred between 1 and 435 days. The minimum follow-up time for survivors was 366 days and maximum follow-up time was 454 days.

		lerivation and validatio	
Characteristics	<b>Derivation Cohort</b>	Validation Cohort	Total n=867
	n (%)	n (%)	n (%)
Male Gender	190 (43.9)	178 (41.0)	368 (42.4)
Comorbid conditions			
Heart Failure	140 (32.3)	154 (35.5)	294 (33.9)
Cancer	37 (8.9)	32 (7.4)	69 (8,0)
Renal Failure	118 (27.3)	131 (30.2)	249 (28.7)
Cognitive Impairment	59 (13.6)	63 (14.5)	122 (14.1)
Symptoms			
Weight loss	89 (20.5)	78 (18.0)	167 (19.3)
Dehydrated	23 (5.3)	23 (5.3)	46 (5.3)
Loss of appetite	143 (33.0)	128 (29.5)	271 (31.3)
Shortness of breath	270 (62.4)	273 (62.9)	543 (62.6)
Care Home Admission in last	19 (4.4)	23 (5.3)	42 (4.8)
6 months	19 (4.4)	25 (5.5)	42 (4.8)
ADL Dependencies			
Mean ADL score (SD)	4.1 (5.1)	3.8 (5.0)	4.0 (5.0)
Mean Age (SD)	78.9 (11.5)	79.0 (11.0)	79.0 (11.5)
MMRI-R Score (SD)	22.78 (10.9)	22.30 (10.3)	22.5 (10.6)
Likelihood of Death (%)	6		
Highly Unlikely	21 (4.9)	23 (5.3)	44 (5.0)
Unlikely	148 (34.2)	152 (35.0)	300 (34.6)
Unsure	121 (27.9)	125 (28.8)	246 (28.4)
Likely	112 (25.9)	101 (23.3)	213 (24.6)
Highly Likely	31 (7.2)	> 33 (7.6)	64 (7.4)
Died	124 (28.6)	125 (28.8)	249 (28.7)

#### Model Development and Specification

In the derivation cohort the MMRI-R and nurses' predictions were associated with mortality both individually and in combination (Table 2). A global test of the proportional hazards assumption in the derivation dataset using scaled Schoenfeld residuals demonstrated no apparent violation.

Kaplan–Meier curves (Supplementary Figure 2 available at Age and Ageing online) demonstrated that low, intermediate and high risk groups were well separated across all equations with greater mortality observed with increasing risk score, suggesting good discrimination. Depending on which equation was used, between 42% and 52% of patients in the high risk group were alive at follow-up with between 79% and 85% survivors in the low risk group.

#### Model Validation

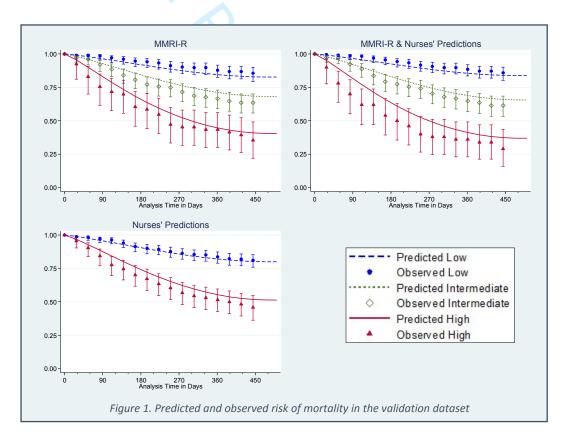
#### Discrimination

Harrell's *c*-index and Royston and Sauerbrei's  $R^2_D$  (Table 2) indicated good discrimination for the MMRI-R and for nurses' predictions. Calculation of the differences in the Harrell's *c*-index between equations, indicated that MMRI-R was not significantly different from the nurses' predictions alone (p=0.83). The addition of MMRI-R significantly improved nurses predictions alone (p=0.01), as did the addition of nurses' predictions to MMRI-R alone (p=0.03).

Model	Derivatio	n Cohort			Validation Cohor	t
	Hazard	95% CI	p-value	Harrell's	Royston & Sau	erbrei's R <sup>2</sup> <sub>D</sub> (SE)
	Ratio			C-index (SE)	R <sup>2</sup>	D
MMRI-R	1.06	1.04-1.08	<0.001	0.706 (0.022)	0.271 (0.049)	1.248 (0.154)
Nurses' Predictions	1.81	1.52-2.16	<0.001	0.701 (0.023)	0.300 (0.051)	1.339 (0.163)
MMRI-R &	1.04	1.02-1.06	<0.001	0 724 (0 022)		1 510 (0 161)
Nurses' Predictions	1.43	1.16-1.76	<0.001	0.734 (0.022)	0.355 (0.049)	1.519 (0.161)
HRs refer to a 1 point incre	ase in each sco	ores: MMRI -2-85;	Nurses' Predic	tions 0-4 (higher score	s indicate worsening	prognosis)

#### Calibration

The predicted mean survival curves were compared with the observed Kaplan-Meier survival curves in each risk group for the validation set (Figure 1). The curves indicated that actual survival was worse than predicted in the high risk group particularly within the combined MMRI-R and Nurses' Predictions equation suggesting a degree of miscalibration. The Kaplan-Meier survival estimates indicated that risk groups are well separated across each equation within the validation set.



# Discussion

Community nurses' predictions are comparable to the MMRI-R at identifying older people who are at low, intermediate and high risk of death. The findings of our analysis indicate that combining nurses' predictions with the MMRI-R improved the accuracy of each model at identifying older people at risk of death within 12 months. A previous study benchmarked the clinical predictions of specialist

palliative care doctors and nurses within a cohort of 1018 patients with advanced cancer against two objective prognostic models [27]. In that study, specialist nurses' predictions demonstrated very good discrimination comparable to both objective models. In our study, the less accurate discrimination of community case-management nurses may be attributable to their unique role within a generalist setting, caring for patients with multiple long-term conditions and less predictable disease trajectories than patients with advanced cancer. In practice, nurses may use subjective estimations of the need to adjust patients' goals of care to initiate discussions about advance planning and end of life care preferences. Our findings suggest that nurses' predictions of mortality present a potentially useful screening tool for assessment of the timeliness of end of life care planning.

We have demonstrated that the MMRI-R retains good performance within this heterogenous cohort of patients. The c-index indicated slightly poorer performance than the original study which reported very good discrimination in predicting six-month mortality among long-term care home residents [20]. Differences in discrimination measures are likely attributable to differences in baseline characteristics of the populations. The MMRI-R was developed within a population of care home residents in the United States, whilst our patient sample were predominantly living in their own homes. Additionally over half of the original sample were aged 85 years or over, compared with one third aged over 85 in this UK sample. However, despite these differences, mortality in each study is similar which suggests a comparable burden of illness and mortality risk.

The MMRI-R may improve community case-management nurses' understanding of the determinants of the clinical course and outcomes of their patients [28]. However, mortality rating indices only provide a snapshot assessment of a patient's mortality risk and do not take into account the rapidity of a patient's deterioration or any potential for reversibility of signs and symptoms. The MMRI-R should not be used to replace clinical judgement or person-centred communication as a trigger to initiate advance care planning, rather as an adjunct to clinical decision making by providing an objective estimate of risk to be considered alongside other relevant clinical information.

We found that community case-management nurses recognise patients who are approaching the end of their lives with reasonably good accuracy, yet the extent to which their patients engaged with advance care planning is unknown. Future studies should explore the impact of community casemanagement nurses on meeting the desired goals and expectations of patients and the cost of healthcare at the end of life through advance care planning. Since the data for this study was collected a sustained increase in the aged population continues to place mounting pressure on primary care resources. Discussing patients' preferences for clinical care and treatment is therefore assuming a greater prominence in order to prevent unnecessary hospitalisation at the end of life and to support people to die in their preferred place of care. The use of a mortality risk index as an adjunct to clinical decision making may help nurses to prioritise patients for whom advance care planning may be appropriate. Mortality risk indices do not appear to have been adopted widely in practice and further research is needed to explore their potential benefits and limitations to identify patients who may benefit from advance care planning.

#### Application of the Prediction Models

The following example illustrates how the MMRI-R and nurses' predictions assesses survival. An 83 year old man with heart failure, chronic kidney disease stage 4 and a creatinine greater than 3mg/dL has shortness of breath on exertion and poor mobility. He is dependent on carers who visit his home three times daily to assist with washing and meal preparation. His appetite has recently declined and

his community case-management nurse thinks that he is likely to be in his last year of life. His MMRI-R is calculated as 39 (Supplementary table 1 available at Age and Ageing online). Both MMRI-R and nurses' predictions place him in the high risk group and estimates that he has between a 48% and 57% chance of survival at 12 months.

#### Limitations

In order to assess the additive value of nurses' predictions to the MMRI-R a cross-validation design was adopted in which the data was partitioned into derivation and validation sets. This approach resulted in two smaller samples and may contribute to an over-optimistic assessment of model performance [29].

# Conclusion

Community case-management nurses can identify patients who are likely to die within 12 months with reasonable accuracy. The MMRI-R demonstrated comparable discrimination in identifying those at high or low risk of mortality and the addition of nurses' predictions improved its performance. Nurses' estimates of survival may trigger an objective assessment of mortality risk among those patients considered to be most likely to die within twelve months. An objective assessment, using the MMRI-R, a simple assessment tool that can be administered in the patient's own home, has the potential to improve nurses' confidence in their clinical decision making and encourage community case-management nurses to initiate timely conversations with frail older people and those close to them to determine end of life care preferences.

# **Conflicts of Interests**

All authors declare they have nothing to disclose.

#### Key Points

Identifying patients who are approaching the end of their lives can facilitate advance care planning and align future care to patients' preferences.

Prognostication is challenging among frail older patients with multiple progressive and fluctuating long-term conditions.

The MMRI-R is comparable to community nurses' intuition at identifying patients approaching end of life.

The MMRI-R and nurses predictions performed better in combination than alone.

Case-management nurses such as UK community matrons should be trusted to initiate the early implementation of anticipatory care interventions.

# Key Words:

Older people, end of life care, community nurse, mortality, advance care planning

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#### Conclusions

Patient mortality is associated with higher MMRI-R scores and nurses' predictions of 12-month mortality. The MMRI-R enhanced nurses' predictions and may improve nurses' confidence in <del>decision</del>-making. Minimal differences in performance suggest that nurses' judgement can be trusted to initiating anticipatory care interventions whilst avoiding an over-reliance on objective measures.

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### Introduction

The identification of people requiring palliative and supportive care underpins national and international strategies to improve the quality of end of life care [1-5]. Identifying patients who are approaching the end of their lives can facilitate advance care planning to align future care to patients' wishes and preferences [6]. Identifying patients in need of supportive or palliative care is challenging, particularly for frail older patients with multiple progressive and fluctuating long-term conditions due to the complex interplay between pathology, clinical findings, therapeutic response, co-morbidities, psychosocial factors, and rate of decline [7]. The evidence suggests that doctors tend to overestimate survival and feel inadequately prepared to make predictions [8].

A number of clinician-administered identification tools, such as the Prognostic Indicator Guide [9], have been developed to support primary care teams to recognise patients who may require palliative care but few have been validated in primary care settings [10]. The Palliative Prognostic Score [11] includes clinicians' prediction of survival as a significant predictor of mortality, although the validity of this has recently been challenged [12]. A small exploratory randomised controlled trial of general practitioners in Australia demonstrated that the use of the Supportive and Palliative Care Indicators Tool as a screening tool was better at predicting actual death than unguided clinical intuition but incurred a higher false positive rate [13]. Most mortality risk indices for community dwelling older adults have limited clinical value as they have been developed on samples with low mortality rates or do not assess short-term mortality [14]are often cumbersome and have not been widely adopted in clinical practice. The Minimum Dataset Mortality Index (MMRI) [10] was developed and validated using routinely collected data from a large sample of nursing home residents in the United States. This bedside prognostic tool has subsequently been revised and simplified with no significant reduction in performance [11].

Few studies have compared prognostic tools with clinicians' predictions of survival [12]. Studies to date have primarily focussed on the prognostic skills of medical staff who tend to overestimate survival and feel inadequately prepared to make predictions [13]. A small exploratory randomised controlled trial of general practitioners in Australia demonstrated that the use of the Supportive and Palliative Care Indicators Tool as a screening tool was better at predicting actual death than unguided clinical intuition but incurred a higher false positive rate [14]. Less attention has been afforded to the role of nurses in identifying patients in need of supportive and palliative care [15, 16] despite the increasing number of primary care nurses who case manage older adults with frailty at high risk of clinical deterioration, hospitalisation and death. In the UK, case-management nurses, known as community matrons, were introduced in 2005 to coordinate and manage the care of older adults in their own homes or care homes [17, 18]. Supporting patient choice around end of life care through anticipatory care planning is considered a key responsibility of community matrons [18] but it is unclear whether this role has been adopted in practice.

With an increasing emphasis on identifying older people with advanced frailty for whom a supportive or palliative care approach to case management may be beneficial we assessed the ability of community nurses to recognise patients who may be approaching the end of life and compared this to an existing mortality risk index. The Minimum Dataset Mortality Index (MMRI) [19] was developed using routinely collected data from a large sample of nursing home residents in the United States of whom 23% died within six months. The index encompasses demographic data, comorbidities,

symptoms, cognition and functional status. It demonstrated very good discrimination and calibration across a wide range of mortality risk levels and has subsequently been simplified [20] and validated within the UK [21] without significant reduction in performance.

#### Methods

#### Objectives, study design and source of data

A prospective cohort study was undertaken to <u>establish the accuracy of community nurses'</u> predictions of mortality among older people with multiple long-term conditions, to compare these with the MMRI-R and to assess the incremental value of nurses' predictions to the MMRI-R. compare the performance of the MMRI-R and nurses' judgement in predicting 12-month mortality among a community-based sample of older people with multiple long-term conditions. All community nurses employed by two neighbouring NHS Trusts in the East Midlands, who provided a case-management service to older people living in the community, were invited to participate.

#### Predictor variables

The nurses completed an electronic questionnaire for each patient on their caseload to provide demographic information (age and gender) and clinical information relating to comorbidities (cancer, renal failure and heart failure), symptoms (shortness of breath, dehydration, loss of appetite and cognition), functional dependency and recent admission to a care home. Nurses were also asked to estimate the likelihood of mortality within 12 months for each patient using a 5-point Likert scale (highly unlikely, unlikely, unsure, likely, highly likely). Data was pseudonymised in a secure NHS data warehouse prior to analysis.

#### Outcome:

The outcome of the study was all-cause mortality determined through cross-reference with a national database of deaths derived from information contained within the electronic patient record. Each patient was followed up for a minimum of 12 months.

#### Sample size:

In the two participating NHS Trusts, 45 case-management nurses had an estimated total caseload of 1400 patients. We estimated non-response from 20% of nurses and that 15% of patients would die over a 12-month period based on retrospective analysis of caseload data. The dataset was stratified according to 12-month survival and randomly split into derivation and validation cohorts of equal size. Based on retrospective analysis of nurses' caseloads we estimated 15% of our patient sample would die over a 12-month period. We calculated that a sample size of 800 was needed to provide a minimum ratio of 10 events (deaths) based on the 12 predictor variables comprising the MMRI-R. The dataset was stratified according to 12-month survival and randomly split into derivation and validation cohorts of equal size.

#### Model derivation:

MMRI-R scores were calculated from the published bedside scoring tool (Supplementary table 1 available at Age and Ageing online) [21]. The maximum score possible was 85 with increasing scores indicating greater mortality risk. Nurses' predictions of death were rated on a five-point scale with five indicating that the patient was highly likely to be in the last year of life. Cox proportional hazard models were used to define three separate risk equations. Coefficients were estimated for the MMRI-

R and the nurses' predictions of survival separately and then in combination. Prognostic indices were calculated from post estimation linear predictions from each fitted model and used to create rRisk groups were created to facilitate the comparison of observed survival probabilities and predicted survival. Cut-points were 25<sup>th</sup> and 75<sup>th</sup> centile of each prognostic index in the derivation cohort indicating low, intermediate and high mortality risk. Kaplan-Meier survival estimates were calculated across risk groups for each model.

#### Model validation:

Discrimination, the ability to separate those at high risk from those at low risk, was assessed using two measures. Royston-Sauerbrei  $R^2_D$  is based on prognostic separation and is a measure of explained variation on the log relative hazard scale [22]. Harrell's *c*-index measures concordance by quantifying the rank correlation between predicted risk and observed survival times [23]. Values of 0.5 indicate no predictive discrimination whilst a value of 1 indicates perfect separation of patients with different outcomes [24]. Differences in Harrell's c-indices were calculated in order to compare predictive performance of each risk equation [25].

Calibration, or how accurately estimated survival reflected actual survival, was assessed according to methods outlined by Royston [26]. Fractional polynomial regression was used to approximate the log baseline cumulative hazard function and estimation of the coefficients by ordinary least squares regression in the derivation dataset. The predicted mean survival curves were compared with the Kaplan-Meier survival curves for each risk group within the validation dataset. Stata Version 14.2 was used for all analyses.

#### Ethical Approval and Funding

Ethical approval was granted by Nottingham 2 Medical Research Ethics Committee (12/EM/0322 09/11/2012). Site specific approval and research and development approval was obtained from both participating NHS Trusts (DHRD/2012/071). This study was supported by a Learning Beyond Registration Doctoral Scholarship awarded to AC by the East Midlands Professional Deanery and was sponsored by the University of Nottingham.

#### Results

#### Participants

A total of 45 nurses case-managing 1330 predominantly older patients with multiple long-term conditions were invited to complete electronic questionnaires for each of their patients. Of these, 10 nurses declined or were unable to participate leaving 35 community case-management nurses (77%) submitting individual questionnaires for a total of 883 patients (mean 25 patients, SD 14.4). Eight questionnaires were duplicated and were eliminated from analysis. A further eight patients with incomplete identification details were lost to follow-up and were also excluded from analysis. A total of 867 patients were randomly assigned to the derivation cohort (n=433) or validation cohort (n=434) (Supplementary figure 1 available at Age and Ageing online). Questionnaires were submitted between July and September 2013. Survival was determined on 1<sup>st</sup> October 2014, one year after receipt of the last questionnaire.

The characteristics of the patients assigned to the derivation and validation cohorts are reported in Table 1. MMRI-R scores ranged from 2 to 62 (mean 22.5, SD 10.6). <u>Characteristics of the matrons are</u>

reported in Supplementary Table 2 available at Age and Ageing online). Almost two thirds of the matrons had worked previously as a district nurse.

#### Outcomes

A total of 249 deaths were recorded giving an overall mortality of 29%. Among those who died deaths occurred between 1 and 435 days. The minimum follow-up time for survivors was 366 days and maximum follow-up time was 454 days.

Chausstanistics	<b>Derivation Cohort</b>	Validation Cohort	Total n=867
Characteristics	n (%)	n (%)	n (%)
Male Gender	190 (43.9)	178 (41.0)	368 (42.4)
Comorbid conditions			
Heart Failure	140 (32.3)	154 (35.5)	294 (33.9)
Cancer	37 (8.9)	32 (7.4)	69 (8,0)
Renal Failure	118 (27.3)	131 (30.2)	249 (28.7)
Cognitive Impairment	59 (13.6)	63 (14.5)	122 (14.1)
Symptoms			
Weight loss	89 (20.5)	78 (18.0)	167 (19.3)
Dehydrated	23 (5.3)	23 (5.3)	46 (5.3)
Loss of appetite	143 (33.0)	128 (29.5)	271 (31.3)
Shortness of breath	270 (62.4)	273 (62.9)	543 (62.6)
Care Home Admission in last	19 (4.4)	23 (5.3)	42 (4.8)
6 months	15 (4.4)	23 (3.3)	42 (4.0)
ADL Dependencies			
Mean ADL score (SD)	4.1 (5.1)	3.8 (5.0)	4.0 (5.0)
Mean Age (SD)	78.9 (11.5)	79.0 (11.0)	79.0 (11.5)
MMRI-R Score (SD)	22.78 (10.9)	22.30 (10.3)	22.5 (10.6)
Likelihood of Death (%)			
Highly Unlikely	21 (4.9)	23 (5.3)	44 (5.0)
Unlikely	148 (34.2)	152 (35.0)	300 (34.6)
Unsure	121 (27.9)	125 (28.8)	246 (28.4)
Likely	112 (25.9)	101 (23.3)	213 (24.6)
Highly Likely	31 (7.2)	33 (7.6)	64 (7.4)
Died	124 (28.6)	125 (28.8)	249 (28.7)

#### Model Development and Specification

In the derivation cohort the MMRI-R and nurses' predictions were associated with mortality both individually and in combination (Table 2). A global test of the proportional hazards assumption in the derivation dataset using scaled Schoenfeld residuals demonstrated no apparent violation.

Kaplan–Meier curves (Supplementary Figure 2 available at Age and Ageing online) demonstrated that low, intermediate and high risk groups were well separated across all equations with greater mortality observed with increasing risk score, suggesting good discrimination. Depending on which equation was used, between 42% and 52% of patients in the high risk group were alive at follow-up with between 79% and 85% survivors in the low risk group.

#### Model Validation

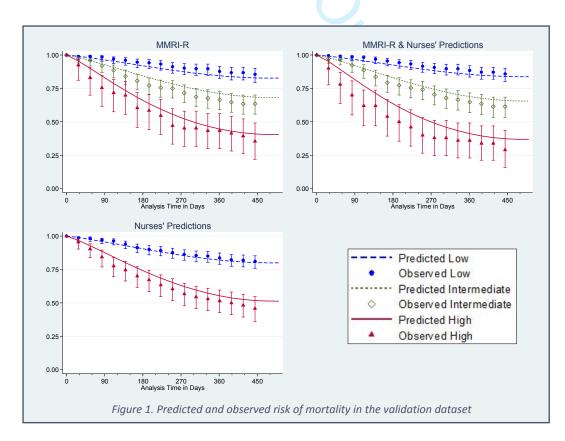
#### Discrimination

Harrell's *c*-index and Royston and Sauerbrei's  $R^2_D$  (Table 2) indicated good discrimination for the MMRI-R and for nurses' predictions. The addition of nurses' predictions conferred a small but significant improvement in discrimination of the MMRI-R. Calculation of the differences in the Harrell's c-index between equations, indicated that MMRI-R was not significantly different from the nurses' predictions alone (p=0.83). The addition of MMRI-R significantly improved nurses predictions alone (p=0.01), as did the addition of nurses' predictions to MMRI-R alone (p=0.03).

Table 2. Hazard Ratio	s and meas	sures of discrin	nination			
Model	Derivatio	n Cohort			Validation Cohor	t
	Hazard	95% CI	p-value	Harrell's	Royston & Sau	erbrei's R <sup>2</sup> <sub>D</sub> (SE)
	Ratio			C-index (SE)	R <sup>2</sup>	D
MMRI-R	1.06	1.04-1.08	<0.001	0.706 (0.022)	0.271 (0.049)	1.248 (0.154)
Nurses' Predictions	1.81	1.52-2.16	<0.001	0.701 (0.023)	0.300 (0.051)	1.339 (0.163)
MMRI-R &	1.04	1.02-1.06	<0.001	0.734 (0.022)	0.355 (0.049)	1.519 (0.161)
Nurses' Predictions	1.43	1.16-1.76	<0.001	0.754 (0.022)	0.555 (0.049)	1.519 (0.101)
HRs refer to a 1 point increa	ase in each sco	ores: MMRI -2-85;	Nurses' Predic	tions 0-4 (higher score	s indicate worsening p	prognosis)

#### Calibration

The predicted mean survival curves were compared with the observed Kaplan-Meier survival curves in each risk group for the validation set (Figure 1). The curves indicated that actual survival was worse than predicted in the high risk group particularly within the combined MMRI-R and Nurses' Predictions equation suggesting a degree of miscalibration. The Kaplan-Meier survival estimates indicated that risk groups are well separated across each equation within the validation set.



## Discussion

Community nurses' predictions are comparable to the MMRI-R at identifying older people who are at low, intermediate and high risk of death. The findings of our analysis indicate that <u>combining</u> nurses' predictions <u>conferred little incremental predictive value to the with the</u> MMRI-R <u>improved the accuracy of each model atfor</u> identifying older people <u>at risk of deathwho are likely to die</u> within 12 months and demonstrated moderately good discrimination of patients' risk of death. A previous study benchmarked the clinical predictions of specialist palliative care doctors and nurses within a cohort of 1018 patients with advanced cancer against two objective prognostic models [27]. In that study, specialist nurses' predictions demonstrated very good discrimination comparable to both objective models. In our study, the less accurate discrimination of community case-management nurses may be attributable to their unique role within a generalist setting, caring for patients with multiple long-term conditions and less predictable disease trajectories than patients with advanced cancer. In practice, nurses may use subjective estimations of the need to adjust patients' goals of care to initiate discussions about advance planning and end of life care preferences. Our findings suggest that nurses' predictions of mortality present a potentially useful screening tool for assessment of the timeliness of end of life care planning.

A recent study in England investigated the performance of an electronic frailty index (eFI) generated from primary care data of 13,149 older people at predicting mortality [24]. The study demonstrated modest discrimination (c-statistic 0.69) and concluded that the eFI is a strong predictor of mortality at a population level but its usefulness for individuals may be limited. The eFI relies upon extraction of electronic data from the patient record and may be compromised by inaccurate or absent coding of clinical signs and symptoms such as weight loss and poor mobility. The c-statistic reported for the eFI is comparable to that of our community case management nurses, suggesting that nurses' clinical assessments may be as good as a risk score generated by an index of frailty. Our MMRI-R [11] model achieved better discrimination than that reported by the eFI [24] and is straightforward to calculate during a patient consultation from twelve easily obtainable variables which may offer a practical advantage over a more complex eFI based on a cumulative deficit model.

We have demonstrated that the MMRI-R retains good performance within this heterogenous cohort of patients. The c-index indicated slightly poorer performance than the original study which reported very good discrimination in predicting six-month mortality among long-term care home residents [20]. Differences in discrimination measures are likely attributable to differences in baseline characteristics of the populations. The MMRI-R was developed within a population of care home residents in the United States, whilst our patient sample were predominantly living in their own homes. Additionally over half of the original sample were aged 85 years or over, compared with one third aged over 85 in this UK sample. However, despite these differences, mortality in each study is similar which suggests a comparable burden of illness and mortality risk.

The MMRI-R may improve community case-management nurses' understanding of the determinants of the clinical course and outcomes of their patients [28]. However, mortality rating indices only provide a snapshot assessment of a patient's mortality risk and do not take into account the rapidity of a patient's deterioration or any potential for reversibility of signs and symptoms. The MMRI-R should not be used to replace clinical judgement or person-centred communication as a trigger to

initiate advance care planning, rather as an adjunct to clinical decision making by providing an objective estimate of risk to be considered alongside other relevant clinical information.

We found that community case-management nurses recognise patients who are approaching the end of their lives with reasonably good accuracy, yet the extent to which their patients engaged with advance care planning is unknown. Future studies should explore the impact of community casemanagement nurses on meeting the desired goals and expectations of patients and the cost of healthcare at the end of life through advance care planning. Since the data for this study was collected a sustained increase in the aged population continues to place mounting pressure on primary care resources. Discussing patients' preferences for clinical care and treatment is therefore assuming a greater prominence in order to prevent unnecessary hospitalisation at the end of life and to support people to die in their preferred place of care. The use of a mortality risk index as an adjunct to clinical decision making may help nurses to prioritise patients for whom advance care planning may be appropriate. Mortality risk indices do not appear to have been adopted widely in practice and further research is needed to explore their potential benefits and limitations to identify patients who may benefit from advance care planning.

#### Application of the Prediction Models

The following example illustrates how the MMRI-R and nurses' predictions assesses survival. An 83 year old man with heart failure, chronic kidney disease stage 4 and a creatinine greater than 3mg/dL has shortness of breath on exertion and poor mobility. He is dependent on carers who visit his home three times daily to assist with washing and meal preparation. His appetite has recently declined and his community case-management nurse thinks that he is likely to be in his last year of life. His MMRI-R is calculated as 39 (Supplementary table 1 available at Age and Ageing online). Both MMRI-R and nurses' predictions place him in the high risk group and estimates that he has between a 48% and 57% chance of survival at 12 months.

#### Limitations

In order to assess the additive value of nurses' predictions to the MMRI-R a cross-validation design was adopted in which the data was partitioned into derivation and validation sets. This approach resulted in two smaller samples and may contribute to an over-optimistic assessment of model performance [29].

# Conclusion

Mortality amongst community based frail older people is associated with an increasing MMRI-R score. Community case-management nurses can identify patients who are likely to die within 12 months with reasonable accuracy. The MMRI-R demonstrated comparable discrimination in identifying those at high or low risk of mortality and the addition of nurses' predictions improved its performance. Nurses' estimates of survival may trigger an objective assessment of mortality risk among those patients considered to be most likely to die within twelve months. An objective assessment, using the MMRI-R, a simple assessment tool that can be administered in the patient's own home, has the potential to improve nurses' confidence in their clinical decision making and encourage community casemanagement nurses to initiate timely conversations with frail older people and those close to them to determine end of life care preferences. However, the difference in performance between MMRI-R and nurses' predictions appears sufficiently small to suggest that nurses' judgement can be trusted to implement anticipatory care interventions at an early stage whilst avoiding an over-reliance on objective measures.

# **Conflicts of Interests**

All authors declare they have nothing to disclose.

#### Key Points

Identifying patients who are approaching the end of their lives can facilitate advance care planning and align future care to patients' preferences.

Prognostication is challenging among frail older patients with multiple progressive and fluctuating long-term conditions.

The MMRI-R is comparable to community nurses' intuition at identifying patients approaching end of life.

The MMRI-R and nurses predictions performed better in combination than alone.

Case-management <u>nurses such as UK community matrons</u> <u>Community nurses' judgement</u> should be trusted to initiate the early implementation of anticipatory care interventions.

# Key Words:

Older people, end of life care, community nurse, mortality, advance care planning

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# age and ageing

# **Revision Sheet**

#### Instructions for authors:

- 1. In the first column please briefly summarise each point raised by the referee or editor.
- 2. In the second column, briefly explain how you have responded to each point.
- 3. In the third column, give the location in the text of the modification with page and paragraph number reference.
- 4. Please upload this form to Manuscript Central alongside your revised paper.

Manuscript title:

Comparing the predictive ability of the Revised Minimum Dataset Mortality Risk Index

(MMRI-R) with nurses' predictions of mortality among frail older people: a cohort study.

# Referee 1

Point raised by referee (please summarise)	Response by author (briefly explain)	Location in text: Page and paragraph reference <u>*</u>
1.1 Nurse predictions conferred little benefit to the MMRI-R tool but the reverse is also true	We have reworded this finding to provide clarity for the reader, emphasising that combining MMRI-R with clinical prediction of survival improved the performance of each model.	Discussion: Page 6, paragraph 2
1.2 There is no critique of the MMRI-R tool which fails to take account of the speed of clinical deterioration or reversible causes. These are serious flaws of the tool.	The reviewer highlights a very important limitation that is applicable to most prognostic indices. We have amended our discussion to emphasise adopting a cautionary approach to using a prognostic index and the potential value of clinical prediction of survival to validate a prognostic score with the aim of prioritising advance care planning conversations.	Discussion: Page 7, paragraph 3
1.3 The risk of using the MMRI-R instead of effective person- centred communication to identify patients as a trigger to discuss advance care planning are not acknowledged.	The importance of person-centred communication as a trigger to advance care planning is acknowledged in the restructured discussion (see point 1.2).	Discussion: Page 7, paragraph 3
1.4 Areas for future research are not identified.	We agree with both reviewers and have amended the discussion section to include areas for future research.	Discussion: Page 7, paragraph 4

\*Page numbers and paragraph references refer to the revision copy with no tracked changes.

# **Referee 2**

Point raised by referee (please summarise)	Response by author (briefly explain)	Location in text: Page and paragraph reference
2.1 Prognostic indices developed on community dwellers are not mentioned	We thank the reviewer for highlighting this point and have amended the introduction to include a reference to a systematic review of prognostic indices by Yourman et al (2012). We have also provided an explanation why the community-based indices were not used in our study.	Introduction: Page 2, paragraph 2
2.2 The MMRI-R is not described in enough detail and justification of its use in the study is not explained.	We do not propose the MMRI-R as a Gold Standard, rather a tool based on a sample that had comparable characteristics and a mortality rate as patients case- managed by community matrons in the UK. Additional information has been provided about the MMRI-R tool in the introduction and mortality rates compared in the discussion.	Introduction: Page 1, paragraph 4 Discussion: Page 7, paragraph 2
2.3 More details about the characteristics of the community nurses, their experience and specialization would be useful since they are not specialty specific.	Community matrons are generalists working within a primary care setting. Additional details about their role has been included in the results section and further details are available in supplementary table 2.	Results: Page 4, paragraph 3 Supplementary Tabl 2
<ul> <li>2.4 The following aspects of the study design were questioned:</li> <li>Why the study was not preceded by a validation</li> </ul>	<ul> <li>While we have not undertaken a validation study of the MMRI-R in this specific population the choice of the MMRI was based on an estimated comparable mortality rate and has been addressed in point 2.2.</li> <li>We have acknowledged the lack of a formal validation study in the limitations section of the discussion.</li> <li>While we would have had greater statistical power had the dataset not been split into derivation and validation samples, the predictive power would have been estimated from the same dataset that the model</li> </ul>	See point 2.2 Limitations: Page 8, paragraph 1
<ul> <li>study</li> <li>The use of derivation and validation cohorts</li> <li>Unclear study aims</li> </ul>	has been optimised for thus contributing to overly optimistic estimates of prediction (Newson 2010). Our focus was to compare the performance of the MMRI-R and nurses' predictions, separately and together, in predicting survival. We accept the reviewer's point that our original aims were under- specified in the manuscript and have added a more precise statement in the methods section. In light of the reviewer's comments and our revisions we have slightly revised the title to take account of this.	Methods: Page 2, paragraph {
2.5 The process of generating risk groups should be described in a more understandable way.	We agree the description of the process was unnecessary technical at times and have endeavoured to provide more clarity. The process of generating risk groups has now been revised.	Methods: Page 3, paragraph 4
2.6 The sample size calculation is to be based on a rule of thumb for regression but seems unjustified as this was not a validation study for the MMRI-R.	The reviewer is correct and the original sample size calculation was based on the original plan that we would reweight the individual items in the MMRI-R. This statement was erroneously left in from an earlier version for which we apologise. We have rewritten this section based on the a priori assumptions of non- response among nurses and estimated mortality.	Methods: Page 3, paragraph 3
2.7 The procedure relating to fractional polynomial regression	In order to provide a graphical assessment of calibration of the Cox model, we followed the	Methods: Page 3, paragraph 6

is not explained in sufficient detail.	procedure proposed by Royston (2015) and Royston & Altman (2013) which approximates the log cumulative hazard function using fractional polynomial regression on the continuous variable of the aggregate prognostic score. On consideration of the reviewer's comments we recognise that a sufficiently technical description of this procedure is not possible within the word limit and might detract from the readability of the paper. We have therefore simplified this section.	
2.8 The rationale for developing a model using a combination of the tool and clinical prediction of survival is unclear.	The reviewer has previously identified that the study aims could have been clearer. We have therefore amended the objectives as outlined in 2.4 to make the assessment of the incremental value of nurses' predictions of mortality on the MMRI-R explicit. We have additionally provided further justification for our decision in the introduction.	Methods: Page 2, paragraph Introduction: Page 2, paragraph
2.9 The authors conclusion in relation to better performance of the combined model are inconsistent with the conclusion that it is sufficient to use nurses prediction alone. This could be rewritten to make the point better.	We thank the reviewer for bringing this inconsistent message to our attention. We have reviewed the abstract, discussion and conclusion in order to provide clarity and consistency in our conclusions.	Abstract: Page 1, paragraph Discussion: Page 7, paragraph Conclusion: Page 8, paragraph
2.10 The last key point does not describe the type of nurse whose judgements the authors refer to specifically.	We have amended the last key point in order to clarify that we are referring to community case-management nurses	Key Point: Page 8, point 5
2.11 Frailty has not been measured in their patients so the discussion about the electronic frailty index , which is not a mortality prediction tool, appears out of place and of no relevance.	Originally we intended to reflect that Stow et al's (2018) dataset was derived electronically from the patient electronic record demonstrated comparable accuracy to our dataset that was clinician-reported. On reconsideration we accept the reviewer and editor's comments that the inclusion of the study by Stow et al (2018) may create confusion for the reader and was designed to predict 3-month mortality rather than mortality at 1 year. We have therefore removed reference to the study from the discussion.	Discussion: Page 7, paragraph
2.12 Areas for future research are not identified.	We agree with both reviewers and have amended the discussion section to include areas for future research.	Discussion: Page 7, paragraph

# **Associate Editor**

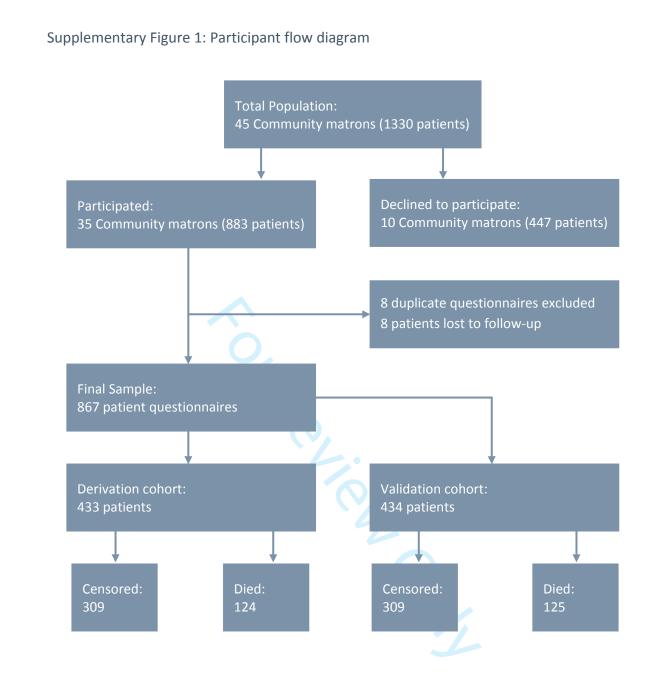
Point raised by referee (please summarise)	Response by author (briefly explain)	Location in text: Page and paragraph reference
3.1 The work validating the MMI-R in a UK population is not referred to within the paper.	We thank the associate editor for highlighting our omission and have made reference to Dutta and Dutta (2015) in the introduction to support our decision to test the MMRI-R in a community-based sample.	Introduction: Page 1, paragraph
3.2 The landscape of community nursing and advanced care planning may have changed since data collection 5 years ago	In response to the associate editor's comments we have revised the discussion section to indicate a greater prominence of advance care planning in light of the ageing population and increasing demands on primary care.	Discussion: Page 7, paragraph
3.3 Frailty has not been measured in their patients so the discussion about the electronic frailty index , which is not a mortality prediction tool, appears out of place and of no relevance.	Originally we intended to reflect that Stow et al's (2018) dataset was derived electronically from the patient electronic record demonstrated comparable accuracy to our dataset that was clinician-reported. On reconsideration we accept the reviewer and editor's comments that the inclusion of the study by Stow et al (2018) may create confusion for the reader and was designed to predict 3-month mortality rather than mortality at 1 year. We have therefore removed reference to the study from the discussion.	Discussion: Page 7, paragraph
3.4 The figures do not project in black and white/greyscale	Figure 1 and supplementary figure 2 have been amended accordingly.	Results: Page 6, Figure 1 Supplementary Dat Page 4, Figure 2
	RZ OS	Page 4, Figure 2

# Supplementary Data

## Supplementary Table 1. MMRI-R Scoring Sheet [1]

Admission to nursing home in the past three months $\Box$ $\Box$ 8Lost weight unintentionally in the last three months $\Box$ $\Box$ 5Renal failure $\Box$ $\Box$ 6Chronic heart failure $\Box$ $\Box$ 4Poor appetite $\Box$ $\Box$ 4Male $\Box$ $\Box$ 4Male $\Box$ $\Box$ 4Short of breath $\Box$ $\Box$ 4Short of breath $\Box$ $\Box$ 8Cancer (if yes – see Age and Cancer worksheet; if no continue) $\Box$ $\Box$ Age of patient/resident at last birthday $Age screwith cancerAge screwith cancerDeteriorated cognitive skills or status in the past three months\Box\BoxADL screwith cancerADL screwith cancerADL screwith cancer$		Yes	No	Points
Lost weight unintentionally in the last three monthsImage: constraint of the set of			-	
Lot rong it content failureImage: content failureImage: content failureRenal failureImage: content failureImage: content failurePoor appetiteImage: content failureImage: content failureMaleImage: content failureImage: content failureDehydratedImage: content failureImage: content failureShort of breathImage: content failureImage: content failureCancer (if yes – see Age and Cancer worksheet; if no continue)Image: content failureAge of patient/resident at last birthdayAge content failureAge of patient/resident at last birthdayImage: content failureDeteriorated cognitive skills or status in the past three monthsImage: content failureADL scorewithout cognitiveContent failureADL scorewith cognitive-2-21declineADL scorewith cognitiveADL scoreWith cognitive-2-21declineADL scoreADL scoreADL scoreADL scoreADL scoreADL scoreADL s				
Chronic heart failure $\Box$ $\Box$ $4$ Poor appetite $\Box$ $\Box$ $4$ Male $\Box$ $\Box$ $5$ Dehydrated $\Box$ $\Box$ $4$ Short of breath $\Box$ $\Box$ $4$ Cancer (if yes – see Age and Cancer worksheet; if no continue) $\Box$ $\Box$ Age of patient/resident at last birthday $\Box$ $Age scoreAge of patient/resident at last birthdayAge score13-20Deteriorated cognitive skills or status in the past three months\Box\BoxADL scorewith out cognitive0-16ADL scorewith cognitive-2-21ADL scor$	Lost weight unintentionally in the last three months			5
Poor appetiteIIIMaleIII5DehydratedII4Short of breathII4Cancer (if yes – see Age and Cancer worksheet; if no continue)IIAge of patient/resident at last birthdayAge scorewithout2-9CancerAge scorewithout2-9CancerAge score13-20Deteriorated cognitive skills or status in the past three monthsIIDeteriorated cognitive skills or status in the past three monthsIIADL scorewith cognitive0-16declineADL scorewith cognitiveADL scorewith cognitive2-221declineADL score2-221declineADL score3-221declineADL score3-221declineADL score3-221declineADL score3-221declineADL score3-221declineADL score3-221declineADL score <t< td=""><td>Renal failure</td><td></td><td></td><td>6</td></t<>	Renal failure			6
Male $\Box$ $\Box$ $\Box$ $\Box$ Dehydrated $\Box$ $\Box$ 4Short of breath $\Box$ $\Box$ 4Cancer (if yes – see Age and Cancer worksheet; if no continue) $\Box$ $\Box$ Age of patient/resident at last birthday $Age score$ $Age score$ Age of patient/resident at last birthday $Age score$ $Age score$ Male $Age score$ $Age score$ $13-20$ Deteriorated cognitive skills or status in the past three months $\Box$ $\Box$ $ADL score$ $without$ $cognitive$ $0-16$ $decline$ $ADL score$ $with cognitive$ $2-221$ $decline$ $ADL score$ $with cognitive$ $-2-21$	Chronic heart failure			4
InstructImage: constraint of the set of	Poor appetite			4
Short of breath $\Box$ $\blacksquare$	Male			5
Cancer (if yes – see Age and Cancer worksheet; if no continue) $\Box$ $\Box$ $\Box$ Age of patient/resident at last birthday	Dehydrated			4
Age of patient/resident at last birthday       Age score       Age score       2-9         Age of patient/resident at last birthday       Age score       13-20         Deteriorated cognitive skills or status in the past three months       I       I         ADL score       without       0-16         decline       ADL score       with cognitive         decline       ADL score       0-16	Short of breath			8
Age of patient/resident at last birthday       2-9         cancer       Age score         Age score       13-20         with cancer       13-20         Deteriorated cognitive skills or status in the past three months $\Box$ $\Box$ ADL score       without       0-16         cognitive       decline       0-16         decline       ADL score       with cognitive         with cognitive       2-9       0-221         decline       -2-21       0-221         decline       -2-21       0-221	Cancer (if yes – see Age and Cancer worksheet; if no continue)			
$\begin{array}{c c c c c } \hline \mbox{with carcer} & 13-20 \\ \hline \mbox{with constraint} & \Box & \Box & 13-20 \\ \hline \mbox{with constraint} & \Box & \Box & 13-20 \\ \hline \mbox{out constraint} & \Box & \Box & 13-20 \\ \hline \mbox{out constraint} & \Box & \Box & 13-20 \\ \hline \mbox{out constraint} & \Box & \Box & 13-20 \\ \hline \mbox{ADL score} & & & & & & & & \\ \hline \mbox{with out constraint} & & & & & & & & & & \\ \hline \mbox{out constraint} & & & & & & & & & & & & & & & & & & &$	Age of patient/resident at last birthday	withou	t	2-9
Image: Construction of the second structure       Image: Constructure       ADL score       Note that the second structure       Note t		-		13-20
without cognitive decline ADL score with cognitive decline -2-21 decline	Deteriorated cognitive skills or status in the past three months			
with cognitive -2-21 decline		<b>withou</b> cogniti	t ve	0-16
TOTAL MMRI-R SCORE 0-85		with co	ognitive	-2-21
	TOTAL MN	/IRI-R S	CORE	0-85

[1] D. Porock, D. Parker-Oliver, G. Petroski, M. Rantz, The MDS Mortality Risk Index: the evolution of a method for predicting 6-month mortality in nursing home residents, BMC Research Notes 3(200) (2010).



Supplementary Table 2: Characteristics of Community Matron Participants

	21
Practice Nurse	21
Specialist Nurse	3
- F	7
Not known	3
Age Group	
20-29	0
30-39	2
40-49	14
50-59	15
60-69	1
Not known	3
Nº Patients on Caseload	
0-19	5
20-29	8
30-39	7
40-49	9
50+	3
Not known	3
Average number of patients	32.4
	9

