



Enhanced neonatal surgical site infection prediction model utilizing statistically and clinically significant variables in combination with a machine learning algorithm



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ABSTRACT

Background: Machine-learning can elucidate complex relationships/provide insight to important variables for large datasets. This study aimed to develop an accurate model to predict neonatal surgical site infections (SSI) using different statistical methods.

Methods: The 2012–2015 National Surgical Quality Improvement Program-Pediatric for neonates was utilized for development and validation models. The primary outcome was any SSI. Models included different algorithms: full multiple logistic regression (LR), *a priori* clinical LR, random forest classification (RFC), and a hybrid model (combination of clinical knowledge and significant variables from RF) to maximize predictive power.

Results: 16,842 patients (median age 18 days, IQR 3–58) were included. 542 SSIs (4%) were identified. Agreement was observed for multiple covariates among significant variables between models. Area under the curve for each model was similar (full model 0.65, clinical model 0.67, RF 0.68, hybrid LR 0.67); however, the hybrid model utilized the fewest variables (18).

Conclusions: The hybrid model had similar predictability as other models with fewer and more clinically relevant variables. Machine-learning algorithms can identify important novel characteristics, which enhance clinical prediction models.

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Introduction

Surgical site infections (SSIs) are the most common complication of surgery in children. SSI results in increased morbidity and mortality, additional procedures, longer hospital stays, and increased healthcare system costs.¹ In neonates, the incidence of SSI ranges between 12 and 17% of all surgeries, and may be more

frequent in the common non-elective thoracic and abdominal procedures.^{1–4} Many factors specific to neonates, including multiple co-morbidities, type of surgery, immunologic immaturity, prematurity, chronic illness and increased length of stay in intensive care units, place this population at increased risk for SSI development.^{2,5,6}

Surgical wound classification was the initial system used to stratify patient risks of SSI; however, multiple studies have demonstrated errors in estimating the risk of SSI with this method, especially in children.^{7–12} Recently, different models offering improved ability to estimate the risk of post-operative complications have been developed.^{2,13–15} However, none specifically

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evaluate SSI in a large neonatal population. The complexity of this population and presence of multiple risk factors for SSI make it especially difficult to determine which risk factors are the most important.

Statistical methods that utilize machine-learning algorithms may be superior at deciphering complex relationships, providing insight into important variables within large datasets, and improving clinical outcome prediction models compared to methods previously used.^{16,17} Random forest classification (RFC) is one example of a machine-learning algorithm that employs multiple classification trees to generate a “forest” of trees with improved outcome prediction.¹⁸ Given the multiple risk factors for SSI in the neonatal population, we hypothesized that combining a machine-learning algorithm with clinical input would result in a more accurate model to predict SSI. This study aimed to determine which patient and clinical characteristics were associated with SSI and develop the most accurate model to predict surgical site infections (SSI) in neonates using RFC and multiple logistic regression.

Material and methods

Data source and patients

The American College of Surgeons National Surgical Quality Improvement Program Pediatric (NSQIP-P) data was queried for a retrospective analysis using Participant Use Files (PUF) from 2012 to 2015. The patient population was narrowed to include all neonates as per NSQIP-P definitions (age <29 days at time of surgery for term infants and <51 weeks post-conceptual age at time of surgery for pre-term infants). All surgical procedures available in the PUF were included.

The NSQIP-P program utilizes trained surgical clinical reviewers (SCR) to abstract patient-level clinical data. Data includes demographics, comorbidities, laboratory values, case type (by surgical specialty and CPT codes) and 30-day outcomes. Stringent variable definitions are adhered to in order to optimize reliability, and random audits are performed by NSQIP-P to check for data validity and definition compliance. Cases are non-consecutive, but systematically sampled across all pediatric surgical specialties at each participating institution per a specified protocol to ensure variability and reduce bias.^{2,19}

Predictor variables

The NSQIP-P provides 218 preoperative and intraoperative variables including demographic information, patient co-morbidities, clinical context and perioperative factors. Preoperative variables missing more than 20% of data were excluded from analysis. Binary or categorical covariates deemed clinically significant with less than 20% of data missing were transformed to a categorical variable with an “unknown” category. Fifty-eight variables were available for model development after exclusion based on the above criteria (Appendix, Table 1). Operative technique reports whether the surgery was laparoscopic, laparoscopic converted to open, open, or unknown. This variable was not collected for the 2012 PUF, thus all patients who underwent surgery in 2012 were designated as “unknown” for the operative technique variable. Preoperative sepsis is a categorical variable with 3 levels that is defined by NSQIP-P as: 1) SIRS – fever, leukocytosis, and elevated heart rate or respiratory rate, 2) Sepsis – SIRS with a source of infection including culture, pus or abscess, 3) Septic shock – sepsis criteria plus cardiovascular dysfunction. This variable is determined by the SCR on the data collection worksheet. Only patients with complete datasets after data cleaning as described above were included in the analysis.

Outcomes

The primary outcome was a composite binary variable of any SSI, including superficial, deep, organ space SSI, and wound dehiscence. Each variable included in the composite outcome is based on the American College of Surgeons NSQIP-P User Guide for the 2015 PUF.

Development and validation cohorts

The cohort was divided based on the outcome variable, SSI, using a random number generator to create a data set for model development or training and another for model validation. Two-thirds of the records were randomized to the model training cohort and 33% of records to the validation group. To ensure comparability, variables of interest were compared amongst the cohorts (Table 1; see Appendix Table 2 for comparison of all variables).

Statistical analysis

Model development applied different statistical algorithms to maximize predictive accuracy. One model utilized a RFC, while 3 of the models utilized multiple logistic regression. The first multiple logistic regression model (full logistic regression) evaluated all perioperative variables that were significant ($n = 42$) on univariate analysis utilizing a p -value threshold of <0.25 . Next, a second model's (clinical model) variables were selected *a priori* based on previous literature and clinical judgement. The final multiple logistic regression model (hybrid model) combined clinical knowledge and the 20 most important variables from RFC to develop the model. Variables were removed from the model when deemed clinically similar to another variable or felt not to be of clinical importance. Review of the adult and pediatric surgical literature guided this process, which was conducted by one author (MBK). Additional clinical expertise was provided by the senior author (KT).

A RFC was selected for its ability to efficiently analyze large data sets with improved accuracy without limiting or excluding the variables in the data set.^{16,18} Additionally, random forest models indicate which predictor variables have the greatest impact on accurate outcome prediction. This machine learning algorithm, developed by Breiman, generates a “forest” of many classification trees using random variable selection.¹⁸ Each tree in the forest “votes” for the best classification for a given observation, and the class receiving the majority of votes results in the prediction for that observation. Using resampling, random forest model training selects random subsets of observations in the training set (66.6% of the training sample). At each decision point in a given tree, the RFC algorithm randomly selects a subset of variables among which it identifies the one that provides the greatest classification accuracy. Each tree is grown to its largest potential with a constant number of variables without pruning. Random resampling of observations permits the algorithm to maintain a running tabulation of predictive accuracy, or “out-of-bag” classification rates, on the remaining one-third of the observations not selected for a given tree. Random variable selection at each potential split in a tree decreases the correlation between trees, thus minimizing the generalization error of the model. Unlike traditional regression analysis, RFC does not have formal distribution criteria and avoids over-fitting the data set. The criticisms of this methodology are primarily related to its complexity, which minimizes the interpretability of the variables included in the model.

K-fold cross validation was used to train the RFC model with 10 folds. This method divides the training data set into 10 partitions at

Table 1
Condensed comparison of variables between the development and validation data sets.

Characteristics	Total	Development dataset	Validation dataset
Total, n (%)	13,589 (100.0)	8970 (66.0)	4619 (34.0)
Median age (days)	18 (IQR 3–58)	18 (IQR 3–57)	19 (IQR 3–60)
Gestational age at birth (weeks)			
<24	164 (1.2)	114 (1.3)	50 (1.1)
24	345 (2.5)	228 (2.5)	117 (2.5)
25–26	700 (5.2)	462 (5.2)	238 (5.2)
27–28	473 (3.5)	311 (3.5)	162 (3.5)
29–30	457 (3.4)	319 (3.6)	138 (3.0)
31–32	489 (3.6)	310 (3.5)	179 (3.9)
33–34	855 (6.3)	561 (6.3)	294 (6.4)
35–36	1548 (11.4)	1011 (11.3)	537 (11.6)
≥37	8556 (63.0)	5653 (63.0)	2903 (62.9)
Male gender	8259 (60.8)	5405 (60.3)	2854 (61.8)
Race			
American Indian/Alaskan	68 (0.5)	48 (0.5)	20 (0.4)
Asian	352 (2.6)	235 (2.6)	117 (2.5)
Black	2514 (18.5)	1660 (18.5)	854 (18.5)
Native Hawaiian/Pacific	41 (0.3)	28 (0.3)	13 (0.3)
White	9338 (68.7)	6178 (68.9)	3160 (68.4)
Hispanic ethnicity	2105 (15.5)	1383 (15.4)	722 (15.6)
Inpatient status	12,949 (95.3)	8532 (95.1)	202 (4.4)
Transferred	6561 (48.3)	4297 (47.9)	2264 (49.0)
Type of case			
Elective	7096 (52.2)	4718 (52.6)	2378 (51.5)
Urgent	2887 (21.3)	1863 (20.8)	1217 (26.4)
Emergent	3606 (26.5)	2389 (26.6)	1024 (22.2)
ASA class			
1	663 (4.9)	439 (4.9)	224 (4.9)
2	3100 (22.8)	2050 (22.9)	1050 (22.7)
3	6606 (48.6)	4341 (48.4)	2265 (49.0)
4	2934 (21.6)	1944 (21.7)	990 (21.4)
5	211 (1.6)	146 (1.6)	65 (1.4)
Surgical site infection (SSI)	542 (4.0)	358 (4.0)	184 (4.0)
Superficial	262 (1.9)	178 (2.0)	84 (1.8)
Deep	36 (0.3)	19 (0.2)	17 (0.4)
Organ space infection	143 (1.1)	92 (1.0)	51 (1.1)
Dehiscence	149 (1.1)	100 (1.1)	49 (1.1)

random and uses each partition to fit the model and the remaining partitions to test the model.^{16,20} Use of this method incorporates the test error from each fold to reduce the risk of a falsely low test error. The H2O package (Mountain View, CA) within R (Vienna, Austria) statistical software was used to complete the RFC analysis.

Stata 14 (College Station, TX) statistical software characterized the sample using descriptive statistics, as well as chi-square, Wilcoxon rank sum test, and logistic regression. A p-value of <0.05 was considered significant for any inferential testing regarding sample characteristics. Univariate and multiple logistic regression were utilized to develop predictive models as described above. Performance of each model was evaluated with receiver operating characteristic (ROC) curves, area under the curve (AUC), goodness of fit and Akaike information criteria. For reference, an AUC of 0.5 is equivalent to the flip of a coin and an AUC greater than 0.7 is considered predictive.²¹ Since the operative technique variable was designated “unknown” for all patients from the 2012 PUF, we conducted a sensitivity analysis for each model utilizing only the patients in the 2013–2015 PUFs.

Results

16,842 patients met age criteria and 3,253 (19.3%) were excluded due to missing data. 13,589 patients were included with a median age of 18 days (IQR 3–58). The majority of patients were Caucasian (68.7%) and male (60.8%). Surgical site infections were observed in 542 patients (4.0%). Significant differences existed

between patients who developed SSI and those who did not (Appendix, Table 3). These differences included obstetric care (type of delivery, location of birth), demographics (race, ethnicity, gender), clinical characteristics (gestational age at birth, weight at surgery, ASA class, co-morbidities), and operative details (operative time, wound class, prior operation, and technique).

The RFC classification utilized 35 out of 58 possible variables to predict SSI in neonates. Days from hospital admission to operation was the most important variable, followed by total operative time, age in days (at surgery), preoperative nutritional supplementation, and weight at surgery. Additional variables of significance were preoperative transfusion, open wound (with or without infection), admission year, presence of a structural central nervous system (CNS) abnormality and presence of bronchopulmonary dysplasia/chronic lung disease. A list of the 20 most significant variables and their relative importance are listed in Appendix, Table 4. Area under the receiver operating curve for the RFC model was 0.65 in model development compared to 0.68 in model validation.

A multiple logistic regression inputting all significant variables resulted in a model with 40 variables after exclusion for collinearity (Appendix, Table 5). Of those variables, 8 were significant after adjustment. Based on this full logistic model, preoperative dialysis is associated with 4 times greater odds of SSI (OR 4.0 95%CI 1.3–12.4) than a patient not requiring preoperative dialysis. Presence of septic shock within 48 h of surgery increased the odds of SSI by 2.5 (95%CI 1.2–5.1) compared to a patient with no signs of SIRS or sepsis. Contaminated wound class was associated with the third

greatest risk of SSI (OR 1.8 95%CI 1.2–2.9). This model had an AUC of 0.71 on the development data set, but only AUC of 0.65 on the validation data set.

The clinical model utilized 23 variables, of which 9 were significant on multiple logistic regression (Appendix, Table 6). Similar to the full logistic model, preoperative dialysis had the greatest association with SSI (OR 4.0 95%CI 1.3–12.0), followed by presence of septic shock within 48 h of surgery (OR 2.3 95%CI 1.2–4.7). Additional variables associated with increased risk of SSI were operative technique (unknown OR 1.5 95%CI 1.3–3.0), preoperative nutritional supplementation (OR 2.3 95%CI 1.2–4.7) and surgical wound class (contaminated OR 1.8 95%CI 1.2–2.8). This model had an AUC of 0.69 on the development data set, and AUC of 0.67 on the validation data set.

Applying clinical knowledge and information gained from the RFC analysis, the hybrid model was designed based on the top 20 most important variables from the RFC and the clinical model. Comorbid conditions without direct evidence of association with SSI or strong biologic plausibility were excluded, which resulted in a model that utilized 18 variables (Appendix, Table 7). Nine of the variables included in the model were significant (Table 2). Like the other logistic regression models, preoperative dialysis (OR 4.0 95%CI 1.3–11.9) and presence of septic shock within 48 h of surgery (OR 2.3 95%CI 1.2–4.5) were associated with greatest risk of SSI. Operative technique (unknown OR 2.0 95%CI 1.3–3.0), preoperative nutritional supplementation (OR 1.8 95%CI 1.4–2.3) and open wound (with or without infection) (OR 1.7 95%CI 1.2–2.5) were also associated with an increased odds of SSI. This model had an AUC of 0.69 on the development data set, and AUC of 0.67 on the validation data set.

Among the four models, the full logistic regression model outperformed the rest on the development data set (Fig. 1). Using validation data, the RFC performed the best with an AUC of 0.68, followed by the hybrid model and clinical model with an AUC of 0.67 (Table 3). The hybrid model was also the least complex with only 18 variables considered and 9 significant predictors. Consistency among several variables was observed between the different models (Table 4). Overall, the models' variables remained largely the same on sensitivity analysis (Appendix, Tables 4–7).

Discussion

The overall rate of SSI observed in this study was 4% suggesting that SSIs are relatively uncommon events. Differences in both

demographic and clinical characteristics were observed between patients who developed SSI and those who did not. The multiple logistic regression models found the greatest association between SSI and preoperative dialysis, preoperative septic shock, operative technique, need for preoperative nutritional support, and open wound with or without infection. The RFC, however, demonstrated that the most important variables associated with SSI were days from hospital admission to operation, total operative time, age in days (at surgery), preoperative nutritional supplementation, and weight at surgery. There was some overlap between the statistical methods with regard to variables identified as risk factors associated with SSI, including operative technique, need for preoperative nutritional support, preoperative blood transfusion, open wound with or without infection, and total operative time. The different risk factors identified by different statistical methods speaks to the complexity of SSI in this population and the need for continued investigation to identify those clinical factors that predict the risk of SSI in the neonatal population.

This study corroborates the findings of prior studies that evaluated risk factors for SSI. In 2015, Stey et al. demonstrated associations between adverse events and dirty or infected wound class (OR 2.1, 95% CI 1.4–3.2) and preoperative dialysis (OR 3.8, 95% CI 1.4–10.2) in neonates undergoing major abdominal or thoracic surgery.¹⁴ Preoperative dialysis was a significant predictor of SSI in this study as well; however, this variable was present in only 0.5% of our patient population, which may explain why it was not significant in the RFC and had a low rank compared to other variables when examining standardized beta coefficients. In 2016, patients requiring preoperative nutritional support (OR 1.6, 95% CI 1.2–2.0), preoperative blood transfusion (OR 2.1, 95% CI 1.6–2.7), and current infection (OR 1.9, 95% CI 1.5–2.5) were found to be associated with SSI in neonates in a study by Fawley et al.⁵ Additionally, increases in SSI were observed in patients with longer hospital stays prior to operation and longer mean operative times in a retrospective review of infants in the neonatal intensive care unit who underwent surgery at a single institution.²² Using the NSQIP-P database, Mainzlin et al. identified associations between preoperative sepsis, preoperative nutritional support, open wound with or without infection, duration of procedure, and younger age at surgery and pediatric SSI.² Identification of all of these factors, which have been associated with SSI in a similar surgical population, supports the validity of our study's findings.

We conducted a sensitivity analysis excluding patients from the 2012 PUF to evaluate the association of operative technique with

Table 2
Odds ratios and 95% confidence intervals for significant variables in the hybrid multiple logistic regression model.

Covariate	Adjusted Odds Ratio	95% Confidence Interval	p-value
Preoperative dialysis	3.97	1.32–11.91	0.01
Preoperative sepsis (ref: none)			
SIRS	1.30	0.59–2.85	0.52
Sepsis	1.31	0.71–2.41	0.39
Septic Shock	2.35	1.22–4.52	0.01
Operative technique (ref: laparoscopic)			
Laparoscopic to Open	0.99	0.45–2.21	0.99
Open	1.54	1.03–2.30	0.04
Unknown	1.96	1.26–3.04	<0.01
Surgical wound class (ref: clean)			
Clean-Contaminated	1.12	0.81–1.54	0.50
Contaminated	1.80	1.16–2.77	0.01
Dirty/Infected	1.39	0.85–2.28	0.18
Nutritional Support	1.78	1.37–2.30	<0.01
Open wound ± infection	1.70	1.17–2.47	0.01
Preoperative transfusion	1.53	1.09–2.14	0.01
Hospital days prior to OR	1.00	1.00–1.01	<0.01
Operative duration (minutes)	1.00	1.00–1.00	0.01

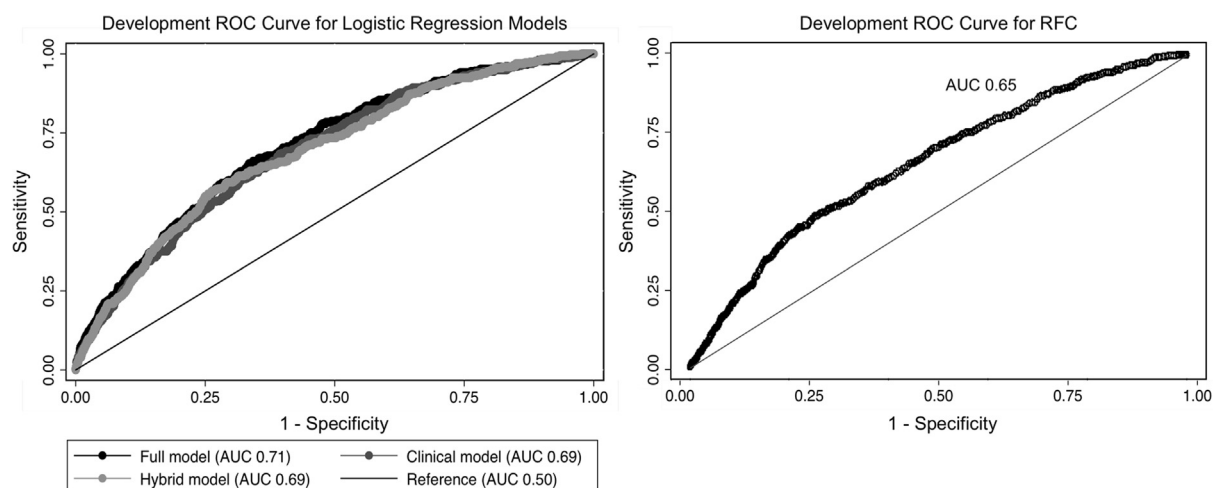


Fig. 1. Receiver operating curves demonstrating area under the curve for each model using the development data.

Table 3

Comparison of SSI prediction model characteristics and performance.

	Random Forest Classification	Full Logistic Regression	Clinical Model	Hybrid Model
n, variables	35	40	23	18
Development AUC	0.650	0.709	0.694	0.694
Validation AUC	0.681	0.649	0.671	0.671
Maximum Accuracy	95.99%	95.88%	95.94%	95.94%

Table 4

Comparison of top variables based on significance by prediction model (importance for RFC and standardized beta coefficients for logistic regression). Number representing ranking with 1 having the greatest association with SSI.

	Random Forest Classification	Full Logistic Regression	Clinical Model	Hybrid Model
Preoperative nutritional support	4	2	1	1
Operative technique - unknown	n/a	n/a	3	2
Operative technique - open	n/a	n/a	10	3
Total operation time	2	4	4	4
Days from Hospital Admission to Operation	1	n/a	n/a	5
Surgical Wound Class - Contaminated	n/a	6	6	6
Open wound \pm infection	7	10	5	7
Preoperative transfusion	6	7	7	8
Preoperative septic shock	n/a	5	9	9
Preoperative dialysis	n/a	11	8	10

SSI. Open surgery remained a risk factor for SSI in all 3 models. Differences in the models were primarily related to the open wound with or without infection, hepatobiliary disease, steroid use 30 days prior to OR, and surgical site variables, all of which had a relatively low prevalence in our patient population, ranging from 0.3% (surgical site: liver) to 7.9%.

Other variables, which have been associated with SSI in pediatric patients, were not significant in our neonatal population. Some of these variables include the presence of cardiac risk factors, the use of steroids, ventilator dependence, and urgent or emergent case type as having greater odds of SSI.^{2,23,24} These factors may play a role in the development of SSI in neonates as well; however, our models identified other risk factors as having a stronger association with SSI when adjusting for all these factors.

We developed 4 prediction models using different statistical

methods (multiple logistic analysis and RFC) and variable strategies (all variables vs. selective). Of these models, the RFC had the greatest ability to predict SSI with an AUC of 0.68. However, the most feasible model was the hybrid that included variables from the RFC and knowledge from the literature and clinical experience. Additionally, the hybrid model had similar predictive ability as the RFC with an AUC of 0.67. It should be noted that although these models were the most accurate of the 4, their ability to predict SSI is still only moderate at best, and we do not advocate for their use in clinical practice. The unbalanced ratio of patients with SSIs compared to those without SSI is referred to as class imbalance. In RFC, this can result in a low rate of prediction because the voting trees tend to favor the majority class, which in this case would be patients without SSI. Over-sampling patients with SSI or under-sampling patients without SSI are possible ways to correct for

class imbalance; however, the low rate of SSI limits the benefit from these methods.²⁵

Alternatively, the reduced predictive ability of the models may be a direct reflection of the multifactorial etiology of SSI in a complex group of patients such as neonates, or it may reflect the relatively low event rate (4%) in our population. In the models developed by Maizlin et al., over 2,000 SSIs were available for model development and validation.² Another explanation is that some important risk factors for neonatal SSI are not currently being captured by NSQIP-P. We hypothesize that perioperative use of antibiotics, appropriate use of prophylactic antibiotics, intraoperative hypothermia, and intraoperative and postoperative hyperglycemia are variables that may play a significant role in the development of SSI. Unfortunately, some variables within this dataset had a large amount of missing data, which we felt best to omit. It is possible that these omitted variables could significantly impact the ability to predict SSI in neonates as well.

Machine-learning algorithms, such as RFCs, are valuable tools for identifying important clinical characteristics from large databases. Despite frequent use of large datasets, these tools are underutilized in pediatric surgery research. Machine-learning algorithms may shed light on some of the risk factors in complex diseases commonly treated by pediatric surgeons. Interpretation of RFC models is difficult given the number of decision tree classifications they are comprised of, which is partly why we chose to use the information provided from the RFC and combine it with clinical knowledge to enhance a logistic regression model. Using this approach, the hybrid model developed had similar ability to the RFC in predicting SSI and was more clinically interpretable. Predictive models can be used to quantify patients' risk for complications and help tailor preoperative counseling to the individual. Ideally, prediction models will help identify modifiable risk factors for SSI so that targeted interventions can be developed to improve outcomes.

There are limitations to this study. The data used in this study was obtained retrospectively; however, the abstraction of this data was performed by trained surgical case reviewers with strict definitions for classification and has been shown to be accurate and reliable.^{26,27} Although NSQIP-P has rigorous methodology for obtaining data, missing data was an issue we encountered; we managed this by excluding incomplete patients from analysis. Another way to potentially manage the missing data would be to use an imputation method; however, we were concerned that imputation would bias the results toward the majority given the low rate of SSI. Despite the above limitations, we felt this dataset was the best available to evaluate our hypothesis. The external validity of these findings is somewhat limited given that the majority of patients included in the dataset are primarily treated at tertiary children's hospitals by pediatric specialists; however, this is partly due to the complexity of disease in the neonatal population. This model should be further tested with a unique cohort to further evaluate its generalizability.

Conclusion

This study demonstrates that utilization of various statistical methods in combination with a large database can identify characteristics predictive of important clinical outcomes such as SSI. Operative technique, need for preoperative nutritional support, preoperative blood transfusion, open wound with or without infection, and total operative time were identified by both statistical methods as risk factors associated with SSI. Future directions include additional model validation with different samples and targeting of modifiable risk factors to reduce the incidence of SSI.

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Appendix A. Supplementary data

Supplementary data related to this article can be found at <https://doi.org/10.1016/j.amjsurg.2018.07.041>.

Appendix

Table 1

Covariates available for model development.

Covariate	Variable Type
Acute renal failure within 7 days prior to surgery	Categorical
ASA classification	Categorical
Asthma	Categorical
Biliary/liver/pancreatic disease	Categorical
Birth weight (<1500 g)	Categorical
Bronchopulmonary dysplasia/Chronic lung disease	Categorical
Cardiac risk factors	Categorical
Case status (elective/urgent/emergent)	Categorical
Cerebral palsy	Categorical
Congenital malformation (binary)	Categorical
Congenital malformation with birthweight	Categorical
Current pneumonia	Categorical
Cystic Fibrosis	Categorical
Developmental delay/impaired cognitive status	Categorical
Do Not Resuscitate (DNR) status	Categorical
Esophageal/Gastric/Intestinal disease	Categorical
Gender	Categorical
Gestational age at birth	Categorical
Hematologic disorder	Categorical
Hispanic ethnicity	Categorical
Hospital status	Categorical
Inotropic support at time of surgery	Categorical
Interventricular hemorrhage	Categorical
Interventricular hemorrhage grade	Categorical
Location of birth	Categorical
Mode of delivery	Categorical
Multiple procedures in same operation	Categorical
Neonate type	Categorical
Neuromuscular Disorder	Categorical
Nutritional support	Categorical
Open wound (with or without infection)	Categorical
Operative technique	Categorical
Ostomy	Categorical
Oxygen support	Categorical
Preoperative dialysis	Categorical
Preoperative transfusion (within 48 h prior to surgery)	Categorical
Previous cardiac surgery	Categorical
Previous CPR within 7 days	Categorical
Prior operation within 30 days	Categorical
Race	Categorical
Seizure Disorder	Categorical
SIRS/sepsis/septic shock within 48 h prior to surgery	Categorical
Small for gestational age	Categorical
Steroid use (within 30 d)	Categorical
Structural pulmonary/Airway abnormalities	Categorical
Structural CNS Abnormality	Categorical
Surgical Site (based on CPT code)	Categorical
Surgical specialty	Categorical
Surgical wound classification	Categorical
Tracheostomy	Categorical
Transfer status	Categorical
Ventilator dependence	Categorical
Weight loss or failure to thrive	Categorical
Admission year	Continuous
Age in days	Continuous

(continued on next page)

Table 1 (continued)

Covariate	Variable Type
Days from Hospital Admission to Operation	Continuous
Total operation time	Continuous
Weight at surgery (pounds)	Continuous

Table 2

Comparison of variables between the development and validation data sets.

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33–34	855 (6.3)	561 (6.3)	294 (6.4)
35–36	1548 (11.4)	1011 (11.3)	537 (11.6)
≥37	8556 (63.0)	5653 (63.0)	2903 (62.9)
Male gender	8259 (60.8)	5405 (60.3)	2854 (61.8)
Race			
American Indian/Alaskan	68 (0.5)	48 (0.5)	20 (0.4)
Asian	352 (2.6)	235 (2.6)	117 (2.5)
Black	2514 (18.5)	1660 (18.5)	854 (18.5)
Native Hawaiian/Pacific	41 (0.3)	28 (0.3)	13 (0.3)
White	9338 (68.7)	6178 (68.9)	3160 (68.4)
Hispanic ethnicity	2105 (15.5)	1383 (15.4)	722 (15.6)
Inpatient status	12,949 (95.3)	8532 (95.1)	202 (4.4)
Transferred	6561 (48.3)	4297 (47.9)	2264 (49.0)
Type of case			
Elective	7096 (52.2)	4718 (52.6)	2378 (51.5)
Urgent	2887 (21.3)	1863 (20.8)	1217 (26.4)
Emergent	3606 (26.5)	2389 (26.6)	1024 (22.2)
ASA class			
1	663 (4.9)	439 (4.9)	224 (4.9)
2	3100 (22.8)	2050 (22.9)	1050 (22.7)
3	6606 (48.6)	4341 (48.4)	2265 (49.0)
4	2934 (21.6)	1944 (21.7)	990 (21.4)
5	211 (1.6)	146 (1.6)	65 (1.4)
Birth location			
Inborn	3581 (26.4)	2425 (27.0)	1156 (25.0)
Outborn	9050 (66.6)	5922 (66.0)	3128 (67.7)
Small for gestational age	494 (3.6)	343 (3.8)	151 (3.3)
Type of delivery			
Vaginal	5302 (39.0)	3489 (38.9)	1813 (39.3)
Scheduled C-section	2284 (16.8)	1505 (16.8)	779 (16.9)
Unscheduled C-section	4274 (31.5)	2829 (31.5)	1445 (31.3)
Congenital malformation	6833 (50.3)	4502 (50.2)	2331 (50.5)
Developmental delay	901 (6.6)	596 (6.6)	305 (6.6)
Neuromuscular disorder	442 (3.3)	299 (3.3)	143 (3.1)
Structural CNS abnormality	2477 (18.2)	1572 (17.5)	905 (19.6)
Seizure disorder	439 (3.2)	299 (3.3)	140 (3.0)
Cerebral palsy	24 (0.2)	15 (0.2)	9 (0.2)
Hematologic disorders	1845 (13.6)	1217 (13.6)	628 (13.6)
Preoperative transfusion	1072 (7.9)	716 (8.0)	356 (7.7)
Asthma	73 (0.5)	42 (0.5)	31 (0.7)
Lung disorder	2232 (16.4)	1469 (16.4)	763 (16.5)
Structural airway abnormality	1796 (13.2)	1178 (13.1)	618 (13.4)
Supplemental oxygen	3688 (27.1)	2403 (26.8)	1285 (27.8)
Ventilator dependent	3412 (25.1)	2219 (24.7)	1193 (25.8)
Tracheostomy	133 (1.0)	85 (1.0)	48 (1.0)
Nutritional Support	5875 (43.2)	3913 (43.6)	1962 (42.5)
Cardiac Risk Factors	5590 (41.1)	3709 (41.4)	1881 (40.7)
Previous cardiothoracic surgery	926 (6.8)	616 (6.9)	310 (6.7)
Inotropic therapy	697 (5.1)	471 (5.3)	226 (4.9)
Esophageal/gastrointestinal disorder	8334 (61.3)	5509 (61.4)	2825 (61.2)
Hepatobiliary disorder	465 (3.4)	313 (3.5)	152 (3.3)
Pre-existing renal failure	94 (0.7)	58 (0.7)	36 (0.8)
Preoperative dialysis	67 (0.5)	36 (0.4)	31 (0.7)
Interventricular hemorrhage	1713 (12.6)	1112 (12.4)	601 (13.0)
Operative duration (minutes)	64 (IQR 35–112)	64 (IQR 34–113)	64 (IQR 35–112)
Weight at surgery (pounds)	6.8 (IQR 5.4–8.4)	6.8 (IQR 5.3–8.4)	6.8 (IQR 5.4–8.4)
Weight loss prior to surgery	550 (4.1)	343 (3.8)	207 (4.5)

Table 2 (continued)

Characteristics	Total	Development dataset	Validation dataset
Hospital days prior to OR	2 (IQR 0–13)	2 (IQR 0–12)	2 (IQR 0–13)
Steroid use 30 days prior to OR	845 (6.2)	538 (6.0)	307 (6.7)
Preoperative CPR	202 (1.5)	137 (1.5)	65 (1.4)
Pre-existing pneumonia	3412 (25.1)	2219 (24.7)	1193 (25.8)
Preoperative sepsis			
SIRS	176 (1.3)	120 (1.3)	56 (1.2)
Sepsis	306 (2.3)	217 (2.4)	89 (1.9)
Septic Shock	191 (1.4)	133 (1.5)	58 (1.3)
OR previous 30 days	420 (3.1)	273 (3.0)	147 (3.2)
Open wound ± infection	1079 (7.9)	687 (7.7)	392 (8.5)
Ostomy present	420 (3.1)	121 (1.4)	50 (1.1)
Surgical wound class			
Clean	4736 (34.9)	3074 (34.3)	1662 (36.0)
Clean-Contaminated	6952 (51.2)	4632 (51.6)	2320 (50.2)
Contaminated	954 (7.0)	618 (6.9)	336 (7.3)
Dirty/Infected	947 (7.0)	646 (7.2)	301 (6.5)
Surgical site infection (SSI)	542 (4.0)	358 (4.0)	184 (4.0)
Superficial	262 (1.9)	178 (2.0)	84 (1.8)
Deep	36 (0.3)	19 (0.2)	17 (0.4)
Organ space infection	143 (1.1)	92 (1.0)	51 (1.1)
Dehiscence	149 (1.1)	100 (1.1)	49 (1.1)

Table 3

Baseline characteristics for neonates that underwent surgery between 2012 and 2015 at NSQIP-P hospitals comparing patients who developed surgical site infection to those who did not.

Characteristics	No SSI (n = 13,589)	SSI (n = 542)	P
Median age (days)	18 (IQR 3–57)	18 (IQR 3–64)	0.42
Prematurity	95.10%	4.90%	<0.01
Gestational age at birth (weeks)			<0.01
<24	1.20%	1.90%	
24	2.50%	4.60%	
25–26	5.10%	7.00%	
27–28	3.40%	5.70%	
29–30	3.30%	4.20%	
31–32	3.60%	5.20%	
33–34	6.30%	6.50%	
35–36	11.40%	10.50%	
≥37	63.30%	54.40%	
Male gender	61.00%	55.70%	0.01
Race			0.02
American Indian/Alaskan	0.50%	0.70%	
Asian	2.60%	3.10%	
Black	18.30%	23.40%	
Native Hawaiian/Pacific	0.30%	0.00%	
White	69.00%	62.90%	
Hispanic ethnicity	15.70%	11.10%	<0.01
Inpatient status	95.10%	99.10%	<0.01
Transferred	48.00%	54.20%	<0.01
Type of case			0.03
Elective	52.00%	57.90%	
Urgent	21.40%	18.60%	
Emergent	26.70%	23.40%	
ASA class			<0.01
1	5.00%	1.90%	
2	23.10%	15.50%	
3	48.30%	55.90%	
4	21.50%	24.00%	
5	1.50%	24.00%	
Birth location			<0.01
Inborn	26.10%	31.90%	
Outborn	66.90%	60.00%	
Small for gestational age	3.70%	3.30%	0.69
Type of delivery			<0.01
Vaginal	39.10%	37.50%	
Scheduled C-section	16.80%	16.60%	
Unscheduled C-section	31.20%	38.80%	
Unknown	13.00%	7.20%	
Congenital malformation	50.30%	50.60%	0.9
Developmental delay	6.60%	7.60%	0.37

(continued on next page)

Table 3 (continued)

Characteristics	No SSI (n = 13,589)	SSI (n = 542)	P
Neuromuscular disorder	3.20%	4.60%	0.07
Structural CNS abnormality	17.80%	27.90%	<0.01
Seizure disorder	3.20%	3.70%	0.54
Cerebral palsy	0.20%	0.20%	0.96
Hematologic disorders	13.30%	19.60%	<0.01
Preoperative transfusion	7.50%	16.40%	<0.01
Asthma	0.50%	0.40%	0.59
Lung disorder	16.20%	22.90%	<0.01
Structural airway abnormality	13.30%	12.00%	0.39
Supplemental oxygen	26.70%	37.50%	<0.01
Ventilator dependent	24.70%	35.20%	<0.01
Tracheostomy	1.00%	1.30%	0.45
Nutritional Support	42.50%	61.10%	<0.01
Cardiac Risk Factors	40.70%	52.40%	<0.01
Previous cardiothoracic surgery	6.70%	8.70%	0.08
Inotropic therapy	5.10%	7.00%	0.04
Esophageal/gastrointestinal disorder	61.30%	62.40%	0.61
Hepatobiliary disorder	3.30%	6.50%	<0.01
Pre-existing renal failure	0.70%	0.70%	0.39
Preoperative dialysis	0.50%	1.10%	0.04
Interventricular hemorrhage	12.50%	15.90%	0.02
Operative duration (minutes)	64 (IQR 34–112)	78 (IQR 46–132)	<0.01
Weight at surgery (pounds)	6.8 (IQR 5.4–8.4)	6.4 (IQR 5.1–7.8)	<0.01
Weight loss prior to surgery	4.00%	4.80%	0.59
Hospital days prior to OR	2 (IQR 0–12)	5 (IQR 1–30)	<0.01
Steroid use 30 days prior to OR	6.10%	10.20%	<0.01
Preoperative CPR	1.50%	1.30%	0.7
Pre-existing pneumonia	24.70%	35.20%	<0.01
Preoperative sepsis			<0.01
SIRS	1.30%	1.70%	
Sepsis	2.10%	5.00%	
Septic Shock	1.30%	3.70%	
OR previous 30 days	3.00%	5.50%	<0.01
Open wound ± infection	7.70%	12.70%	<0.01
Ostomy present	1.20%	2.20%	0.08
Laparoscopic technique	18.50%	10.30%	<0.01
Surgical wound class			<0.01
Clean	35.00%	30.80%	
Clean-Contaminated	51.40%	46.30%	
Contaminated	6.80%	11.30%	
Dirty/Infected	6.80%	11.60%	

Table 4

Relative importance for all variables in the random forest classification. The full dataset refers to the original random forest classification, while the second set of data is a sensitivity analysis using only NSQIP-P PUFs from 2013 to 2015 to reduce data missing from the operative technique variable.

Covariate	Full dataset				PUFs 2013-2015			
	Rank	Relative Importance	Scaled Importance	%	Rank	Relative Importance	Scaled Importance	%
Hospital days prior to OR	1	1168.39	1.00	0.14	1	483.84	1.00	0.14
Operative duration (minutes)	2	1098.66	0.94	0.14	2	469.59	0.97	0.14
Age at surgery (days)	3	797.10	0.68	0.10	5	284.78	0.59	0.08
Nutritional Support	4	755.02	0.65	0.09	3	284.78	0.59	0.08
Weight at surgery (pounds)	5	747.84	0.64	0.09	4	303.89	0.63	0.09
Preoperative transfusion	6	388.26	0.33	0.05	6	147.41	0.30	0.04
Open wound ± infection	7	363.68	0.31	0.04	7	147.01	0.30	0.04
Admission year	8	245.47	0.21	0.03	8	87.30	0.18	0.03
Central Nervous System abnormality	9	187.02	0.16	0.02	11	77.00	0.16	0.02
Lung disorder	10	168.17	0.14	0.02	10	78.57	0.16	0.02
Hepatobiliary disease	11	162.84	0.14	0.02	12	68.82	0.14	0.02
Steroid use 30 days prior to OR	12	161.01	0.14	0.02	13	66.54	0.14	0.02
Tracheostomy	13	127.85	0.11	0.02	15	60.79	0.13	0.02
Inotropic support	14	120.22	0.10	0.01	14	61.63	0.13	0.02
Cardiac Risk Factors*	15	110.90	0.09	0.01	21*	43.95	0.09	0.01
Supplemental oxygen	16	109.94	0.09	0.01	16	45.16	0.09	0.01
Esophageal/gastrointestinal disorder*	17	101.21	0.09	0.01	20*	44.08	0.09	0.01
Asthma	18	99.25	0.08	0.01	17	44.89	0.09	0.01
Neonate type (Preterm)*	19	94.30	0.08	0.01	28*	33.94	0.07	0.01
Transferred from outside hospital	20	93.90	0.08	0.01	22	40.04	0.08	0.01
Congenital malformation	21	92.78	0.08	0.01	19	44.14	0.09	0.01
Ventilator dependent*	22	92.61	0.08	0.01	27*	34.40	0.07	0.01

Table 4 (continued)

Covariate	Full dataset				PUFs 2013–2015			
	Rank	Relative Importance	Scaled Importance	%	Rank	Relative Importance	Scaled Importance	%
Hematologic disorder	23	88.35	0.08	0.01	25	35.80	0.07	0.01
Preoperative pneumonia	24	81.80	0.07	0.01	23	39.55	0.08	0.01
Interventricular hemorrhage	25	77.01	0.07	0.01	24	39.37	0.08	0.01
Preoperative CPR within 7 days*	26	76.97	0.07	0.01	34*	22.35	0.05	0.01
Multiple procedures*	27	76.44	0.07	0.01	18*	44.54	0.09	0.01
Previous Cardiac Surgery*	28	75.43	0.06	0.01	32*	24.49	0.05	0.01
Neuromuscular disorder	29	66.87	0.06	0.01	29	33.11	0.07	0.01
Hispanic ethnicity*	30	63.10	0.05	0.01	26*	34.94	0.07	0.01
Structural airway abnormality	31	60.89	0.05	0.01	30	30.08	0.06	0.01
Developmental delay	32	60.20	0.05	0.01	31	28.83	0.06	0.01
Seizure disorder	33	54.88	0.05	0.01	33	24.38	0.05	0.01
Small for gestational age	34	29.37	0.03	0.00	35	10.42	0.02	0.00
Operation year*	n/a	n/a	n/a	n/a	9*	79.68	0.16	0.02

*denotes a difference in importance ranking greater than 2 positions between the two models.

Table 5

Odds ratios and 95% confidence intervals for the full multiple logistic regression model. The full dataset refers to the original full multiple logistic regression model, while the second set of data is a sensitivity analysis using only NSQIP-P PUFs from 2013 to 2015 to reduce data missing from the operative technique variable.

Covariate	Full Dataset		PUFs 2013–2015	
	Odds Ratio	95% Confidence Interval	Odds Ratio	95% Confidence Interval
Surgical site				
Abdomen	1.67	0.67–4.13	2.22	0.86–5.67
Anus	omitted		omitted	
Bariatric	0.63	0.08–5.12	0.66	0.08–5.62
Biliary	0.54	0.10–2.79	1.06	0.16–6.85
Otolaryngology	omitted		omitted	
Endocrine	1.47	0.88–2.47	1.56	0.86–2.83
Esophagus	omitted		omitted	
Genital	1.38	0.93–2.05	1.38	0.88–2.17
Intestines	7.03	1.89–26.22	9.08	2.34–35.24
Liver	0.36	0.03–4.46	0.58	0.04–9.08
Musculoskeletal	3.65	0.49–27.27	1.05	0.04–26.30
Nervous	3.68	0.41–33.34	omitted	
Pancreas	omitted		omitted	
Skin/soft tissue	1.21	0.73–2.00	1.46	0.82–2.59
Stomach	0.98	0.30–3.27	0.85	0.20–3.70
Thoracic	1.01	0.17–5.97	1.34	0.22–8.16
Urinary	omitted		omitted	
Preoperative dialysis	4.01	1.30–12.42	4.65	1.40–15.42
Preoperative sepsis (ref: none)				
SIRS	1.35	0.61–3.00	1.40	0.59–3.36
Sepsis	1.36	0.73–2.53	1.05	0.49–2.23
Septic Shock	2.49	1.21–5.12	2.93	1.33–6.43
Surgical wound class (ref: clean)				
Clean-Contaminated	1.11	0.80–1.53	1.04	0.72–1.51
Contaminated	1.84	1.19–2.85	2.01	1.22–3.29
Dirty/Infected	1.46	0.88–2.42	1.71	0.97–2.99
Nutritional Support	1.65	1.26–2.16	1.55	1.14–2.09
Open wound ± infection*	1.63	1.12–2.39	1.32	0.83–2.10
Operative technique (ref: laparoscopic)				
Laparoscopic to Open	1.00	0.45–2.23	1.08	0.48–2.44
Open	1.57	1.05–2.36	1.67	1.09–2.55
Unknown	1.53	0.78–2.99	n/a	n/a
Preoperative transfusion	1.50	1.05–2.14	1.74	1.17–2.59
Operative duration (minutes)	1.00	1.00–1.00	1.00	1.00–1.00
Hospital days prior to OR*	1.00	0.99–1.01	1.00	1.00–1.01
Hepatobiliary disease*				
Yes	1.27	0.32	1.26	0.73–2.18
Unknown	1.48	0.06	1.51	1.01–2.26
Race				
American Indian or Alaska Native	1.67	0.50–5.59	1.57	0.50–5.59
Asian	1.44	0.80–2.60	1.56	0.80–2.60
Black or African American	1.10	0.84–1.45	1.07	0.84–1.45
Hawaiian or Other Pacific Islander	omitted		omitted	
Unknown/Not Reported	1.16	0.79–1.70	1.01	0.79–1.70
Hispanic ethnicity	0.78	0.55–1.10	0.78	0.53–1.16
Admission year	0.82	0.51–1.31	0.75	0.45–1.25
Outpatient hospital status	0.53	0.19–1.47	0.29	0.07–1.22
Transferred from outside hospital	0.94	0.70–1.25	0.92	0.67–1.26

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Table 5 (continued)

Covariate	Full Dataset		PUFs 2013–2015	
	Odds Ratio	95% Confidence Interval	Odds Ratio	95% Confidence Interval
Ventilator dependent	1.03	0.74–1.42	1.21	0.84–1.74
Chronic lung disease	1.18	0.84–1.65	1.18	0.80–1.74
Supplemental oxygen	0.96	0.72–1.27	0.98	0.71–1.35
Preoperative pneumonia	omitted		omitted	
Tracheostomy	1.51	0.61–3.73	1.12	0.37–3.38
Hematologic disorder	0.82	0.60–1.12	0.78	0.54–1.11
Steroid use 30 days prior to OR	1.33	0.90–1.97	1.38	0.89–2.14
Cardiac Risk Factors	1.19	0.93–1.53	1.09	0.82–1.45
Previous Cardiac Surgery	0.81	0.52–1.24	0.81	0.50–1.33
Inotropic support	0.95	0.56–1.61	0.97	0.53–1.75
Weight at surgery (pounds)	0.97	0.92–1.02	0.94	0.89–1.00
Neonate type (Preterm)	1.08	0.49–2.39	1.10	0.44–2.73
Gestational Age (ref: term)				
<24 weeks	omitted		omitted	
24 weeks	0.89	0.36–2.22	0.85	0.30–2.44
25–26 weeks	0.82	0.36–1.88	0.84	0.33–2.15
27–28 weeks	1.07	0.46–2.53	1.30	0.50–3.38
29–30 weeks	1.00	0.41–2.46	0.88	0.31–2.49
31–32 weeks	0.96	0.38–2.42	1.08	0.38–3.04
33–34 weeks	0.73	0.30–1.80	0.61	0.21–1.75
35–36 weeks	0.84	0.36–1.99	0.91	0.34–2.43
Unknown	omitted		omitted	
Mode of delivery				
Unscheduled C-Section	1.22	0.86–1.74	1.00	0.53–1.91
Vaginal delivery	1.20	0.85–1.68	1.29	0.86–1.95
Unknown	0.80	0.44–1.44	1.30	0.87–1.93
Birth location				
Outborn	1.02	0.74–1.40	1.15	0.81–1.64
Unknown	1.47	0.89–2.41	1.42	0.80–2.52
Birthweight				
>1500 g	1.08	0.75–1.55	1.39	0.91–2.13
Unknown	1.11	0.63–1.95	1.38	0.72–2.65
Surgery in previous 30 days				
Yes	1.23	0.69–2.20	1.35	0.72–2.51
Unknown	1.440522	0.86–2.42	1.35	0.65–2.82
Ostomy (ref: none)				
Present	1.08	0.51–2.26	1.08	0.51–2.31
Unknown	omitted		omitted	
Case Type (ref: elective)				
Urgent	0.82	0.60–1.12	0.78	0.54–1.12
Emergent	0.91	0.67–1.25	0.76	0.53–1.08
Weight loss				
Yes	1.42	0.86–2.33	1.19	0.65–2.18
Unknown	1.10	0.62–1.94	1.58	0.87–2.84
Multiple procedures	0.92	0.72–1.17	0.86	0.65–1.13
ASA class (ref: 1)				
2	0.83	0.38–1.79	1.06	0.41–2.75
3	0.79	0.37–1.70	1.07	0.42–2.74
4	0.60	0.27–1.33	0.65	0.24–1.75
5	0.31	0.09–1.02	0.43	0.11–1.66
None assigned	1.21	0.34–4.28	1.68	0.40–7.08
Interventricular hemorrhage				
Grade 2	0.91	0.41–2.00	1.01	0.38–2.66
Grade 3	0.72	0.33–1.58	0.91	0.36–2.28
Grade 4	0.48	0.23–1.03	0.69	0.29–1.64
IVH reported but no grade assigned	0.50	0.14–1.79	0.76	0.20–2.91
No IVH reported	0.82	0.48–1.39	1.02	0.53–1.97
Central Nervous System abnormality	1.23	0.86–1.76	1.23	0.82–1.86
Surgical Specialty				
Cardiovascular-Thoracic	omitted		omitted	
General Surgery	2.09	0.54	2.60	0.25–27.10
Neurosurgery	0.60	0.74	2.48	0.05–131.23
Orthopedics	omitted		omitted	
Otolaryngology (ENT)	2.07	0.66	1.35	0.04–40.59
Pediatric Cardiovascular-Thoracic	omitted		omitted	
Pediatric Neurosurgery	0.94	0.97	2.94	0.06–136.27
Pediatric Orthopedic Surgery	omitted		omitted	
Pediatric Otolaryngology (ENT)	5.46	0.21	3.06	0.18–51.41
Pediatric Plastics	5.92	0.23	6.68	0.30–150.21
Pediatric Surgery	1.64	0.65	1.51	0.18–12.78
Pediatric Urology	omitted		omitted	
Plastics	omitted		omitted	
Urology	omitted		omitted	

*denotes a difference in significance between the two models. Bolded variables are statistically significant.

Table 6

Odds ratios and 95% confidence intervals for the clinical multiple logistic regression model. The full dataset refers to the original clinical multiple logistic regression model, while the second set of data is a sensitivity analysis using only NSQIP-P PUFs from 2013 to 2015 to reduce data missing from the operative technique variable.

Covariate	Full Dataset		PUFs 2013–2015	
	Odds Ratio	95% Confidence Interval	Odds Ratio	95% Confidence Interval
Preoperative dialysis	3.98	1.32–12.00	5.44	1.72–17.21
Preoperative sepsis (ref: none)				
SIRS	1.28	0.58–2.83	1.31	0.55–3.11
Sepsis	1.29	0.70–2.38	0.96	0.46–2.03
Septic Shock	2.35	1.16–4.76	2.67	1.25–5.69
Operative technique (ref: laparoscopic)				
Laparoscopic to Open	0.98	0.44–2.18	1.04	0.46–2.34
Open	1.54	1.03–2.30	1.63	1.07–2.48
Unknown	1.96	1.26–3.05	n/a	n/a
Nutritional Support	1.85	1.43–2.40	1.73	1.29–2.31
Surgical wound class (ref: clean)				
Clean-Contaminated	1.14	0.83–1.57	1.08	0.75–1.55
Contaminated	1.82	1.17–2.81	2.00	1.23–3.27
Dirty/Infected	1.36	0.83–2.23	1.48	0.86–2.56
Open wound ± infection*	1.67	1.15–2.42	1.37	0.87–2.15
Preoperative transfusion	1.57	1.12–2.21	1.76	1.20–2.57
Operative duration (minutes)	1.00	1.00–1.00	1.00	1.00–1.00
Steroid use 30 days prior to OR*	1.46	0.99–2.14	1.54	1.01–2.36
Surgical site				
Abdomen	0.70	0.08–6.40	1.09	0.32–3.73
Anus	1.12	0.40–12.02	2.31	0.55–9.76
Bariatric	omitted		omitted	
Biliary	0.49	0.02–10.08	0.83	0.08–8.76
Otolaryngology	1.14	0.11–11.37	2.28	0.61–8.47
Endocrine	omitted		omitted	
Esophagus	1.07	0.11–10.26	1.69	0.49–5.88
Genital	omitted		omitted	
Intestines	0.98	0.11–9.16	1.49	0.45–4.90
Liver*	5.27	0.41–67.46	10.11	1.82–56.22
Musculoskeletal	0.45	0.02–9.02	1.08	0.11–11.05
Nervous	1.48	0.16–13.96	2.27	0.66–7.85
Pancreas	2.98	0.14–65.45	omitted	
Skin/soft tissue	omitted		omitted	
Stomach	0.94	0.10–8.93	1.78	0.51–6.17
Thoracic	0.66	0.05–8.04	0.90	0.14–5.65
Urinary	0.45	0.04–5.53	omitted	
Vascular	omitted		omitted	
Race				
American Indian or Alaska Native	1.75	0.53–5.82	1.66	0.39–7.17
Asian	1.40	0.78–2.52	1.54	0.83–2.87
Black or African American	1.16	0.88–1.52	1.15	0.85–1.56
Hawaiian or Other Pacific Islander	omitted		omitted	
Unknown/Not Reported	1.16	0.79–1.69	1.03	0.65–1.61
Hispanic ethnicity	0.77	0.55–1.09	0.79	0.54–1.17
Outpatient hospital status	0.49	0.18–1.33	0.27	0.07–1.12
Transferred from outside hospital	0.96	0.77–1.21	0.99	0.77–1.28
Ventilator dependent	1.03	0.76–1.40	1.17	0.83–1.65
Cardiac Risk Factors	1.24	0.96–1.58	1.14	0.86–1.51
Previous Cardiac Surgery	0.90	0.60–1.37	0.93	0.58–1.49
Inotropic support	0.87	0.52–1.45	0.89	0.50–1.59
Weight at surgery (pounds)	0.98	0.94–1.05	0.98	0.93–1.03
Neonate type (Preterm)	0.99	0.78–1.25	1.00	0.76–1.30
Interventricular hemorrhage	1.01	0.73–1.40	0.96	0.66–1.38
Ostomy (ref: none)				
Present	1.25	0.61–2.58	1.27	0.61–2.64
Unknown	0.78	0.58–1.05	0.79	0.59–1.07
Multiple procedures	0.95	0.75–1.21	0.88	0.67–1.16
ASA class (ref: 1)				
2	0.89	0.41–1.90	1.11	0.43–2.87
3	0.90	0.43–1.92	1.19	0.47–3.04
4	0.64	0.29–1.43	0.66	0.24–1.77
5	0.35	0.11–1.14	0.44	0.12–1.66
None assigned	1.24	0.35–4.33	1.67	0.40–6.90

*denotes a difference in significance between the two models. Bolded variables are statistically significant.

Table 7
Odds ratios and 95% confidence intervals for the hybrid multiple logistic regression model. The full dataset refers to the original hybrid multiple logistic regression model, while the second set of data is a sensitivity analysis using only NSQIP-P PUFs from 2013 to 2015 to reduce data missing from the operative technique variable.

Covariate	Full Dataset		PUFs 2013–2015	
	Odds Ratio	95% Confidence Interval	Odds Ratio	95% Confidence Interval
Preoperative dialysis	3.97	1.32–11.91	5.53	1.77–17.33
Preoperative sepsis (ref: none)				
SIRS	1.30	0.59–2.85	1.32	0.56–3.11
Sepsis	1.31	0.71–2.41	0.96	0.46–2.02
Septic Shock	2.35	1.22–4.52	2.72	1.33–5.54
Operative technique (ref: laparoscopic)				
Laparoscopic to Open	0.99	0.45–2.21	1.06	0.47–2.37
Open	1.54	1.03–2.30	1.63	1.07–2.49
Unknown	1.96	1.26–3.04	n/a	n/a
Surgical wound class (ref: clean)				
Clean-Contaminated	1.12	0.81–1.54	1.04	0.72–1.50
Contaminated	1.80	1.16–2.77	1.94	1.20–3.16
Dirty/Infected	1.39	0.85–2.28	1.48	0.86–2.55
Nutritional Support	1.78	1.37–2.30	1.65	1.23–2.22
Open wound ± infection*	1.70	1.17–2.47	1.41	0.90–2.21
Preoperative transfusion	1.53	1.09–2.14	1.71	1.17–2.49
Hospital days prior to OR	1.00	1.00–1.01	1.01	1.00–1.01
Operative duration (minutes)	1.00	1.00–1.00	1.00	1.00–1.00
Surgical site				
Abdomen	0.84	0.10–7.08	1.14	0.34–3.88
Anus	1.41	0.14–13.99	2.67	0.64–11.18
Bariatric	omitted		omitted	
Biliary	0.56	0.03–10.71	0.81	0.08–8.57
Otolaryngology	1.19	0.13–10.86	2.03	0.55–7.51
Endocrine	omitted		omitted	
Esophagus	1.21	0.14–10.65	1.70	0.49–5.90
Genital	omitted		omitted	
Intestines	1.14	0.13–9.75	1.50	0.46–4.95
Liver*	6.02	0.51–71.20	9.90	1.79–54.66
Musculoskeletal	0.49	0.03–9.20	1.01	0.10–10.26
Nervous	1.73	0.20–14.95	2.29	0.67–7.85
Pancreas	3.99	0.20–81.45	omitted	
Skin/soft tissue	omitted		omitted	
Stomach	1.04	0.12–9.14	1.70	0.49–5.89
Thoracic	0.80	0.07–9.13	0.98	0.16–6.13
Urinary	0.52	0.05–5.94	omitted	
Vascular	omitted		omitted	
Ventilator dependent	1.78	0.76–1.39	1.16	0.82–1.63
Steroid use 30 days prior to OR	1.38	0.94–2.02	1.46	0.96–2.22
Cardiac Risk Factors	1.17	0.92–1.49	1.07	0.82–1.41
Ostomy (ref: none)				
Present	1.05	0.51–2.18	1.04	0.50–2.17
Unknown	0.79	0.58–1.06	0.79	0.59–1.07
Weight at surgery (pounds)	0.97	0.93–1.02	0.96	0.92–1.02
Neonate type (Preterm)	0.92	0.73–1.17	0.92	0.70–1.21
Hispanic ethnicity	0.77	0.56–1.08	0.77	0.53–1.13
ASA class (ref: 1)				
2	0.89	0.41–1.91	1.12	0.43–2.87
3	0.87	0.41–1.84	1.13	0.44–2.88
4	0.62	0.28–1.38	0.64	0.24–1.72
5	0.33	0.10–1.08	0.42	0.11–1.59
None assigned	1.18	0.34–4.13	1.60	0.39–6.59

*denotes a difference in significance between the two models. Bolded variables are statistically significant.

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