

Track patient recovery in real-time by processing streaming data

BIOMEDICAL DATA DESIGN

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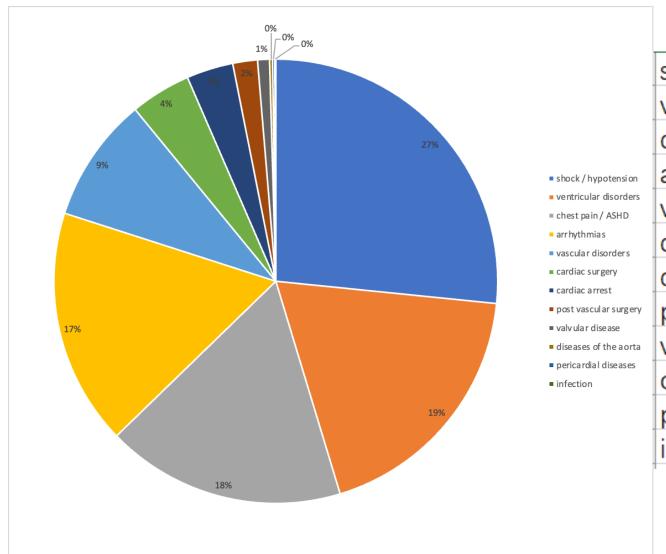
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Main symptoms of the disease:

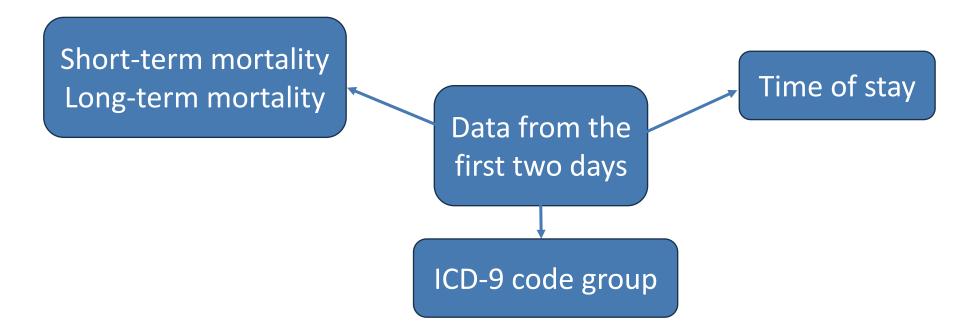


shock / hypotension	1815
ventricular disorders	1277
chest pain / ASHD	1189
arrhythmias	1174
vascular disorders	622
cardiac surgery	301
cardiac arrest	232
post vascular surgery	123
valvular disease	59
diseases of the aorta	15
pericardial diseases	11
infection	5
	I

- 1.Heart rate
- 2.Blood pressure
- 3. Cardic Output
- 4. Coronary Artery Blood Flow
- 5.Body Temperature (BT)
- 6.Oxygen Saturation
- 7. Peripheral Capillary Oxygen Saturation (SpO2)
- 8. Oxygen Saturation (O2sat)
- 9.C-reactive Protein (CRP)
- 10.Lipid Profile
- 11.Age, sex, smoke history, etc...

Benchmarking deep learning models on large healthcare datasets

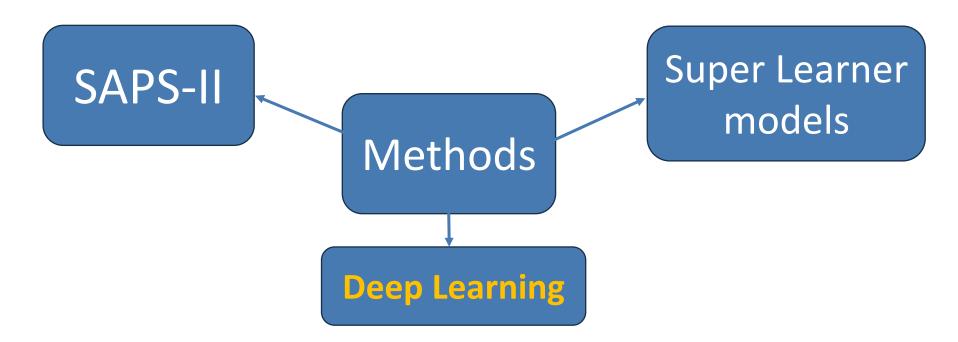
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Sanjay Purushotham ^{a \, 1} \boxtimes , Chuizheng Meng ^{b \, 1} \boxtimes , Zhengping Che ^{a} \boxtimes , Yan Liu ^{a} \supseteq \boxtimes  Show more \checkmark
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Benchmarking deep learning models on large healthcare datasets

<u>Sanjay Purushotham</u> ^{a 1} ⋈, <u>Chuizheng Meng</u> ^{b 1} ⋈, <u>Zhengping Che</u> ^a ⋈, <u>Yan Liu</u> ^a 😃

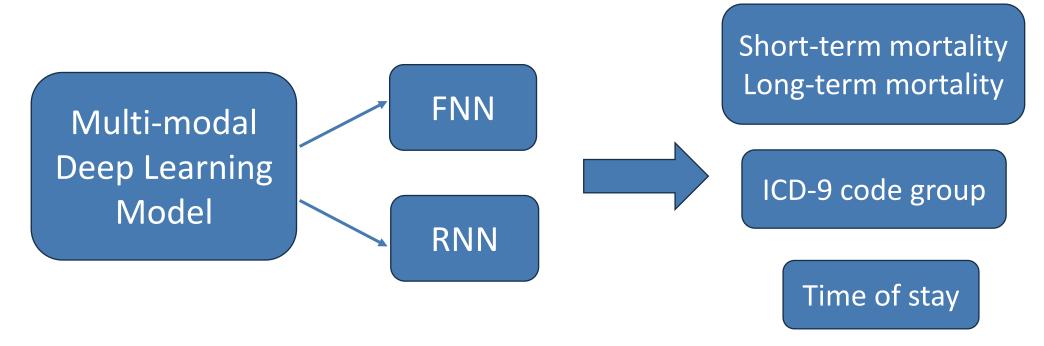
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Benchmarking deep learning models on large healthcare datasets

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Mortality

Algorithm	Feature Set A, 24-h data		Feature Set A, 48-h data	
	AUROC score	AUPRC score	AUROC score	AUPRC score
SuperLearner-II	0.8673±0.0045	0.4968±0.0097	0.8595±0.0035	0.4422±0.0200
FFN	0.8496±0.0047	0.4632±0.0074	0.8375±0.0041	0.4090±0.0169
RNN	0.8544+0.0053	0.4519±0.0145	0.8618+0.0059	0.4458±0.0144
MMDL	0.8664±0.0056	0.4776±0.0162	0.8737 ± 0.0045	0.4714 ± 0.0176
	SuperLearner-II FFN RNN	AUROC score SuperLearner-II 0.8673±0.0045 FFN 0.8496±0.0047 RNN 0.8544+0.0053	AUROC score AUPRC score SuperLearner-II 0.8673±0.0045 0.4968±0.0097 FFN 0.8496±0.0047 0.4632±0.0074 RNN 0.8544±0.0053 0.4519±0.0145	AUROC score AUPRC score AUROC score SuperLearner-II 0.8673±0.0045 0.4968±0.0097 0.8595±0.0035 FFN 0.8496±0.0047 0.4632±0.0074 0.8375±0.0041 RNN 0.8544±0.0053 0.4519±0.0145 0.8618±0.0059

ICD-9 code & Time of stay



Analysis | Open Access | Published: 17 June 2019

Multitask learning and benchmarking with clinical time series data

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41k Accesses 231 Citations 18 Altmetric Metrics

elCU Real time Time of stay



03 Dataset Analysis

Dataset Analysis

vitalPeriodic & vitalAperiodic & nurseCharting

Vital signs — Time series

diagnosis & patient

Patients' information —— Labels

Dataset Analysis

diagnosisid	patientunitstayid	activeupondischarge	diagnosisoffset	diagnosisstring
4035907	143870	TRUE	10	cardiovascular chest pain / ASHD coronary artery disease
3843251	143870	TRUE	10	cardiovascular post vascular surgery s/p cartoid endarterectomy
3460672	143870	TRUE	10	cardiovascular arrhythmias bradycardia
3717065	151179	FALSE	29	cardiovascular shock / hypotension septic shock
4102418	151179	FALSE	120	cardiovascular shock / hypotension septic shock
3885168	151179	TRUE	3929	cardiovascular shock / hypotension septic shock
4053934	151179	TRUE	3929	cardiovascular shock / hypotension hypotension
3850876	151900	FALSE	148	cardiovascular shock / hypotension septic shock
3707280	151900	FALSE	939	cardiovascular shock / hypotension septic shock
4192192	151900	FALSE	939	cardiovascular chest pain / ASHD acute coronary syndrome
3379776	151900	TRUE	2895	cardiovascular chest pain / ASHD acute coronary syndrome
3892141	151900	TRUE	2895	cardiovascular shock / hypotension septic shock
3678632	152954	FALSE	39	cardiovascular shock / hypotension signs and symptoms of sepsis (SIRS)
3977729	152954	FALSE	39	cardiovascular ventricular disorders congestive heart failure
4144394	152954	FALSE	219	cardiovascular shock / hypotension signs and symptoms of sepsis (SIRS)
3757248	152954	FALSE	219	cardiovascular ventricular disorders congestive heart failure

O4 Next Step

Next Step

- 1.Extract more meaningful data
- 1.1 Extraction
- 1.2 Interpolation, Correction

2. Replicate the deep learning model as baseline and try new models.



References

[1] Rocheteau, Emma, et al. "Predicting Patient Outcomes with Graph Representation Learning." arXiv preprint arXiv:2101.03940 (2021).

[2] The eICU Collaborative Research Database, a freely available multi-center database for critical care research. Pollard TJ, Johnson AEW, Raffa JD, Celi LA, Mark RG and Badawi O. Scientific Data (2018).