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Mortality Predicting Calculators used in ICU

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ABSTRACT

Diseases are constantly present, like a persistent companion, and their effects serve as a reminder of their presence. We all have to address medical needs in some way or other. Disease can be classified as general illness, severe or chronic disease. General illness needs normal attention though severe or chronic disease needs much attention because critically ill patients have a potential risk of death. It happens that chronic disease may lead to patient admission in ICU.

The stay of the patient into ICU and likelihood of mortality can be predicted in many ways ranging from manual to automated prediction. Manual prediction requires experienced doctors though if supplied with right parameters, the same can be done by system i.e., application-based procedure.

Predicting the likelihood of mortality has been a cornerstone of medical decision-making for centuries. With advancements in healthcare technology and data science, we can now leverage sophisticated models to predict a patient's likelihood of mortality, length of ICU stay and usage of mechanical ventilator with increased accuracy.

Different Medical calculators like SOFA, SAPS, APACHE, MPM, GCS etc. are in use to predict the likelihood of mortality these days. Their accuracy can be calculated before manually and later automatically and now AI-assisted.

Keywords: ICU, Predicting Calculators, Mortality.

1. INTRODUCTION

An intensive care unit (ICU) in the hospital are for those patients facing severe illness or injuries. Patients in intensive care unit (ICU) are mostly critically ill, so they need special attention because of presenting high mortality risk compared with other departments in the hospital. Death is the primary clinical outcome while readmission and prolonged length of stay (LOS) are common clinical outcomes that indicates patient's health conditions. So, in most of the cases patients require special attention with

medical equipment such as mechanical ventilation to ensure that the body functions normally and because of this they need to be observed continuously and thoroughly.

For thorough observation Intensive care units (ICUs) depend on scoring systems such as APACHE (Acute physiology and chronic health evaluation) and SAPS (Simplified acute physiology score) to understand patient severity of illness and basically these scoring systems predict the short-term outcomes, including fatality. These continuous evolving tools are designed for mainstream ICU use. Scoring systems have a very important position in ICU for gauging the severity of the disease and

anticipating the likelihood of mortality, Prolonged length of stays etc. These tools like APACHE and SAPS are updated regularly for common ICU use so that they can provide useful observation for patient management.

So, our objective is to identify the principal characteristics that can predict the likelihood of mortality. By using these principal characteristics, we will try to build a simplified and effective scoring system for predicting the likelihood of mortality

Family of Evaluation Assessment Scoring System Which are Present in ICU While Taking Care of a Patient

The main prognostic models for assessing the overall severity of illness in critically ill adults are Acute Physiology and Chronic Health Evaluation (APACHE), Sequential organ failure assessment (SOFA), Simplified Acute Physiology Score (SAPS), and Mortality Probability Model (MPM). Simplified Acute Physiology Score and Mortality Probability Model have been updated to their third versions and Acute Physiology and Chronic Health Evaluation to its fourth version as shown in Figure 1.

Related Work

William A. Knaus is recognized as the founder of the APACHE (Acute Physiology and Chronic Health Evaluation) system, which he developed in collaboration with Jack E. Zimmerman, Douglas P. Wagner, and Elizabeth A. Draper. APACHE was established in 1978 during Knaus's tenure as an attending physician at George Washington University

Hospital [9]. The APACHE scoring system was developed to address critical needs in managing patients in intensive care units (ICUs). One of its primary objectives was to standardize the assessment of severity, as there was no uniform method to quantify the illness severity of critically ill patients. By incorporating objective physiological data, APACHE provided a systematic approach that ensured consistent evaluations across different healthcare settings [6,9].

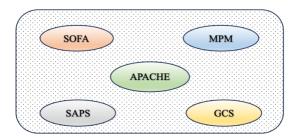


Figure 1: Family of Evaluation Assessment Scoring System

Additionally, APACHE played a crucial role in quality assurance and outcome monitoring. By utilizing a scoring system, healthcare providers could track patient outcomes and assess the effectiveness of treatment strategies, leading to improved clinical practices [6,9]. Another significant feature of APACHE was its ability to facilitate risk stratification, allowing clinicians to predict mortality rates and tailor treatment plans accordingly. This ensured that high-risk patients received the appropriate level of care [6]. Moreover, the system supported research and comparative analysis by enabling researchers to evaluate outcomes across different ICUs and studies. Such comparisons were essential for identifying best practices and advancing patient care in critical care environments [6,9]. Lastly, APACHE was designed to integrate clinical knowledge with modern statistical techniques, enhancing its precision and reliability. This combination allowed clinicians to make data-driven decisions rather than relying solely on anecdotal evidence, thereby improving patient outcomes and healthcare efficiency [9].

Acute Physiology Score (APS) [10]

The APS is based on 12 physiological variables assessed within the first 24 hours of ICU admission. These variables include:

Body Temperature: Measures hypothermia or hyperthermia.

Mean Arterial Pressure (MAP): Evaluates blood pressure stability.

Heart Rate: Monitors pulse rate.

Respiratory Rate: Counts breaths per minute.

Oxygenation: Assessed through arterial oxygen tension (PaO2).

Serum Sodium Level: Measures sodium concentration.

Serum Potassium Level: Evaluates potassium concentration.

Creatinine Level: Indicates kidney function, with higher levels contributing more points.

Hematocrit: Measures the proportion of blood volume occupied by red blood cells.

White Blood Cell Count: Assesses immune response.

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Glasgow Coma Scale (GCS): Evaluates consciousness level; lower scores indicate more severe impairment.

Blood pH: Assesses acidity or alkalinity.

Age Points

Additional points are assigned based on the patient's age [3,10]:

0 points for ages < 44 years

2 points for ages 45–54 years

3 points for ages 55-64 years

5 points for ages 65-74 years

6 points for ages \geq 75 years

Chronic Health Status

Points are added based on the patient's chronic health conditions before admission:

Elective postoperative patients with severe organ insufficiency or immunocompromised status receive additional points.

Nonoperative patients or emergency postoperative patients with similar conditions receive even more points.

The total APACHE score is calculated by summing the acute physiology score, age points, and chronic health status points. The overall score can range from 0 to 71, with higher scores indicating greater severity of illness and increased risk of mortality [3,13].

Why did you develop the APACHE system?

In the 1970s, when William began developing the APACHE system, Diagnosis-Related Groups (DRGs) were newly introduced and focused on healthcare financing. DRGs classified ICU patients using simple tests like blood lactate levels with fixed thresholds, which were ineffective for continuous measures. Bryan Jennett later developed the Glasgow Coma Scale, which was successful but limited to head injuries and emergencies. The innovation of using ICU patients' physiological data to create a comprehensive severity measure was a breakthrough and was unexpectedly well-received [4,10]. Over time, the APACHE system evolved to incorporate many variables; however, the available technology at the time could not efficiently process large-scale computations. Simple equations used in APACHE would take an entire weekend to compute, leading to strategic discussions on whether to halt and focus solely on updating the database. It became clear that classification systems do not improve over time unless continuously updated with current data. The release of APACHE II demonstrated improvements in critical care outcomes across various areas, highlighting that outdated databases fail to reflect contemporary results [4,10,11].

During this period, technology was advancing rapidly—computers became faster, and there was increasing optimism about the future of data collection and scoring systems. The vision was for computers to be easily accessible for medical consultations, much like how Google's algorithms function today. However, decades after APACHE II's release, progress in healthcare technology, particularly regarding interoperability and modern computing, has been slow. Despite ambitious goals, expected advancements have not been fully realized. While the APACHE system continues to evolve, its latest version, APACHE IV, with an improved algorithm and database, is still not widely adopted compared to APACHE II [4,10]. In hindsight, if the slow development of healthcare technology had been anticipated, a stronger focus on updating the APACHE II database rather than creating newer versions might have been preferred. Although APACHE IV is more sensitive and capable, automatic data integration remains a challenge. Thus, when utilizing APACHE, it is essential to ensure that the database—whether proprietary or external—incorporates contemporary patient data, as the relationship between scores and patient outcomes changes over time. Using the same scoring system is acceptable, but maintaining up-to-date data is critical for accuracy and relevance [4,10].

Evolution of the APACHE Score Evaluation System

The Acute Physiology and Chronic Health Evaluation (APACHE) scoring system has evolved significantly since its inception in 1981 to enhance the accuracy of predicting patient mortality in intensive care units (ICUs). The first version, APACHE I, consisted of 34 variables and involved complex calculations for risk assessment. To simplify the process, APACHE II was introduced in 1985 with 12 variables, incorporating the Glasgow Coma Scale (GCS) and chronic health status. Further advancements led to APACHE III (1991), which expanded to 20 variables, adding comorbidities, diagnostic categories, and refined prediction models. Finally, APACHE IV (2006) became the most comprehensive model, integrating physiological, demographic, and hospital-specific data for improved ICU mortality predictions. This evolution, as shown in Figure 2. reflects continuous efforts to enhance clinical decision-making and patient outcome assessments.

Transition from APACHE I to APACHE II

The transition from APACHE I (1981) to APACHE II (1985) involved simplifying the model by removing several parameters as shown in Figure 3, while enhancing mortality prediction accuracy. APACHE I originally utilized 34 physiological variables mentioned in Figure 3, making data collection time-consuming and impractical for routine ICU use. Additionally, some variables contributed minimal predictive power, rendering them redundant. To address these issues, APACHE II reduced the number of key physiological variables to 12, selecting those with the most significant impact on patient outcomes. This reduction streamlined the scoring process, making it easier and faster, which encouraged widespread adoption [6,10].

Transition from APACHE II to APACHE III

Further refinements were made in the transition from APACHE II (1985) to APACHE III (1991) to enhance accuracy, specificity, and usability in ICU mortality prediction. While APACHE II considered 12 physiological variables as mentioned in Figure 3, it sometimes lacked specificity for certain ICU patients and did not fully differentiate between various ICU admission types, such as medical, surgical, or trauma cases. APACHE III addressed these limitations by expanding the number of physiological parameters for improved prediction. Additionally, it incorporated the patient's admission source (e.g., Emergency Room, Operating Room) to enhance ICU-specific risk assessment, allowing for more detailed patient evaluation and outcome prediction [7,10].

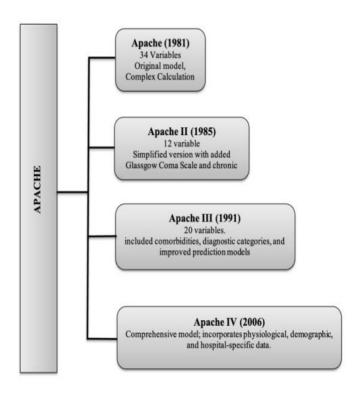


Figure 2: Apache Version Progression and key features.

Transition from APACHE III to APACHE IV

The transition from APACHE III (1991) to APACHE IV (2006) was necessary to enhance ICU mortality prediction accuracy, adjust for medical advancements, and improve ICU benchmarking. One of the key limitations of APACHE III was its reliance on outdated mortality prediction models from the early 1990s, which did not account for significant advancements in ICU treatments. Additionally, some variables used in APACHE III as mentioned below in Figure 4 became less predictive due to changes in clinical practices. To address these issues, APACHE IV introduced updated statistical models for mortality prediction, utilizing data from over 100,000 ICU patients. These improvements allowed APACHE IV to better adjust for modern ICU treatments, including advanced ventilation strategies and new sepsis protocols, thereby enhancing its predictive accuracy and clinical relevance [8,10].

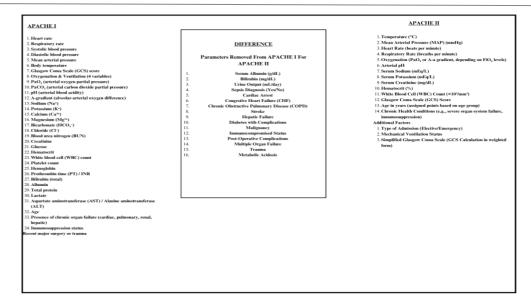


Figure 3: Parameters of Apache I and Apache II



Figure 4: Parameters of Apache III and Apache IV

2. APACHE SCORE CALCULATION [6,8,10]

Apache II Score = [APS (12 physiological parameter) + AGE POINTS + CHRONIC HEALTH POINT]

This is taken in Initial 24 hours of the patient admission

Temperature

In this we Measures hypothermia or hyperthermia 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 1.

Table 1: Apache II Score on The Basis of Temperature

POINTS:	+4	+3	+1	0	+1	+2	+3	+4
TEMP:	>41	39-40.9	38.5-38.9	36-38.4	34-35.9	32-33.9	30-31.9	<29.9

Mean Arterial Pressure (mm Hg)

In this we Evaluates blood pressure stability of a patient. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 2.

Table 2: Apache II Score on The Basis of MAP

POINTS:	+4	+3	+2	0	+2	+3	+4
MAP:	>160	130-159	110-129	70-109	50-69	30-31.9	<49

Heart Rate

In this we Monitors pulse rate of a patient. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 3

Table 3: Apache-2 Score on The Basis of Heart Rate

POINTS:	+4	+3	+2	0	+2	+3	+4
HEART RATE:	>180	190-179	110-139	70-109	55-69	40-54	<39

Respiratory Rate

In this we count breaths per minute a patient. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 4.

Table 4: Apache II Score on The Basis of Respiratory Rate

POINTS:	+4	+3	+1	0	+1	+2	+4
RESPIRATORY RATE:	>50	35-49	25-34	10-24	10-11	6-9	<5

Oxygenation if FIO2 > 50% then A-ADO2 / PIO2 < 50PAO2

In this we Assessed through arterial oxygen tension (PaO2). 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 5.

Table 5: Apache II Score on The Basis of Oxygenation

POINTS:	+4	+3	+2	0	+1	+3	+4
OXYGENATION:	>500	350-499	200-349	400-70	61-70	55-60	<55

To Calculate the AaDO2: AaDO2 ((FIO2 * 713) - PaO2) - PaCO2

Arterial PH

Measure of the acidity of the blood in the arteries. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 6.

Table 6: Apache II Score on The Basis of Arterial PH

ARTERIAL PH:	>7.7	7.6-7.69	7.5-7.59	7.33- 7.49	7.25-7.32	7.15-7.24	<7.15
				,			

Bicarbonate in Venous Blood

Measure of the amount of the bicarbonate in the blood. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 7.

Table 7: Apache II Score on The Basis of Bicarbonate in Venous Blood

POINTS:	+4	+3	+1	0	+2	+3	+4
BVB:	> 52	41-51.9	32-40.9	22-31.9	18-21.9	15-17.9	< 15

If ABG is not available we can send venous blood

Potassium (MEQ / L)

Potassium levels are measured in millie equivalents per liter (mEq/L). A normal potassium level for adults is between 3.5 and 5.2 mEq/L. Low-level termed as hypokalemia and high-level potassium termed as hyperkalemia. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 8.

Table 8: Apache II Score on The Basis of Potassium

POINTS:	+4	+3	+1	0	+1	+2	+4
POTASSIUM:	>7	6-6.9	5.5-5.9	3.5-5.4	3-3.4	2.5-2.9	<2.5

Sodium (MEQ / L)

Sodium levels are measured in milliequivalents per liter (mEq/L). A level below 135 mEq/L is called hyponatremia. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 9.

Table 9: Apache II Score on The Basis of Sodium

POINTS:	+4	+3	+2	+1	0	+2	+3	+4
SODIUM:	>180	160-170	155-159	150-154	130-149	120-129	111-119	<110

Serum Creatinine (MG / DL)

In this Indicates kidney function, with higher levels contributing more points. 0 being the normal score and +4 being the highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 10.

Table 10: Apache II Score on The Basis of Serum Creatinine

POINTS:	+4	+3	+2	0	+2
SERUM CREATININE:	>3.5	2-3.4	1.5-1.9	0.6-1.4	<0.6

Hematocrit (%)

In this we measure the proportion of blood volume occupied by red blood cells. 0 being the normal score and +4 being the

highest and risky score of a patient in both the cases either it is increasing or decreasing as mentioned below in Table 11.

Table 11: Apache II Score on The Basis of Hematocrit

POINTS:	+4	+3	+2	+1	0	+2	+4
Hematocrit (%):	>60	59.1- 59.9	50-59	46-49.9	30-45.9	20-29.9	<20

TLC (IN 1000)

(TOTAL LEUKOCYTE COUNT)

GCS (Glasgow Coma Scale):

It is a neurological scale designed to assess a patient's level of consciousness following a traumatic brain injury or other condition affecting the brain.

The GCS is a score based on three components:

Eye Opening €

Verbal Response (V)

Motor Response(M)

The total GCS score is the sum of these three components.

Ranging from 3(Deep unconsciousness) to 15 (fully alert person) as mentioned below in Table 12.

In this we Evaluates consciousness level; lower scores indicate more severe impairment.

Table 12: Apache-2 Score on The Basis of GCS

GCS SCORE	SEVERITY INDICATOR
15	0
14	1
13	2
12	3
11	4
10	5
9	6
8	7
7	8
6	9
5	10
4	11
3	12

3. AGE CHRONIC HEALTH POINTS:

- 2 Points if an elective postoperative patient with immunocompromised or a history of severe organ insufficiency as mentioned below in Table 13.
- 5 Points for non-operative patient emergency postoperative patients with immunocompromised or severe organ insufficiency as mentioned below in Table 13.

Table 13: Apache-2 Score on The Basis of Age

POINTS:	0	2	3	5	6
AGE:	<44	45-54	55-64	65-74	>75

Mortality Prediction:

If the overall score of a patient is according to the range given in the table 14 then the chances of mortality in percentage written in percentage respectively.

Table 14: Apache-2 Score on The Basis of Mortality Prediction

SCORE RANGE	MORTALITY PERCENTAGE (%)
0-4	4%
5-9	8%
10-14	15%
15-19	25%
20-24	40%
25-29	55%
30-24	75%
>34	85%

Patient 1 Input Parameters in APACHE Scoring System

The APACHE (Acute Physiology and Chronic Health Evaluation) scoring system utilizes multiple physiological parameters to assess the severity of a patient's illness and predict ICU mortality. As shown in Figure 5: Patient with Different Input Parameters of APACHE Scoring System, key variables such as temperature (38.2°C), heart rate (110 bpm), respiratory rate (22 breaths/min), mean arterial pressure (85 mmHg), oxygenation level (250 mmHg), and Glasgow Coma Scale (GCS = 14) are used in the calculation. Additionally, serum potassium (4.2 mEq/L), serum sodium (140 mEq/L), hematocrit (42%), and serum creatinine (1.2 mg/dL) play a crucial role in evaluating organ function and overall patient status. These input parameters help clinicians determine the severity of illness and guide critical care decisions.

Patient with different input parameters of APACHE scoring system

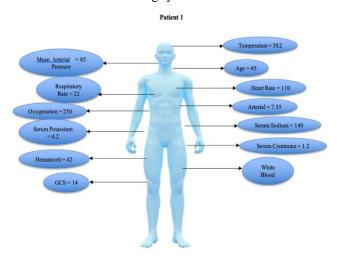


Figure 5: Patient-1 Stats of 45 year old male with pneumonia and chronic heart disease

Total APACHE score is 9, predicted Low ICU mortality risk (~5-10%)

Patient 2 Parameters in APACHE Scoring System

The APACHE (Acute Physiology and Chronic Health Evaluation) scoring system considers multiple physiological factors to assess the severity of illness in critically ill patients. As illustrated in Figure: 6 Patient with Different Input Parameters of APACHE Scoring System, the patient's vital signs include temperature (36°C), heart rate (130 bpm), respiratory rate (30 breaths/min), and mean arterial pressure

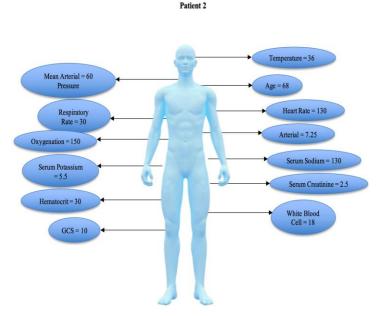


Figure 6: Patient-2 (A 68 year old patient with sepsis and no chronic health condition)

Total APACHE score is 50, Very high ICU mortality risk (>80%)

(60 mmHg). Additionally, blood chemistry parameters such as serum potassium (5.5 mEq/L), serum sodium (130 mEq/L), serum creatinine (2.5 mg/dL), and hematocrit (30%) play a role in determining organ function. The Glasgow Coma Scale (GCS = 10) and oxygenation level (150 mmHg) further contribute to assessing neurological status and respiratory function. These parameters collectively aid in determining the severity of illness and guiding ICU management.

Table 15: Apache-2 Score of Patient-1 And Patient-2

	PATIENT 1		PATIENT 2		
CONDITION:	A 45-year-old male with pneumonia and chronic heart disease		A 68-year-old male with sepsis and no chronic health conditions		
PARAMETERS					
Parameters	Value (Patient one) APACHE-2 POINT		Value (Patient two)	APACHE-2 POINT	
Age:	45	3	68	9	
Temperature:	38.2	0	36.0	1	
Mean Arterial Pressure:	85 0		60	10	
Heart Rate (bpm):	110 2		130	5	
Respiratory Rate (bpm):	22	1	30	4	
Oxygenation:	250 2		150	5	
Arterial PH:	7.35	0	7.25	3	

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Serum Sodium (mEq/L):	140	0	130	2
Serum Creatinine (mg/dL):	1 2	0	2.5	4
Hematocrit (%):	42	0	30	2
White Blood Cell Count:	12	1	18	2
Glasgow Coma Scale (GCS):	14	1	1 0	6
Total APACHE II Score:	Low Risk	9	Very High Risk	50

Overview

Patient-1, with an APACHE II score of 9, is at low risk with a favorable prognosis. Patient-2, with a score of 50, is at very high risk due to advanced age, severe physiological disturbances, and sepsis. Immediate intensive care is required for Patient-2. The APACHE II scoring system accurately differentiates patient severity an assists in clinical decision-making.

How APACHE IV Calculates Mortality Risk

APACHE IV (2006) is the most advanced version of the APACHE scoring system, using 142 variables to provide the most accurate ICU mortality predictions. It uses a logistic regression model that incorporates multiple patient factors.

Steps in APACHE IV Mortality Calculation[12]:

Acute Physiology Score (APS) Calculation

17 key physiological variables are assessed within the first 24 hours of ICU admission.

Each variable is assigned a weighted score based on how much it deviates from normal values.

Example variables:

Heart Rate

Respiratory Rate

Blood Pressure

Glasgow Coma Scale (GCS)

Serum Sodium, Potassium, Creatinine

Arterial pH, PaO₂, PaCO₂

Higher deviations = Higher APS Score (worse condition).

Chronic Health Condition Adjustment

APACHE IV considers whether the patient has a severe chronic illness (e.g., liver failure, COPD, immunosuppression).

Patients with pre-existing chronic diseases receive additional risk points.

ICU Admission Diagnosis (115 Categories)

APACHE IV includes 115 different ICU admission diagnoses, each with its own mortality risk weight.

Examples:

Acute Myocardial Infarction (AMI)

Sepsis or Septic Shock

Pneumonia or Acute Respiratory Distress Syndrome (ARDS)

Acute Kidney Injury (AKI)

More severe conditions = Higher base mortality risk.

Time Before ICU Admission & Length of Stay Before ICU

Unlike APACHE III, APACHE IV considers how long a patient was in the hospital before ICU admission.

If a patient was in the hospital for several days before ICU transfer, their risk increases.

Mechanical Ventilation Status

If a patient requires mechanical ventilation, their mortality risk is significantly higher.

The impact of ventilation is adjusted based on the patient's other conditions.

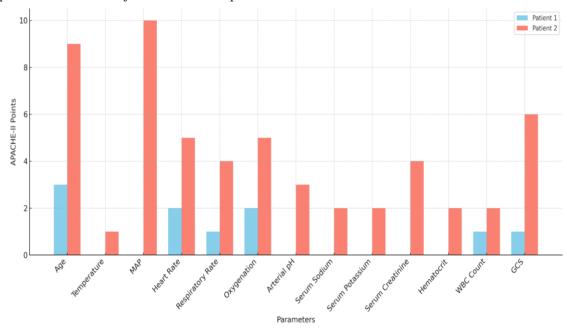


Figure 7: Apache-2 Score of Patient-1 And Patient-2.

Final APACHE IV Mortality Calculation (Logistic Regression Model) as mentioned below in Figure 8. Once all factors are collected, a logistic regression formula is used:

$$ext{Mortality Probability} = rac{e^{(b_0 + b_1 X_1 + b_2 X_2 + ... + b_n X_n)}}{1 + e^{(b_0 + b_1 X_1 + b_2 X_2 + ... + b_n X_n)}}$$

Figure 8: Formula of Mortality Probability

Where:

 $X1,X2,...XnX_1,X_2,...XnX_1,X2,...Xn = Patient-specific factors (APS score, chronic disease, ICU admission reason, etc.)$

b0,b1,b2,...bnb 0, b 1, b 2, ... b nb0,b1,b2,...bn = Predefined coefficients from APACHE IV research.

The output is the predicted probability of in-hospital mortality (% chance of death).

The output is the predicted probability of in-hospital mortality (% chance of death) as mentioned below in Table 16.

Table 16: APACHE IV Mortality Score Calculation

Variable	Patient Value	Score Contribution
Age	70 years	+5 points
Heart Rate	120 bpm	+4 points
Mean Arterial Pressure	60 mmHg	6 points

Glasgow Coma Scale (GCS)	10	+10 points
Serum Creatinine	2.5 mg/dL	+7 points
Mechanical Ventilation	Yes	+8 points
Primary ICU Diagnosis	Septic Shock	High Risk
Total Score (APS + Adjustments)	87	
Mortality Risk (%)	52%	(Based On APACHE IV Formula)

Table 17: Literature Review

S.No.	Author/Publication/Year	Research Aim	Methodology	Significant Finding	Limitation or Recommendation
1	W. A. Knaus, Archives of Surgery, 2002	Discuss the development of the APACHE system from 1978 to 2001	Review of system evolution and application	APACHE impacted ICU care quality and outcome prediction	Retrospective review; lacks new empirical data
2	Wong & Knaus, Can J Anaesth, 1991	Evaluate APACHE system accuracy in outcome prediction	Comparative review with other ICU systems	APACHE II was reliable for ICU mortality prediction	Recommends ongoing updates and validation
3	Rapsang & Shyam, IJCCM, 2014	Summarize ICU scoring systems including APACHE	Narrative review and compendium	Strengths and weaknesses of ICU scores outlined	Contextual use based on local ICU settings advised
4	W. Knaus, MDCalc, accessed 2023	Provide an online APACHE II calculator	Web-based implementation from algorithm	Enables standardized bedside score calculations	Manual input required; not real- time
5	Dimensions of Critical Care Nursing, 1986	Introduce APACHE II to nursing professionals	Educational summary article	Raised awareness of APACHE II in nursing community	Lacks technical and validation data
6	Knaus et al., Crit Care Med, 1985	Present the APACHE II scoring system	Empirical multicenter study	Established APACHE II as a validated model	Needs further population validation
7	Knaus et al., Chest, 1991	Introduce and validate APACHE III	Development using large ICU dataset	Improved predictive accuracy over APACHE II	Complex and proprietary system
8	Zimmerman et al., Crit Care Med, 2006	Develop the APACHE IV model	Multicenter dataset with updated regression modeling	Enhanced predictive performance over APACHE III	Needs regular updates; hospital- specific performance varies

9	Knaus et al., Crit Care Med, 1981	Introduce original APACHE system	Developed from physiological ICU data	First model using physiological variables for ICU classification	Limited by available data and technology of the time
10	Knaus et al., Chest, 1991 (Duplicate Listing)	Discuss APACHE III system development	Empirical study with extensive data	Confirmed robust hospital mortality prediction	Duplicated in citation; reflects importance but may confuse citation tracking
11	Wang Z et al., Heart Lung, 2023	Develop an explainable ML model to predict mechanical ventilation duration in ARDS patients	Machine learning with SHAP explainability on ICU dataset	ML model accurately predicted ventilation duration with interpretability	Requires external validation; limited to ARDS cases
12	Aczon MD et al., Pediatr Crit Care Med, 2021	Continuous mortality prediction in PICU using RNN	Recurrent neural network on pediatric ICU dataset	Continuous model outperformed traditional static models	Single-center data limits generalizability
13	Kim JH et al., J Clin Med, 2021	Predict 30-day mortality in mechanically ventilated patients using ML	Comparative analysis of ML algorithms (RF, SVM, XGBoost)	XGBoost achieved highest predictive accuracy	Small dataset; lacked prospective validation

4. FUTURE SCOPE

The future of the APACHE scoring system lies in integrating AI and machine learning to improve predictive accuracy and automate real-time data collection. IoT-enabled ICU devices can further streamline patient monitoring by feeding live physiological data into the system, reducing manual errors and improving decision-making. The integration with electronic health records (EHRs) will enable automated scoring and remote access for healthcare professionals. Additionally, expanding APACHE to emergency rooms and general wards can help in the early identification of high-risk patients. Future versions, such as APACHE V, may incorporate genetic markers and AI-driven risk assessments to further refine predictions and improve patient outcomes.

5. CONCLUSION

The APACHE scoring system has revolutionized critical care by providing a standardized method for assessing the severity of illness and predicting patient outcomes in ICUs. Over the years, its evolution has led to improved accuracy in mortality prediction, aiding healthcare professionals in making informed decisions. By integrating physiological parameters, chronic health conditions, and demographic factors, APACHE helps optimize treatment plans, resource allocation, and patient monitoring. With advancements in AI, automation, and real-time data integration, its predictive capabilities will continue to improve, ensuring more precise risk assessment. As technology progresses, APACHE will remain a crucial tool in ICU management, adapting to new medical innovations and enhancing overall patient care

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