

1 **Comparing childhood asthma prediction models in terms of risk classification: suggestions for future**
2 **research**

3 Dear Editor,

4 We would like to thank Owora et al. for their interest in our systematic review of childhood asthma
5 prediction models and welcome their subsequent analyses and recommendations for future research
6 regarding childhood asthma predictions.

7 In their letter, Owora et al. rightly highlight that prediction models can be classified as either diagnostic
8 or prognostic models, with diagnostic models predicting the presence of a disease of interest at the
9 time of prediction whilst prognostic models predict the presence of the disease of interest at a future
10 time-point. We agree with the authors that studies often do not make it clear as to which class of
11 prediction model they have developed or are reviewing. Having acknowledged that the authors also
12 found the scope of our systematic review to be unclear, we would like to clarify that our systematic
13 review was limited to prognostic prediction models for childhood asthma. As detailed in our inclusion
14 criteria, we only considered studies which offered predictions for children aged ≤ 5 years on their risk
15 of developing the future outcome of childhood asthma or wheeze persistence at school-age (6-13
16 years old)¹.

17 Following their meta-analysis of externally validated regression-based models, Owora et al. concluded
18 that based on their pooled estimates of sensitivity, specificity and AUC, existing models offer poor
19 predictive performance, particularly among clinically relevant ranges of sensitivity and specificity. We
20 are pleased that Owora et al. were able to offer quantitative support for our initial inference that
21 existing models offer moderate performance with modest generalisability. Furthermore, their
22 bivariate-random effects meta-regression analysis identified external validation and the use of
23 machine learning approaches as significant moderators of prognostic model sensitivity. These findings
24 corroborate our recommendations for the exploration of novel methodologies and/or biomarkers in
25 the development of future prediction models in order to improve their predictive power, to provide

26 clinical utility. Although Owora et al. identified that externally validated models had a significantly
27 higher sensitivity than non-validated models, we note that external validation is ultimately performed
28 at the discretion of the researchers with regard to i) the chosen study design; ii) the model's potential
29 based on its initial performance in the developmental population; and iii) the availability of
30 independent data. Therefore, we interpret this finding as external validation acting as an important
31 milestone in the developmental process, to not only assess the true generalisability of a model, but
32 also offer confidence in its performance and potential clinical viability. Indeed, as mentioned in our
33 systematic review, the lack of external validation limits studies as exploratory, with little clinical utility,
34 thus, we again emphasise the need for external validation in all future studies.

35 Owora et al. propose the use of reclassification tables to stratify the results of asthma prediction
36 models into distinct risk groups corresponding to different clinical decisions. Reclassification tables,
37 and net reclassification indices, were initially proposed to assess whether the addition of a marker
38 provides any clinically relevant improvement to an existing prediction model^{2,3}. However, we agree
39 that the use of reclassification tables could offer additional insight for physicians when choosing
40 between models, particularly given the trade-off between model sensitivity and specificity. To provide
41 an illustrative example of the potential insight offered by reclassification tables, we compared two
42 existing asthma prediction models for which data was available in the Isle of Wight Birth Cohort
43 (described elsewhere⁴) - the updated PIAMA⁵ and PARS⁶ models. Of the 1456 individuals enrolled in
44 the IOWBC, 837 individuals who had complete data for all predictors used in the two models, and
45 whom had a defined asthma outcome at age 10, were included in the analysis. First, we evaluated
46 overall model performance based on the commonly used measure of area under the receiver
47 operating curve (AUC). We then used a reclassification table to compare the risk classification
48 predicted by each model (as referred to by Owora et al. in reference 3), and reported the results
49 separately for asthmatics and non-asthmatic children at age 10. Risk classification, into low (<15%
50 risk), moderate (15-39% risk) and high (\geq 40% risk) categories, were matched between the two scores
51 as detailed in their original publications (Table 1).

52 In our analysis, both prognostic prediction models indicated approximately equivalent model
53 discrimination (Figure 1), with AUC of 0.76 (95% CI: 0.70-0.81) and 0.77 (95% CI: 0.71-0.82) for the
54 updated PIAMA and PARS models, respectively. However, using the reclassification table (Table 1), we
55 uncovered a substantial inconsistency in the risk categorisation of asthmatic and non-asthmatic
56 individuals between the two models. With respect to the risk classifications made by the updated
57 PIAMA model, 45% of asthmatic children at age 10 were reassigned into a more appropriate (higher)
58 risk category by the PARS model. However, this was accompanied by 13% of non-asthmatic children
59 also being reclassified into a less appropriate (higher) risk category by the PARS model compared to
60 that made by the updated PIAMA model. Therefore, by providing insight into predictions made by
61 prognostics models on an individual-level, and evaluating them separately for asthmatic cases and
62 non-asthmatic controls, reclassification tables can be useful tools for comparing between models
63 rather than using AUC alone. For example, although not evident from the ROC plot, the reclassification
64 table suggests that predictions made by the updated PIAMA model are skewed towards
65 underestimating asthma risk in the IOWBC when compared against the PARS model. This could
66 potentially be informative for policy decision-making depending on whether the clinical intention is to
67 detect true future asthmatics at high risk (PARS preferred), or discount true future non-asthmatics at
68 low risk (updated PIAMA preferred), for primary or secondary interventions.

69 Whilst we agree with Owora et al. and promote the use of reclassification tables, their use requires
70 prediction models to be compared within a single population. In our systematic review, we highlighted
71 the need for a single cohort, of sufficient size, with data available for all predictors used by the existing
72 models. However, this remains an unrealistic ask, which, at present, hinders an objective comparison
73 between existing asthma prediction models. Owora et al.'s second recommendation to estimate and
74 meta-analyse the predicted outcome probabilities across different time-points could offer beneficial
75 insight into the consistency and reliability of a model's predictive power over time. However, the
76 execution of this may also be hindered by the need for a longitudinal cohort with data available for all
77 predictors and the asthma outcome across multiple time-points, with high retention rates.

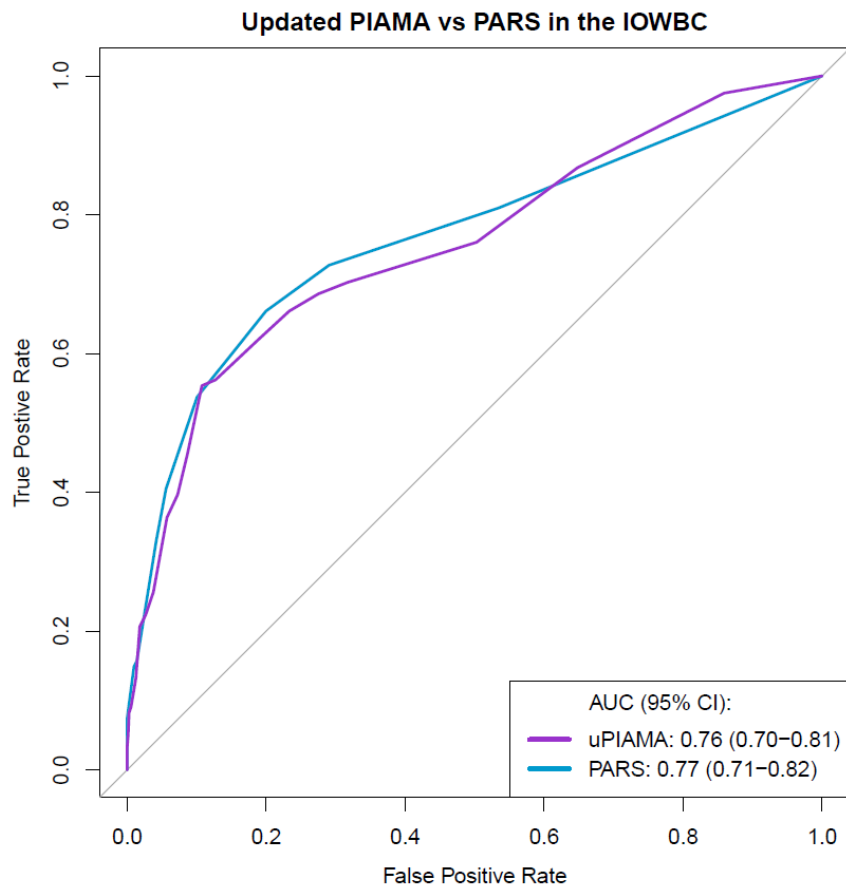
78 The final recommendation for future studies offered by Owora et al. highlighted the importance of
79 developing prognostic models that predict beneficial or adverse responses to primary and secondary
80 interventions. Whilst this was beyond the scope of our systematic review, which focused on models
81 that predicted the development of childhood asthma, we agree that this is a logical progression of
82 research in the field, which could offer beneficial clinical insight. Where data availability permits, we
83 welcome these recommendations for future exploration in childhood asthma predictions.

84 **Tables and Figures**

85 Table 1. Reclassification of predicted asthma risk using the PARS risk score instead of the updated PIAMA risk score

Predicted risk (updated PIAMA)	Predicted risk (PARS)				Reclassified by PARS (%)			NRI
	Low	Moderate	High	Total	Increased risk	Decreased risk	Correctly reclassified	
No asthma at age 10 (n=716)								
Low	605	82	2	689				
Moderate	1	18	7	26	91(13%)	1(<1%)	1(<1%)	-0.13
High	0	0	1	1				
Total	606	100	10	716				
Asthma at age 10 (n=121)								
Low	48	42	0	90				
Moderate	0	12	13	25	55(45%)	0(0%)	55(45%)	0.45
High	0	0	6	6				
Total	48	54	19	121				
					Total	146	1	56

86 Using the updated PIAMA as the reference prognostic model, the table shows the changes in individual
87 risk classification if the PARS model (which offers marginally higher model discrimination) was used
88 instead. For both the updated PIAMA and the PARS models, risk strata were categorised as low=<15%
89 risk (updated PIAMA score: 0-12, PARS score: 0-4); moderate=15-39% risk (updated PIAMA score: 13-
90 19, PARS score: 5-8); high= \geq 40% risk (updated PIAMA score: 20-23, PARS score: 9-14) as defined in
91 their original publications. Results are presented separately for individuals who were asthmatic and
92 non-asthmatic at age 10. Values in bold identify the number of individuals who were reclassified into
93 a more appropriate (green) or less appropriate (red) risk group by the PARS model with respect to the
94 risk classifications made by the updated PIAMA model. NRI=net reclassification index.



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Figure 1. ROC plot comparing the performance of the updated PIAMA (uPIAMA) and PARS models in the IOWBC.

96 **References**

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