

# Towards the development of a continuous model for the assessment of bacteremia, bloodstream infection and sepsis

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Joint work with Bernard Hernandez, Cristina Soguero-Ruiz, Inmaculada Mora-Jiménez, and A. G. Marques

June 21st, 2024

**Rey Juan Carlos University (URJC), Madrid, Spain**

- B.Sc. in Telecommunications
- M.Sc. in Telecommunications
- Ph.D. in Artificial Intelligence and Data Science

**Università Campus Bio-Medico (UCBM), Roma, Italy**

- Visiting researcher (PhD)

**Imperial College London (ICL), London, United Kingdom**

- Visiting researcher (PhD)

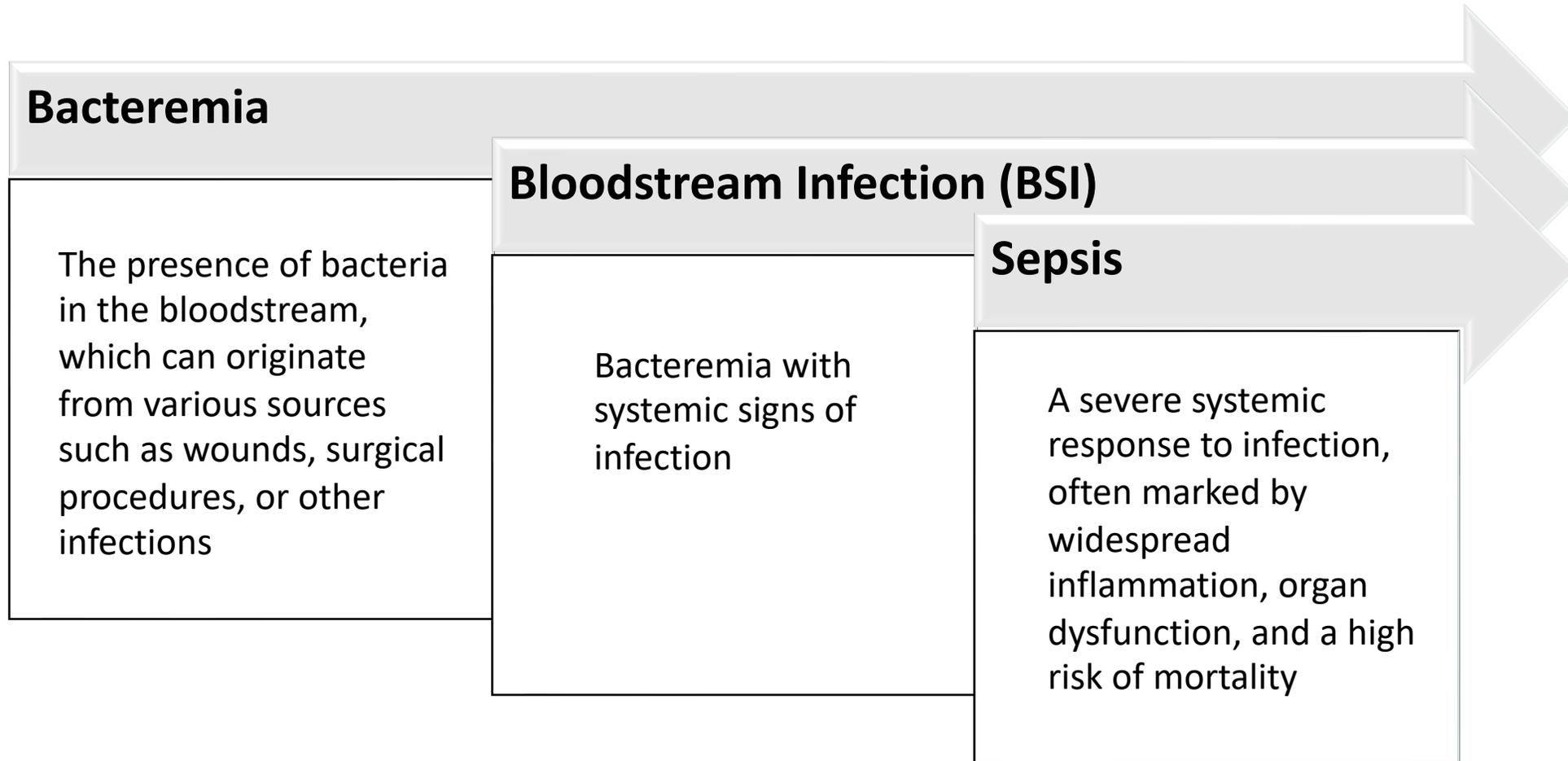


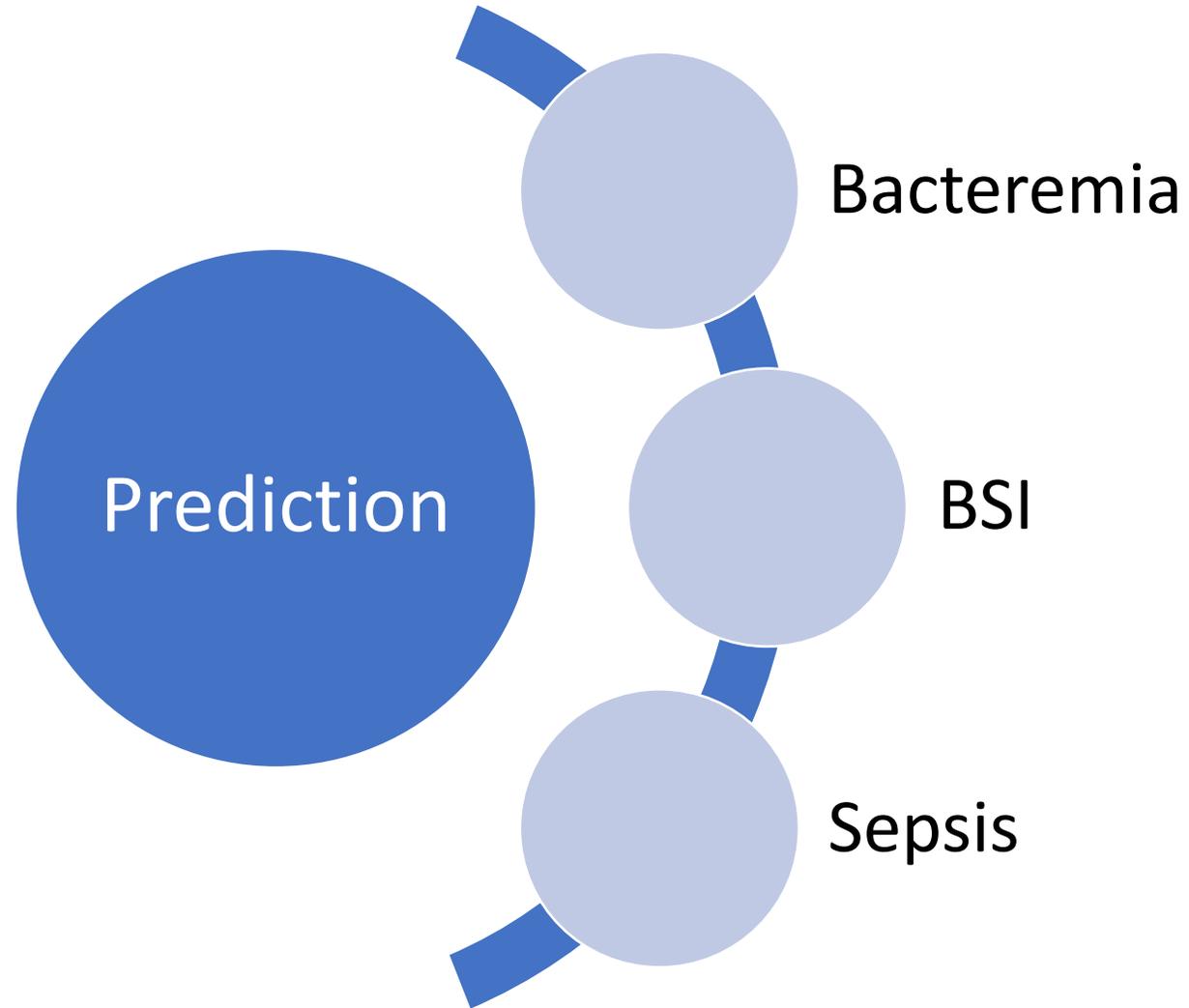
**Rey Juan Carlos University (URJC), Madrid, Spain. Ph.D. in Artificial Intelligence and Data Science**

- Electronic Health Record
- Irregular Multivariate Time Series
- Distance measures and kernel methods
- Multimodal Architectures
- Interpretable Recurrent Neural Networks
- Interpretable Spatio-Temporal Graph Neural Networks

**Clinical Task**

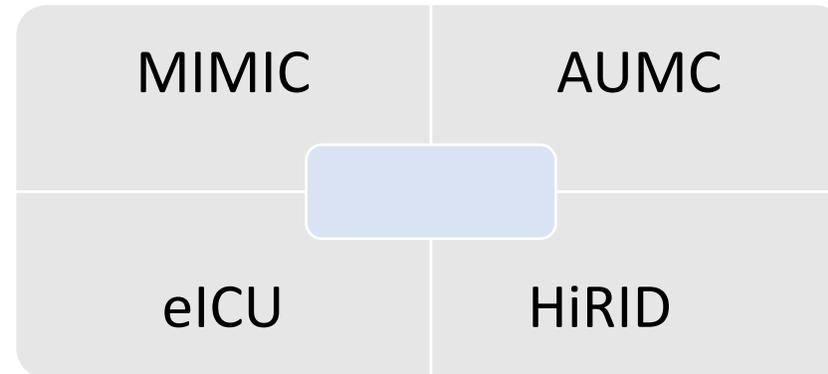
- [1] Escudero-Arnanz Ó, Mora-Jiménez I, Martínez-Agüero S, Álvarez-Rodríguez J, Soguero-Ruiz C. Temporal Feature Selection for Characterizing Antimicrobial Multidrug Resistance in the Intensive Care Unit. InAAI4H@ ECAI 2020 Sep (pp. 54-59)
- [2] Escudero-Arnanz, Óscar, Rodríguez-Álvarez, J., Mikalsen, K. Ø., Jenssen, R., & Soguero-Ruiz, C. On the Use of Time Series Kernel and Dimensionality Reduction to Identify the Acquisition of Antimicrobial Multidrug Resistance in the Intensive Care Unit. KDD 2021 Health Day and 2021 KDD Workshop on Applied Data Science for Healthcare.
- [3] Escudero-Arnanz Ó, Mora-Jiménez I, Martínez-Agüero S, Álvarez-Rodríguez J, Soguero-Ruiz C. Feature Selection and Tree-based Models to Predict Multidrug-Resistance. CASEIB 2020.
- [4] Oscar Escudero-Arnanz, Antonio G. Marques, Cristina Soguero-Ruiz, Inmaculada Mora-Jiménez, Gregorio Robles. dtwParallel: A Python Package to Efficiently Compute Dynamic Time Warping Between Time Series. SoftwareX, 2023.
- [5] Oscar Escudero-Arnanz, Antonio G. Marques, Rosa Sicilia, Cristina Soguero-Ruiz. Low-Rank Tensor Completion for Heart Failure Detection in Multivariate Time Series with Missing Data. IEEE 37th International Symposium on Computer Based Medical Systems 2024.
- [6] Oscar Escudero-Arnanz, Antonio G. Marques, Inmaculada Mora-Jiménez, Joaquín Álvarez-Rodríguez, Cristina Soguero-Ruiz. Leveraging Multivariate Time Series Analysis and Machine Learning for the Characterization of Antimicrobial Resistance in the Intensive Care Unit. Engineering Applications of Artificial Intelligence. (submitted, 2024, under review)
- [7] Oscar Escudero-Arnanz, Cristina Soguero-Ruiz, Inmaculada Mora-Jiménez, Joaquín Álvarez-Rodríguez, Antonio G. Marques. Explainable AI Techniques for Irregular Temporal Prediction of Antimicrobial Multidrug Resistance Acquisition in Intensive Care Unit Patients. Engineering Applications of Artificial Intelligence. (to submit on 30<sup>th</sup> in June 2024)
- [8] Oscar Escudero-Arnanz, Cristina Soguero-Ruiz, Inmaculada Mora-Jiménez, Joaquín Álvarez-Rodríguez, Antonio G. Marques. Explainable Spatio-Temporal Graph Architecture for Irregular Multivariate Time Series in Inference Tasks. IEEE TSIPN. (to submit on 30<sup>th</sup> in July 2024)





## Data Sources

- We will use data from four public datasets...

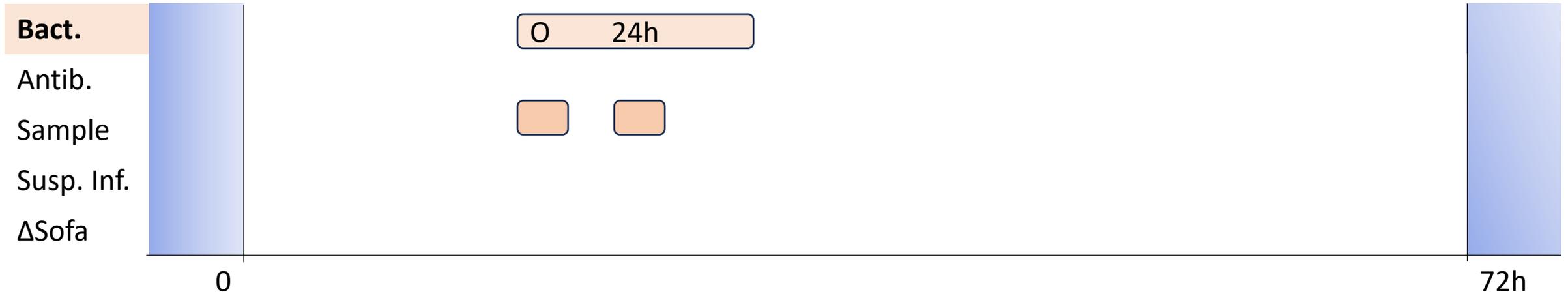


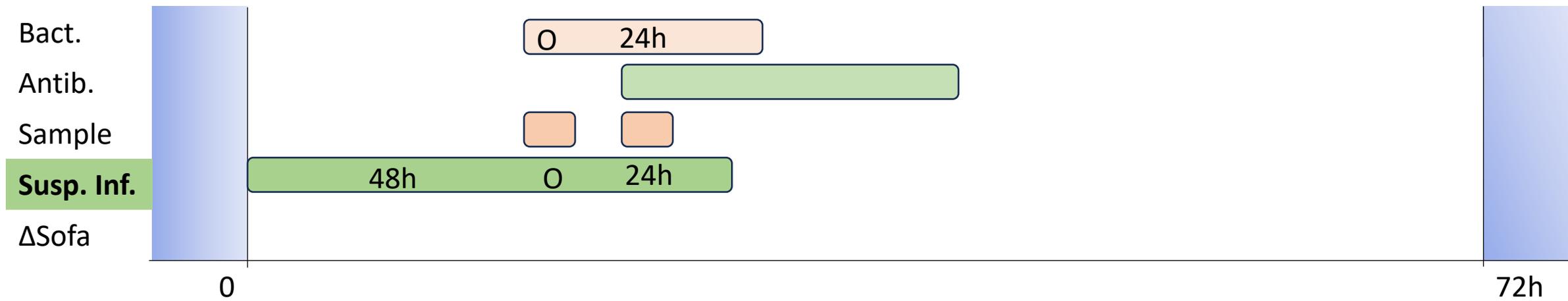
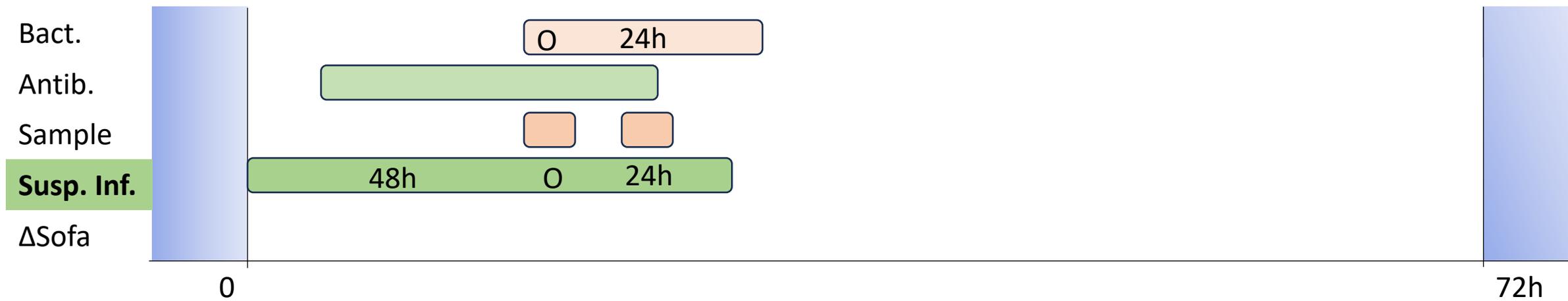
- Focusing on patients in the Intensive Care Unit (ICU).

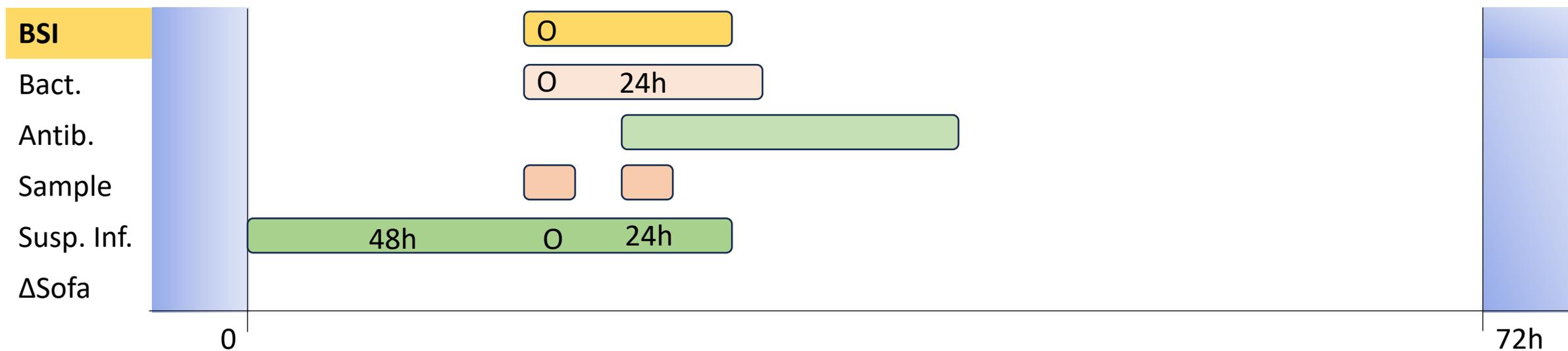
## Inclusion Criteria for Patients

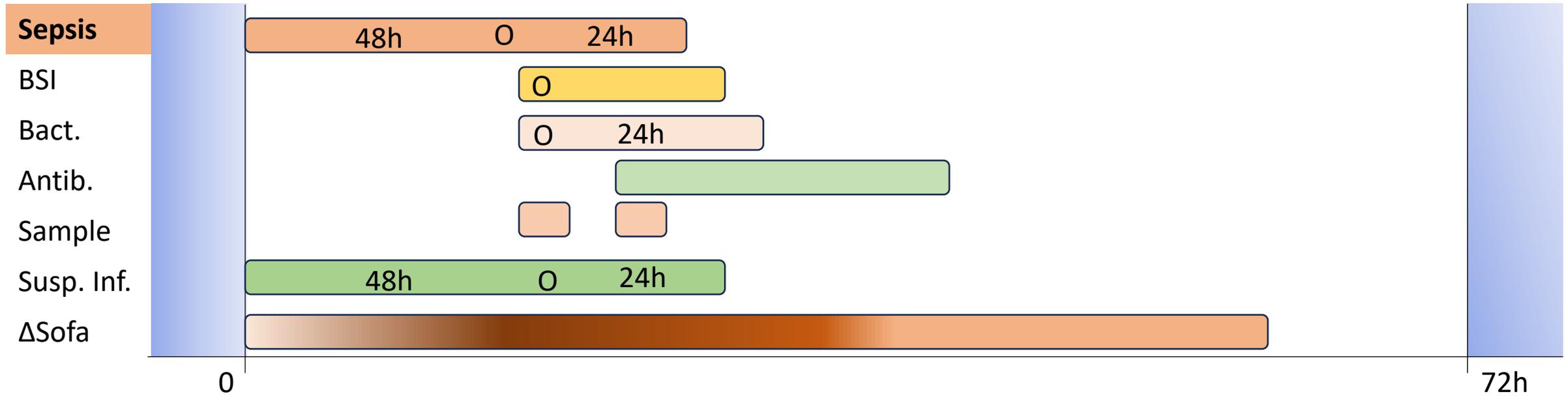
- Exclude patients with more than 85% missing values.











### Pre-processing

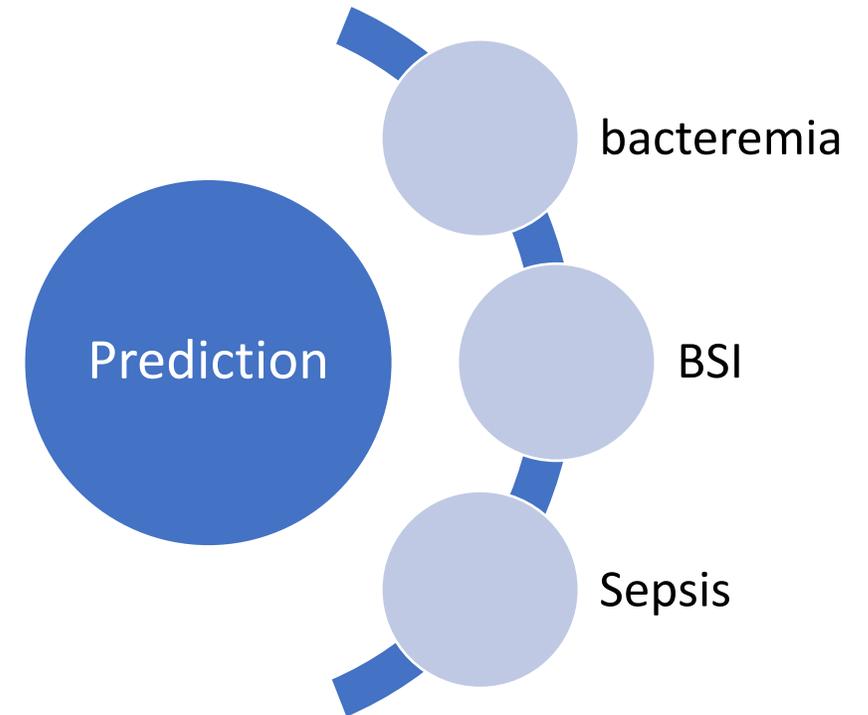
Each patient has been segmented into windows

- **Window Size:** Fixed at 6 hours
- **Look Back Period:** 6 hours
- **Look Ahead Period:** 1 hour

### Prediction Goal

The aim is to predict within a 6-hour look back and 1-hour look ahead window whether the patient will develop:

- Bacteriemia
- BSI
- Sepsis



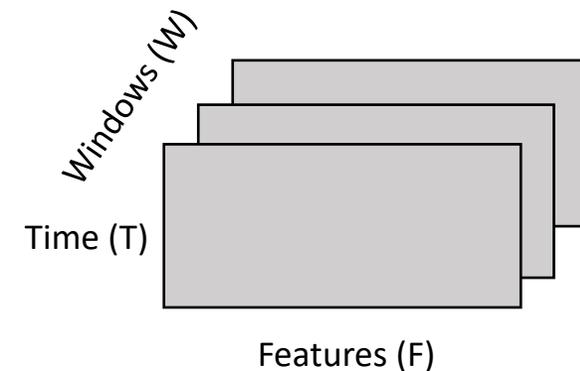
## Post Preprocessing

$$D_{vs}^t = \{(\mathbf{X}_p)\}_{p=1}^P, \text{ with } \mathbf{X}_p \in \mathbb{R}^{W \times F \times T}$$

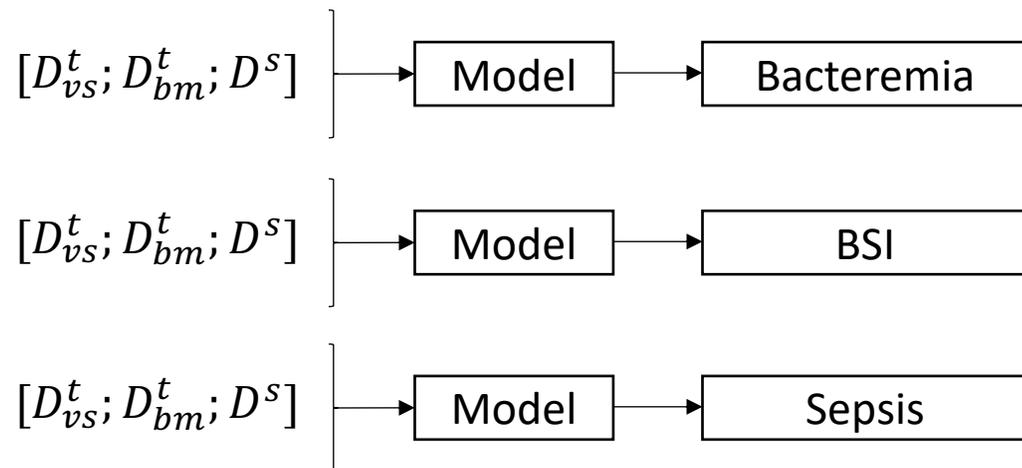
$$D_{bm}^t = \{(\mathbf{X}_p)\}_{p=1}^P, \text{ with } \mathbf{X}_p \in \mathbb{R}^{W \times F \times T}$$

$$D^s = \{(\mathbf{x}_p)\}_{p=1}^P, \text{ with } \mathbf{x}_p \in \mathbb{R}^{W \times F}$$

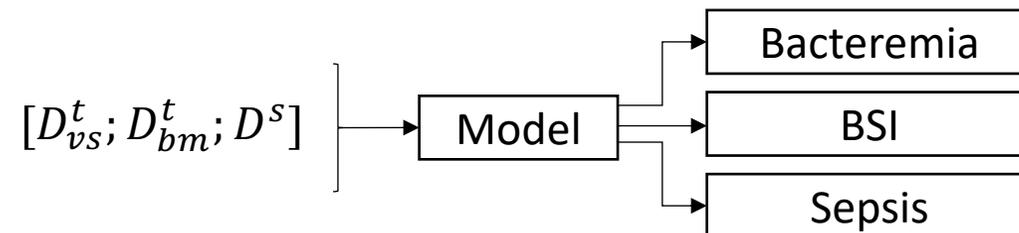
$$D_t = \{(\mathbf{Y}_p)\}_{p=1}^P, \text{ with } \mathbf{Y}_p \in \mathbb{R}^{W \times L \times T}, \text{ being } L = \{Bac, BSI, Sepsis\}$$



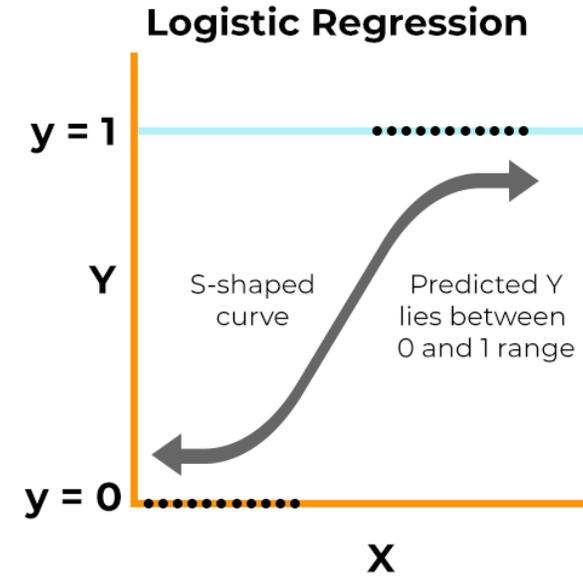
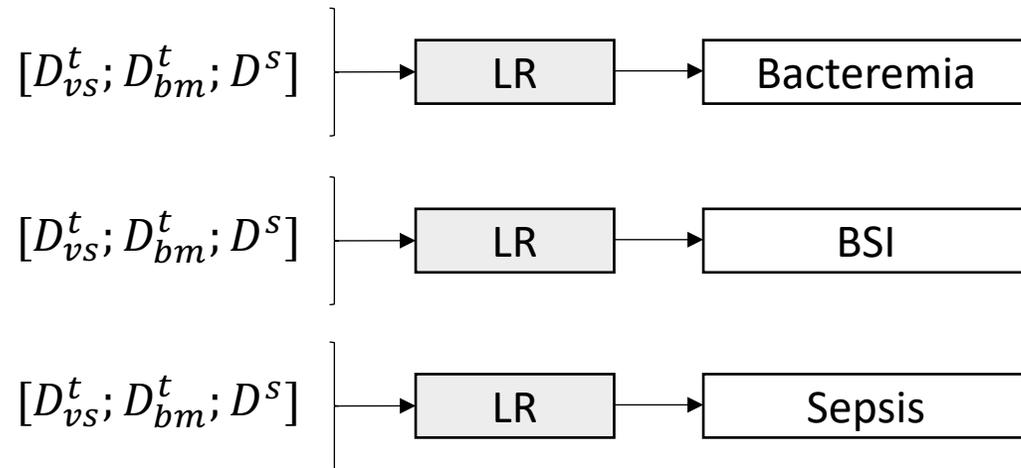
## Approach 1



## Approach 2



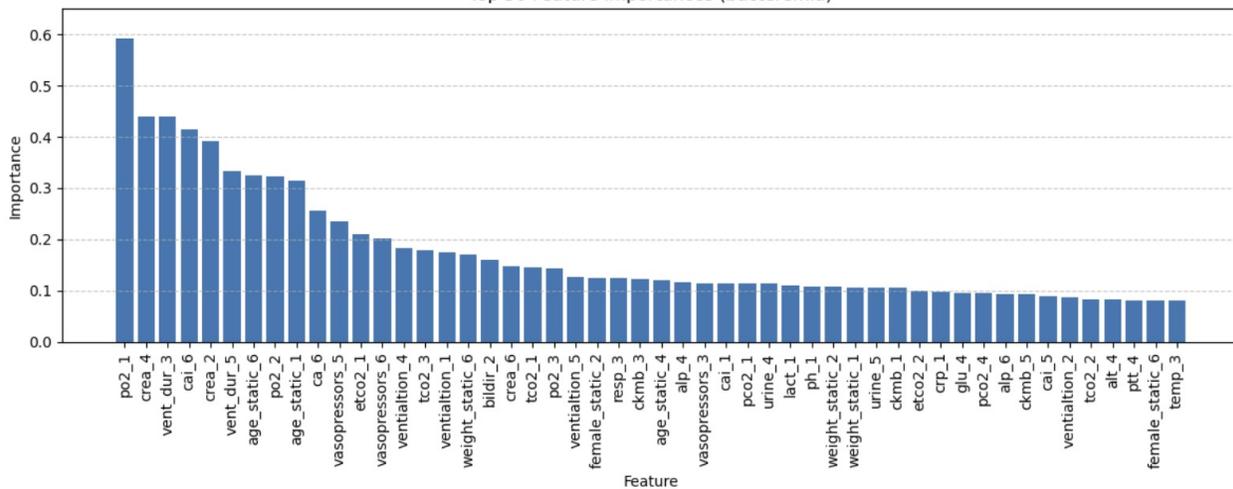
## Approach 1



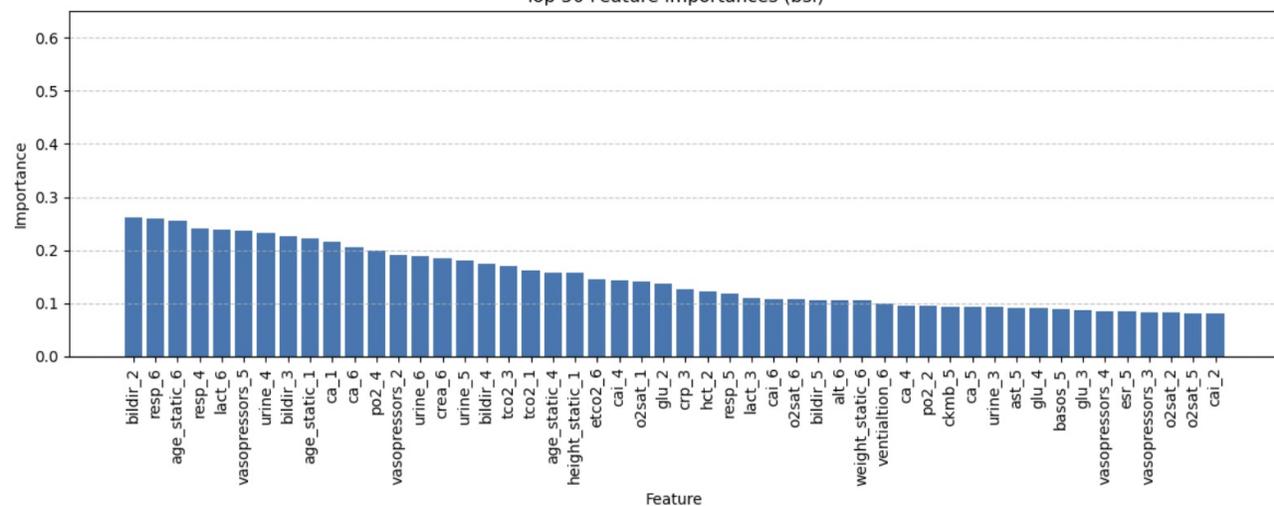
## Results

Bacteremia	BSI	Sepsis
Specificity: 74.6 +- 2.58	Specificity: 69.91 +- 2.65	Specificity: 70.15 +- 2.65
Sensitivity: 61.44 +- 6.88	Sensitivity: 48.39 +- 7.2	Sensitivity: 55.58 +- 11.45
AUC: 76.97 +- 2.5	AUC: 64.99 +- 1.37	AUC: 67.17 +- 4.22

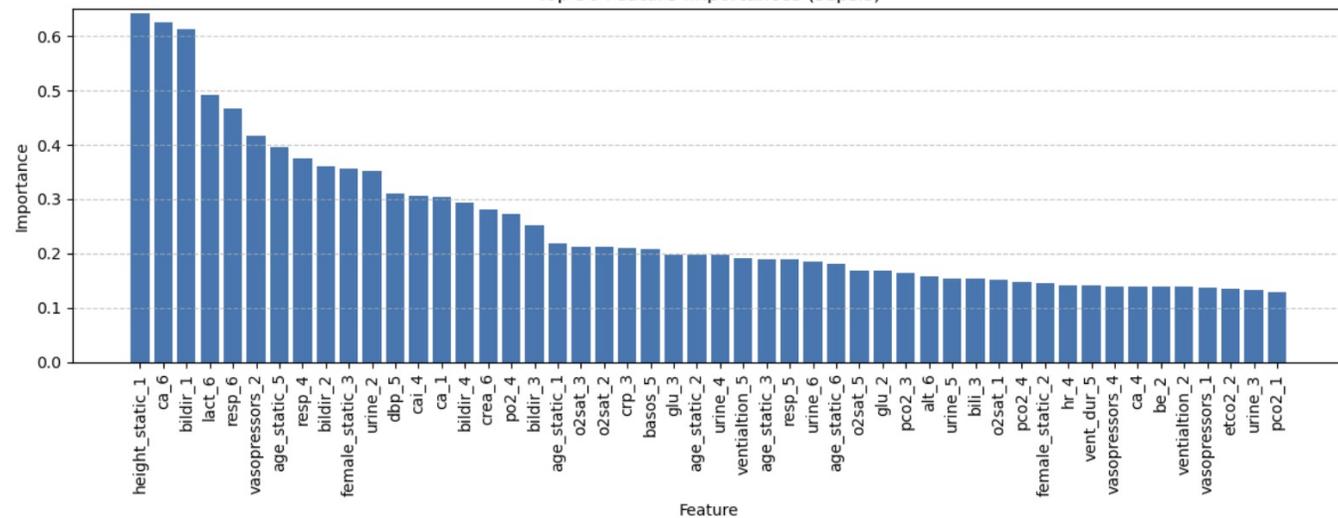
Top 50 Feature Importances (bacteremia)



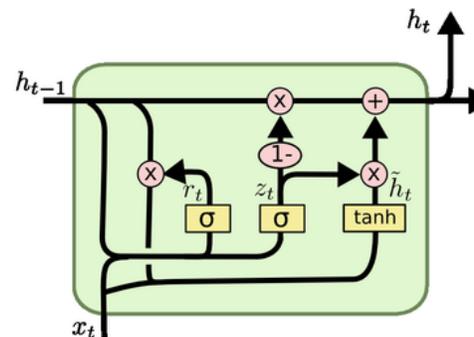
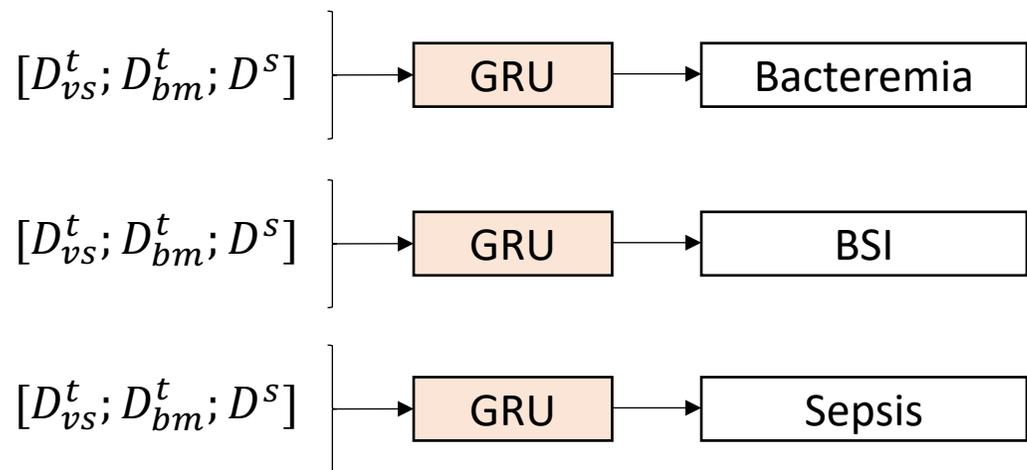
Top 50 Feature Importances (bsi)



Top 50 Feature Importances (sepsis)



## Approach 1



$$z_t = \sigma(W_z \cdot [h_{t-1}, x_t])$$

$$r_t = \sigma(W_r \cdot [h_{t-1}, x_t])$$

$$\tilde{h}_t = \tanh(W \cdot [r_t * h_{t-1}, x_t])$$

$$h_t = (1 - z_t) * h_{t-1} + z_t * \tilde{h}_t$$

## Results

Bacteremia

Running...

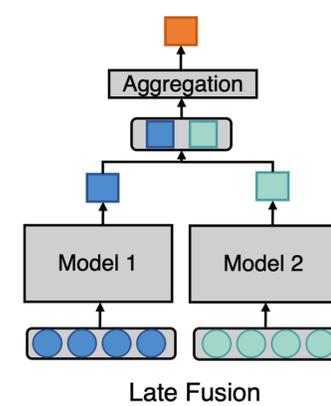
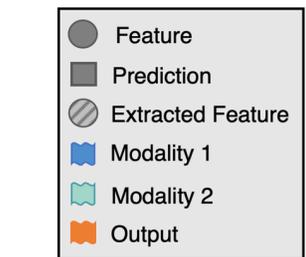
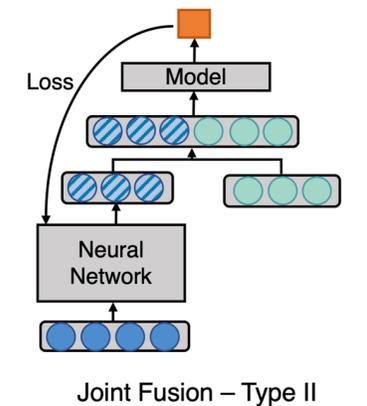
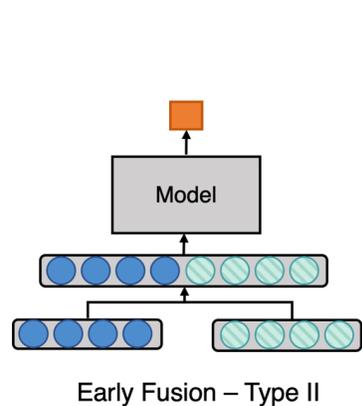
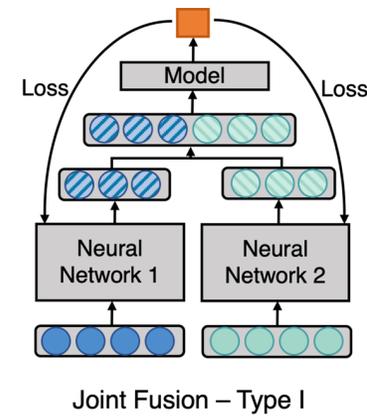
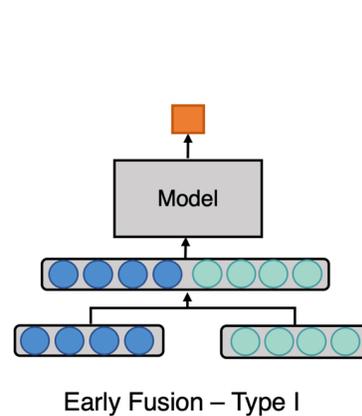
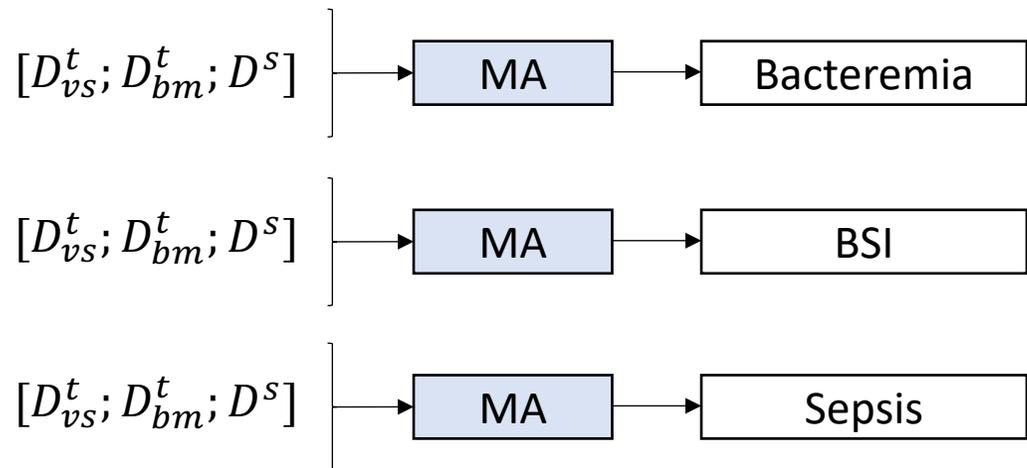
BSI

Running...

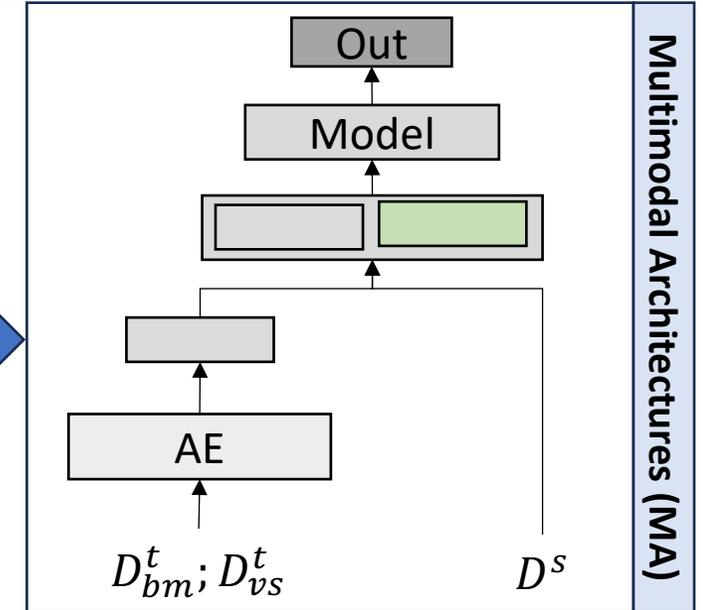
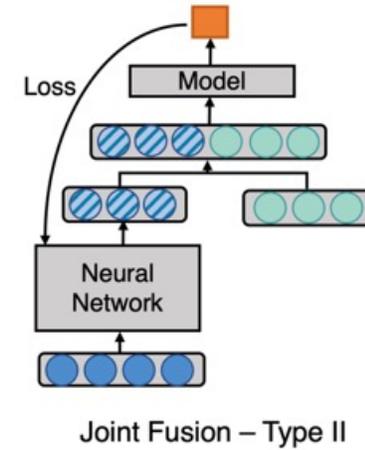
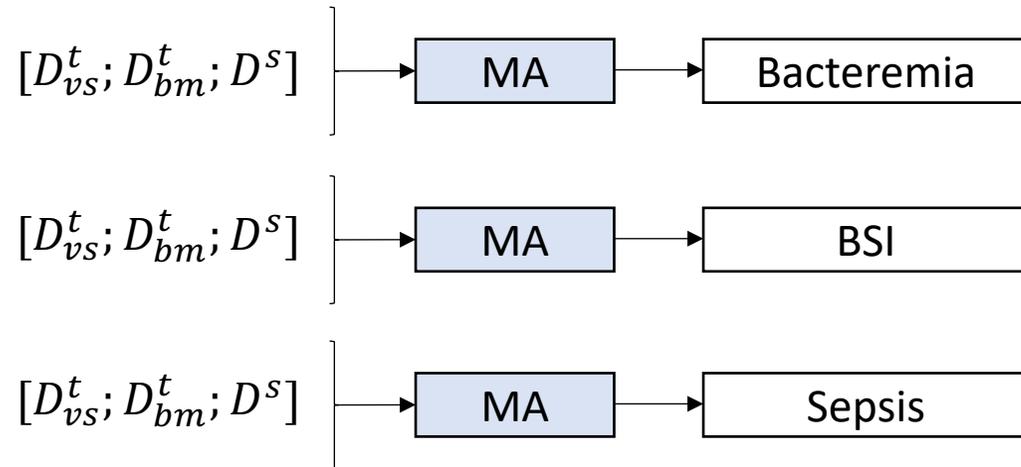
Sepsis

Running...

## Approach 1



## Approach 1



## Results

### Bacteremia

Specificity: 71.29 +- 3.27  
 Sensitivity: 64.26 +- 7.86  
 AUC: 75.67 +- 2.72

### BSI

Specificity: 69.67 +- 3.1  
 Sensitivity: 49.49 +- 7.98  
 AUC: 65.3 +- 1.92

### Sepsis

Specificity: 70.02 +- 3.39  
 Sensitivity: 56.93 +- 10.25  
 AUC: 68.34 +- 3.2

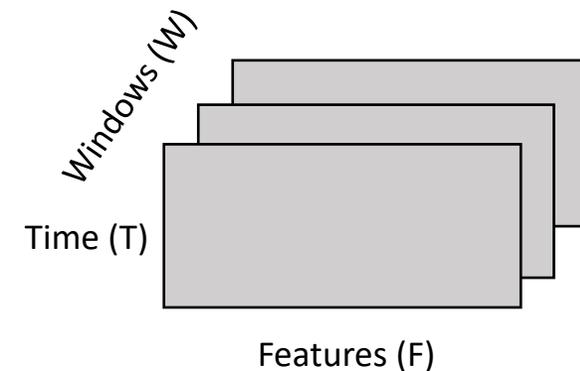
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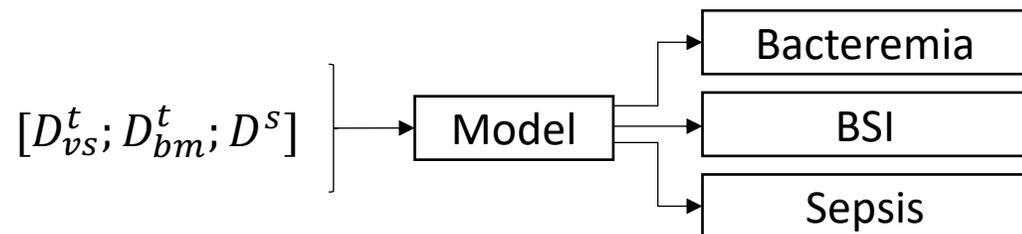
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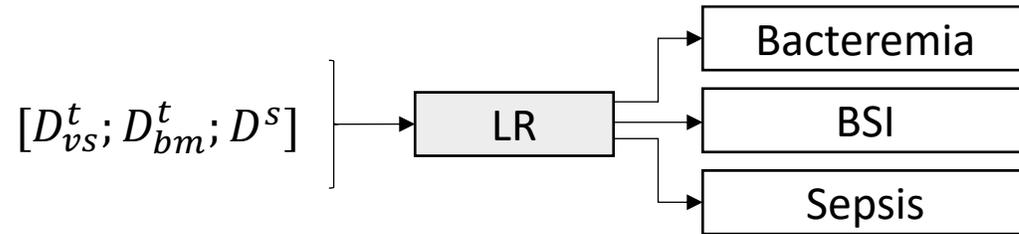
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## Approach 2

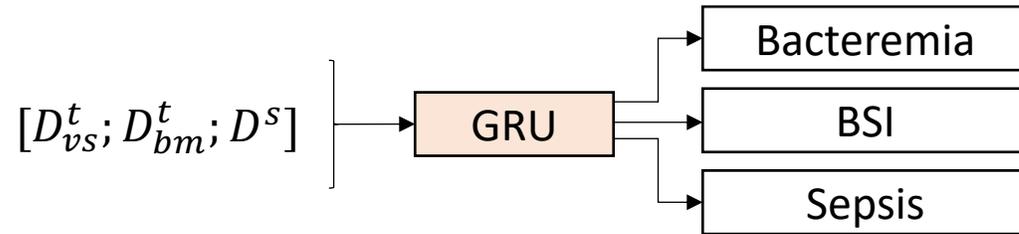


## Approach 2



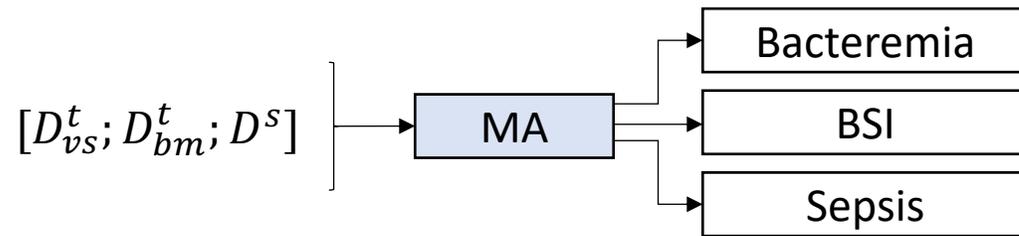
Results		
<div data-bbox="300 1005 621 1068" data-label="Text"> <p>Bacteremia</p> </div> <div data-bbox="384 1139 580 1182" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1098 1005 1419 1068" data-label="Text"> <p>BSI</p> </div> <div data-bbox="1166 1139 1363 1182" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1888 1005 2209 1068" data-label="Text"> <p>Sepsis</p> </div> <div data-bbox="1956 1139 2153 1182" data-label="Text"> <p>Running...</p> </div>

## Approach 2



Results		
<div data-bbox="295 1005 621 1071" data-label="Text"> <p>Bacteremia</p> </div> <div data-bbox="377 1135 580 1186" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1095 1005 1421 1071" data-label="Text"> <p>BSI</p> </div> <div data-bbox="1159 1135 1363 1186" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1880 1005 2206 1071" data-label="Text"> <p>Sepsis</p> </div> <div data-bbox="1944 1135 2147 1186" data-label="Text"> <p>Running...</p> </div>

## Approach 2



Results		
<div data-bbox="295 1005 621 1071" data-label="Text"> <p>Bacteremia</p> </div> <div data-bbox="377 1133 580 1185" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1093 999 1419 1065" data-label="Text"> <p>BSI</p> </div> <div data-bbox="1156 1130 1360 1182" data-label="Text"> <p>Running...</p> </div>	<div data-bbox="1877 999 2204 1065" data-label="Text"> <p>Sepsis</p> </div> <div data-bbox="1933 1136 2142 1188" data-label="Text"> <p>Running...</p> </div>

- Implement multitask learning as a third approach
- Predict outcomes at  $T$  time steps look forward
- Evaluate different window sizes and sliding intervals (look backward)
- Assess interpretable architectures to identify the most relevant variables for the prediction task

# THANK YOU FOR YOUR ATTENTION!

For any further doubt or suggestion

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## Biochemical markers

1. Albumin (alb) [g/dL]
2. Alkaline phosphatase (alp) [IU/L]
3. Alanine aminotransferase (alt) [IU/L]
4. Aspartate aminotransferase (ast) [IU/L]
5. Band form neutrophils (bnd) [%]
6. Base excess (be) [mEq/L]
7. Basophils (basos) [%]
8. Bicarbonate (bicar) [mEq/L]
9. Bilirubin direct (bildir) [mg/dL]
10. Blood urea nitrogen (bun) [mg/dL]
11. Calcium (ca) [mg/dL]
12. Carboxyhemoglobin (hbco) [-]
13. Chloride (cl) [mEq/L]
14. CO<sub>2</sub> partial pressure (pcO<sub>2</sub>) [mmHg]
15. Creatine kinase (ck) [IU/L]
16. Creatine kinase MB (ckmb) [ng/mL]
17. Creatinine (crea) [mg/dL]
18. C-reactive protein (crp) [mg/L]
19. Endtidal CO<sub>2</sub> (etcO<sub>2</sub>) [mmHg]
20. Eosinophils (eos) [%]

20. Eosinophils (eos) [%]
21. Erythrocyte distribution width (rdw) [%]
22. Erythrocyte sedimentation rate (esr) [mm/hr]
23. Fibrinogen (fgn) [mg/dL]
24. Glucose (glu) [mg/dL]
25. Hematocrit (hct) [%]
26. Hemoglobin (hgb) [g/dL]
27. Lactate (lact) [mmol/L]
28. Lymphocytes (lymph) [%]
29. Magnesium (mg) [mg/dL]
30. Mean cell hemoglobin (mch) [pg]
31. Mean corpuscular hemoglobin concentration (mchc) [%]
32. Mean corpuscular volume (mcv) [fL]
33. Methemoglobin (methb) [%]
34. Neutrophils (neut) [%]
35. O<sub>2</sub> partial pressure (po<sub>2</sub>) [mmHg]
36. Partial thromboplastin time (ptt) [sec]
37. Phosphate (phos) [mg/dL]
38. pH of blood (ph) [-]
39. Platelet count (plt) [K/ul]
40. Potassium (k) [mEq/L]

40. Potassium (k) [mEq/L]
41. Prothrombin time (inrpt) [-]
42. Prothrombine time (pt) [K/ul]
43. Red blood cell count (rbc) [m/uL]
44. Sodium (na) [mEq/L]
45. Total bilirubin (bili) [mg/dL]
46. Total CO<sub>2</sub> (tco<sub>2</sub>) [mEq/L]
47. Troponin I (tri) [ng/mL]
48. Troponin T (tnt) [ng/mL]
49. White blood cell count (wbc) [K/ul]
50. Respiratory rate (resp) [insp/min]
51. Endtidal CO<sub>2</sub> (etco<sub>2</sub>) [mmHg]
52. O<sub>2</sub> partial pressure (po<sub>2</sub>) [mmHg]
53. CO<sub>2</sub> partial pressure (pco<sub>2</sub>) [mmHg]
54. Urine output (urine) [mL]
55. Duration of ventilation (vent\_dur) [sec]

**Vital signs**

- 1.mean arterial pressure (map) [mmHg]
- 2.heart rate (hr) [bpm]
- 3.temperature (temp) [C]
- 4.systolic blood pressure (sbp) [mmHg]
- 5.diastolic blood pressure (dbp) [mmHg]
- 6.oxygen saturation (o2sat) [%]
- 7.calcium ionized (cai) [mmol/L]

**Statics**

- 1.age [years]
- 2.sex [-]
- 3.weight [kg]
- 4.height [cm]

Biochemical markers				Vital signs		Statics
Albumin ( <b>alb</b> ) [g/dL]	alkaline phosphatase ( <b>alp</b> ) [IU/L]	alanine aminotransferase ( <b>alt</b> ) [IU/L]	aspartate aminotransferase ( <b>ast</b> ) [IU/L]	band form neutrophils ( <b>bnd</b> ) [%]	mean arterial pressure ( <b>map</b> ) [mmHg]	age [years]
base excess ( <b>be</b> ) [mEq/L]	Basophils ( <b>basos</b> ) [%]	Bicarbonate ( <b>bicar</b> ) [mEq/L]	bilirubin direct ( <b>bil_dir</b> ) [mg/dL]	blood urea nitrogen ( <b>bun</b> ) [mg/dL]	heart rate ( <b>hr</b> ) [bpm]	sex []
Calcium ( <b>ca</b> ) [mg/dL]	Carboxyhemoglobin ( <b>hbco</b> ) [-]	Chloride ( <b>cl</b> ) [mEq/L]	CO2 partial pressure ( <b>pcO2</b> ) [mmHg]	creatine kinase ( <b>ck</b> ) [IU/L]	temperature ( <b>temp</b> ) [C]	weight [kg]
creatine kinase MB ( <b>ckmb</b> ) [ng/mL]	Creatinine ( <b>crea</b> ) [mg/dL]	C-reactive protein ( <b>crp</b> ) [mg/L]	endtidal CO2 ( <b>etcO2</b> ) [mmHg]	Eosinophils ( <b>eos</b> ) [%]	systolic blood pressure ( <b>sbp</b> ) [mmHg]	height [cm]
erythrocyte distribution width ( <b>rdw</b> ) [%]	erythrocyte sedimentation rate ( <b>esr</b> ) [mm/hr]	Fibrinogen ( <b>fgn</b> ) [mg/dL]	Glucose ( <b>glu</b> ) [mg/dL]	Hematocrit ( <b>hct</b> ) [%]	diastolic blood pressure ( <b>dbp</b> ) [mmHg]	
Hemoglobin ( <b>hgb</b> ) [g/dL]	Lactate ( <b>lact</b> ) [mmol/L]	Lymphocytes ( <b>lymph</b> ) [%]	Magnesium ( <b>mg</b> ) [mg/dL]	mean cell hemoglobin ( <b>mch</b> ) [pg]	oxygen saturation ( <b>o2sat</b> ) [%]	
mean corpuscular hemoglobin concentration ( <b>mchc</b> ) [%]	mean corpuscular volume ( <b>mcv</b> ) [fL]	Methemoglobin ( <b>methb</b> ) [%]	Neutrophils ( <b>neut</b> ) [%]	O2 partial pressure ( <b>po2</b> ) [mmHg]	calcium ionized ( <b>cai</b> ) []	
partial thromboplastin time ( <b>ptt</b> ) [sec]	Phosphate ( <b>phos</b> ) [mg/dL]	pH of blood ( <b>ph</b> ) [-]	platelet count ( <b>plt</b> ) [K/uL]	Potassium ( <b>k</b> ) [mEq/L]		
prothrombin time ( <b>inrpt</b> ) [-]	prothrombine time ( <b>pt</b> ) [K/uL]	red blood cell count ( <b>rbc</b> ) [m/uL]	respiratory rate ( <b>resp</b> ) [insp/min]	Sodium ( <b>na</b> ) [mEq/L]		
total bilirubin ( <b>bili</b> ) [mg/dL]	total CO2 ( <b>tco2</b> ) [mEq/L]	troponin I ( <b>tri</b> ) [ng/mL]	troponin t ( <b>tnt</b> ) [ng/mL]	white blood cell count ( <b>wbc</b> ) [K/uL]		