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Hospital Workforce Engagement and Inpatient Mortality Rate: Findings from the English National Health Service Staff Surveys

Badgett RG, MD¹; Jonker L, PhD²; Xirasagar S, MBBS, PhD³

- 1. Departments of Internal Medicine and Population Health. University of Kansas School of Medicine-Wichita. Wichita, KS. USA
- 2. Research & Development Department, North Cumbria Integrated Care NHS Foundation Trust, Carlisle, CA2 7HY, UK
- 3. Department of Health Services Policy and Management. Arnold School of Public Health. University of South Carolina. Columbia, SC. USA

Corresponding author:

Robert Badgett, MD Department of Internal Medicine University of Kansas School of Medicine-Wichita 1010 N Kansas Wichita, KS 67214 <u>rbadgett@kumc.edu</u> 316-293-2650

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Running title: hospital workforce engagement and in-patient mortality

Word count: 2226 (without abstract) Tables: 3 Figures: 3

Abstract

Background: Healthcare workforce engagement may represent a proactive approach against provider burnout, a widely prevalent condition that is associated with poor patient outcomes.

Objective: We examine whether workforce engagement is associated with better hospital performance, measured as lower inpatient mortality, in English National Health Services (NHS) acute Trusts.

Design: Panel study using cross-lagged regression, applying an optimally time-lagged value of the dependent variable as covariate to account for unmeasured Trust characteristics.

Participants: NHS acute Trusts and respondents to the NHS Staff Surveys, 2012-2018.

Main Measures: We measured engagement using three survey questions corresponding to validated engagement factors, and hospital performance using the summary Hospital-level Mortality Indicator (SHMI). In the first analyses, associations of SHMI (dependent variable) with workforce engagement in the current, prior, and subsequent years were studied to find the optimum lag period for lagged regression analysis. In the subsequent cross-lagged regression analysis, bi-directional associations between SHMI and engagement were studied. Heterogeneity in engagement components across Trusts was studied in detail for the year 2017.

Key Results: In the first analyses, current SHMI was negatively associated with engagement in the current year ($\beta = -0.044$; p = 0.035) more than with the prior year ($\beta = -0.037$; p = 0.049). In the second analysis: a) engagement predicted same-year SHMI after controlling for prior year SHMI ($\beta = -0.044$; p = 0.035). A 1-unit higher engagement score was associated with 4.4% lower SHMI) SHMI predicted engagement in the same year ($\beta = -0.066$; p = 0.001) after controlling for prior-year engagement. More in-depth analysis showed high inter-trust heterogeneity on all three engagement factors ($l^2 > 85\%$).

Conclusion: Higher workforce engagement predicts lower mortality which in turn predicts engagement. Heterogeneity in workforce well-being suggests an opportunity to foster mutual learning across Trusts.

BACKGROUND

Healthcare management research has focused on burnout as nearly half of physicians and one-third of the U.S. working population report burnout.¹ Similarly, about 40% of consultants in the National Health Service (NHS) report burnout.² Burnout also affects the worker's role performance and may reduce organizational performance.³ Research focusing on positive workforce states may be useful to facilitate better outcomes for both provider well-being and organizational outcomes. Aspirational goals to improve workforce morale could be 1) engagement, a validated construct defined as "a positive fulfilling, work-related state of mind characterized by vigor, dedication, and absorption"⁴ and 2) thriving, which is a combination of engagement and a drive for learning or self-improvement. These higher states are associated with extra-role performance and prosocial actions^{5,6}, reduced burnout, and improved employee and organizational performance in business organizations.^{4,7,8} There are no data sources that capture healthcare workforce measures of thriving; however engagement is measured annually in the English Trusts of the National Health Service (NHS). This study examines whether workforce engagement in the NHS is associated with lower inpatient mortality.

METHODS

We executed a panel study using the annual NHS Staff Surveys from 2012 to 2018. We limited the study to the Trusts of the English NHS because availability of the adjusted mortality indicator is limited to English hospitals, the organizational outcome of interest. Due to prior findings that engagement varies by Trust type and survey respondents' profession⁹, we limited our study to acute, non-specialty Trusts. Acute trusts are organizations within the English NHS that include one or more hospitals providing secondary health services with financial and managerial autonomy to improve organizational performance and care quality. The survey was administered by each Trust using national guidance from the NHS, Through the year 2015, the NHS required all staff to be surveyed unless a Trust had more than 600 staff in which case a random sample of staff beyond 600 were

sampled (Table 1). Since 2016, the threshold for allowing a supplemental random sample was a staff size of 1,250.

The NHS Staff Survey considers engagement to have three dimensions: psychological motivation, involvement in the organization, and advocacy for the organization¹⁰. We used Schaufeli's more widely used and validated definition of engagement which captures only the motivation dimension⁴. In this view, advocacy and involvement are considered outcomes rather than components of engagement^{5,6}. In the Staff Surveys, engagement was measured by Likert scale responses (1 never to 5 always) to three questions representing dedication ("I am enthusiastic about my job"), absorption ("time passes quickly when I am working"), and vigor ("I look forward to going to work") adapted from Schaufeli's validated UWES-9 survey.⁴ Individual mean scores across all three questions were aggregated into a Trust level workforce 'engagement score'. The Summary Hospital-level Mortality Indicator (SHMI) data were obtained from NHS Digital.¹¹ Each Trust's SHMI is the ratio of observed deaths during hospitalization to expected number of deaths based on the average mortality rate for English hospitals adjusted for the Trust's case-mix.

Our goal was to study whether same year engagement predicts SHMI and in- turn, whether SHMI predicts engagement.¹² The associations between engagement and mortality were assessed in a two-step process. As shown in Figure 1, we first identified the optimal lag between workforce engagement and mortality by studying SHMI prediction by the Trust-level engagement score in the same year and prior years (Analysis 1). Using the optimal lag, we used cross-lagged regression, a technique to adjust for baseline mortality of the Trust by including the optimum lag SHMI as an independent variable in the model when assessing the net association of engagement with the mortality outcome (Analysis 2). The associations tested are shown in Figure 2. Crossed-lagged regression allows estimation of the "unique effects of job attitudes on later performance (with baseline performance controlled) and of performance on later job attitudes (with baseline job attitudes controlled)".¹² Thus, crossed-lagged regression, by controlling for the baseline status of the

dependent variable, accounts for measured and unmeasured Trust characteristics that may confound the association of interest.

We weighted all regression analyses by the number of respondents at each Trust who reported having frequent contact with patients, assuming frequent patient contact to be a a proxy for caregiver staff. The appendix Table displays the results of alternate approaches to weighting that were considered but not used. We used the number rather than the proportion of clinical respondents as using the proportion would ignore Trust size. The additional contribution of each independent variable in predicting the dependent variable beyond the effect of the other variables in the regression was assessed by dominance analysis¹³. Each variable's contribution to total variance is designated R_{A}^{2} .

We studied 2017 engagement data in depth, using each factor of engagement separately, to assess the role of heterogeneity in workforce engagement components across Trusts in predicting mortality. We used random effects analyses with the Hartung-Knapp estimator,¹⁴ with responses dichotomized by always/often versus other. Heterogeneity in the rates of engagement across Trusts was measured with the l² statistic. The l² statistic is the proportion of total variation across Trusts that is due to heterogeneity between Trusts.¹⁵ The l² value is interpreted as: "0% to 40%: may not be important; 30% to 60%: moderate heterogeneity; 50% to 90%: substantial heterogeneity; 75% to 100%: considerable heterogeneity"¹⁵. All statistical analyses were performed with the R Programming Language¹⁶ version 3.6.2 using the procedure programs for dominance analysis¹⁷ and meta¹⁸.

RESULTS

The number of Trusts participating in the survey with data on workforce engagement varied between 142 in 2012 and 132 in 2018. Average number of staff respondents per Trust varied from 404

in 2012 to 5721 in 2018 (Table 1). However, engagement data are available for fewer than the total number of participating Trusts because not all Trusts asked engagement questions every year (Table 1). The average staff response rate across all Trusts was 43% (range: 41% in 2015 to 49% in 2013).

The proportion of staff per Trust reporting frequent contact with patients varied between 68% and 71%. Pooling data for all years with engagement data available, Trust SHMI ranged between 0.9996 and 1.0044, a variation range of -0.04% and +0.44% relative to expected mortality. The median SHMI was similar across study years. The mean engagement score across Trusts (all staff types included) ranged between 3.83 and 3.94, with no significant differences in median engagement scores by year.

In analysis 1, using 2013-2017 pooled data, SHMI in a given year was negatively associated with engagement in the same year (β =-0.044; p = 0.035), i.e. each unit difference in engagement score was associated with a 4.4% lower SHMI in the same year. Same year association was greater than the association in the prior year which, though the latter was borderline statistically significant (Table 2). Thus, a short lag of one year was deemed optimal to use in the second analysis (to control for unmeasured Trust-level effects on SHMI). In analysis 2, using the short lag, workforce engagement predicted same-year SHMI after controlling for prior year SHMI (β =-0.043; p = 0.035), i.e. after adjusting for prior year SHMI, a 1-unit higher engagement score (for a simplistic example, if a Trust moved all of its responses from 'sometimes' to 'often') was associated with 4.3% lower SHMI in the same year (β =-0.066; p = 0.001) after controlling for prior-year engagement (Table 3). The reverse association persisted: index year SHMI predicted engagement in the ensuing year (β = -0.186; p < 0.001).

Dominance analysis showed that engagement in the same year accounted for an additional 1% of SHMI variability after adjusting for prior year SHMI (Table 3; R^2_{Δ} = 0.011). Engagement in a given year accounted for 43% of the variation in engagement in the ensuing year (data not shown).

In a detailed, univariate analysis of 2017 data, the proportions of all-staff respondents at each Trust who responded often/always to the three engagement questions were: dedication 74% (range 67% to 80%), absorption 77% (72% to 84%), and vigor 58% (50% to 67%) with the l^2 statistic indicating considerable¹⁹ inter-Trust heterogeneity on all three questions ($l^2 > 80\%$).¹⁹ The correspondence of

engagement with SHMI is presented in Figure 3 showing the percentage of respondents reporting always feeling dedicated and the corresponding Trust SHMI in Trusts with the 5 highest and 5 lowest scores on all-staff dedication, a component of engagement.

DISCUSSION

Our finding that healthcare workforce engagement is associated with lower mortality is consistent with previously documented findings in business settings showing an association between engagement and organizational performance.^{6,8,12} This association was reported in an in-house, NHS organizational report using two years of NHS data.⁹ Our study, using data across seven years shows that this association is stable across multiple years as we had the ability to treat each year's data on a Trust as a unique observation, and therefore changes in engagement, positive or negative contributed to the observed associations in our study. Our adjustment for prior year SHMI mitigates the association- inflating effect of treating repeated measures as unique, independent observations in pooled data analysis. A new finding of our study is the reverse association, that current SHMI predicts current engagement after adjusting for prior engagement. Another new finding is that of 'considerable' heterogeneity in staff engagement across Trusts (l² > 80%).

The engagement effect in the NHS appears to be similar to its effect on business measures of organizational performance that is documented in the business literature. First, the lag period for which engagement continues to impact performance is short, documented to be less than a year.¹² Second, engagement levels appear to be durable over time. Our dominance analysis found that 43% of variance in engagement in the ensuing year is explained by the engagement in the current year. This supports Schaufeli's finding that 31% to 53% of engagement in a second measurement is explained by the first measurement. The less than 100% persistence of engagement levels suggests that organizational leaders and culture may affect engagement.⁴

The reciprocal directions of association, that engagement affects mortality and vice versa suggests that organizations can create a virtuous positive cycle. This cycle may represent a Matthew effect.^{20,21} The Matthew effect describes preferential attachment and cumulative advantage in which success-breeds-success, and advantage tends to beget further advantage.²⁰ This may contribute to the considerable heterogeneity of engagement that we found across Trusts. Whether this is mediated by active human resource policies to increase employee engagement through soliciting employee participation, decentralized decision making, and team-building as postulated in the literature should be explored in future studies in order to identify organizational approaches that are effective in increasing workforce engagement.⁷

The observed heterogeneity in workforce engagement across Trusts suggests opportunities to foster mutual learning and sharing of organizational development strategies between Trusts with high engagement and those with lower than average engagement levels. Rose has documented successful efforts at created learning and improvement across units of a dispersed organization using positive deviance.^{22,23}

Our findings for engagement agree with prior findings reported for burnout. In a recent review of burnout and the quality of care ³, two studies of emotional exhaustion and adjusted mortality in over 50 clinical sites found a similar proportion of inter-site variation in mortality being explained by emotional exhaustion ²⁴, ²⁵.

Our study has some limitations. First, we had access to aggregated data at the level of the Trusts rather than responses of individual staff or staff categories. We did not assess other organizational factors that have been studied or postulated to affect organizational performance such as mortality. These factors include high-performance human resource management (HRM)⁷, research activity ²⁶, openness²⁷, and a Freedom to Speak Up Index²⁸. As HRM is considered an antecedent to engagement, and extra-role behavior is considered an expected outcome of employee wellbeing^{5,6}, we did not include them in our assessment of engagement. However, the additional impact of these

factors merits future research. Additional factors, currently not captured in the Staff Survey, merit inclusion in future Surveys. These factors include thriving⁵ and calling²⁹. As thriving (ongoing workforce learning and improvement combined with engagement) and calling (defined as "work that a person perceives as his purpose in life"²⁹) may better predict organizational performance, addition of validated measures of the learning dimension to the NHS surveys may improve the utility of survey findings for improving Trust performance.^{30–32}

Our second limitation is the lack of availability of shorter lags to study. The strongest lag between work attitude and job performance may be less than a year.^{12,33} This is consistent with our finding that engagement and SHMI effect each other the most when measured during the same year. Presumably, as the year progresses, the impact of engagement and SHMI start impacting each other before the year is complete. The NHS might survey their staff every six months as done by Google.³⁴

In conclusion, our findings indicate that staff engagement is important to maximize patient outcomes, particularly in-hospital mortality. Heterogeneity in workforce engagement across Trusts suggests an opportunity for improvement, specifically through mutual sharing of recipes for success and learning between Trusts. With increasing roles of hospitalists in the care management of inpatients in the United States and internationally, our findings present opportunities for hospital and multi-hospital system leaders to initiate human resource initiatives to improve healthcare workforce well-being, prevent and combat burnout and improve patient outcomes.

ADDENDUM

Data files and analytic code are available at <u>https://ebmgt.github.io/nhs</u>. This project received no funding.

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Figure 1. Analysis 1: determining the length of the optimal lag between engagement and SHMI

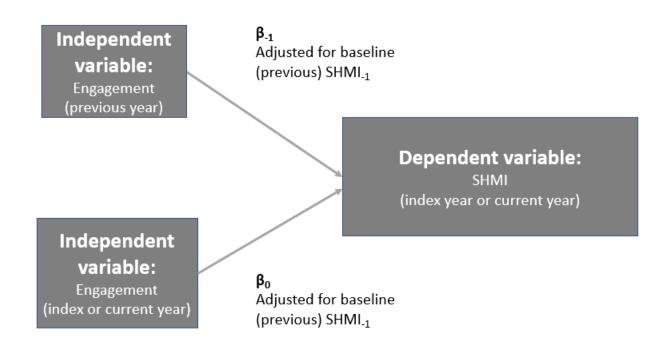
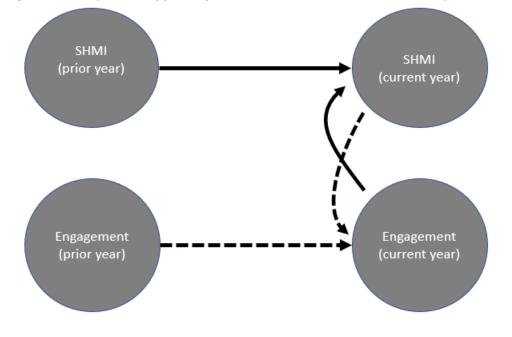


Figure 2. Analysis 2: lagged regression based on the results of analysis 1.



Measuring the impact of engagement on SHMI, controlling for prior SHMI.

---- Measuring the impact of SHMI on engagement, controlling for prior engagement.

	Number of Trusts with public data	Number of Trusts with public engagement data	Trust staff size up to which full census participation required*	Surveys distributed	Response rate (range across Trusts) ‡	Proportion reporting frequent patient contact (range across trusts)	Mean engagement score§ (standard deviation)
2012	142	142	600	116,455	49% (27% to 71%)	71% (64% to 85%)	3.83 (0.086)
2013	141	140	600	249,592	49% (30% to 78%)	71% (61% to 80%)	3.86 (0.081
2014	138	136	600	378,437	42% (23% to 82%)	70% (62% to 80%)	3.85 (0.093)
2015	137	134	600	469,219	41% (25% to 78%)	68% (57% to 80%)	3.94 (0.075)
2016	138	134	1,250	643,320	43% (29% to 77%)	68% (59% to 80%)	3.94 (0.066)
2017	136	132	1,250	748,678	43% (27% to 73%)	67% (60% to 78%)	3.92 (0.085)
2018	132	130	1,250	755,206	43% (25% to 72%)	68% (58% to 78%)	3.93 (0.064)

Table 1. Characteristics of surveyed populations in the English, acute Trusts.

* If trust staff size was less than the maximum size, a census was required. If the staff size was more than the maximum size, additional random sampling was allowed. Since 2016, the NHS has encouraged surveying all staff. This information was gathered from each year's technical or guidance document at https://www.nhsstaffsurveys.com/.

† This information was gathered from each year's technical or guidance document at https://www.nhsstaffsurveys.com/.

‡ Response rates gathered from the yearly Staff Surveys themes or key findings reports from https://www.nhsstaffsurveys.com/.

§ For engagement score: 5=always; 4 = often; 3 = sometimes; 2 = rarely; 1 = never.

Table 2. Regression analyses results showing associations of workforce engagement, in the previous and current years, with current SHMI* pooled across the years 2013-2017†,

(n= 683 Trust-years from 143 Trusts).

	Previous year engagement ('long lag')	Same year engagement ('short lag')		
B estimate	-0.037	-0.044		
Significance (p)	0.049	0.035		
R ² (proportion of SMHI variation explained	0.792	0.791		
by model) ‡				
* SHMI. Summary Hospital-level Mortality Indicator.				
† Models include the previous year's SMHI to a	djust unmeasured confo	unders and are weighted by the number		

of clinical respondents in each Trust.

 $\ddagger R^2$ is adjusted for the number of predictors in the model

Table 3. Cross-lagged regression analyses showing SHMI* predicting current engagement adjusted for prior year engagement, and current engagement predicting current SHMI adjusted for prior year SHMI, 2013-2017†.

SHMI in the current year predicted by	Engagement in the current year predicted by		
current engagement	current SHMI		
(Covariate previous SHMI)	(Covariate previous engagement)		
-0.044	-0.066		
p = 0.035	P = 0.001		
$R_{\Delta}^{2} = 0.011 \ddagger$	$R_{\Delta}^2 = 0.428 \ddagger$		
* SHMI. Summary Hospital-level Mortality Indicator.			
† Models are weighted by the number of clinical respondents in each Trust.			
$\ddagger R_{\Delta}^2$ is the change in the R^2 due to the independent variabl	e alone.		

Figure **3.** Five NHS Trusts with respective highest and lowest reported rates for the dedication component of engagement.

Trust*	SHMI+		Proportion	95%-CI
	Trusts with hig	ghest dedication		
RHM RAP RBA RRF RD3	0.95 0.86 0.89 1.17 0.86	+ 	0.78 0.78 0.79	[0.77; 0.79] [0.76; 0.80] [0.76; 0.80] [0.75; 0.83] [0.78; 0.82]
	Trusts with lo	west dedication		
RE9 RXH RM3 RJL RQ6	1.20 1.00 0.89 1.19 1.05	+	0.69 0.69 0.69	[0.66; 0.70] [0.68; 0.70] [0.67; 0.71] [0.67; 0.71] [0.67; 0.71]
Random effects mo Heterogeneity: $I^2 = 859$	6 0 0.2	0.4 0.6 0.8 of 'always' dedication	0.74	[0.74; 0.74] (all Trusts)

Notes:

* Trust organisational code

[†] Summary Hospital-level Mortality Indicator

Appendix Table. Cross-lagged regression analyses using different weighting, showing current engagement predicting current SHMI adjusted for prior year SHMI, 2013-2017†.

Weighting method	SHMI in the current year predicted by
	current engagement
	(Covariate previous SHMI)
Number of clinical respondents	-0.044
(same analysis as Table 2)	p = 0.035
	R ² (full model) ‡ = 0.791
	$R_{\Delta}^2 = 0.011$ §
Number of all respondents	-0.043
-	p = 0.780
	R^2 (full model) $\ddagger = 0.790$
	$R_{\Delta}^2 = 0.011$ §
No weighting	-0.006
	p = 0.040
	R ² (full model) ‡ = 0.751
	$R_{A}^{2} = 0.011$ §

 $\ddagger R^2$ is adjusted for the number of predictors in the model § R_{Δ}^2 is the change in the R² due to the independent variable alone.

STROBE Statement—Checklist of items that should be included in reports of *cohort studies*

	Item No	Recommendation	Paragraph No.
Title and abstract	1	 (a) Indicate the study's design with a commonly used term in the title or the abstract (b) Provide in the abstract an informative and balanced 	a) Title b) NA
		summary of what was done and what was found	
Introduction			
Background/rationale	2	Explain the scientific background and rationale for the investigation being reported	Background, para 1
Objectives	3	State specific objectives, including any prespecified hypotheses	Background, para 1
Methods			
Study design	4	Present key elements of study design early in the paper	Methods para 1
Setting	5	Describe the setting, locations, and relevant dates, including periods of recruitment, exposure, follow-up, and data collection	Methods para 1
Participants	 6 (a) Give the eligibility criteria, and the sources and methods of selection of participants. Describe methods of follow-up (b) For matched studies, give matching criteria and number of exposed and unexposed 		Methods para 1
Variables	7	Clearly define all outcomes, exposures, predictors, potential confounders, and effect modifiers. Give diagnostic criteria, if applicable	Methods para 2
Data sources/ measurement	8*	For each variable of interest, give sources of data and details of methods of assessment (measurement). Describe comparability of assessment methods if there is more than one group	Methods paras 1,2
Bias	9	Describe any efforts to address potential sources of bias	Methods para 3,4
Study size	10	Explain how the study size was arrived at	Methods para 1
Quantitative variables	uantitative variables 11 Explain how quantitative variables were handled in the analyses. If applicable, describe which groupings were cho and why		Methods para 2,3,4
Statistical methods	12	 (a) Describe all statistical methods, including those used to control for confounding (b) Describe any methods used to examine subgroups and interactions (c) Explain how missing data were addressed (d) If applicable, explain how loss to follow-up was addressed 	Methods para 3,4, 5
		(<u>e</u>) Describe any sensitivity analyses	
Results Participants	13*	 (a) Report numbers of individuals at each stage of study—eg numbers potentially eligible, examined for eligibility, confirmed eligible, included in the study, completing follow-up, and analysed (b) Give reasons for non-participation at each stage (c) Consider use of a flow diagram 	Results para 1 and Table 1

Descriptive data		14*	(a) Give characteristics of study participants (eg demographic, clinical, social) and information on exposures and potential confounders	Results para 1 and Table 1	
			 (b) Indicate number of participants with missing data for each variable of interest (c) Summarise follow-up time (eg, average and total amount) 		
Outcome data		15*	Report numbers of outcome events or summary measures over time	Results para 1	
Main results	16	estimat which c (b) Rep categor (c) If re	e unadjusted estimates and, if applicable, confounder-adjusted tes and their precision (eg, 95% confidence interval). Make clear confounders were adjusted for and why they were included port category boundaries when continuous variables were rized elevant, consider translating estimates of relative risk into te risk for a meaningful time period	Results para 2,3	
Other analyses	17	Report	Report other analyses done—eg analyses of subgroups and interactions, and sensitivity analyses		
Discussion					
Key results	18	Summa	arise key results with reference to study objectives	Discussion para 1	
Limitations	19	Discuss limitations of the study, taking into account sources of potential bias or imprecision. Discuss both direction and magnitude of any potential bias		Discussion paras 7,8	
Interpretation	20	Give a limitation other re	Discussion para 9		
Generalisability	21	Discus	Discuss the generalisability (external validity) of the study results		
Other information					
Funding	22	study a	e source of funding and the role of the funders for the present ind, if applicable, for the original study on which the present is based	Addendum, para 1	

*Give information separately for exposed and unexposed groups.

Note: An Explanation and Elaboration article discusses each checklist item and gives methodological background and published examples of transparent reporting. The STROBE checklist is best used in conjunction with this article (freely available on the Web sites of PLoS Medicine at http://www.plosmedicine.org/, Annals of Internal Medicine at http://www.annals.org/, and Epidemiology at http://www.epidem.com/). Information on the STROBE Initiative is available at http://www.strobe-statement.org.