Development and validation of the Surgical Outcome Risk Tool (SORT)

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Background: Existing risk stratification tools have limitations and clinical experience suggests they are not used routinely. The aim of this study was to develop and validate a preoperative risk stratification tool to predict 30-day mortality after non-cardiac surgery in adults by analysis of data from the observational National Confidential Enquiry into Patient Outcome and Death (NCEPOD) Knowing the Risk study. **Methods:** The data set was split into derivation and validation cohorts. Logistic regression was used to construct a model in the derivation cohort to create the Surgical Outcome Risk Tool (SORT), which was tested in the validation cohort.

Results: Prospective data for 19097 cases in 326 hospitals were obtained from the NCEPOD study. Following exclusion of 2309, details of 16 788 patients were analysed (derivation cohort 11 219, validation cohort 5569). A model of 45 risk factors was refined on repeated regression analyses to develop a model comprising six variables: American Society of Anesthesiologists Physical Status (ASA-PS) grade, urgency of surgery (expedited, urgent, immediate), high-risk surgical specialty (gastrointestinal, thoracic, vascular), surgical severity (from minor to complex major), cancer and age 65 years or over. In the validation cohort, the SORT was well calibrated and demonstrated better discrimination than the ASA-PS and Surgical Risk Scale; areas under the receiver operating characteristic (ROC) curve were 0.91 (95 per cent c.i. 0.88 to 0.94), 0.87 (0.84 to 0.91) and 0.88 (0.84 to 0.92) respectively (P < 0.001).

Conclusion: The SORT allows rapid and simple data entry of six preoperative variables, and provides a percentage mortality risk for individuals undergoing surgery.

Paper accepted 25 July 2014 Published online in Wiley Online Library (www.bjs.co.uk). **DOI:** 10.1002/bjs.9638

Introduction

Approximately 8 million surgical procedures are performed in the UK each year, and 230 million worldwide¹. Accurate risk stratification facilitates meaningful informed patient consent and shared decision-making. It might also identify high-risk patients who could benefit from targeted interventions including goal-directed fluid therapy², postoperative respiratory support³ and admission to critical care⁴. There is some evidence that appropriately targeted interventions⁵ can reduce mortality^{6,7}, morbidity and length of hospital stay⁸.

Clinical judgement alone is not a reliable predictor of adverse outcome⁹. Thus a variety of risk assessment tools have been developed to help clinicians calculate perioperative risk¹⁰ that complement investigations for identifying high-risk patients, such as cardiopulmonary exercise testing^{11,12} and biomarker assays¹³. Exercise testing facilities are not available routinely^{14,15} and are inappropriate in urgent or emergency surgical patients. The potential value of biomarkers such as N-terminal pro-B-type natriuretic peptide is still emerging^{16–18}. Risk stratification tools remain the most readily and widely available means of determining perioperative risk.

Some risk stratification tools consist entirely of preoperative variables, such as the American Society of Anesthesiologists Physical Status (ASA-PS) grade¹⁹ and the Surgical Risk Scale (SRS)²⁰. Other tools combine preoperative data with intraoperative/postoperative variables, such as the Physiological and Operative Severity Score for the enUmeration of Mortality and morbidity

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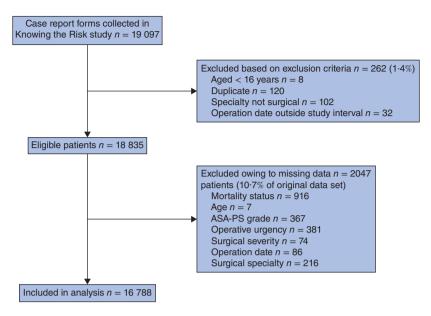


Fig. 1 Reasons for exclusion from the study. ASA-PS, American Society of Anesthesiologists Physical Status

(POSSUM)²¹, and the subsequent Portsmouth version (P-POSSUM)²². Clinical experience suggests that their incorporation into clinical practice is variable. Reluctance to use them may relate to concerns over the accuracy, complexity²³ and/or accessibility of data, for example if a blood test is required. A recent systematic review²³ of risk stratification tools validated in heterogeneous patient cohorts found that P-POSSUM and the SRS were the most widely validated and accurate risk stratification tools available. However, both have their limitations. Overall, clinical experience suggests that existing risk stratification tools are not widely used in the UK or elsewhere.

In 2011, the National Confidential Enquiry into Patient Outcome and Death (NCEPOD) published Knowing the Risk, a report assessing perioperative care¹⁴. Key recommendations included the introduction of a national system to identify patients at high risk of morbidity and death after surgery; and that a mortality risk assessment should be made explicit to patients before surgery, and documented on the consent form. It was also recommended that high-risk patients be identified before operation to aid planning and provision of critical care resources. An important step towards meeting these recommendations would be achieved by a risk stratification tool that consisted entirely of readily available preoperative variables which enabled easy calculation of a predicted percentage mortality. Thus, a post hoc analysis of data collected in the Knowing the Risk study was conducted, to develop and validate a risk stratification tool that met these requirements, and compare it with existing preoperative risk stratification tools.

Methods

The Knowing the Risk study was a prospective, multicentre, observational cohort study. The initial data collection was undertaken for 7 days from midnight on 1 March until midnight on 8 March 2010. National Health Service (NHS) hospitals in England, Wales and Northern Ireland, public hospitals in the Isle of Man, Guernsey and Jersey, and independent-sector hospitals in all these regions, were invited to participate. Perioperative data were collected on paper forms by anaesthetists at the time of surgery. Details of the study design and data collection methodology have been described previously¹⁴ and an extract of the form can be found in Appendix S1 (supporting information). Data were collected without obtaining patient consent or ethical approval as the study was not defined as research under the Health Research Authority (formerly National Research Ethics Service). However, Section 251 approval had been obtained from the Health Research Authority Confidentiality Advisory Group (formerly the National Information Governance Board, NIGB) to collect identifiable data without consent. For the work described here, further NIGB approval was obtained to extend the data-holding interval for England and Wales. Data Access Agreements were drawn up for each Northern Ireland Trust during the Knowing the Risk study. Approval was subsequently obtained from the Privacy Advisory Committee for Northern Ireland to extend the data-holding interval.

Table 1	Descriptive	data for par	tients exc	luded for 1	nissing
mortali	ty status in c	omparison	with the	whole data	ı set

	Mortality data	Mortality data
	missing	known
	(n = 916)	(<i>n</i> = 16788)
Mean age (years)	60.5	55.8
Missing data	59 (6.4)	-
Sex ratio (M : F)	428:488	7481:9307
ASA-PS grade		
I. I.	253 (27.6)	5416 (32.3)
II	370 (40.4)	7585 (45.2)
III	179 (19.5)	3339 (19.9)
IV	28 (3.1)	417 (2.5)
V	1 (0.1)	31 (0.2)
Missing data	85 (9.3)	-
Urgency of surgery		
Elective	569 (62·1)	10 987 (65.4)
Expedited	95 (10·4)	2136 (12.7)
Urgent	176 (19·2)	3424 (20.4)
Immediate	8 (0.9)	241 (1.4)
Missing data	68 (7.4)	-
Severity		
Minor	72 (7.9)	1423 (8.5)
Intermediate	188 (20.5)	4134 (24.6)
Major	281 (30.7)	5488 (32.7)
Xmajor/complex	299 (32.6)	5743 (34-2)
Missing data	76 (8.3)	-
Co-morbidities		
None documented	551 (60.2)	9472 (56.4)
Arrhythmia	85 (9.3)	1177 (7.0)
Cancer	84 (9·2)	1649 (9.8)
Cirrhosis	10 (1.1)	123 (0.7)
Congestive cardiac failure	26 (2.8)	276 (1.6)
Smoker	71 (7.8)	1689 (10.1)
Diabetes (insulin-dependent)	15 (1.6)	445 (2.7)
Diabetes (non-insulin-dependent)	58 (6.3)	1128 (6.7)
Ischaemic heart disease	99 (10.8)	1635 (9.7)
Respiratory disease	109 (11.9)	2082 (12.4)
TIA/stroke	34 (3.7)	677 (4.0)

Values in parentheses are percentages. ASA-PS, American Society of Anesthesiologists Physical Status; Xmajor, extra major; TIA, transient ischaemic attack.

Inclusion/exclusion criteria

Data were analysed for patients aged 16 years or over who underwent surgery of any urgency (immediate, urgent, expedited or elective) and required a planned overnight admission. Surgery was defined as a procedure performed in an operating theatre by a surgeon. Patients undergoing day-case surgery, obstetric procedures, neurosurgery, cardiac or transplant surgery were excluded. If a patient had more than one procedure during the study interval, data for the most complex procedure were used. Patients were excluded if any of the following key variables were missing: age/date of birth, operative procedure or urgency, operation date and surgical severity, ASA-PS grade and mortality status.

Data set

In the prospectively collected Knowing the Risk data set, the surgical procedure was described using free text. In the analysis described below, data coders used this to categorize each procedure according to surgical type (for example abdominal - gastrointestinal) and severity. Where the free-text surgical procedure was missing, the operation was identified from OPCS coding obtained retrospectively using patient identifiers provided during the initial study. Severity coding into four categories (minor, intermediate, major or extra major (Xmajor)/complex) was based on the reference manual for the AXA Specialist Procedure Codes²⁴, which is used to grade the magnitude of surgical procedures in UK independent hospitals. A comparison of the surgical severity for all orthopaedic and gastrointestinal surgical procedures (the two most common surgical specialty groups in the study sample) was made to assess similarities between the AXA and British United Provident Association (BUPA) coding schedules, as the BUPA schedule has been used in previous studies²⁰. Clinical judgement was used for procedures falling under more than one surgical type, and where there was no listing for a procedure in the AXA schedule. Two clinical reviewers reached agreement on how to classify such procedures.

Data on deaths within 30 days were provided retrospectively by each hospital. In addition, following publication of the Knowing the Risk report, further data cleaning and collection were undertaken to complete missing data. Data linkage with the Office for National Statistics (ONS) (for England and Wales) and the Northern Ireland Statistics and Research Agency (NISRA) was used to validate dates of deaths and provide further mortality data. The final data linkage exercises took place on 8 February 2013 (ONS) and 26 March 2013 (NISRA), approximately 3 years after the initial data collection.

Statistical analysis

Univariable analyses were performed initially using χ^2 testing to assess the relationship between each independent variable and 30-day mortality. Variables that were not available for more than 10 per cent of patients were omitted from the analyses. In addition, 44.7 per cent of height and 31.7 per cent of weight data were estimated, so body mass index was not entered as a variable owing to the likelihood of inaccurate estimates²⁵.

Subsequently, the data set was divided randomly into two cohorts: a derivation cohort of approximately two-thirds of the sample, and a validation cohort consisting of the remainder, as described previously^{26,27}. Logistic regression

 Table 2 Descriptive data for the study, including 30-day mortality

o. of patients 11 219 (100) 6989 (62-3) 3073 (27-4) 1157 (10-3) 3616 (32-2) 5108 (45-5) 2190 (19-5) 286 (2-5) 10 (2)	Mortality (%) 1.3 0.4 1.6 6.1 0.1 0.5 3.1	No. of patients 5569 (100) 3506 (63.0) 1495 (26.8) 568 (10.2) 1800 (32.3) 2477 (44.5)	Mortality (%) 1.6 0.7 1.5 7.0 0.0 0.5	No. of patients 16 788 (100) 10 495 (62·5) 4568 (27·2) 1725 (10·3) 5416 (32·3)	Mortality (%) 1·4 0·5 1·6 6·4 0·0
6989 (62·3) 3073 (27·4) 1157 (10·3) 3616 (32·2) 5108 (45·5) 2190 (19·5) 286 (2·5)	0·4 1·6 6·1 0·1 0·5 3·1	3506 (63.0) 1495 (26.8) 568 (10.2) 1800 (32.3) 2477 (44.5)	0.7 1.5 7.0 0.0	10 495 (62·5) 4568 (27·2) 1725 (10·3) 5416 (32·3)	0·5 1·6 6·4
3073 (27·4) 1157 (10·3) 3616 (32·2) 5108 (45·5) 2190 (19·5) 286 (2·5)	1.6 6.1 0.5 3.1	1495 (26-8) 568 (10-2) 1800 (32-3) 2477 (44-5)	1.5 7.0 0.0	4568 (27·2) 1725 (10·3) 5416 (32·3)	1.6 6.4
3073 (27·4) 1157 (10·3) 3616 (32·2) 5108 (45·5) 2190 (19·5) 286 (2·5)	1.6 6.1 0.5 3.1	1495 (26-8) 568 (10-2) 1800 (32-3) 2477 (44-5)	1.5 7.0 0.0	4568 (27·2) 1725 (10·3) 5416 (32·3)	1.6 6.4
1157 (10·3) 3616 (32·2) 5108 (45·5) 2190 (19·5) 286 (2·5)	6·1 0·1 0·5 3·1	568 (10·2) 1800 (32·3) 2477 (44·5)	7·0 0·0	1725 (10·3) 5416 (32·3)	6.4
3616 (32·2) 5108 (45·5) 2190 (19·5) 286 (2·5)	0·1 0·5 3·1	1800 (32·3) 2477 (44·5)	0.0	5416 (32.3)	
5108 (45·5) 2190 (19·5) 286 (2·5)	0.5 3.1	2477 (44.5)		· · ·	0.0
5108 (45·5) 2190 (19·5) 286 (2·5)	0.5 3.1	2477 (44.5)		· · ·	0.0
2190 (19·5) 286 (2·5)	3.1	· · /	0.5		
286 (2.5)			0.9	7585 (45.2)	0.5
. ,		1149 (20.6)	3.5	3339 (19.9)	3.2
10 (0 0)	14.7	131 (2.4)	20.6	417 (2.5)	16.5
19 (0.2)	63	12 (0.2)	58	31 (0.2)	61
7374 (65.7)	0.3	3613 (64.9)	0.4	10987 (65.4)	0.3
1415 (12.6)	2.1	721 (12.9)	1.4	2136 (12.7)	1.9
2282 (20.3)	3.2	1142 (20.5)	4.6	3424 (20.4)	3.7
148 (1.3)	15.5	93 (1.7)	11	241 (1.4)	13.7
935 (8.3)	0.7	488 (8.8)	1.0	1423 (8.5)	0.8
2743 (24.4)	0.8	1391 (25.0)	0.7	4134 (24.6)	0.8
3680 (32.8)	1.1	1808 (32.5)	1.4	5488 (32.7)	1.2
3861 (34.4)	2.0	1882 (33.8)	2.4	5743 (34.2)	2.1
6363 (56.7)	0.3	3109 (55.8)	0.4	9472 (56.4)	0.3
795 (7.1)	5.7	382 (6.9)	6.3	1177 (7.0)	5.9
1111 (9.9)	3.5	538 (9.7)	4.6	1649 (9.8)	3.9
84 (0.7)	5	39 (0.7)	10	123 (0.7)	6.5
177 (1.6)	11.3	99 (1.8)	9	276 (1.6)	10.5
1106 (9.9)	0.9	583 (10.5)	1.7	1689 (10.1)	1.2
303 (2.7)	3.3	142 (2.5)	2.1	445 (2.7)	2.9
718 (6.4)	2.8	410 (7.4)	2.7	1128 (6.7)	2.7
1079 (9.6)	3.0	556 (10.0)	5.0	1635 (9.7)	3.7
1362 (12.1)	3.5	720 (12.9)	4.3	2082 (12.4)	3.7
468 (4.2)	5.3	209 (3.8)	4.8	677 (4·0)	5.2
	19 (0.2) 7374 (65-7) 1415 (12-6) 2282 (20-3) 148 (1-3) 935 (8-3) 2743 (24-4) 3680 (32-8) 3861 (34-4) 6363 (56-7) 795 (7-1) 1111 (9-9) 84 (0-7) 177 (1-6) 1106 (9-9) 303 (2-7) 718 (6-4) 1079 (9-6) 1362 (12-1)	19 (0.2) 63 7374 (65·7) 0.3 1415 (12·6) 2·1 2282 (20·3) 3·2 148 (1·3) 15·5 935 (8·3) 0·7 2743 (24·4) 0·8 3680 (32·8) 1·1 3861 (34·4) 2·0 6363 (56·7) 0·3 795 (7·1) 5·7 1111 (9·9) 3·5 84 (0·7) 5 177 (1·6) 11·3 1106 (9·9) 0·9 303 (2·7) 3·3 718 (6·4) 2·8 1079 (9·6) 3·0 1362 (12·1) 3·5 468 (4·2) 5·3	$19 (0.2)$ 63 $12 (0.2)$ $7374 (65-7)$ 0.3 $3613 (64-9)$ $1415 (12-6)$ $2\cdot 1$ $721 (12-9)$ $2282 (20.3)$ 3.2 $1142 (20-5)$ $148 (1.3)$ $15-5$ $93 (1-7)$ $935 (8.3)$ 0.7 $488 (8.8)$ $2743 (24-4)$ 0.8 $1391 (25-0)$ $3680 (32.8)$ 1.1 $1808 (32-5)$ $3861 (34.4)$ 2.0 $1882 (33-8)$ $6363 (56-7)$ 0.3 $3109 (55-8)$ $795 (7-1)$ $5-7$ $382 (6-9)$ $1111 (9-9)$ 3.5 $538 (9-7)$ $84 (0.7)$ 5 $39 (0.7)$ $177 (1-6)$ 11.3 $99 (1-8)$ $1106 (9-9)$ $0-9$ $583 (10-5)$ $303 (2-7)$ 3.3 $142 (2-5)$ $718 (6-4)$ 2.8 $410 (7-4)$ $1079 (9-6)$ 3.0 $556 (10-0)$ $1362 (12-1)$ 3.5 $720 (12-9)$	19(0.2) 63 $12(0.2)$ 58 $7374(65.7)$ 0.3 $3613(64.9)$ 0.4 $1415(12.6)$ 2.1 $721(12.9)$ 1.4 $2282(20.3)$ 3.2 $1142(20.5)$ 4.6 $148(1.3)$ 15.5 $93(1.7)$ 11 $935(8.3)$ 0.7 $488(8.8)$ 1.0 $2743(24.4)$ 0.8 $1391(25.0)$ 0.7 $3680(32.8)$ 1.1 $1808(32.5)$ 1.4 $3861(34.4)$ 2.0 $1882(33.8)$ 2.4 $6363(56.7)$ 0.3 $3109(55.8)$ 0.4 $795(7.1)$ 5.7 $382(6.9)$ 6.3 $1111(9.9)$ 3.5 $538(9.7)$ 4.6 $84(0.7)$ 5 $39(0.7)$ 10 $177(1.6)$ 11.3 $99(1.8)$ 9 $1106(9.9)$ 0.9 $583(10.5)$ 1.7 $303(2.7)$ 3.3 $142(2.5)$ 2.1 $718(6.4)$ 2.8 $410(7.4)$ 2.7 $1079(9.6)$ 3.0 $556(10.0)$ 5.0 $1362(12.1)$ 3.5 $720(12.9)$ 4.3	19(0.2) 63 $12(0.2)$ 58 $31(0.2)$ $7374(65.7)$ 0.3 $3613(64.9)$ 0.4 $10987(65.4)$ $1415(12.6)$ 2.1 $721(12.9)$ 1.4 $2136(12.7)$ $2282(20.3)$ 3.2 $1142(20.5)$ 4.6 $3424(20.4)$ $148(1.3)$ 15.5 $93(1.7)$ 11 $241(1.4)$ $935(8.3)$ 0.7 $488(8.8)$ 1.0 $1423(8.5)$ $2743(24.4)$ 0.8 $1391(25.0)$ 0.7 $4134(24.6)$ $3680(32.8)$ 1.1 $1808(32.5)$ 1.4 $5488(32.7)$ $3861(34.4)$ 2.0 $1882(33.8)$ 2.4 $5743(34.2)$ $6363(56.7)$ 0.3 $3109(55.8)$ 0.4 $9472(56.4)$ $795(7.1)$ 5.7 $382(6.9)$ 6.3 $1177(7.0)$ $1111(9.9)$ 3.5 $538(9.7)$ 4.6 $1649(9.8)$ $84(0.7)$ 5 $39(0.7)$ 10 $123(0.7)$ $177(1.6)$ 11.3 $99(1.8)$ 9 $276(1.6)$ $1106(9.9)$ 0.9 $583(10.5)$ 1.7 $1689(10.1)$ $303(2.7)$ 3.3 $142(2.5)$ 2.1 $445(2.7)$ $718(6.4)$ 2.8 $410(7.4)$ 2.7 $1128(6.7)$ $1079(9.6)$ 3.0 $556(10.0)$ 5.0 $1635(9.7)$ $1362(12.1)$ 3.5 $720(12.9)$ 4.3 $2082(12.4)$

Values in parentheses are percentages. ASA-PS, American Society of Anesthesiologists Physical Status; Xmajor, extra major; TIA, transient ischaemic attack.

was used to construct a model to predict 30-day mortality in the derivation cohort. All variables that were significant on univariable analysis at P < 0.100 were entered into the initial model. Age was treated as a categorical variable, with cut-offs at 65 and 80 years, based on inspection of a locally weighted scatter plot curve. All co-morbidity variables were included as indicator variables. High-risk surgical specialty was included as a binary indicator variable, with vascular, gastrointestinal and thoracic surgery defined as the high-risk specialties²⁶. In addition, interactions between each co-morbidity and each age category were tested, with each of three ASA-PS categories: high (ASA-PS IV or V), medium (III) or low (I or II). The interaction between cancer and the expedited surgery category was also tested. Variables were dropped from the model sequentially on repeated regression analyses, initially at P > 0.100, and then P > 0.050.

The final restricted model constructed from the derivation cohort was then tested on the validation cohort. A risk score was derived for each patient by summating the model coefficients for risk factors present. This was then converted into a percentage risk using the formula:

$$\ln (R/(1-R)) = \text{constant} + \text{risk score}$$

where R is the risk of 30-day mortality, and the constant was derived from the logistic regression model.

The accuracy of this novel risk stratification tool, the Surgical Outcome Risk Tool (SORT), was then assessed in the validation cohort by calculating the area under the receiver operating characteristic curve (AUROC) to measure discrimination and using the Hosmer–Lemeshow test to assess calibration. The AUROC has a value between 0.5 and 1.0, where 0.5 is equivalent to guessing, and 1.0 indicates perfect predictive accuracy. Previous work has described an AUROC of less than 0.7 to indicate poor performance, 0.7-0.9 to indicate moderate performance, and over 0.9 to indicate high performance^{23,28}. The

	I	No. of patients	
	Derivation	Validation	Total
Abdominal (bariatric)	46	28	74
Abdominal (endocrine)	4	0	4
Abdominal (gastrointestinal)	1802	864	2666
Abdominal (hepatobiliary)	440	222	662
Body surface (breast)	561	274	835
Body surface (other)	389	200	589
Endocrine	142	72	214
Gynaecology	1164	594	1758
Head and neck	848	409	1257
Ophthalmology	126	73	199
Orthopaedics	3945	1958	5903
Thoracic	120	71	191
Urology	1135	569	1704
Vascular	497	235	732
Total	11219	5569	16788

 Table 3 Specialty case mix in the derivation and validation cohorts

Hosmer–Lemeshow test compares observed and predicted event rates across a range of predicted risk. A non-significant test result indicates that a model is well calibrated.

Finally, χ^2 testing was used to compare the accuracy of the SORT (as measured by AUROC) and two previously published risk stratification tools: ASA-PS¹⁹ and a modified version of the SRS (with the 4 classes of surgical severity described above, rather than 5 as originally defined)²⁰. The calibration of the SRS was tested using the Hosmer–Lemeshow test.

All data were analysed using Microsoft[®] Excel 2010 (Microsoft, Redmond, Washington, USA) and Stata[®]

Table 4	Restricted	model	of six	variables	follo	owing	the analy	yses
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InterCooled 12 (StataCorp LP, College Station, Texas, USA). An app and web-based calculator were developed to facilitate risk calculation at the bedside.

Results

A total of 19097 case report forms were collected from 326 hospitals (Fig. 1). Characteristics of the 916 patients whose mortality data were unobtainable were similar to those of the final sample used in the analysis (Table 1). The mortality rate was lower among the 2047 patients who were excluded owing to missing data (Fig. 1) than in the included patients (0.6 versus 1.4 per cent; P = 0.003). After exclusions, the sample size for analysis was 16788, of which the derivation cohort comprised 11 219 patients and the validation cohort 5569. Descriptive data for these cohorts are summarized in Tables 2 and 3. Comparison of the surgical severity of 510 different surgical procedures in the AXA and BUPA schedules encompassed 5903 orthopaedic and 2666 gastrointestinal surgical procedures. Only 18 procedures (3.5 per cent) were coded differently, and in all instances the difference in grading was between consecutive categories.

Model development and derivation

Creatinine and haemoglobin results were excluded from analyses because of the large proportion of missing data (27·2 and 28·3 per cent respectively). Neither sex nor smoking history was associated significantly with 30-day mortality on univariable analysis, and so these were excluded from the multivariable analysis. Forty-five variables were included in the initial model (*Table S1*, supporting information). Stepwise sequential analyses based on significance testing led to a final model of six variables

IV 2.388 0.290 1.821, 2.956 8.25 <	Р
IV 2.388 0.290 1.821, 2.956 8.25 < V 4.081 0.596 2.911, 5.251 6.84 <	
V 4.081 0.596 2.911, 5.251 6.84 < Urgency of surgery Expedited 1.236 0.296 0.657, 1.812 4.18 <	<0.001
Urgency of surgery Expedited 1.236 0.296 0.657, 1.812 4.18 <	<0.001
Expedited 1.236 0.296 0.657, 1.812 4.18 <	<0.001
Urgent 1-657 0-259 1-149, 2-164 6-40 <	<0.001
	<0.001
Immediate 2:452 0:410 1:649, 3:256 5:98 <	<0.001
Specialty	
High-risk specialty (gastrointestinal, thoracic or vascular) 0.712 0.188 0.344, 1.081 3.79 <	<0.001
Severity of surgery	
Xmajor/complex 0.381 0.185 0.019, 0.744 2.06	0.039
Cancer 0.667 0.211 0.253, 1.081 3.16	0.002
Age (years)	
65-79 0.777 0.258 0.272, 1.281 3.02	0.003
≥ 80 1·591 0·260 1·082, 2·010 6·12 <	<0.001

ASA-PS, American Society of Anesthesiologists Physical Status; Xmajor, extra major.

that were independent predictors of 30-day mortality (*Table 4*).

The model coefficients were then used to develop a formula for a risk score as follows, where each variable is assigned a value of 1 if present and 0 if absent:

 $Risk score = (ASA - PS III \times 1.411)$

- + (ASA PS IV \times 2.388)
- $+ (\text{ASA} \text{PS V} \times 4.081)$
- + (urgency 'expedited' \times 1.236)
- + (urgency 'urgent' \times 1.657)
- + (urgency 'immediate' \times 2.452)
- + (high-risk specialty \times 0.712)
- + (severity 'Xmajor complex' \times 0.381)
- + (cancer \times 0.667)
- + (age 65-79 years $\times 0.777$)
- + (age \geq 80 years \times 1.591)

This score was then converted into a percentage risk of 30-day mortality using the formula described in the methods, with a constant of -7.366.

When these calculations were used to produce a percentage predicted mortality for each patient in the validation cohort, the AUROC for 30-day mortality was 0.91, indicating high discrimination. The *P* value for the Hosmer–Lemeshow χ^2 test was 0.204, indicating that the new model was well calibrated in the validation cohort. Observed and predicted mortality rates for the SORT are shown in *Fig. 2* and *Table 5*.

The discrimination of the SORT was also tested in seven surgical specialty subgroups for which there were at least 100 patients and at least one death in the validation

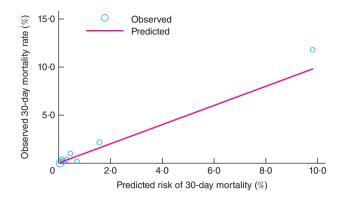


Fig. 2 Observed *versus* predicted 30-day mortality at varying levels of risk in the validation cohort of 5569 patients undergoing non-cardiac surgery. Circle size corresponds to the proportion of patients at each level of risk

cohort, to enable AUROC curves to be calculated (*Table 6*). Discrimination in these subgroups varied between moderate (AUROC 0.82 for hepatobiliary surgery) and excellent (AUROC 0.96 for head and neck surgery).

Comparison with previously validated risk stratification tools

Both the ASA-PS and the SRS demonstrated moderately good discrimination when tested in the validation cohort: AUROC 0.87 (95 per cent c.i. 0.84 to 0.91) and 0.88 (0.84 to 0.92) respectively. However, the SORT was significantly more accurate, with an AUROC of 0.91 (0.88 to 0.94) (P < 0.001) (*Fig. 3*). Furthermore, the SRS was poorly calibrated (P < 0.001, Hosmer–Lemeshow χ^2 test). Examination of observed : predicted ratios at different levels of risk demonstrated that the SRS overestimated risk in all but the highest-risk patients.

Table 5 Observed versus predicted mortality in validation cohort in nine quantiles, with Hosmer-Lemeshow statistic

Quantile	No. of patients	Mean SORT estimated probability of death (%)	Observed deaths at 30 days	Predicted deaths at 30 days	Hosmer-Lemeshow statistic
1	1220	0.063	0 (0)	0.8 (0.1)	0.77
2	496	0.093	0 (0)	0.5 (0.1)	0.46
3	523	0.134	2 (0.4)	0.7 (0.1)	2.41
4	572	0.206	1 (0.2)	1.2 (0.2)	0.03
5	617	0.309	2 (0.3)	1.9 (0.3)	0.00
6	510	0.463	5 (1.0)	2.4 (0.5)	2.96
7	527	0.723	1 (0.2)	3.8 (0.7)	2.09
8	561	1.595	12 (2.1)	8.9 (1.6)	1.07
9	543	9.818	64 (11.8)	53.3 (9.8)	2.37
Total	5569	1.319	87 (1.6)	73.4 (1.3)	12.16

Values in parentheses are percentages. Nine quantiles were used (rather than 10) because of ties within groups. SORT, Surgical Outcome Risk Tool. $\chi^2 = 12 \cdot 16$, P = 0.204.

	No. of patients in validation cohort	AUROC of SORT
Orthopaedics	1958	0.93
Gastrointestinal	864	0.88
Urology	569	0.95
Head and neck	409	0.96
Vascular	235	0.84
Hepatobiliary	222	0.82
Body surface (other)	200	0.87

AUROC, area under the receiver operating characteristic curve; SORT, Surgical Outcome Risk Tool.

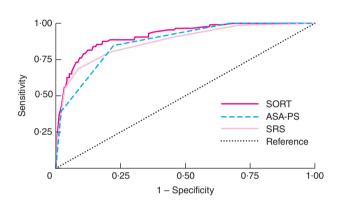


Fig. 3 Receiver operating characteristic (ROC) curves for the Surgical Outcome Risk Tool (SORT), American Society of Anesthesiologists Physical Status (ASA-PS) grade and Surgical Risk Scale (SRS) for the validation cohort of 5569 patients undergoing non-cardiac surgery. The areas under the ROC curves were 0.91 (95 per cent c.i. 0.88 to 0.94), 0.87 (0.84 to 0.91) and 0.88 (0.84 to 0.92) respectively

Discussion

A novel risk stratification tool comprising six preoperative variables was developed and validated internally to predict 30-day mortality in adults undergoing non-cardiac non-neurological inpatient surgery. Internal validation demonstrated the SORT to be more accurate than the ASA-PS and the SRS. There remain barriers to the routine use of risk stratification, including extensive data collection and data entry, the use of intraoperative and postoperative variables, and the inability to calculate an individual percentage mortality risk. By addressing these issues in this model, and developing an app and web-based calculator, the SORT has the potential to become used more widely. Although developed as a risk prediction tool before surgery, the SORT may also be valid as a risk adjustment tool in post hoc analyses of clinical performance, and thus aid epidemiological research and comparative audit.

All the variables in the SORT are known predictors of adverse outcome after surgery^{21,26,29-33}. Surgical severity is a significant contributor to postoperative mortality risk. The AXA-PPP system is an objective and widely used measure of operative severity, which employs the same classification method (Clinical Coding and Schedule Development group), and is very similar to the BUPA schedule used in the SRS²⁰. Although both have their limitations, an objective method of defining surgical severity is likely to reduce inter-rater variability and should therefore improve the accuracy of the SORT in clinical practice.

Some variables previously associated with risk of surgical mortality, such as raised body mass index, low haemoglobin and raised creatinine level $^{34-37}$, were omitted from the analyses owing to poor data collection rates during the initial study. For the SORT to be used routinely, the speed and simplicity of collecting variables were important features to retain. Therefore, a choice was made not to use imputation to derive missing data. The SORT demonstrated high performance (AUROC over 0.9) in internal validation, so the inclusion of other variables is likely to have had only a minimal impact on performance, at the risk of making the tool less easy to use. Including a larger number of variables, such as described previously³⁸, would greatly increase the time taken to collect data, and thereby decrease the usability of the tool. The inclusion of haematological and biochemical variables would have prevented the SORT being used when blood results were not available. With respect to body mass index, there is evidence that other measurements such as waist-hip ratio, waist-to-height ratio and waist circumference are better predictors of risk^{39,40}.

One co-morbidity that did not reach significance in the final model, despite being highly predictive of mortality on univariable analysis, was congestive cardiac failure. This may seem surprising given the known association between cardiac failure and outcome in surgical patients. Adjustment for ASA-PS grade, which is a reflection of functional capacity, is likely to explain this finding.

Systematic review²³ has previously identified P-POS-SUM and SRS as the most accurate methods of perioperative risk stratification in heterogeneous cohorts²³. Even though P-POSSUM has been validated multiply, it is still not used widely. The SORT has a number of advantages. First, it is a parsimonious model, consisting of only six preoperative variables, compared with 18 preoperative, intraoperative and postoperative variables for P-POSSUM. Second, POSSUM was originally designed as a *post hoc* audit tool. When risk assessment is arguably the most important (in the preoperative assessment clinic or emergency department), many variables required to compute the preoperative part of P-POSSUM (for example blood results) may not be available. Furthermore, P-POSSUM contains subjective variables, such as interpretation of a chest radiograph, and potential interobserver variability may affect its accuracy.

Conversely, the population-based ASA-PS is a widely used measure of perioperative risk, and has face validity as a measure of functional capacity. Although it was demonstrated to be a moderately accurate predictor of outcome in the present analysis, at least four previous studies found that it lacked accuracy when tested in heterogeneous cohorts²³. The lack of discrimination between patients in ASA-PS grade III is a particular limitation. A cut-off of 10 per cent predicted short-term mortality risk has been recommended when considering strategies to improve perioperative care and outcome, including planned critical care admission⁴¹. The population mortality of patients with ASA-PS grade III in this study was 3.2 per cent, and that of the ASA-PS grade IV population was 16.5 per cent; these findings are broadly consistent with other studies³⁰. Thus there is a need to be able to discriminate between patients in the ASA-PS grade III population, which can be achieved only by using a tool with more variables.

The SRS was also designed as a risk adjustment tool for comparative audit. Although it has been identified as a promising alternative to more complex risk stratification tools²³, previous validation studies are limited to two analyses from the same collaborators^{20,33}, and a subsequent external validation that included only urgent or emergency surgery in a single hospital⁴². The analysis undertaken in the present study demonstrated that the SRS overestimated risk in all but the highest-risk patients.

Encouraging the widespread and routine use of risk prediction tools is a challenge. Ease of use and face validity are two important factors that may influence this, and which have been addressed for the SORT. With easily available preoperative data, and an app or web-based calculator, it is hoped that adoption of the SORT may exceed that of other models.

The present study has strengths and some weaknesses. It is the largest study validating risk stratification tools prospectively in a heterogeneous cohort of patients undergoing non-cardiac surgery in the UK²³. Despite the large sample size, some selection bias may have occurred as the data capture did not encompass all patients undergoing surgery in the study week. Furthermore, it was not possible to determine whether 4.9 per cent of patients who met the inclusion criteria were dead or alive, despite linkage with national databases. In all instances, this was because key demographic information required for data linkage (such as NHS number) was missing. Despite these limitations, the sample analysed was representative, and had face validity,

as the prevalence of co-morbid risk factors and 30-day mortality statistics were broadly similar to those reported previously^{32,43}. The results are generalizable owing to the broad representation of size and types of hospital, the inclusion of surgery of all urgencies, and a wide range of specialties for inpatient surgery in England, Wales and Northern Ireland. The generalizability of the model in international cohorts remains unknown.

The SORT could be used in conjunction with clinical judgement to aid decision-making and facilitate informed consent. External validation of the SORT is necessary to test its validity further, as is the periodic recalibration and re-evaluation of the model to maintain validity as healthcare delivery changes^{44,45}. Studies evaluating the impact of risk stratification on improving patient outcomes through individual care planning should be a research priority as there is an opportunity to improve outcomes substantially^{14,43,46,47}.

Acknowledgements

The categorization of the surgical severity and specialty data was undertaken by staff at University College London/University College London Hospitals Surgical Outcomes Research Centre (SOuRCe): C. Dougall, D. Wyndham, E. Bettini, C. M. Oliver and P. Das in conjunction with J.C.S. This final manuscript was reviewed by all authors as well as M. Mason, G. P. Findlay, A. P. L. Goodwin, M. C. Juniper and M. J. Gough of NCEPOD. M. Mason supported K.L.P. with this initiative and provided the necessary resources. Mortality data were obtained from participating hospitals, the ONS and NISRA. The Knowing the Risk study published in 2011 was designed by G. P. Findlay and A. P. L. Goodwin in conjunction with a nominated Expert Group. NCEPOD local reporters in participating hospitals and healthcare professionals who completed study forms assisted with the data collection from which the data set for this paper was derived. Full details can be found in the report, which is available at http://www.ncepod.org.uk.

This work was supported by the Healthcare Quality Improvement Partnership (HQIP) on behalf of NHS England, NHS Wales, the Northern Ireland Department of Health, Social Services and Public Safety, the States of Jersey, Guernsey and the Isle of Man. The data were collected under the Clinical Outcome Review Programme into Medical and Surgical Care, which was commissioned by HQIP. The contribution of SOuRCe was funded by the National Institute for Health Research University College London Hospitals Biomedical Research Centre, where S.R.M. is a member of the Faculty.

Disclosure: The authors declare no conflict of interest.

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Supporting information

Additional supporting information may be found in the online version of this article:

Appendix S1 Clinical form completed prospectively for Knowing the Risk study (pdf)

Table S1 Unrestricted model of 45 variables (Word document)