

Clinical risk scores for predicting stroke-associated pneumonia: a systematic review

Amit K Kishore¹, Andy Vail², Benjamin D Bray³, Angel Chamorro⁴, Mario Di Napoli⁵, Lalit Kalra⁶, Peter Langhorne⁷, Joan Montaner^{8,9}, Christine Roffe¹⁰, Anthony G Rudd¹¹, Pippa J Tyrrell¹, Diederik van de Beek¹², Mark Woodhead¹³, Andreas Meisel¹⁴, Craig J Smith¹

¹Stroke and Vascular Research Centre, University of Manchester,
Institute of Cardiovascular Sciences, Greater Manchester
Comprehensive Stroke Centre Salford Royal Foundation Trust, UK

²Centre for Biostatistics, University of Manchester, Salford Royal
Foundation Trust, UK

³King's College, London, UK

⁴Comprehensive Stroke Centre, Department of Neuroscience, Hospital
Clinic, University of Barcelona, Barcelona, Spain;

⁵Neurological Service, San Camillo de' Lellis General Hospital, Rieti,
Italy

⁶Clinical Neurosciences, King's College Hospital NHS Foundation Trust
London, UK

⁷Institute of Cardiovascular and Medical Sciences, University of
Glasgow, Glasgow Royal Infirmary, Glasgow, UK

⁸Laboratorio de Investigación Neurovascular, Unidad Neurovascular,
Servicio de Neurología Hospital Vall d' Hebron, Barcelona

⁹IBIS Stroke Programme, Hospital Virgen del Rocío, Sevilla, Spain

¹⁰Keele University Institute for Science and Technology in Medicine,
Guy Hilton Research Centre, Stoke-on-Trent, UK

¹¹Department of Health and Social Care, Kings College, London, UK

¹²Department of Neurology, Centre for Infection and Immunity
Amsterdam (CINIMA), Academic Medical Centre, University of
Amsterdam, Amsterdam, Netherlands

¹³Faculty of Medical and Human Sciences, University of Manchester &
Department of Respiratory Medicine, Central Manchester University
Hospitals NHS Foundation Trust, Manchester Academic Health Science
Centre, Manchester, UK

¹⁴NeuroCure Clinical Research Centre, Centre for Stroke Research
Berlin, Department of Neurology Charité Universitaetsmedizin Berlin,
Germany

Corresponding author:

Dr Amit K Kishore, Greater Manchester Neurosciences Centre, Salford
Royal NHS Foundation Trust, Stott lane, Salford, M6 8HD, UK.

E-mail: Amit.Kishore@manchester.ac.uk

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1 **Purpose**

2 Several risk stratification scores for predicting stroke-associated pneumonia (SAP) have been
3 derived. We aimed to evaluate the performance and clinical usefulness of such scores for
4 predicting SAP.

5 **Method**

6 A systematic literature review was undertaken in accordance with the Preferred Reporting
7 Items for Systematic Reviews and Meta-Analyses (PRISMA) statement, with application of
8 the Quality Assessment of Diagnostic Accuracy (QUADAS)-2 tool. Published studies of
9 hospitalised adults with ischaemic stroke, intracerebral haemorrhage, or both, which derived
10 and validated an integer-based clinical risk score, or externally validated an existing score to
11 predict occurrence of SAP, were considered and independently screened for inclusion by two
12 reviewers.

13 **Findings**

14 We identified 9 scores, from 8 derivation cohorts. Age was a component of all scores, and the
15 NIHSS score in all except one. 6 scores were internally validated and 5 scores were
16 externally validated. The A2DS2 score (Age, Atrial fibrillation, Dysphagia, Severity
17 [NIHSS], Sex) was the most externally validated in 8 independent cohorts. Performance
18 measures were reported for 8 scores. Discrimination tended to be more variable in the
19 external validation cohorts (C statistic 0.67-0.83) than the derivation cohorts (C statistic 0.74-
20 0.85).

21 **Discussion**

22 Overall, discrimination and calibration were similar between the different scores. No study
23 evaluated influence on clinical decision-making or prognosis.

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25

1 **Conclusion**

2 The clinical prediction scores varied in their simplicity of use and were comparable in
3 performance. Utility of such scores for preventive intervention trials and in clinical practice
4 remains uncertain and requires further study.

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1 **Introduction**

2 Stroke-associated pneumonia (SAP) is a common and serious complication after acute stroke,
3 associated with increased length of hospital stay, mortality and worse outcomes in
4 survivors.¹⁻⁶ A recent systematic review reported that SAP occurs in 14.3%, although the
5 frequency varies widely depending on definition of SAP and patient characteristics.⁷ Several
6 features of SAP such as varied clinical manifestation,⁸ uncertain role of blood biomarkers⁹
7 and absence of definitive diagnostic criteria make it challenging to diagnose in clinical
8 practice. As a first step, the recently convened Pneumonia In Stroke ConsEnsuS (PISCES)
9 group proposed operational diagnostic criteria for SAP based on Center of Disease Control
10 criteria (CDC).⁹

11
12 Numerous baseline clinical factors such as age, dysphagia, severity of stroke, low conscious
13 level, type and location of stroke may pre-dispose individuals to SAP.¹⁰⁻¹³ Predictive risk
14 models derived using these routinely available variables may help in identifying patients at an
15 increased risk of pneumonia for targeted preventive measures and may also provide
16 opportunities for novel interventions for monitoring or therapy. However, clinical prediction
17 scores have several potential weaknesses such as differences in derivation, inconsistent
18 external validation and complexity thus making choice of score and application to clinical
19 practice challenging.^{14,15} We therefore undertook a systematic review to identify scores used
20 in predicting risk of SAP, with the aim of evaluating performance, usability and utility for
21 clinical practice and research.

22

1 **Methods**

2 A systematic literature review was undertaken in accordance with the Preferred Reporting
3 Items for Systematic Reviews and Meta-Analyses (PRISMA) statement.¹⁶

4

5 *Data sources and searches*

6 Searches were undertaken in MEDLINE (1946-15th September 2015) and EMBASE (1947-
7 15th September 2015) using pre-defined search criteria and terms (**Online only Table I**).

8 Hand searching of reference lists for additional eligible articles was also carried out, and the
9 PISCES group were invited to provide any other potentially eligible articles.

10

11 *Study selection*

12 Published studies (English and Non-English) of hospitalised adults with ischaemic stroke,
13 intracerebral haemorrhage (ICH), or both, which derived and validated an integer-based
14 clinical risk score, or externally validated an existing score to predict occurrence of
15 pneumonia after stroke, were independently screened for eligibility by two reviewers (AKK
16 and CJS), using the study title and abstract (**Online Only Table II**). Lead or corresponding
17 authors of studies under consideration were contacted by e-mail to resolve any issues relating
18 to assessment of eligibility or data extraction. Discrepancies relating to eligibility or data
19 extraction were resolved by discussion between the same two study investigators.

20

21 *Data Extraction*

22 Data were independently extracted by two reviewers (AKK and CJS) and included study
23 design, clinical environment, country, stroke subtype (ischaemic or ICH), mean age, mean
24 National Institutes of Health Stroke Scale (NIHSS) score, components of score and

1 weighting, measures of discrimination and calibration, co-morbidities, criteria used in
2 diagnosis of pneumonia and proportion of patients diagnosed with pneumonia.

3

4 *Assessment of quality: risk of bias and applicability*

5 Quality was assessed in terms of risk of bias and concerns regarding applicability, using the
6 Quality Assessment of Diagnostic Accuracy (QUADAS)-2 tool.¹⁷ In brief, judgement of
7 applicability and risk of bias are made across 4 domains using relevant signalling questions;
8 patient selection, index risk score, reference standard (diagnosis of SAP) and flow and
9 timing. The QUADAS-2 tool was applied for each score within the identified validation
10 cohorts by two reviewers (AKK and AV) independently.

11

12 *Risk score performance*

13 For the discriminative ability of scores, we extracted information on the area under the
14 receiver operating characteristic curve (AUROC) or C-statistic, their 95% confidence
15 intervals, and the p-value for comparison between models if they were available for both the
16 derivation and validation cohorts. C-statistic values range from 0.5 (no discrimination, no
17 better than chance) to 1.0 (perfect discrimination). A C-statistic of 0.7-0.8 indicates modest
18 discriminative ability, while a C-statistic greater than 0.8 indicates good discriminative
19 ability. To describe score calibration, we similarly extracted data on the difference between
20 the observed and predicted rates of pneumonia if available, as well as the ‘goodness of fit’
21 statistic and p-value of the corresponding test statistic. Calibration was considered better
22 when the observed to predicted ratio was closer to 1.

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1 *Clinical usefulness*

2 We noted the complexity of application and use at the bedside, whether prediction scores
3 incorporated categories of risk-stratification (usability), and whether scores had been used to
4 evaluate clinical management or clinician behaviors (utility). We also evaluated the
5 generalisability of each prediction model by determining whether it had been externally
6 validated in an independent patient population, either in the original or subsequent
7 publication.

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1 **Findings**

2 *Search results*

3 The electronic search yielded 2493 publications. After screening, exclusion of duplicates and
4 applying eligibility criteria, 46 full texts and abstracts were reviewed (**Figure 1**). No
5 additional articles were identified through hand-searching of major stroke journals or by the
6 PISCES group. 12 fully published studies were finally considered eligible for inclusion.<sup>11, 18-
7 28</sup>

9 *Clinical risk scores for predicting SAP*

10 14 separate cohorts were identified which had either derived or validated risk scores for
11 predicting SAP (**Online only Table III**). 9 clinical risk-scores (**Table 1**) were identified
12 from 8 derivation cohorts. The risk scores identified were: The Pneumonia Score,¹⁸
13 Veteran's Health Administration cohort score,¹¹ A2DS2 (Age, Atrial fibrillation, Dysphagia,
14 Severity [NIHSS], Sex) score,¹⁹ PANTHERIS (Preventive Antibacterial Therapy in Acute
15 Ischaemic Stroke) score,²⁰ AIS-APS (Acute Ischaemic Stroke-Associated Pneumonia
16 Score),²¹ ICH-APS (Intracerebral Haemorrhage-Associated Pneumonia Score) A,²² ICH-
17 APS score B,²² Pneumonia (PNA) prediction score²⁵ and ISAN (Prestroke Independence,
18 Sex, Age, NIHSS) score.²⁷ The number of components of each score varied between 3 and
19 11. Age was a component of all scores, and the NIHSS score in all except the PANTHERIS
20 score.²⁰ Dysphagia (55%) and pre-morbid modified Rankin Scale (mRS, 36%) were other
21 commonly used variables. 2 scores also included routine laboratory evaluation^{20,21} and 2
22 scores incorporated neuroimaging features.²² The majority of the scores were derived only in
23 ischaemic stroke patients (56%). The ISAN and the PNA prediction score included ICH
24 patients in the derivation cohorts,^{25, 27} while the ICH-APS scores²² were developed
25 exclusively in ICH.

1 *Study and participant characteristics*

2 Characteristics of the derivation and validation cohorts are summarised in **Online only Table**

3 **III**. Median age was 71y (range 61y-76y) and median NIHSS was 5 (range 4-13). All studies

4 adequately described selection of study sample. Except one study which was prospective,²⁶

5 all were retrospective evaluations of existing prospective cohorts. Of the 14 separate

6 derivation or validation cohorts, 8 (61%) were multicentre or national stroke registries and 6

7 (39%) were single-centre hospital-based stroke registries. The majority of the 14 studies

8 (80%) evaluated only acute ischaemic stroke. Definition and ascertainment of risk factors for

9 model derivation was varied and often limited by availability of data, particularly in existing

10 national registries. For example, dysphagia assessment was not described among several

11 studies,^{21, 22, 25} and pre- stroke disability was described in different ways.^{21,22,27} Some

12 studies did not record pre-existing disability.^{18,25} Diagnostic approach to pneumonia varied

13 between cohorts; clinician reported diagnosis of pneumonia (36%) and the CDC criteria for

14 pneumonia (36%) were the most commonly used approaches. The other methods include

15 adhoc objective criteria (14%) and Chinese Consensus criteria (14%).

16

17 *Quality assessment*

18 Overall, risk of bias and concerns regarding applicability were judged as generally low

19 (Online only **Table IV**). In some validation cohorts, risk of bias was judged as high based on

20 patient selection (exclusions based on incomplete baseline data²⁷, or selected higher-risk

21 cohort²⁰), reference standard (non-standardised criteria for diagnosis of SAP^{11,19,27,28}) and

22 flow and timing (verification bias, related to differences in applying the same reference

23 standard by the study group^{11,19,27,28}).

24

25

1 *Performance and validation of the risk-scores*

2 Performance and validation of the clinical risk scores is summarized in **Table 2** and **3**. C-
3 statistics ranged from 0.74-0.85 in the derivation cohorts, indicating a degree of
4 discriminative performance that varied from modest to good. Only one score, the Pneumonia
5 score,¹⁸ **did** not report performance. Calibration was reported in 6 models using varying
6 goodness of fit models or net reclassification index.

7

8 6 of the risk-scores were validated internally through split samples (**Table 2**). All reported the
9 C-statistic for the internal validation cohort, which ranged from 0.73 to 0.88, with 5 models
10 reporting calibration metric. 5 of the 9 scores were validated externally (**Table 3**), with C-
11 statistic ranging from 0.68 to 0.83. The A2DS2 score¹⁹ has been evaluated most extensively,
12 in the largest derivation sample (n=15,335), and in 8 separate external validation cohorts. The
13 A2DS2 score performed consistently across these cohorts (C statistic 0.73 to 0.84), with good
14 calibration.

15

16 *Clinical usefulness*

17 The risk-scores varied in their complexity and ease of use (**Table 1**), although most scores
18 incorporated clinical variables readily available at baseline. 2 scores require admission
19 laboratory variables^{20,21} and one of the scores developed exclusively for ICH requires
20 quantitative measurement of hematoma volume.²² Several of the scores were stratified into
21 integer-based risk categories^{11,19,21,22,27} (e.g. low, moderate, high risk), facilitating usability
22 by clinicians. The role of implementing the models as prediction rules in terms of risk
23 stratification, decision-making or improved patient outcomes was not evaluated for any of the
24 scores.

25

1 **Discussion**

2 An ideal risk score for predicting SAP would incorporate variables readily available at stroke
3 presentation, be quick to apply, provide meaningful risk categories with performance
4 acceptable to the particular application (and to clinicians), and have impact on clinical
5 decision-making and clinical outcomes. In this systematic review, we identified nine clinical
6 risk scores for predicting SAP, and assessed their performance metrics, clinical usability and
7 utility. We sought to identify whether any of the scores could be applied for use in clinical
8 care or research.

9

10 The scores varied considerably in their complexity, component variables, derivation cohort
11 characteristics, approach to defining SAP, ease of application, consistency of external
12 validation, and performance evaluation. Substantial heterogeneity between the studies was
13 therefore anticipated and precluded meta-analyses. As previously acknowledged,²⁷ the
14 prevalence of SAP varied between the cohorts, most likely related to underlying differences
15 in patient characteristics and definitions used,¹ potentially contributing to outcome reporting
16 bias. Several of the scores were derived from relatively small single centre cohorts^{18, 20, and 25}
17 limiting their generalisability. As all of the scores were derived using retrospective analyses
18 of registry-based studies, model-building was limited by the baseline characteristics recorded
19 in the different cohorts. Therefore, potentially important baseline characteristics (e.g.
20 smoking, medication, chronic lung disease), medications (e.g. statin therapy or beta-
21 blockade), laboratory variables (e.g. leukocyte count or C-reactive protein [CRP]) or
22 interventions (e.g. mechanical intervention, type of swallow screen), which may have
23 influenced SAP risk, were not available in the majority of the derivation cohorts.

24

1 For the studies reporting performance metrics, the discriminative ability and calibration of the
2 scores ranged from moderate to good. However, several of the scores have not yet undergone
3 external validation to our knowledge.^{20,22,25} Some of the scores performed similarly in the
4 external validation and derivation cohorts,^{19,21,27} despite differences in patient characteristics,
5 supporting generalisability. Importantly, the majority of the validation studies were unable to
6 compare the performance of more than one score concurrently due to limitations imposed by
7 data routinely collected in the registry-based cohorts. One study compared four scores
8 concurrently,²¹ and found no material difference in the performance metrics of the four
9 scores (Pneumonia score, VHA score, AIS-APS, A2DS2) tested. Most scores were derived
10 only in ischaemic stroke cohorts, although 2 scores with comparable performance were
11 available for ICH. The ISAN and A2DS2 were evaluated in both ischaemic stroke and ICH,
12 and performance metrics tended to be superior in ischaemic stroke rather than ICH, most
13 likely due to ceiling effects.²⁷ The only scores derived exclusively for ICH (ICH-APS A and
14 B) are less practical to apply, requiring baseline imaging parameters, and have not been
15 externally validated to date.²² Considering the high-rate of early neurological deterioration
16 and conflicting risk of death after ICH, the ISAN, A2DS2 and ICH-APS scores each
17 performed better, and comparably, in sensitivity analyses stratifying for survival beyond 48-
18 72 h after ICH.^{22,27}

19
20 The role of clinical risk-scores for predicting SAP in clinical care or research remains
21 uncertain. None of the studies investigated utility in terms of clinician behaviours (for
22 example, the time taken to administer the risk scores) or impact analysis on clinical
23 outcomes. The current levels of sensitivity and specificity for given cut-offs on the scores^{19,}
24 ^{21, 22, 24} may be unacceptable to clinicians, although this may depend on the particular
25 application of the score. For example, for a cut-off of ≥ 4 on the A2DS2 score, sensitivity is

1 91% but specificity is 57%.¹⁹ This means that only 9% of actual SAP cases are not identified
2 as high-risk (false negative rate), yet 43% of the patients who do not get SAP are incorrectly
3 identified as being at high-risk (false positive rate). For a safe, inexpensive and well-tolerated
4 intervention to prevent SAP (e.g. enhanced monitoring or oral hygiene protocol) this extent
5 of exposure to unnecessary interventions may be acceptable. However, for more expensive
6 and complex preventive interventions with adverse effects, which are challenging to
7 administer, then such low specificity may make clinical trials impractical and more difficult
8 to justify.

9

10 Further large, multi-centre prospective studies of consecutive patients, with adjudicated
11 diagnosis of SAP using standardised and validated criteria are required to evaluate
12 comparative performance and utility of the available scores. Refining the existing scores,
13 including the addition of laboratory biomarkers such as CRP to improve performance,²⁶
14 warrants further consideration. Finally, evaluating clinical utility of the scores is an essential
15 step to determine effects on clinician behaviours, impact on clinical decision-making, clinical
16 outcomes and feasibility of implementation.

17

18 **Conclusion**

19 We identified several clinical risk scores for predicting SAP which varied in their simplicity
20 and consistency of validation. When recorded, performance metrics were comparable
21 between scores, and no single score consistently performed better than others. However,
22 interpretation was limited by heterogeneity and some risk of bias. The utility of risk scores
23 for predicting SAP remains uncertain and requires further study in prospective cohorts with
24 standardised criteria for definition of SAP.

25

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1 **Table 1: Components of clinical risk scores for predicting stroke-associated pneumonia**

	ISAN	A2DS2	AIS-APS	PANTHERIS	VHA cohort	Pneumonia score	PNA prediction score	ICH-APS-A	ICH-APS-B
Age	✓	✓	✓	✓	✓	✓	✓	✓	✓
NIHSS	✓	✓	✓		✓	✓	✓	✓	✓
GCS			✓	✓				✓	
Sex	✓	✓				✓			
Dysphagia		✓			✓	✓		✓	✓
Mechanical ventilation						✓			
Dysphasia			✓						
OCSF subtype			✓						
“Found-down”					✓				
Increase in systolic BP > 200mmHg				✓					
Comorbidities									
Pre-stroke dependence (mRS)	✓		✓					✓	✓
Atrial fibrillation		✓	✓						
Congestive cardiac failure			✓						
COPD			✓					✓	✓
Current smoking			✓					✓	✓
Excess alcohol consumption								✓	✓
Previous pneumonia					✓				
Diabetes							✓		
Laboratory									
Blood glucose (mmol/l)			✓						
WBC count /µl				✓					
Radiology									
Infratentorial location								✓	✓
Extension into ventricles								✓	
Haematoma volume									✓

2 *ISAN indicates Independence Prestroke, Sex, Age, National Institutes of Health Stroke Scale; A2DS2, Age, Atrial Fibrillation, Dysphagia*
3 *Sex, Severity; AIS-APS, Acute Ischaemic Stroke-Associated Pneumonia Score; PANTHERIS, Preventive Antibacterial Therapy in Acute*
4 *Ischaemic Stroke; VHA, Veterans Health Administration; PNA, Pneumonia Prediction; ICH-APS, Intracerebral haemorrhage-Associated*
5 *Pneumonia Score; NIHSS: National Institutes of Health Stroke Scale; GCS: Glasgow coma scale; OCSF: Oxfordshire community stroke*
6 *project; mRS: modified Rankin scale; COPD: chronic obstructive pulmonary disease; WBC: White blood cell.*
7
8
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Table 2: Performance of the clinical risk scores for predicting SAP in the derivation and internal validation cohorts

Score	Derivation Cohort	Size (n)	Frequency of SAP (%)	SAP diagnostic criteria	Discrimination metric; C statistic (95% CI)	Calibration metric
ISAN score	SSNAP	11551	6.7	Clinician reported	0.79 (0.77 to 0.81)	'well calibrated across all 22 levels in validation cohort; ceiling effect for score > 15 among ICH patients Cox and Snell R ² =0.106, Nagelkerke R ² =0.259, McFadden R ² = 0.213
A2DS2	BSR	15335	7.2	Clinician reported	0.84 (0.83-0.85)	
AIS-APS	CNSR	8820	11.4	CDC	0.80 (0.78–0.81)	
PANTHERIS	Berlin NICU	223	30	CDC	0.85 (0.80–0.91)	Nagelkerke's R ² 0.46
VHA score	VHA	925	10.4	Clinician reported	0.76 (NR)	2.1% misclassification
Pneumonia score	Seoul	286	10.5	Adhoc objective	NR	NR
PNA score	New-Orleans	568	11.4	Adhoc objective	0.79 (NR)	NR
ICH-APS(A)	CNSR	2998	17	CDC	0.75 (0.72–0.77)	Hosmer–Lemeshow goodness of fit test, P=0.20
ICH-APS (B)	CNSR	2998	17	CDC	0.74 (0.71–0.76)	Hosmer–Lemeshow goodness of fit test, P=0.10
Score	Internal Validation Cohort	Size (n)	Frequency of SAP (%)	SAP diagnostic criteria	Discrimination metric; C statistic (95% CI)	Calibration metric
ISAN score	SSNAP	11648	6.7	Clinician reported	0.78 (0.76-0.80)	NR
AIS-APS	CNSR	5882	11.3	CDC	0.79 (0.77–0.80)	The Hosmer–Lemeshow test was not significant P=0.22
PANTHERIS	Berlin NICU	112	33.9	CDC	0.88 (0.81–0.95)	
VHA score	VHA	438	10.5	Clinician reported	0.78	0.9% reclassification
ICH-APS(A)	CNSR	2000	16.7	CDC	0.76 (0.71–0.79)	Hosmer–Lemeshow goodness of fit test, P=0.66
ICH-APS (B)	CNSR	2000	16.7	CDC	0.73 (0.70–0.76)	Hosmer–Lemeshow goodness of fit test, P=0.17

SAP indicates Stroke-Associated Pneumonia; ISAN, Independence Prestroke, Sex, Age, National Institutes of Health Stroke Scale; A2DS2, Age, Atrial, Fibrillation, Dysphagia, Sex, Severity; AIS-APS, Acute Ischemic Stroke-Associated Pneumonia Score; PANTHERIS, Preventive Antibacterial Therapy in Acute Ischaemic Stroke; VHA, Veterans Health Administration; PNA, Pneumonia Prediction; ICH-APS, Intracerebral Haemorrhage-Associated Pneumonia Score; SSNAP, Sentinel Stroke National Audit Programme; BSR, Berlin Stroke Register; CNSR, Chinese National Stroke Registry; Berlin NICU, Berlin Neurological Intensive Care Unit; CDC, Centers for Disease Control and Prevention; NR, Not Reported; CI, Confidence Interval.

Table 3: Performance of the clinical risk scores for predicting SAP in the external validation cohorts

Score	External Validation cohort	Size (n)	Frequency of SAP (%)	SAP diagnostic criteria	Discrimination metric; C statistic (95% CI)	Calibration metric
ISAN	Athens	3204	12.8	Clinician reported	0.83 (0.81-0.85)	The Hosmer–Lemeshow goodness of fit test (Cox and Snell $R^2 = 0.243$)
A2DS2	NWGSR	45085	7.8	Clinician reported	0.83 (0.83-0.84)	
	SSNAP	11648	6.7	Clinician reported	0.79 (0.77-0.81)	Cox and Snell $R^2 = 0.112$, Nagelkerke $R^2 = 0.264$, McFadden $R^2 = 0.215$
	CNSR	8820	11.4	CDC	0.74 (0.73–0.75)	NR
	CNSR	5882	11.3	CDC	0.73 (0.72-0.74)	NR
	CICAS	3037	7.3	CDC	0.76 (0.74-0.77)	NR
	HNSR	1142	18.8	CDC	0.83 (0.8-0.87)	Cox and Snell $R^2 = 0.243$
	WCH	1279	24	Chinese Expert Consensus	NR	NR
	Shanghai	101	50.5	Chinese Expert Consensus	0.82(0.74-0.9)	NR
AIS-APS	CICAS	3 037	7.3	CDC	0.79 (0.76–0.82)	The Hosmer–Lemeshow goodness of fit test; $P=0.30$
VHA score	CNSR	8820	11.4	CDC	0.75 (0.74–0.76)	NR
	CNSR	5882	11.3	CDC	0.73 (0.72–0.74)	NR
Pneumonia score	CNSR	8820	11.4	CDC	0.71 (0.70–0.72)	NR
	CNSR	5882	11.3	CDC	0.69 (0.68–0.71)	NR
	CICAS	3037	7.3	CDC	0.68 (0.66–0.69)	NR

SAP indicates Stroke-Associated Pneumonia; ISAN, Independence Prestroke, Sex, Age, National Institutes of Health Stroke Scale; A2DS2, Age, Atrial Fibrillation, Dysphagia, Sex, Severity; AIS-APS, Acute Ischaemic Stroke-Associated Pneumonia Score; VHA, Veterans Health Administration; NWGSR, North West Germany Stroke Register; SSNAP, Sentinel Stroke National Audit Programme; CNSR, Chinese National Stroke Registry; CICAS, Chinese Intracranial Atherosclerosis Study; HNSR, Henan Province Stroke Registry; WCH; Wuhan Central Hospital; CDC, Centers for Disease Control and Prevention; NR, Not Reported; CI, Confidence Interval.