Clinical Data Science – Final Assignment

One Year Survival Analysis of Sepsis Patients

Gal Shlomo Ben David Ohayone

Abstract

In this work I am trying to reproduce the published study of Zhao et al. Zhao have shown that platelets count in sepsis patients can be a predictor for one year survival after admission at the ICU. The statistical analysis using MIMIC-III data. In this work we took into consideration the population demographic details, vital signs, Laboratory test results, calculated risk scores and some admission outcomes.

Introduction

Background

Sepsis is "life-threatening organ dysfunction caused by a dysregulated host response to infection" (Singer et al, 2016) When body's immune system repones to the infection is exaggerated or not effective enough, this incidence can lead to multi-organ failure, shock, and death.

Sepsis causes many deaths worldwide, with a very high (16.7 billion US dollars) cost of caring to those patients that suffer from this incidence. (Walkey et al, 2015) Those hospitalized partienents conservative estimated death rate reach 30%. (Reinhart et al, 2017)

Objectives

Zhao et al have created a Proportional hazards regression model, our goal is to create a logistic regression model that will outperform Zhao model, and will dill with some of its limitations. In our work we would like to be based on data that can be obtained within first 24 hours of ICU admission. We will validate our model with 10 folds cross-validation on train data (70%) and test our model on test data (30%).

Methods

Source of data

For the research we have used the MIMIC (Medical Information Mart for Intensive Care) Database, the same one that were used in the original research. This database include data of patients that have admitted to Intensive Care Unit of Beth Israel Care Unit in 2001 until 2012. I have extracted the data using SQL with google Big Query platform and processed with python data science libraries using Google's Collaboratory online notebook. The MIMIC III database (version 1.4) is publicly available in https://mimic.physionet.org/ under filling the access requirements.

Participants

As explained above, The population in this research have admitted to Intensive Care Unit, therefore they can represent the whole population only partially.

Outcome

- One year survival –

I have extracted the date of death (DOD) for each patient (when available) and calculated the distance in time[days] from this admission. When DOD wasn't available, I have assumed that the patients is still alive and fill the data cell with a 'very big number' in the terms of the problem (10,000). After that, I have screened the values to less than one year (0) and one year or more (1).

- Length of stay in hours
- In hospital mortality (binary)

Predictors

- Partial thromboplastin time (PTT)
- International normalized ratio (INR)
- Prothrombin time (PT)
- White blood cell count (WBC)
- Haemoglobin
- Platelet
- Blood urea nitrogen (BUN)
- Creatinine (Cr)
- Spo2

For all of the above I have made an assumption that their mean value of lab test is the mean between their max value and their mean value of the first 24h of staying at the ICU.

- Sodium (minimal and maximal from first 24 hours of ICU stay)
- Potassium (minimal and maximal from first 24 hours of ICU stay)
- Glucose (minimal and maximal from first 24 hours of ICU stay)

For those 3 I have also calculated "mean"

- Admission type (Emergency, Elective, or Urgent) 2. Age Sex
 - Race (White, African American, Asian, or Others)
- Heart rate (HR)
 - Systolic blood pressure (SBP)
- Diastolic blood pressure (DBP)
- Respiratory rate (RR) Temperature

- Glasgow coma scale (GCS) 'GCS', 'GCSMotor', 'GCSVerbal', 'GCSEyes'
 - o (I have used more than one feature of GCS in order to try and improve results)
- Patients' simplified acute physiology score (SAPSII)
- Sequential organ failure assessment (SOFA)

Sample size

There were 5152 admission that were taken into consideration, (I had another try with less restricted excluding rules and were able to create better model in some way)

Missing data

I have eliminated rows with missing data at the lab columns.

For censored data, like 'days untill death'— I have filled by myself as mention above.

Statistical analysis methods

I have used Kaplan-Meier curve in order to compare survival groups, I have created a logistics regression model in order to forecast one year survival.

Risk groups

blood platelets level \geq 100 (×109/L) and \leq 300 (×109/L) were considered Normal. (the final risk group were :Thrombocytosis, Normal, and Thrombocytopenia)

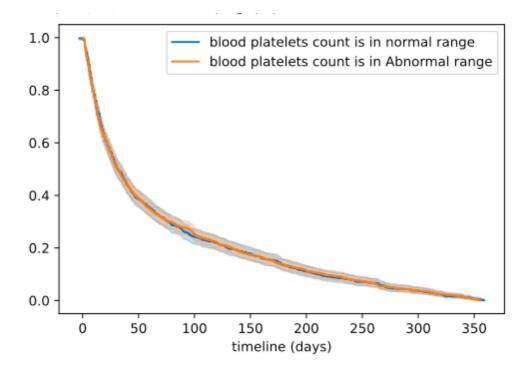
Development vs. validation

Results

Baseline patient characteristics and outcomes. I have painted the features that I believe that have big enough p-value in order to be correlated to our risk group.

	Table 1									
	Grouped by group	ped by								
		Missing	Overall	Normal	Thrombocytopeni a	Thrombocytosis	P-Value			
n (00)		22	5152	3063	962	1127	0.000			
GCS, mean (SD) GCSMotor, mean (SD)		28 86	13.5 (2.8) 5.0 (1.6)	13.4 (2.9) 4.9 (1.7)	13.6 (2.6) 5.2 (1.5)	13.6 (2.6) 5.1 (1.6)	0.009 <0.001			
, , ,		71	2.9 (2.2)	2.8 (2.2)	3.3 (2.1)	2.8 (2.2)	<0.001			
GCSVerbal, mean (SD) GCSEyes, mean (SD)		44	3.0 (1.1)	2.9 (1.2)	3.1 (1.1)	3.0 (1.1)	<0.001			
sapsii, mean (SD)		0	43.6 (15.4)	43.2 (14.9)	46.4 (16.8)	42.1 (15.0)	<0.001			
sapsii_prob, mean (SD)		0	0.3 (0.3)	0.3 (0.3)	0.4 (0.3)	0.3 (0.3)	<0.001			
HEART_RATE_HR, mean (SD)		910	85.6 (18.9)	84.6 (18.8)	87.0 (19.5)	86.9 (18.3)	0.001			
SYSTOLIC_BLOOD_P RESSURE_SBP, mean (SD)		442	118.3 (33.7)	118.8 (34.0)	116.5 (42.0)	118.5 (23.2)	0.224			
DIASTOLIC_BLOOD_P RESSURE_DBP, mean (SD)		442	64.8 (34.7)	64.3 (32.6)	65.7 (37.3)	65.3 (37.7)	0.495			
SOFA, mean (SD)		0	6.8 (3.5)	6.5 (3.2)	9.0 (3.9)	5.7 (3.0)	<0.001			
sepsis, mean (SD)		0	1.0 (0.0)	1.0 (0.0)	1.0 (0.0)	1.0 (0.0)	nan			
organ_failure, mean (SD)		0	0.9 (0.3)	0.9 (0.3)	0.9 (0.3)	0.9 (0.3)	<0.001			
respiratory, mean (SD)		0	0.4 (0.5)	0.4 (0.5)	0.4 (0.5)	0.4 (0.5)	0.161			
cardiovascular, mean (SD)		0	0.4 (0.5)	0.4 (0.5)	0.5 (0.5)	0.4 (0.5)	0.19			
renal, mean (SD)		0	0.7 (0.5)	0.7 (0.5)	0.7 (0.5)	0.6 (0.5)	0.005 <0.001			
hepatic, mean (SD) hematologic, mean		0	0.1 (0.3)	0.1 (0.3)	0.2 (0.4)	0.1 (0.2)	<0.001			
(SD) metabolic, mean (SD)		0	0.2 (0.4)	0.2 (0.4)	0.3 (0.4)	0.2 (0.4)	<0.001			
metabolic, mean (SD)		0	0.2 (0.4)	0.2 (0.4)	0.3 (0.4)	0.2 (0.4)	0.256			
neurologic, mean (SD)	F	0	` ′	1256 (41.0)	382 (39.7)	502 (44.5)	0.053			
SEX, n (%)	M	U	2140 (41.5) 3012 (58.5)	1807 (59.0)	580 (60.3)	625 (55.5)	0.053			
AGE, mean (SD)		0	64.5 (15.2)	65.6 (15.1)	59.5 (14.9)	65.6 (14.5)	<0.001			
RACE, n (%)	ASIAN	0	119 (2.3)	86 (2.8)	17 (1.8)	16 (1.4)	0.001			
	BLACK OTHER		574 (11.1) 751 (14.6)	365 (11.9) 431 (14.1)	84 (8.7) 178 (18.5)	125 (11.1) 142 (12.6)				
	WHITE		3704 (71.9)	2178 (71.1)	682 (70.9)	844 (74.9)				
ADMISSION_TYPE, n (%)	ELECTIVE	0	199 (3.9)	130 (4.2)	29 (3.0)	40 (3.5)	0.201			
	EMERGENCY		4816 (93.5)	2856 (93.2)	900 (93.6)	1060 (94.1)				
	URGENT		137 (2.7)	77 (2.5)	33 (3.4) 252291.3	27 (2.4)				
icustay_id, mean (SD)		0	250229.4 (28903.8)	250058.8 (28866.5)	(28894.9)	248932.8 (28945.2)	0.026			
CREATININE, mean (SD)		12	2.1 (2.0)	2.2 (2.1)	2.0 (1.7)	2.0 (2.0)	0.007			
GLUCOSE, mean (SD)		10	147.9 (62.0)	148.6 (62.3)	144.1 (61.2)	149.1 (62.1)	0.107			
HEMOGLOBIN, mean (SD)		13	10.3 (1.8)	10.5 (1.8)	9.8 (1.7)	10.0 (1.7)	<0.001			
PLATELET, mean (SD)		14	216.5 (134.1)	192.1 (54.6)	62.8 (23.6)	413.8 (113.0)	<0.001			
PTT, mean (SD)		378	41.5 (19.9)	41.3 (20.0)	44.4 (20.1)	39.5 (18.9)	<0.001			
INR, mean (SD) PT, mean (SD)		351 350	1.8 (1.2) 17.9 (8.4)	1.8 (1.2) 17.8 (8.8)	1.8 (0.9) 18.3 (6.3)	1.8 (1.4) 17.8 (8.9)	0.264 0.245			
BUN, mean (SD)		12	36.7 (25.5)	36.4 (24.9)	39.7 (27.0)	35.0 (25.7)	<0.001			
WBC, mean (SD)		17	13.9 (14.0)	13.6 (8.4)	10.8 (26.3)	17.4 (9.5)	<0.001			
POTASSIUM_min, mean (SD)		8	3.8 (0.6)	3.8 (0.6)	3.7 (0.6)	3.8 (0.6)	<0.001			
POTASSIUM_max, mean (SD)		8	4.7 (1.0)	4.7 (1.0)	4.5 (0.9)	4.7 (0.9)	<0.001			
SODIUM_min, mean (SD)		9	136.1 (5.5)	136.3 (5.4)	135.7 (6.1)	136.0 (5.5)	0.006			
SODIUM_max, mean (SD)		9	140.0 (5.4)	140.2 (5.2)	139.6 (6.0)	139.6 (5.2)	<0.001			
tempc_mean, mean (SD)		58	36.9 (0.8)	36.9 (0.8)	36.8 (0.8)	36.9 (0.7)	<0.001			
tempc_max, mean (SD)		58	37.7 (1.0)	37.8 (1.0)	37.6 (1.0)	37.7 (0.9)	0.001			
tempc_min, mean (SD)		58	36.1 (0.9)	36.1 (0.9)	35.9 (1.0)	36.1 (0.8)	<0.001			
resprate_mean, mean (SD)		20	20.6 (4.7)	20.4 (4.6)	21.0 (5.3)	20.5 (4.7)	0.005			
resprate_max, mean (SD)		20	29.2 (7.5)	29.1 (7.3)	29.6 (8.2)	29.2 (7.3)	0.181			
resprate_min, mean (SD)		20	13.3 (4.2)	13.2 (4.1)	13.6 (4.4)	13.2 (4.3)	0.025			
RRT, mean (SD)		0	0.2 (0.4)	0.2 (0.4)	0.3 (0.4)	0.2 (0.4)	<0.001			

Survival analysis



CoxPHFitter for the two main risk group. It is impossible to infer from this graph if a patients is having more chance to survive due to the fact that he have normal or abnormal blood platelents counts.

One year survival analysis

Table 2 – univariate and multivariate survivals analysis

I have colored the features that have the most statistical significance.

Table2

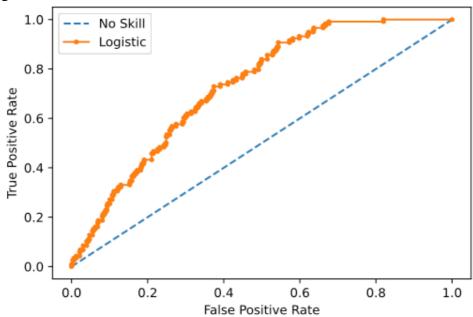
		1 able2 Univariate analysis				Multivariate analysis				
			Ollivariat	lower	upper		Willivaria	lower	upper	
#	Feature name	Odds Ratio	p-value		bound 95%	Odds Ratio	p-value			
0	ADMISSION_TYPE_ ELECTIVE	0.95	0.50	-0.19	0.09	1.06	0.69	-0.21	0.32	
1	ADMISSION_TYPE_ EMERGENCY	1.06	0.34	-0.06	0.16	0.99	0.96	-0.26	0.24	
2	ADMISSION_TYPE_ URGENT	0.94	0.50	-0.23	0.11	0.94	0.67	-0.34	0.22	
3	AGE	1.00	1.00	0.00	0.00	1.72	0.00	0.46	0.63	
4	BUN	1.00	0.07	0.00	0.00	1.00	0.09	0.00	0.00	
5	cardiovascular	1.17	0.00	0.10	0.21	1.03	0.37	-0.03	0.08	
6	CREATININE	1.00	0.48	-0.02	0.01	0.97	0.00	-0.05	-0.01	
7	DIASTOLIC_BLOOD _PRESSURE_DBP	1.00	0.76	0.00	0.00	1.00	0.20	0.00	0.00	
8	GCS	1.00	0.35	-0.01	0.01	1.01	0.04	0.00	0.02	
9	GCSEyes	0.99	0.25	-0.04	0.01	0.99	0.55	-0.04	0.02	
10	GCSMotor	0.99	0.13	-0.03	0.00	1.01	0.22	-0.01	0.04	
11	GCSVerbal	1.00	0.56	-0.02	0.01	0.99	0.20	-0.03	0.01	
12	GLUCOSE	1.00	0.43	0.00	0.00	1.00	0.11	0.00	0.00	
13	group_Normal	0.92	0.01	-0.13	-0.02	0.98	0.75	-0.14	0.10	
14	group_Thrombocytop enia	1.12	0.00	0.04	0.18	1.05	0.48	-0.09	0.18	
15	group_Thrombocytos is	1.01	0.73	-0.05	0.08	0.98	0.82	-0.15	0.12	
16	HEART_RATE_HR	1.00	0.00	0.00	0.00	1.00	0.03	0.00	0.00	
17	hematologic	1.12	0.00	0.05	0.18	1.00	0.91	-0.07	0.06	
18	HEMOGLOBIN	0.99	0.08	-0.03	0.00	0.97	0.00	-0.04	-0.01	
19	hepatic	1.21	0.00	0.10	0.28	1.03	0.60	-0.07	0.12	
20	INR	1.02	0.13	-0.01	0.04	1.08	0.00	0.04	0.11	
21	LOS_IN_HOURS	1.00	0.25	0.00	0.00	1.00	0.04	0.00	0.00	
22	metabolic	1.14	0.00	0.06	0.20	0.97	0.32	-0.11	0.03	
23	neurologic	1.18 1.15	0.00	0.09 0.06	0.25 0.23	0.91 1.06	0.02	-0.18 -0.04	-0.02 0.16	
25	organ_failure PLATELET	1.00	0.00	0.00	0.23	1.00	0.24	0.00	0.00	
26	POTASSIUM max	1.00	0.77	-0.02	0.00	1.02	0.32	-0.02	0.05	
27	POTASSIUM min	1.02	0.77	-0.02	0.07	1.07	0.01	0.02	0.03	
28	PT PT	1.00	0.01	0.00	0.01	0.99	0.00	-0.01	0.00	
29	PTT	1.00	0.59	0.00	0.00	1.00	0.01	0.00	0.00	
30	RACE ASIAN	1.16	0.10	-0.03	0.33	1.03	0.75	-0.18	0.24	
31	RACE_BLACK	0.99	0.73	-0.10	0.07	0.93	0.32	-0.22	0.07	
32	RACE_CARIBBEAN_I SLAND	1.46	0.60	-1.01	1.76	0.67	0.54	-1.71	0.90	
33	RACE_HISPANIC_or _LATINO_HONDURA N	1.46	0.71	-1.58	2.34	0.63	0.62	-2.30	1.38	
34	RACE_HISPANIC_or _LATINO_MEXICAN	1.46	0.71	-1.58	2.34	0.75	0.77	-2.14	1.57	
35	RACE_OTHER	1.02	0.61	-0.06	0.10	1.02	0.82	-0.12	0.16	
36	RACE_WHITE	0.98	0.48	-0.08	0.04	1.02	0.73	-0.11	0.16	
37	renal	1.11	0.00	0.04	0.16	1.00	0.96	-0.07	0.07	
38	respiratory	1.09	0.00	0.03	0.14	1.03	0.35	-0.03	0.09	
39	resprate_max	1.01	0.00	0.00	0.01	1.00	0.31	0.00	0.01	
40	resprate_mean	1.01	0.00	0.01	0.02	0.99	0.15	-0.02	0.00	
41	resprate_min	1.00	0.81	-0.01	0.01	1.01	0.09	0.00	0.01	
42	RRT	1.08	0.03	0.01	0.15	1.19	0.00	0.10	0.26	
43	sapsii sapsii prob	1.00 1.35	0.00	0.00 0.19	0.01	1.01 1.36	0.00	0.00	0.01 0.54	
44	SEX M	1.02	0.00	-0.03	0.41	1.36	0.01	-0.03	0.08	
46	SODIUM max	1.00	0.43	-0.03	0.00	1.00	0.54	-0.03	0.08	
47	SODIUM min	1.00	0.06	-0.01	0.00	1.01	0.05	0.00	0.01	
48	SOFA	1.02	0.00	0.02	0.00	0.99	0.03	-0.02	0.00	
49	SYSTOLIC_BLOOD_ PRESSURE SBP	1.00	0.10	0.00	0.00	1.00	0.67	0.00	0.00	
50	tempc max	1.00	1.00	-0.03	0.03	0.96	0.07	-0.09	0.00	
51	tempc_max	0.97	0.15	-0.03	0.03	0.99	0.07	-0.09	0.00	
52	tempc_mcan	0.97	0.09	-0.06	0.00	0.98	0.34	-0.07	0.02	
53	WBC	1.00	0.60	0.00	0.00	1.00	0.91	0.00	0.00	

Multivariate model – I have created a logistic regression model based on those features.

Multivariate Model discrimination

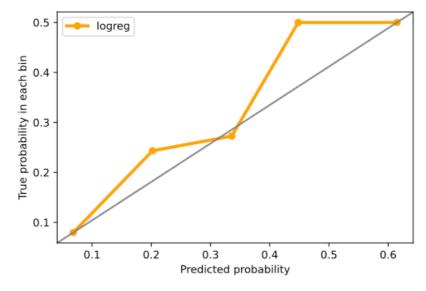
the model preform better than no skilled classifier but doesn't outperform the original article model.

No Skill: ROC AUC=0.500 Logistic: ROC AUC=0.729



Multivariate Model calibration

As it is possible to see, the model is quite calibrated when being divided into 7 beans, it is possible to see that the calculated risk is very similar to the real risk.



Discussion

Limitations- the population was chosen under very limited criteria as mention above, it might be interesting to make the same research with bigger population. I have followed the instructions and haven't increased the population group. I suggest that that problem of sepsis survival analysis can be defined with less restriction on the population clinical data, and maybe more defined on the demographic data.

It might be interesting to explore other features that appears in this database.

Interpretation- we have gotten some similar results in our bassline table, model performance and on the data analysis odds ratio – to the information that appears in the article.

Implications- I don't think that my model will be more relevant than the one that appears in the article, but exploring the data was an instructive experience.

Bibliography

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Walkey AJ, Lagu T, Lindenauer PK Ann Am Thorac Soc. 2015 Feb; 12(2):216-20. Those

I am sorry but I didn't manage to extract my python file as PDF. I have tried to do it in many different ways.

Here is a colab link:

https://colab.research.google.com/drive/1iYTBqcXe8eAJxkwKJ-

hNoa9N2UJsXOwj?usp=sharing

another link (git):

https://github.com/galbendavids/project1/blob/master/CDS ex final.ipynb