

RESEARCH ARTICLE

Risk factors for surgical site infections using a data-driven approach

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Abstract

Objective

The objective of this study was to identify risk factors for surgical site infection from digestive, thoracic and orthopaedic system surgeries using clinical and data-driven cut-off values. A second objective was to compare the identified risk factors in this study to risk factors identified in literature.

Summary background data

Retrospective data of 3 250 surgical procedures performed in large tertiary care hospital in The Netherlands during January 2013 to June 2014 were used.

Methods

Potential risk factors were identified using a literature scan and univariate analysis. A multivariate forward-step logistic regression model was used to identify risk factors. Standard medical cut-off values were compared with cut-offs determined from the data.

Results

For digestive, orthopaedic and thoracic system surgical procedures, the risk factors identified were preoperative temperature of $\geq 38^{\circ}\text{C}$ and antibiotics used at the time of surgery. C-reactive protein and the duration of the surgery were identified as a risk factors for digestive surgical procedures. Being an adult (age ≥ 18) was identified as a protective effect for thoracic surgical procedures. Data-driven cut-off values were identified for temperature, age and CRP which can explain the SSI outcome up to 19.5% better than generic cut-off values.

Conclusions

This study identified risk factors for digestive, orthopaedic and thoracic system surgical procedures and illustrated how data-driven cut-offs can add value in the process. Future

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studies should investigate if data-driven cut-offs can add value to explain the outcome being modelled and not solely rely on standard medical cut-off values to identify risk factors.

Introduction

Surgical site infections (SSI), as defined by the European Centre for Disease Prevention and Control (ECDC) [1], make up 19.6% of the total number of healthcare-associated infections (HAIs) in Europe. With an estimated 81 089 patients in Europe having an HAI on any given day, almost 16 000 people in Europe are suffering from some form of SSI at any given time [2]. The burden of SSI can be measured in terms of increased length of stay in hospital, additional (surgical) procedures required, increased morbidity and mortality, as well as in economic terms [3].

Risk factors relating to the patient, procedure and the environment alter the odds of an SSI occurring. Research has been done to identify risk factors for SSI with the aim to identify preventative actions to reduce the incidence rate of SSI [4–10]. Patient-related risk factors for SSI, such as obesity, diabetes, surgery duration and the American Society of Anaesthesiologists (ASA) score are risk factors for digestive system, thoracic and orthopaedic surgical procedures [11–22]. Risk factors in low-income countries also include unemployment and level of education due to the disparity in socioeconomic status [14]. Risk factors can be modifiable or non-modifiable [23]. Modifiable risk factors are most interesting of the two since they can be changed preoperatively to reduce the risk of SSI.

The Segmentation of surgical procedures into homogenous groups makes it possible to find useful and relevant risk factors unique to each segment. Digestive system surgical procedures are more prone to SSI as they are generally clean-contaminated or dirty surgeries which make deep space SSI more likely. The occurrence of SSI after thoracic and orthopaedic surgeries are both relatively low because they are both typically clean surgeries, but the probability of attracting a deep space SSI after thoracic surgery is much higher compared to orthopaedic surgeries [15]. Because of these differences, we focus on digestive system, thoracic and orthopaedic surgical procedures for this study.

Multivariate logistic regression is the most common statistical model used to identify risk factors in longitudinal study design data [16]. Not all studies report the discriminatory power of the multivariate logistic regression model fitted. Risk factor identification studies do not usually specify how continuous variables cut-offs are determined. Cut-off values for variables such as age (≥ 18) or patient temperature (37°C) may seem intuitive or standard for clinical practice, but they may not statistically be the best cut-offs values determined by the data [17].

The objective of this study is to identify risk factors for SSI from digestive, thoracic and orthopaedic system surgeries using clinical and data-driven cut-off values. A second objective is to compare the identified risk factors in this study to risk factors identified in the literature.

Materials and methods

Literature search

A literature search was performed to identify known risk factors for SSI associated with digestive system surgical procedures, thoracic surgery and orthopaedic procedures using the corresponding medical subject headings (MeSH) linked data representation and the MEDLINE database.

Search strings used for MEDLINE literature search:

1. “Surgical Wound Infection”[Mesh] AND “Risk Factors”[Mesh] AND “Digestive System Surgical Procedures”[Mesh]
2. “Surgical Wound Infection”[Mesh] AND “Risk Factors”[Mesh] AND “Orthopaedic Procedures”[Mesh]
3. “Surgical Wound Infection”[Mesh] AND “Risk Factors”[Mesh] AND “Thoracic Surgery”[Mesh]

The search results were sorted, using the *Best Match* algorithm [18] developed by PubMed. Search results were deemed relevant using title and abstract screening. Risk factors were extracted if they were significant in a multivariable analysis until data saturation was achieved [19]. Risk factors identified, which were common to all three groups of surgeries, were defined as “general risk factors” in this study.

Setting and data collection

The Erasmus MC University Medical Centre in Rotterdam is the largest university medical hospital in the Netherlands with more than 1 300 beds [15]. The data used for this study were anonymised in accordance with the Dutch Personal Data Protection Act (WBP). Approval from the Medical Ethical Research Committee was obtained (MEC-2018-1185).

A weekly prevalence survey was performed by infection control practitioners (ICP) from January 2013 until December 2013 and two-weekly until June 2014 using a semi-automated algorithm proposed by Streefkerk et al. [20, 21]. This algorithm was used to calculate a nosocomial infection index (NII) which was then verified by ICP in case of a positive outcome to determine whenever an HAI was present or not. An ICP verified all patients with an NII > 7, and a definite SSI outcome was concluded by the ICP using the electronic patient data system. This outcome was used in this study as the occurrence of SSI outcome variable.

Data were extracted from a centralised database, containing cross-departmental data, clinical synopsis reports, infectious disease consultation reports, laboratory results and imaging reports. Data regarding the prescription of antimicrobials, in the J01 class of the Anatomical Therapeutic Chemical (ATC) classification system [22], were also included. Surgeries were included if they were part of the three groups of surgeries under investigation in this study and had a point prevalence measurement within 30 days after the surgery took place. If a second surgery took place within 30 days after an included surgery, then the recent surgery was excluded. All emergency surgeries were excluded to avoid possible undesirable confounding effects relating to the urgency and necessity of the surgeries.

Statistical analysis

The differences in the averages of variables with missing values and those without were evaluated using t-tests and were found statistically significant. These tests, together with Little’s MCAR test, convinced us that the missing values were not completely randomly missing and that we could not make use of more simple imputation methods. Therefore, we chose to use conditional Markov chain Monte Carlo (MCMC) with multiple imputations for the imputation process [24, 25].

Two methods were used to discretise continuous measurement variables: 1) standard medical cut-offs as used by Erasmus MC and 2) recursive partitioning [17]. Recursive partitioning is a data-driven, supervised discretisation method, used to group continuous values with similar outcomes optimally. The data-driven method was used to test and confirm if the standard medical cut-offs were the best way to explain the outcome variable for the groups of surgical procedures considered.

To build a prognostic prediction model for SSI, Hosmer et al. suggest fitting a univariate logistic regression model to each variable separately and if the p-value is less than a specific p-value, 0.1 in this case, then consider the variable good enough to include in the multivariate logistic regression model [26]. A univariate analysis was performed for each of the three groups of surgeries using the variables identified from the literature search. Significant variables ($p < 0.1$) in the univariate analysis were added to the list of variables associated with each group of surgery, together with the variables identified from the literature search. This resulted in an extended list of general risk factors as more risk factors were common across the three groups of surgeries.

A multivariate logistic regression model was built using a forward stepwise approach for each of the three groups of surgeries [27]. The general risk factors were first added to the model and then the risk factors unique to each surgery group in the order of the Akaike information criterion (AIC) until convergence was reached. In this case, we chose the conversion of the model to imply that there are no additional variables which can be added which will be statistically significant with a p-value of less than 0.05 or an AIC of 3.8415. Model performance was determined using the Gini coefficient after each step of the multivariate model, and the difference is reported as the marginal contribution of surgery group-specific risk factors for this study [19, 28]. Model performance was cross-validated using 5-fold cross-validation to estimate how the model would perform on new data [29]. R [30] was used in this study together with packages mice (multiple imputation) [31], smbinning (recursive partitioning) [32], dplyr (data wrangling) [33], finalfit (formatting of tables) [34] and scorecard (cross-validation) [35].

Approval was obtained from the Medical Ethical Committee of Erasmus MC (MEC-2018-1185) to perform this study. Data were analysed anonymously, and thus no further consent was obtained.

Results

Literature search

The literature search resulted in 1 422 research papers (as at 5 March 2020) using the MeSH headings in the PubMed search engine. We identified 24 research papers, published from 2008 until 2019, which contained statistically significant results from a multivariate analysis. A total of 79 risk factors were identified for the three groups of surgical procedures [11–13, 16, 23, 36–54] (S1 Table). Age, ASA class, body mass index (BMI), preoperative length of stay and diabetes were identified as general risk factors from the literature search. In total, 29 risk factors for digestive system surgical procedures, 31 for orthopaedic procedures and 19 for thoracic surgeries were identified. This amounted to 59 unique risk factors, of which 15 were present in more than one group of surgeries.

Risk factor identification

A total of 21 of the 59 unique risk factors could be replicated using our own data. The variable describing the type of surgery was used to create three homogenous groups of surgical procedures. The emergency classification variable was used to exclude emergency surgeries from the study such that 19 risk factors remained (Table 1). We observed 3 250 surgeries over the study period and excluded 526 (16.2%) emergency surgeries to be left with 2 724 surgical observations. CRP and temperature data were available for 52.55% (60.47% for in-patients) and 96.88% of all surgeries respectively.

The significant univariate results of digestive system, orthopaedic and thoracic surgical procedures are shown in Table 2. Antibiotic use, CRP and temperature were added to the list of

Table 1. Variable names and definitions used to investigate the occurrence of SSI in this study.

Variable	Surgery group	Definition
Demographic		
Gender	D,O	Gender of patient (Male/Female)
Age	D,O,T	Age of patient on the day of surgery (Years)
ASA class	D,O,T	ASA class of patient (I-V)
BMI	D,O,T	BMI of patient at the time of surgery.
Behavioural		
Alcohol use	O	Alcohol use of patient at the time of surgery (Current/Never/Past).
Smoking	D,O	Smoking status of patient at the time of surgery (Current/Never/Past).
Comorbidities		
Heart disease	O,T	Patient has a history of heart disease at the time of surgery (Yes/No).
Liver disease	D	Patient has a history of liver disease at the time of surgery (Yes/No).
Hypertension	O	Patient has a history of hypertension (Yes/No).
Diabetes	D,O,T	Patient has diabetes Type I or II at the time of surgery (Yes/No).
Measurement		
Temperature	D	Highest temperature of patient in the past 7 days before surgery.
CRP	O	Highest CRP of patient in the 7 days before surgery.
Leukocyte	D	Highest leukocyte level of patient in the 7 days before surgery.
Serum total protein	D	Highest serum total protein of patient in the 7 days before surgery.
Glucose	D	Highest glucose level of patient in the 7 days before surgery.
Haemoglobin	D	Highest haemoglobin level of patient in the 7 days before surgery.
Operative		
Preoperative length of stay	D,O,T	Preoperative length of hospital stay of patient at the time of surgery (Days).
Antibiotic use	T	Antibiotic (WHO ATC code J01 [22]) use of patient at the time of surgery (Yes/No).
Duration of surgery	D,O	Duration of the surgical procedure (Minutes).

D, Digestive system surgical procedures; O, Orthopaedic system surgical procedures; T, Thoracic system surgical procedures; ASA, American Society of Anaesthesiologists; CRP, C-reactive protein; BMI, Body Mass Index; SSI, Surgical Site Infection; ATC, Anatomical Therapeutic Chemical; WHO, World Health Organization.

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general risk factors after being found statistically significant in the univariate analysis—increasing the number of general risk factors to 8. Diabetes was identified as a general risk factor from our literature search but was not found significant in any of the three univariate analyses in our own study. For digestive system surgical procedure and thoracic procedures, the data-driven cut-off for age was obtained as 23 years and both the standard cut-off (18 years) and the data-driven cut-off were statistically significant with p-values of less than 0.001 which resulted in rejecting the null hypothesis that the coefficient associated with the age of the patient is zero. For orthopaedic procedures, the data-driven cut-off for the temperature (39 degrees) was found statistically significant, but the standard medical cut-off not. A data-driven CRP cut-off of 8.1 was identified for orthopaedic surgical procedures as opposed to a standard medical CRP cut-off of 10; both cut-offs are statistically significant.

The multivariate results using standard medical cut-offs and data-driven cut-offs are shown in Tables 3 and 4, respectively. The temperature variable was statistically significant in the multivariate analysis using the data-driven cut-offs for all three groups of surgeries, but not in one of the multivariate analysis using the medical standard cut-offs. The duration of the surgery was the only statistically significant variable in the multivariate analyses which was not

Table 2. Digestive system surgical procedures: univariate analysis of risk factors for the future occurrence of SSI.

Variable		SSI = No (2 600)	SSI = Yes (124)	Univariate OR (95%CI, P-value)
Digestive System Surgical Procedures				
Gender	Female	359 (43.9) ²	24 (33.8)	Reference
	Male	458 (56.1)	47 (66.2)	1.54 (0.93–2.60, p = 0.099)
Age ¹	≤18	246 (30.1)	8 (11.3)	Reference
	>18	571 (69.9)	63 (88.7)	3.39 (1.70–7.77, p<0.001)
Age (data-driven)	≤23	258 (31.6)	8 (11.3)	Reference
	>23	559 (68.4)	63 (88.7)	3.63 (1.82–8.32, p<0.001)
Antibiotic use	No	496 (60.7)	17 (23.9)	Reference
	Yes	321 (39.3)	54 (76.1)	4.91 (2.85–8.86, p<0.001)
Temperature ¹	≤36.5	0 (0.0)	0 (0.0)	NA
	(36.5,37.5]	98 (12.0)	2 (2.8)	Reference
	>37.5	719 (88.0)	69 (97.2)	4.70 (1.44–28.91, p = 0.033)
Temperature (data-driven)	≤38	535 (65.5)	20 (28.2)	Reference
	(38,39]	187 (22.9)	25 (35.2)	3.58 (1.95–6.66, p<0.001)
	>39	95 (11.6)	26 (36.6)	7.32 (3.94–13.79, p<0.001)
CRP ¹	≤10	397 (48.6)	21 (29.6)	Reference
	>10	420 (51.4)	50 (70.4)	2.25 (1.35–3.89, p = 0.003)
CRP (data-driven)	≤8.1	365 (44.7)	18 (25.4)	Reference
	>8.1	452 (55.3)	53 (74.6)	2.38 (1.39–4.24, p = 0.002)
Preoperative length of stay (Days)	Mean Days (SD)	6.6 (24.1)	12.1 (37.3)	1.01 (1.00–1.01, p = 0.092)
Duration of surgery	Mean Minutes (SD)	243.6 (143)	330.4 (190.8)	1.00 (1.00–1.01, p<0.001)
Orthopaedic Procedures				
ASA class	ASA CLASS I	196 (26.8)	6 (33.3)	
	ASA CLASS II	339 (46.4)	6 (33.3)	0.58 (0.18–1.87, p = 0.348)
	ASA CLASS III	182 (24.9)	4 (22.2)	0.72 (0.18–2.55, p = 0.612)
	ASA CLASS ≥ IV	13 (1.8)	2 (11.1)	5.03 (0.69–24.47, p = 0.062)
Alcohol use	Current	327 (44.8)	6 (33.3)	Reference
	Never	339 (46.4)	8 (44.4)	1.29 (0.44–3.94, p = 0.645)
	Past	64 (8.8)	4 (22.2)	3.41 (0.85–12.26, p = 0.063)
Antibiotic use	No	591 (81.0)	8 (44.4)	Reference
	Yes	139 (19.0)	10 (55.6)	5.31 (2.06–14.16, p<0.001)
Temperature (data-driven)	≤39	695 (95.2)	14 (77.8)	Reference
	>39	35 (4.8)	4 (22.2)	5.67 (1.55–16.79, p = 0.003)
Thoracic Surgery				
Age ¹	≤18	232 (22.0)	16 (45.7)	Reference
	>18	821 (78.0)	19 (54.3)	0.34 (0.17–0.67, p = 0.002)
Age (data-driven)	≤23	226 (21.5)	16 (45.7)	Reference
	>23	827 (78.5)	19 (54.3)	0.32 (0.16–0.65, p = 0.001)
BMI	Mean (SD)	24.5 (5.3)	22.1 (4.2)	0.91 (0.85–0.98, p = 0.010)
Alcohol use	Current	534 (50.7)	11 (31.4)	Reference
	Never	422 (40.1)	18 (51.4)	2.07 (0.98–4.57, p = 0.061)
	Past	97 (9.2)	6 (17.1)	3.00 (1.01–8.09, p = 0.034)
Antibiotic use	No	705 (67.0)	18 (51.4)	Reference
	Yes	348 (33.0)	17 (48.6)	1.91 (0.97–3.77, p = 0.060)
Temperature ¹	≤36.5	0 (0.0)	0 (0.0)	NA
	(36.5,37.5]	302 (28.7)	3 (8.6)	Reference
	>37.5	751 (71.3)	32 (91.4)	4.29 (1.52–17.94, p = 0.017)

(Continued)

Table 2. (Continued)

Variable	SSI = No (2 600)	SSI = Yes (124)	Univariate OR (95%CI, P-value)	
Temperature (data-driven)	≤38	882 (83.8)	20 (57.1)	Reference
	>38	171 (16.2)	15 (42.9)	3.87 (1.91–7.67, p<0.001)
CRP ¹	≤10	684 (65.0)	17 (48.6)	Reference
	>10	369 (35.0)	18 (51.4)	1.96 (1.00–3.88, p = 0.050)
Haemoglobin ¹	≤8.6	665 (63.2)	21 (60.0)	Reference
	(8.6,10.5]	358 (34.0)	11 (31.4)	0.97 (0.45–2.00, p = 0.942)
	>10.5	30 (2.8)	3 (8.6)	3.17 (0.72–9.85, p = 0.074)

CRP, C-reactive protein; OR, Odds Ratio; BMI, Body Mass Index; NA, Not Applicable; CI, Confidence Interval; SSI, Surgical Site Infection; OR, Odds ratio; Data-driven, cut-off values determined using recursive partitioning.

¹Standard Erasmus MC clinical cut-offs.

²The percentage distribution of the SSI outcome is provided in brackets next to the frequency for each variable.

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identified as a general risk factor to increase the odds of SSI by approximately 6% for every 30 minutes spent in surgery. For digestive surgical procedures, the addition of duration of surgery to the multivariate model increased the Gini coefficient from 0.46 to 0.52 based on standard medical cut-offs and from 0.57 to 0.62 for the multivariate model based on the data-driven cut-offs. This increase translates into a 12.5% and 8.8% increase in the Gini coefficient, respectively. Neither the orthopaedic nor the thoracic group of surgical procedures had any statistically significant risk factors which are not part of the general risk factors group of surgeries. The Gini coefficient of the data-driven multivariate model is 19.5% (0.62 vs 0.52) higher than the multivariate model based on the standard medical cut-offs. The 5-fold cross-validated 95% confidence intervals for the Gini coefficients based on the validation samples of the data-driven models are (0.49, 0.72) for digestive procedures, (0.21, 0.86) for orthopaedic procedures and (0.21,0.70) for thoracic procedures.

An overview of the study results (Table 5) shows that 10 of the 19 risk factors, identified during the literature search, were not statistically significant in the univariate or multivariate analysis for any of the surgery groups. BMI and diabetes were identified across all three groups of surgeries and multiple studies as risk factors for SSI but were not statistically significant in this study. Temperature and the duration of the surgery were confirmed as risk factors for digestive system surgeries, and similarly, antibiotic use and age were confirmed as risk factors

Table 3. Multivariate analysis risk factors for the occurrence of SSI by group of surgeries using standard medical cut-offs.

Risk factor by surgery group ¹	Coefficient	Multivariate OR (95%CI)	P-value
Digestive System Surgical Procedures			
Antibiotic use	1.240	3.455 (1.951–6.384)	<0.001
Duration of surgery (Minutes)	0.003	1.003 (1.001–1.004)	<0.001
CRP >10	0.803	2.232 (1.302–3.951)	0.004
Orthopaedic Surgical Procedures			
Antibiotic use	1.670	5.315 (2.059–14.158)	<0.001
Thoracic Surgical Procedures			
Age >18	-4.195	0.146 (0.058–0.351)	<0.001
Antibiotic use	1.311	4.849 (2.035–12.266)	<0.001

CRP, C-reactive protein; CI, Confidence Interval; OR, Odds ratio.

¹The multivariate analysis was performed using Erasmus MC clinical cut-offs.

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Table 4. Multivariate analysis risk factors for the occurrence of SSI by group of surgeries using data-driven cut-offs.

Risk factor by surgery group ¹	Coefficient	Multivariate OR (95%CI)	P-value
Digestive System Surgical Procedures			
Temperature [38,39]	1.067	2.907 (1.556–5.497)	<0.001
Temperature >39	1.732	5.650 (2.952–10.947)	<0.001
Antibiotic use	1.201	3.322 (1.856–6.200)	<0.001
Duration of surgery (Minutes)	0.002	1.002 (1.001–1.004)	0.003
CRP >8.1	0.639	1.894 (1.062–3.510)	0.035
Orthopaedic Surgical Procedures			
Antibiotic use	1.552	3.665 (1.370–10.006)	0.009
Temperature >39	1.224	5.120 (1.316–16.387)	0.009
Thoracic Surgical Procedures			
Age >17	-1.847	0.158 (0.055–0.426)	<0.001
Antibiotic use	1.597	4.939 (1.896–14.043)	0.002
Temperature >38	0.824	2.280 (1.098–4.653)	0.024

Data-driven, cut-off values determined using recursive partitioning; CRP, C-reactive protein; CI, Confidence Interval; OR, Odds ratio.

¹The multivariate analysis was performed using data-driven cut-offs.

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for thoracic surgeries. Antibiotic use and CRP were identified as risk factors for digestive surgeries from the multivariate analysis, which were identified during the literature search for thoracic and orthopaedic surgeries, respectively. Antibiotic use and temperature were

Table 5. Statistical significance of risk factors and the source which lead them to be considered by surgical procedure.

Risk Factor	Significance ¹	Digestive System ²	Orthopaedic ²	Thoracic ²
Age	D _U ,T _M	[38, 11, 43, 47]	[16]	[12]
Alcohol use	O _U ,T _U		[51]	
Antibiotic use	D _M ,O _M ,T _M			[40]
ASA Class	O _U	[37, 39, 41, 43, 54]	[16, 51, 53]	[16]
BMI	None	[44]	[51–53]	[42]
CRP	D _M		[16]	
Diabetes	None	[38, 47, 50]	[16, 45, 51, 53]	[13]
Duration of surgery	D _M	[36, 38, 41, 43, 44, 49, 54]	[16, 45, 51, 53]	
Gender	D _U	[38, 11, 43]	[16, 51]	
Glucose	None	[47]		
Haemoglobin	None	[11, 44, 54]		
Heart Disease	None		[51]	[12]
Hypertension	None		[51]	
Leukocyte	None	[55]		
Liver disease	None	[54]		
Preoperative length of stay	D _U	[41, 50]	[16, 52]	[12, 13, 40]
Serum total protein	None	[36, 49]		
Smoking	None	[49]	[51–53]	
Temperature	D _M ,O _M ,T _M	[55]		

D, Digestive system surgical procedures; O, Orthopaedic system surgical procedures; _U, Significant in univariate analysis; _M, Significant in multivariate analysis; T, Thoracic system surgical procedures; ASA, American Society of Anaesthesiologists; CRP, C-reactive protein; SSI, Surgical Site Infection; BMI, Body Mass Index.

¹During which part of the analysis the risk factor was found statistically significant.

²References to the literature which had the risk factor as a multivariate result for each group of surgeries.

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statistically significant for all three groups of surgeries and were included because of two studies regarding thoracic and digestive system surgeries, respectively [40, 55].

Discussion

We identified temperature and antibiotics used at the time of surgery as risk factors for digestive, orthopaedic and thoracic system surgical procedures in this study. The duration of the surgery was identified as a risk factor for digestive surgical procedures. Being an adult (age ≥ 18) was identified as a protective effect for thoracic surgical procedures. Data-driven cut-offs were identified for temperature, CRP and age, which differ from the standard medical cut-offs. Temperature would not have been identified as a risk factor if only standard medical cut-offs were considered. From our literature search, we identified age, ASA class, BMI, preoperative length of stay and diabetes as general risk factors, while CRP, temperature and antibiotic use were identified as general risk factors because of this study.

The identified risk factors may be classified as modifiable or non-modifiable, depending upon the circumstances of the patient like the complexity of his condition. For instance, the temperature of a patient may be high because of an existing infection, which is why the surgery is needed in the first place and may not be modifiable before surgery. Age, on the other hand, may be a modifiable risk factor if the surgery can be postponed for several years, e.g. due to a heart defect. This study revealed that children are more likely to be diagnosed with an SSI after thoracic surgery than adults. There are studies which identify risk factors for children after thoracic surgeries, but none found that being a child is a risk factor for SSI [42, 48] after undergoing thoracic surgery. We segmented the thoracic surgeries between adults and children and obtained multivariate results for children and adults separately. The multivariate model based only on children (age ≤ 18) did not reveal any significant results, contrary to the results of the thoracic study which found age to be a risk factor for children [12]. This absence could be partly due to the small study population size of 248. Antibiotic usage was the only significant factor in the multivariate analysis of thoracic surgeries based on adults. The other two groups of surgical procedures were consistent in terms of their statistical significance of risk factors based on adults.

The data-driven cut-offs confirmed the existing standard medical cut-offs. On average the clinical cut-off for temperature was one degree Celsius lower, while for digestive system surgical procedures, the clinical cut-off for CRP (10) was just less than two units more than the data-driven cut-off of 8.1. This means that there is a greater difference between the occurrence of SSI for patients with a CRP below and above 8.1 than below and above 10. The data-driven cut-offs improved the ability of the statistical model to explain the occurrence of SSI. The performance of the digestive system surgical procedure prediction model increased by 19.5% due to using data-driven cut-offs rather than the standard medical cut-offs. Using data-driven cut-offs, we were able to identify temperature as a risk factor for all three groups of surgical procedures. If standard clinical cut-offs were used, temperature would not have been significant from the multivariate analysis. This potential oversight illustrates the importance of evaluating the cut-offs used for continuous variables against the data before identifying risk factors.

Antibiotic use, temperature and CRP were added to the list of general risk factors by incorporating the statistically significant results of the univariate analysis. These risk factors might have been overlooked when the focus was on only one type of surgery. Temperature was identified as a risk factor in the multivariate results for all three groups of surgical procedures, whereas the literature search identified it only for digestive surgeries. Antibiotic use was not found during our literature search for digestive or orthopaedic surgical procedures but was found significant for both groups of surgeries in the multivariate analysis of our study.

The Centres for Disease Control and Prevention (CDC), the European centre for disease prevention and control (ECDC), World Health Organisation (WHO) and Netherlands National Institute for Public Health and the Environment (RIVM) suggest maintaining normothermia intraoperatively to prevent undesirable hypothermia (during some thoracic and neurosurgeries, hypothermia may be desirable). [56–58] A lower intraoperative bound for temperature of 35.5°C to 36°C is explicitly mentioned, and only the RIVM mention an upper bound of 38°C which is consistent with the risk factors identified in our study. An upper limit for preoperative temperature should, therefore, be investigated instead of only the lower limit. The four health organisations refer to the proper administration and timing of surgical antimicrobial prophylaxis, but not to the proper preoperative use of standard prescription antibiotics. Systemic antibiotics are typically prescribed to stabilise patients before undergoing surgery. A possible explanation for the increased occurrence of SSI associated with antimicrobials prescribed before surgery could be that these patients were not completely stabilised before surgery which increased their risk of SSI. The proper preoperative use of antibiotics should be well defined, and the reason why antibiotic-use was identified as a risk factor for SSI should be further investigated.

Limitations

This is a retrospective, single-centre study, and therefore the data were not collected for the purpose of this study. Even though cross-validation was performed to estimate model performance on new data, the models were not externally validated. Surgeries were aggregated into three broad groups of surgical procedures which serve as a proxy for the reason for surgery but leads to the loss of information regarding the exact reasons for the surgery. Some measurements, like temperature and CRP, were not always present and was partly overcome using imputation. Patient information concerning smoking and drinking habits may be understated due to incomplete medical records. The literature search used for this study was not exhaustive but rather based on the principal on data saturation. A comprehensive list of variables related to the nutritional and immunological alterations of the patients was not included in the analyses as they were not available from the data. We used a 30-day outcome period in which we observe if an SSI was present or not, but according to the CDC definition, this outcome period should be one year for surgical implantation procedures. Since our data only spans over 18 months, it was not possible to use a 12-month outcome window for all surgical implantation procedures, which is a limitation of this study. The administration of prophylaxis and the optimal timing thereof is an important risk factor for the occurrence of SSI. However, these data were not available.

Future work

Future work will investigate the modifiability of the risk factors identified in this study in more detail, as the circumstances under which this occurs are hitherto unclear. The exact purpose of the use of antibiotics over the time of surgery was not investigated in depth, which can be done in future studies. Future research can also investigate differences between adults and children, which lead to the occurrence of SSI among children. Another opportunity for future research is to investigate which risk factors are predictive for the occurrence of SSI over different periods. Doing this will enable healthcare workers to identify which risk factors explain the occurrence of SSI soon after surgery, towards the end of the 30 days and even later for implantation surgeries. These insights can help set guidelines to determine the vigilance necessary to mitigate the risk of SSI on a patient level.

Conclusion

This study shows that data-driven cut-offs can be used to identify risk factors which would not have been identified by only using standard medical cut-offs. Preoperative temperature and antibiotic use were identified as risk factors for digestive, orthopaedic, thoracic system surgeries, while the duration of surgery and age were identified as risk factors for orthopaedic and thoracic system surgeries, respectively. In contrast with literature, this study found that an SSI is more likely to occur in children (age < 18) than in adults after thoracic system surgeries. Statistical modelling has been important to quantify important risk factors and indicate their significance. Clinical studies using retrospective data are important to carry out, despite limitations in the data sets. To this end, future studies should use both standard medical cut-offs and data-driven cut-offs to investigate risk factors.

Supporting information

S1 Table. Risk factors identified from multivariate analysis during literature search.
(DOCX)

S1 Formulae. The multivariate logistic regression equations based on the data-driven cut-offs.
(DOCX)

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References

1. European Centre for Disease Prevention and Control. Surveillance of surgical site infections in European hospitals—HAISSE protocol [Internet]. 2012. 1–47 p. Available from: http://www.ecdc.europa.eu/en/publications/Publications/120215_TED_SSI_protocol.pdf

2. Zarb P, Coignard B, Griskeviciene J, Muller A, Vankerckhoven V, Weist K, et al. point prevalence survey of healthcare-associated infections and antimicrobial use in European acute care hospitals. Vol. 17, *Eurosurveillance*. 2012. 1–16 p.
3. Roy S, Patkar A, Daskiran M, Levine R, Hinoul P, Nigam S. Clinical and Economic Burden of Surgical Site Infection in Hysterectomy. *Surg Infect (Larchmt)*. 2014; 15(3):266–73. <https://doi.org/10.1089/sur.2012.163> PMID: 24801549
4. Leung Wai Sang S, Chaturvedi R, Alam A, Samoukovic G, de Varennes B, Lachapelle K. Preoperative hospital length of stay as a modifiable risk factor for mediastinitis after cardiac surgery. *J Cardiothorac Surg [Internet]*. 2013; 8(1):45. Available from: *Journal of Cardiothoracic Surgery* <https://doi.org/10.1186/1749-8090-8-45> PMID: 23497663
5. Triantafyllopoulos G, Stundner O, Memtsoudis S, Poultides LA. Patient, surgery, and hospital related risk factors for surgical site infections following total hip arthroplasty. *Sci World J*. 2015; 2015:1–9.
6. Abuzaid A, Zaki M, Al Tarief H. Potential risk factors for surgical site infection after isolated coronary artery bypass grafting in a Bahrain Cardiac Centre: A retrospective, case-controlled study. *Hear Views*. 2015; 16(3):79.
7. Schimmel JJP, Horsting PP, De Kleuver M, Wonders G, Van Limbeek J. Risk factors for deep surgical site infections after spinal fusion. *Eur Spine J*. 2010; 19(10):1711–9. <https://doi.org/10.1007/s00586-010-1421-y> PMID: 20445999
8. Guohua X, cheng keping, Li J, Kong Q, Wang C, Nanyuan Y. Risk factors for surgical site infection in a teaching hospital: a prospective study of 1,138 patients. *Patient Prefer Adherence*. 2015; 9:1171. <https://doi.org/10.2147/PPA.S86153> PMID: 26316722
9. Gomila A, Carratalà J, Biondo S, Badia JM, Fracalvieri D, Shaw E, et al. Predictive factors for early- and late-onset surgical site infections in patients undergoing elective colorectal surgery. A multicentre, prospective, cohort study. *J Hosp Infect*. 2018; 99(1):24–30. <https://doi.org/10.1016/j.jhin.2017.12.017> PMID: 29288776
10. Martin ET, Kaye KS, Knott C, Nguyen H, Santarossa M, Evans R, et al. Diabetes and risk of surgical site infection: A systematic review and meta-analysis. *Infect Control Hosp Epidemiol*. 2016; 37(1):88–99. <https://doi.org/10.1017/ice.2015.249> PMID: 26503187
11. Isik O, Kaya E, Dundar HZ, Sarkut P. Surgical Site Infection: Re-assessment of the Risk Factors. *Chirurgia (Bucur) [Internet]*. 2015; 110(5):457–61. Available from: <http://www.ncbi.nlm.nih.gov/pubmed/26531790> PMID: 26531790
12. Chen LF, Arduino JM, Sheng S, Muhlbaier LH, Kanafani ZA, Harris AD, et al. Epidemiology and outcome of major postoperative infections following cardiac surgery: Risk factors and impact of pathogen type. *Am J Infect Control*. 2012; 40(10):963–8. <https://doi.org/10.1016/j.ajic.2012.01.012> PMID: 22609237
13. Jolivet S, Lescure FX, Armand-Lefevre L, Raffoul R, Dilly MP, Ghodbane W, et al. Surgical site infection with extended-spectrum β -lactamase-producing Enterobacteriaceae after cardiac surgery: incidence and risk factors. *Clin Microbiol Infect [Internet]*. 2018; 24(3):283–8. Available from: <https://doi.org/10.1016/j.cmi.2017.07.004> PMID: 28698036
14. Di Gennaro F, Marotta C, Pisani L, Veronese N, Pisani V, Lippolis V, et al. Maternal caesarean section infection (MACSI) in Sierra Leone: a case-control study. *Epidemiol Infect*. 2020;1–6. <https://doi.org/10.1017/S0950268820000370> PMID: 32102721
15. European Centre for Disease Prevention and Control. Healthcare-associated infections: surgical site infections. *ECDC Annu Epidemiol Rep 2016 Stock ECDC*. 2018;(May).
16. KUNUTSOR SK, WHITEHOUSE MR, BLOM AW, BESWICK AD. Systematic review of risk prediction scores for surgical site infection or periprosthetic joint infection following joint arthroplasty. *Epidemiol Infect*. 2017; 145(9):1738–49. <https://doi.org/10.1017/S0950268817000486> PMID: 28264756
17. Strobl C, Malley J, Gerhard Tutz. Characteristics of Classification and Regression Trees, Bagging and Random Forests. *Psychol Methods*. 2009; 14(4):323–48. <https://doi.org/10.1037/a0016973> PMID: 19968396
18. Fiorini N, Canese K, Starchenko G, Kireev E, Kim W, Miller V, et al. Best Match: New relevance search for PubMed. *PLoS Biol*. 2018; 16(8):1–12. <https://doi.org/10.1371/journal.pbio.2005343> PMID: 30153250
19. Aldiabat KM, Le Navenec CL. Data saturation: The mysterious step in grounded theory methodology. *Qual Rep*. 2018; 23(1):245–61.
20. Streefkerk RHRA, Moorman PW, Parlevliet GA, van der Hoeven C, Verbrugh HA, Vos MC, et al. An Automated Algorithm to Preselect Patients to Be Assessed Individually in Point Prevalence Surveys for Hospital-Acquired Infections in Surgery. *Infect Control Hosp Epidemiol [Internet]*. 2014; 35(7):886–7. Available from: <http://www.jstor.org/stable/info/10.1086/676868> <https://doi.org/10.1086/676868> PMID: 24915221

21. Streefkerk RHRA, Borsboom GJJM, van der Hoeven CP, Vos MC, Verkooijen RP, Verbrugh HA. Evaluation of an Algorithm for Electronic Surveillance of Hospital-Acquired Infections Yielding Serial Weekly Point Prevalence Scores. *Infect Control Hosp Epidemiol* [Internet]. 2014 [cited 2017 Jul 18]; 35(07):888–90. Available from: https://www.cambridge.org/core/product/identifier/S0899823X00192372/type/journal_article <https://doi.org/10.1086/676869> PMID: 24915222
22. World Health Organization. Guidelines for ATC classification and DDD assignment 2020 [Internet]. 2020. Available from: https://www.whooc.no/filearchive/publications/2020_guidelines_web.pdf
23. Silvestri M, Dobrinja C, Scomersi S, Giudici F, Turoldo A, Princic E, et al. Modifiable and non-modifiable risk factors for surgical site infection after colorectal surgery: a single-center experience. *Surg Today*. 2018; 48(3):338–45. <https://doi.org/10.1007/s00595-017-1590-y> PMID: 28948367
24. Asendorpf JB, Van De Schoot R, Denissen JJA, Hutteman R. Reducing bias due to systematic attrition in longitudinal studies: The benefits of multiple imputation. *Int J Behav Dev*. 2014; 38(5):453–60.
25. van Buuren S. Multiple imputation of discrete and continuous. *Stat Methods Med Res*. 2007; 16(3):219–42. <https://doi.org/10.1177/0962280206074463> PMID: 17621469
26. Hosmer DW, Lemeshow S. *Applied logistic regression* [Internet]. Wiley. 2000. 118–128 p. Available from: <http://onlinelibrary.wiley.com/book/10.1002/9781118548387>
27. In Lee K, Koval JJ. Determination of the best significance level in forward stepwise logistic regression. *Commun Stat—Simul Comput*. 2007; 26(2):559–75.
28. Hastie T, Tibshirani R, Friedman J. *The Elements of Statistical Learning*. *Math Intell* [Internet]. 2009 [cited 2017 Aug 23]; 27(2):764. Available from: <http://www.springerlink.com/index/10.1007/b94608> PMID: 19443179
29. Wong T-T. Performance evaluation of classification algorithms by k-fold and leave-one-out cross validation. *Pattern Recognit*. 2015; 48(9):2839–46.
30. R Core Team. *R: A Language and Environment for Statistical Computing* [Internet]. Vienna, Austria; 2020. Available from: <http://www.r-project.org/>
31. van Buuren S, Groothuis-Oudshoorn K. {mice}: Multivariate Imputation by Chained Equations in R. *J Stat Softw* [Internet]. 2011; 45(3):1–67. Available from: <https://www.jstatsoft.org/v45/i03/>
32. Jopia H. *smbinning: Scoring Modeling and Optimal Binning* [Internet]. 2019. Available from: <https://cran.r-project.org/package=smbinning>
33. Wickham H, François R, Henry L, Müller K. *dplyr: A Grammar of Data Manipulation* [Internet]. 2019. Available from: <https://cran.r-project.org/package=dplyr>
34. Harrison E, Drake T, Ots R. *finalfit: Quickly Create Elegant Regression Results Tables and Plots when Modelling* [Internet]. 2019. Available from: <https://cran.r-project.org/package=finalfit>
35. Xie S. *scorecard: Credit Risk Scorecard* [Internet]. 2020. Available from: <https://cran.r-project.org/package=scorecard>
36. Takahashi Y, Takesue Y, Fujiwara M, Tatsumi S, Ichiki K, Fujimoto J, et al. Risk factors for surgical site infection after major hepatobiliary and pancreatic surgery. *J Infect Chemother* [Internet]. 2018; 24(9):739–43. Available from: <https://doi.org/10.1016/j.jiac.2018.05.007> PMID: 30001844
37. Fukuda H. Patient-related risk factors for surgical site infection following eight types of gastrointestinal surgery. *J Hosp Infect* [Internet]. 2016; 93(4):347–54. Available from: <https://doi.org/10.1016/j.jhin.2016.04.005> PMID: 27209057
38. Kokudo T, Uldry E, Demartines N, Halkic N. Risk factors for incisional and organ space surgical site infections after liver resection are different. *World J Surg*. 2015; 39(5):1185–92. <https://doi.org/10.1007/s00268-014-2922-3> PMID: 25561190
39. Moreno Elola-Olaso A, Davenport DL, Hundley JC, Daily MF, Gedaly R. Predictors of surgical site infection after liver resection: A multicentre analysis using National Surgical Quality Improvement Program data. *Hpb*. 2012; 14(2):136–41. <https://doi.org/10.1111/j.1477-2574.2011.00417.x> PMID: 22221576
40. Lola I, Levidiotou S, Petrou A, Arnaoutoglou H, Apostolakis E, Papadopoulos GS. Are there independent predisposing factors for postoperative infections following open heart surgery? *J Cardiothorac Surg* [Internet]. 2011; 6(1):151. Available from: <https://doi.org/10.1186/1749-8090-6-151> PMID: 22082355
41. Araki T, Okita Y, Uchino M, Ikeuchi H, Sasaki I, Funayama Y, et al. Risk factors for surgical site infection in Japanese patients with ulcerative colitis: A multicenter prospective study. *Surg Today*. 2014; 44(6):1072–8. <https://doi.org/10.1007/s00595-013-0809-9> PMID: 24337501
42. Ben-Ami E, Levy I, Katz J, Dagan O, Shalit I. Risk factors for sternal wound infection in children undergoing cardiac surgery: a case-control study. *J Hosp Infect* [Internet]. 2008; 70(4):335–40. Available from: <https://doi.org/10.1016/j.jhin.2008.08.010> PMID: 18951662

43. Ejaz A, Schmidt C, Johnston FM, Frank SM, Pawlik TM. Risk factors and prediction model for inpatient surgical site infection after major abdominal surgery. *J Surg Res [Internet]*. 2017; 217:153–9. Available from: <https://doi.org/10.1016/j.jss.2017.05.018> PMID: 28595819
44. Giri S, Kandel BP, Pant S, Lakhey PJ, Singh YP, Vaidya P. Risk factors for surgical site infections in abdominal surgery: A study in Nepal. *Surg Infect (Larchmt)*. 2013; 14(3):313–8. <https://doi.org/10.1089/sur.2012.108> PMID: 23672239
45. Hijas-Gómez AI, Egea-Gámez RM, Martínez-Martín J, González-Díaz R, Losada-Viñas JI, Rodríguez-Caravaca G. Surgical Wound Infection Rates and Risk Factors in Spinal Fusion in a University Teaching Hospital in Madrid, Spain. *Spine (Phila Pa 1976)*. 2017; 42(10):748–54.
46. Liu S, Miao J, Wang G, Wang M, Wu X, Guo K, et al. Risk factors for postoperative surgical site infections in patients with Crohn's disease receiving definitive bowel resection. *Sci Rep [Internet]*. 2017; 7(1):1–6. Available from: <https://doi.org/10.1038/s41598-016-0028-x> PMID: 28127051
47. McKenzie Stancu S, Iordache F. Thrombocytosis Is a Risk Factor for Surgical Site Infections after Colon Resection: A Prospective Observational Study. *Surg Infect (Larchmt) [Internet]*. 2019; 20(1):39–44. Available from: <http://www.ncbi.nlm.nih.gov/pubmed/30256728> <https://doi.org/10.1089/sur.2018.146> PMID: 30256728
48. Mehta PA, Cunningham CK, Colella CB, Alferis G, Weiner LB. Risk factors for sternal wound and other infections in pediatric cardiac surgery patients. *Pediatr Infect Dis J*. 2000; 19(10):1000–4. <https://doi.org/10.1097/00006454-200010000-00012> PMID: 11055604
49. Morikane K, Honda H, Suzuki S. Factors associated with surgical site infection following gastric surgery in Japan. *Infect Control Hosp Epidemiol*. 2016; 37(10):1167–72. <https://doi.org/10.1017/ice.2016.155> PMID: 27430979
50. Perkins JD. Techniques to ensure adequate portal flow in the presence of splenorenal shunts. *Liver Transplant*. 2007; 13(5):767–8. PMID: 17563933
51. Shao J, Zhang H, Yin B, Li J, Zhu Y, Zhang Y. Risk factors for surgical site infection following operative treatment of ankle fractures: A systematic review and meta-analysis. *Int J Surg [Internet]*. 2018; 56(May):124–32. Available from: <https://doi.org/10.1016/j.ijss.2018.06.018> PMID: 29929022
52. Su J, Cao X. Risk factors of wound infection after open reduction and internal fixation of calcaneal fractures. *Med (United States)*. 2017; 96(44):e8411. <https://doi.org/10.1097/MD.00000000000008411> PMID: 29095273
53. Xing D, Ma JX, Ma XL, Song DH, Wang J, Chen Y, et al. A methodological, systematic review of evidence-based independent risk factors for surgical site infections after spinal surgery. *Eur Spine J*. 2013; 22(3):605–15. <https://doi.org/10.1007/s00586-012-2514-6> PMID: 23001381
54. Zhang JF, Zhu HY, Sun YW, Liu W, Huo YM, Liu DJ, et al. Pseudomonas aeruginosa infection after pancreatoduodenectomy: Risk factors and clinic impacts. *Surg Infect (Larchmt)*. 2015; 16(6):769–74. <https://doi.org/10.1089/sur.2015.041> PMID: 26237502
55. Isik O, Kaya E, Sarkut P, Dundar HZ. Factors affecting surgical site infection rates in hepatobiliary surgery. *Surg Infect (Larchmt)*. 2015; 16(3):281–6. <https://doi.org/10.1089/sur.2013.195> PMID: 25830815
56. Berríos-Torres SI, Umscheid CA, Bratzler DW, Leas B, Stone EC, Kelz RR, et al. Centers for Disease Control and Prevention Guideline for the Prevention of Surgical Site Infection, 2017. *JAMA Surg [Internet]*. 2017; 152(8):784–91. Available from: <https://doi.org/10.1001/jamasurg.2017.0904> PMID: 28467526
57. Leaper DJ, Edmiston CE. World Health Organization: global guidelines for the prevention of surgical site infection. *J Hosp Infect [Internet]*. 2017; 95(2):135–6. Available from: <http://linkinghub.elsevier.com/retrieve/pii/S0195670116305874> <https://doi.org/10.1016/j.jhin.2016.12.016> PMID: 28139389
58. Werkgroep infectiepreventie. Preventie van postoperatieve wondinfecties. 2011;1–26. Available from: http://www.rivm.nl/Documenten_en_publicaties/Professioneel_Praktisch/Richtlijnen/Infectieziekten/WIP_Richtlijnen/Actuele_WIP_Richtlijnen/Ziekenhuizen/WIP_richtlijn_Postoperatieve_wondinfecties_ZKH