1	A simple asthma prediction tool for pre-school children with wheeze or cough
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- Asthma, wheeze, cough, children, prediction, prognosis, persistence, longitudinal,
- 36 cohort study
- 37

38 Clinical Implications

- 39 The proposed asthma prediction tool is simple and uses information that is non-
- 40 invasive and easy to assess. This makes it an ideal instrument for use in clinical
- 41 practice and research.

42

43 Capsule summary

- 44 We have developed a simple tool to predict later asthma in preschool children
- suffering from wheeze or cough. Its simplicity and internal validity facilitate use in
- 46 clinical practice and epidemiological research.

47

48 Abbreviations

- 49 ROC curve: receiver operating characteristic curve
- 50 AUC: area under the ROC curve

- 51 HL test: Hosmer-Lemeshow goodness-of-fit-test
- 52 OR: odds ratio

53

54 Abstract

Background: Many preschool children suffer from wheeze or cough, but only some
 have asthma later. Existing prediction tools are difficult to apply in clinical practice or
 exhibit methodological weaknesses.

58 **Objective:** To develop a simple and robust tool for predicting asthma at school-age 59 in pre-school children with wheeze or cough.

Methods: From a population-based cohort in Leicestershire, UK, we included 1-3 60 year-olds seeing a doctor for wheeze or cough, and assessed prevalence of asthma 61 62 five years later. We considered only non-invasive predictors that are easy to assess in primary care: demographic and perinatal data, eczema, upper and lower 63 respiratory symptoms and family history of atopy. We developed a model using 64 logistic regression, avoided over-fitting with LASSO-penalty, and then simplified it to 65 a practical tool. We performed internal validation and assessed its predictive 66 performance using the scaled Brier score and the area under receiver operating 67 characteristic curve (AUC). 68 Results: Of 1226 symptomatic children with follow-up information, 345 (28%) had 69

asthma 5 years later. The tool consists of 10 predictors yielding a total score

between 0 and 15: sex, age, wheeze without colds, wheeze frequency, activity

disturbance, shortness of breath, exercise-related and aeroallergen-related

wheeze/cough, eczema, and parental history of asthma/bronchitis. The scaled Brier

scores for the internally validated model and tool were 0.20 and 0.16, and the AUCs

vere 0.76 and 0.74, respectively.

76 **Conclusion:**

- 77 This tool represents a simple, low-cost and non-invasive method to predict the risk
- for later asthma in symptomatic pre-school children, which is ready to be tested in
- 79 other populations.

80 Introduction

Many preschool children present to primary care with recurrent wheeze or cough. 81 These symptoms are a burden to families and lead to treatment with inhalers, 82 antibiotics or cough mixtures, hospitalizations and considerable health care costs.¹ In 83 this age-group, wheezing illness is heterogeneous and includes different phenotypes 84 with varying prognoses.²⁻⁵ Fortunately, only some children will have persistent 85 problems till school-age. The ability to predict persistence of wheeze up to school-86 age would allow preventative and therapeutic efforts to be directed to those most in 87 need⁶ and would reassure parents of children with transient problems. It would also 88 help to select children for intervention studies aiming to alter the course of disease.⁷ 89 Several groups have presented tools for prediction of later asthma in preschool 90 children⁸⁻¹⁶, but their use for primary care is limited.¹⁷ Some tools were developed in 91 study populations untypical for primary care. For instance, they included 92 asymptomatic children,^{8, 10, 14, 16} children with mild symptoms, who never visited their 93 doctor,^{13, 15} or only high-risk children hospitalized for bronchiolitis.¹² Several studies 94 excluded children with chronic cough,^{13, 15} who might actually suffer from a variant of 95 asthma.^{4, 18} Some tools included predictors, such as parental education, that are not 96 easily generalizable to other populations.⁹ Other tools involve invasive 97 measurements (blood tests or skin prick tests) that might not be accepted by all 98 families in primary care.^{8, 11, 13, 14} Finally, the methods commonly used to develop the 99 prediction tools are prone to over-fitting the data.^{9, 11, 13} Over-fitting leads to reduced 100 performance when tools are applied to other populations.^{19, 20} 101 In this study we aimed to develop a simple tool to predict asthma at school-age in 102 preschool children with wheeze or chronic cough. We designed the tool for 103 application in clinical practice, particularly primary care, by: a) studying a population 104

of symptomatic children, who had presented to the doctor for wheeze or cough; b)
 defining a clinically relevant outcome; c) considering only predictive factors easily
 assessed during a single consultation (a detailed symptom history, but no blood or
 skin prick tests and no repeated observations); d) developing a robust model that
 performs well in internal validation and relevant sensitivity analyses but does not
 over-fit the data and is therefore likely to be transferable to other populations.

111

112 Methods

113 Study population

We analyzed data from a population-based childhood cohort from Leicestershire, 114 UK, described in detail elsewhere.^{21, 22, 23} In brief, we recruited a representative 115 population-based sample of 6808 children of white and south Asian ethnic origin, 116 born in 1993-97. Perinatal data were collected at birth; data on growth and 117 development were acquired prospectively during childhood. Upper and lower 118 respiratory morbidity, treatments and health care utilization, family history of atopic 119 disease and individual and family-related exposures were assessed by repeated 120 questionnaires (1998, 1999, 2001, 2003, 2006, 2010). The study was approved by 121 the Leicestershire Health Authority Research Ethics Committee. 122

123 Presentation at baseline (inclusion criteria)

Our analysis included all cohort children aged 1-3 years at baseline with parentreported wheeze or chronic cough (cough without colds or cough at night) with one or more visits to the doctor for wheeze or cough during the past 12 months (Fig 1, highlighted in grey). The original questions are provided in the online repository. We included chronic cough, because some children with chronic cough might suffer from a variant of asthma and be at risk for asthma later in life.^{4, 18} Information on

symptoms at baseline was taken from the 1998 or the 1999 questionnaire, favoring
the questionnaire when children were closest to age 2.0 years.

132 Any asthma at school-age (definition of outcome)

We defined a clinically relevant outcome as the combination of current wheeze *plus* use of asthma medication during the past 12 months at the age of 6-8 years, i.e. 5 years later (see online repository for original questions). Asthma medication included short- or long-acting beta-2-agonists, inhaled corticosteroids, leukotriene receptor antagonists or oral corticosteroids.

138 We used Fisher's exact test to compare characteristics of children with and without the outcome (Table E1, Table I) as well as to compare characteristics of children by 139 availability of follow-up information (Table E2). Choice of potential predictive factors 140 We used the following approach to compile the list of potential predictors. First, we 141 reviewed the literature to identify relevant risk factors for incidence or persistence of 142 childhood asthma.^{3, 24-31} From these, we only selected factors that are readily 143 available in primary care and do not require repeated observations or additional 144 investigations like blood or skin prick tests. The final list contained 24 potential 145 predictors (Table E1): demographic and perinatal data; eczema; upper and lower 146 respiratory symptoms, particularly those reflecting triggers and severity of wheeze; 147 and parental history of wheeze, asthma, bronchitis or hay fever (see online 148 149 repository for original questions). We did not include environmental or socioeconomic information, because their prevalence and interpretation is likely to 150 vary between populations and, thus, their inclusion might reduce the generalizability 151 of the tool. 152

153 *Model development*

¹⁵⁴ We used LASSO-penalized logistic regression to develop the prediction model.^{32, 33}

This approach allows to identify important predictors and to estimate their influence 155 on later asthma without over-fitting the data. Traditional methods used for selecting 156 predictors, such as stepwise backward or forward selection, tend to over-fit the data, 157 resulting in models that predict outcomes in the current dataset well, but become 158 unreliable in other datasets.²⁰ For our analysis, we recoded all potential predictors 159 with >2 response categories into multiple binary variables. Thus, 38 binary variables 160 derived from the 24 questions entered the variable selection process (see online 161 repository for details). LASSO regression selects predictors in the order of their 162 163 predictive importance. The final prediction model allows calculation of a prediction score and the probability of later asthma for each child. 164

165 *Model performance*

We assessed our prediction model in terms of overall performance, discrimination 166 and calibration. To assess overall performance we calculated the scaled Brier 167 score,²⁰ a measure of the discrepancy between the predicted probability and the 168 actual outcome. A scaled Brier score with a value of zero means that the model does 169 not predict later asthma in an individual better than if it had been informed only by 170 the average prevalence of asthma at school-age; the maximal value of one indicates 171 perfect prediction. To determine the *discriminative ability* of the model (i.e. its ability 172 to distinguish between children with and without later asthma) we plotted the receiver 173 operating characteristics (ROC) curve and calculated the area under this curve 174 (AUC), also known as c-statistic.^{20, 34} The AUC can take on values from 0 to 1, with 1 175 being a perfectly discriminating model. Discrimination is considered not better than 176 chance if AUC=0.5, moderate if AUC is 0.6 to 0.8, and good if AUC>0.8.³⁴ 177 Calibration of the model (how well the predicted probabilities agree with the 178 prevalence of the outcome in subgroups of children) was tested using the Hosmer-179

Lemeshow goodness-of-fit-test (HL test)^{20, 35} and visualized using a calibration plot.²⁰
 An HL test result of less than 0.05 indicates that the predicted probabilities and the

An HL test result of less than 0.05 indicates that the predicted probabilities and the

actual outcome agree poorly. In the calibration plot, a perfect calibration curve would

lie exactly on the diagonal line.

184 Internal validity

A prediction model can be validated internally to provide a more accurate estimate of model performance in other populations. As an internal validation of our model, we used the leave-one-out cross-validation method^{20, 34} assessing overall performance

188 (Brier), discrimination (AUC), and calibration (see online repository for further

189 explanations).

190 Sensitivity analyses

191 To test the robustness of the model developed in our original study population (P0),

we performed sensitivity analyses using modified inclusion criteria at baseline or

193 modified definitions of the outcome, resulting in slight changes of the study

194 populations (P1 to P4, described in more detail in Tables E3 and E4 of the online

195 repository).

196 We first applied our existing prediction model to these modified populations and

calculated the scaled Brier score and AUC (Sensitivity analysis I). Second, we

developed new models within the slightly modified study populations P1 to P4, and

assessed their performance (Sensitivity analysis II).

200 Clinical prediction tool

To simplify our model to a practical tool, we considered three different approaches:

a) multiplying regression coefficients by factors 10, 5 and 3 and rounding them to the

nearest integer;²⁰ b) setting the penalty of the LASSO-penalized logistic regression

so that only a few important predictors (5 or 3) were retained, and c) considering a

model with frequency of wheeze as the only predictor.¹⁹ All these approaches aimed
to reduce the number of variables while maintaining a comparable predictive
performance.

208

209 Results

210 Study population

At the baseline survey, 5878 of 6808 children were aged 1-3 years. Figure 1 shows 211 how many of the 1-3 year old children reported episodes of wheeze, cough without 212 213 colds or cough at night in the past 12 months and in addition reported visits to a doctor (N=2444), making them eligible for the study. For 1226 we had information on 214 any asthma five years later. Their characteristics are shown in Table I for the 215 variables selected by the main model and in Table E1 (online repository) for all 216 potential predictors considered. At baseline, 336 children (27.4%) were aged one 217 year, 702 (57.3%) two years and 188 (15.3%) three years. The mean prediction 218 interval from baseline to outcome was 4.5 (± SD 0.5) years. At school-age, 345 219 (28.1%) had any asthma. 220 Table E2 in the online repository compares eligible children with and without follow-221 up information. The groups were comparable in many aspects (chronic cough, upper 222 respiratory infections, eczema and parental history), but those with follow-up 223

information were more likely to be of white ethnicity and less likely to have wheeze atbaseline.

226 Main prediction model

Of the 38 binary predictors that entered variable selection, the LASSO-penalized logistic regression retained 22 (Table II). The 5 most important predictors were, in order of importance, shortness of breath, frequent wheeze, wheeze without colds,

activity disturbance by wheeze and wheeze/cough triggered by exercise. In addition,

the model included aeroallergen-related wheeze/cough, male sex, age, birth weight,

232 gestational age, eczema, upper respiratory symptoms, and parental history of

wheeze, asthma, bronchitis or hay fever.

In the original study population, the overall performance of the main model measured

by the scaled Brier score was 0.23 and its discriminative ability (AUC) was 0.78. In

internal validation, these measures were comparable, 0.20 and 0.76 respectively.

237 The calibration plot (Fig 2) shows good agreement between the predicted

probabilities of later asthma and the observed frequencies in internal validation. The

same was indicated by the Hosmer-Lemeshow test (p=0.6).

240 Sensitivity analyses

241 Sensitivity analyses I: The main model was robust to changes in baseline criteria

242 (P1, P2 in Table E3). When the outcome definition was changed to wheeze plus a

doctor's diagnosis of asthma (P3) or to moderately severe asthma (≥4 attacks plus

inhaled corticosteroids; P4), the AUC improved to 0.80 and 0.87 respectively (P3

and P4 in Table E3). Sensitivity analyses II: The performance of new models

developed in these alternative study populations was comparable to the main model

for P1-P3 and slightly improved for P4 (Table E4). The selected predictors and

estimated coefficients in the newly developed models (Table E5) were comparable to

those of the main model. Severity-related predictors (wheeze without colds, frequent

attacks, shortness of breath, activity disturbance) gained comparatively more weight

when predicting moderately severe asthma (P4).

252 Clinical prediction tool

253 We then simplified the model using the three planned approaches. Our preferred

simplification includes 10 variables (13 binary predictors), each of which contributes

with one of 3 values (1, 2 or 3) to the prediction score (Fig 3; an online version of the prediction tool is available on www.leicestercohorts.org).

257 This tool was derived from the original model by multiplying all regression

coefficients with 3 and rounding them to the nearest integer, dropping variables with

coefficients rounded to zero.²⁰ It had almost the same discriminative ability

260 (AUC=0.775) as the main model (AUC=0.782) (Fig.4). Other approaches to

simplification retained more predictors (making the tool complicated with little benefit)

or had reduced discriminative ability (Table E6), particularly the model with

frequency of wheeze only.

In internal validation, the prediction tool showed only a minor decrease in

performance compared to the main model: the scaled Brier score was 0.16 and the

AUC 0.74.

The maximum score a child can attain using the prediction tool is 15, corresponding

to a 95% probability of having any asthma 5 years later (Fig 3). Sensitivity and

specificity of the tool are 0.72 and 0.71 for a score of 5, and 0.22 and 0.98 for a

score of 10 (additional performance measures are reported in Table E7). In our study

sample, 840 (69%) children were at low risk (score ≤5), 288 (23%) at medium risk

(score ≥ 6 and ≤ 9) and 98 (8%) at high risk (score ≥ 10) of any asthma 5 years later.

The percentage of children with any asthma at school age was 16%, 48% and 79%

in the low, medium and high risk groups respectively.

275

276 Discussion

277 Summary of findings

We have developed a new tool for predicting asthma at school-age in preschool children who see a doctor for wheeze or cough. Our tool includes 10 predictors

representing wheeze severity and triggers, male sex, age, eczema and parental
 respiratory history. It showed good internal validity and is distinguished by ease of

use in primary care and epidemiological studies.

283 Comparison with previous prediction models

Several prediction models have been proposed for estimating the risk of persistent 284 asthma in preschool children.⁸⁻¹⁶ Table III summarizes inclusion criteria, outcome, 285 methods used to derive the tool, predictors and performance for three tools that used 286 a similar prediction interval as ours and had a sample size of >300. In short, Castro-287 Rodriguez (Tucson Children's Respiratory Study) used data from 2-3 year-olds with 288 and without respiratory symptoms to develop two prediction tools for asthma at 289 school-age (loose and stringent asthma predictive index, API; Table III).⁸ 290 Kurukulaaratchy (Isle of Wight birth cohort) proposed a score for persistence of early 291 wheeze up to age 10.¹³ Caudri (PIAMA birth cohort), developed a clinical risk score 292 for 0-4 year-olds with wheeze or cough to predict asthma at age 7-8.9 293 The performance of these tools was comparable or slightly less than ours (Table III). 294 with a Youden index³⁶ (sensitivity + specificity -1) varying from 0.32^8 to 0.38^{13} 295 (calculated based on the maximal sum of sensitivity and specificity reported in the 296 respective studies) compared to 0.43 in our study. The Youden index ranges 297 between 0 and 1. Values close to 1 indicate large predictive effectiveness and values 298 299 close to 0 limited effectiveness. The method used to derive the APIs is difficult to replicate,⁸ while methods used for 300 the other tools ^{9, 13} (logistic regression with stepwise variable selection) tend to over-301

302 fit the data, i.e. the models might be overly influenced by the random variation in the

data used to develop them. This limits the application of the models to other

304 populations.

Only Caudri et al. performed an internal validation of their prediction model and reported a similar AUC (0.72) to the one we obtained (0.74). They included 8 predictors with exact regression coefficients, while our model includes 10 predictors with simplified regression coefficients that facilitate calculation of individual risks in a clinical setting. The PIAMA risk score and the API have been tested in a small external population.^{19, 37}

In comparison to our tool, previous asthma prediction rules included at most two
descriptors of wheeze (out of frequency, duration or wheeze without colds).^{8-10, 14} In
addition, they relied on blood or skin prick tests,^{8, 11-13, 15} which are more time
consuming, costly and cumbersome than a detailed symptom history.
Socioeconomic position is a proxy measure for a variety of exposures and health

care access and might have a variable impact in different populations.⁹

317 Strengths and limitations

The main strengths of our tool are the objective approach used for its development 318 and its clinical applicability. We used a population-based sample of an adequate size 319 to develop the model. We included only children with health care visits for wheeze or 320 cough, assuring that the sample represents a clinically relevant population. We 321 defined a clinically relevant outcome measure (wheeze needing treatment). When 322 defining a more severe outcome (moderately severe asthma, defined as ≥4 attacks 323 324 per year and inhaled corticosteroid treatment) the tool performed even better. All predictors are obtained routinely when taking a respiratory history for a child 325 presenting with chronic cough or wheeze and predictors are easy to assess even 326 during a short primary care consultation or in a questionnaire survey. We used a 327 method that minimizes over-fitting and is less affected by sampling variability 328 compared to stepwise variable selection procedures,³⁸ and we did an internal 329

validation. Finally, our model predicts a range of probabilities rather than predicting
 only a low or high risk as the API.⁸

Like other studies,^{8, 9, 11, 13} ours relies on parent-reported questionnaire data. 332 However, it uses standardized questions, mostly from the ISAAC-study³⁹ and reflects 333 to some extent the clinical situation, where parents report respiratory symptoms. The 334 applied questionnaire showed good repeatability.⁴⁰ We did not use objective 335 measurements to define our outcome. However, for a subsample of our study 336 population (N=451), we assessed bronchodilator response in a later survey 337 conducted in 2006 (Table E8). Using the same outcome definitions (any asthma and 338 moderately severe asthma), mean percent change in forced expiratory volume in the 339 1st second (FEV₁) was significantly higher in children with any asthma compared to 340 those without (5.5% (95% CI 3.6-7.3) vs 2.6% (2.0-3.2), p<0.001). For maximal 341 expiratory flow at 50% of vital capacity (MEF₅₀), mean percent change was 16.7% 342 (12.8-20.5) and 10.7% (8.8-12.5) respectively (p=0.003). This is less than the cut-343 offs recommended for clinical situations.⁴¹ However, our measurements came not 344 from hospital-based children referred when they were unwell, but from community-345 based children with very mild asthma who were usually asymptomatic when 346 measured. Our results are in line with data from Galant et al, where bronchodilator 347 responses for FEV₁ were 7.3% (4.2-10.4) in mild persistent asthmatics and 7.6% 348 (5.8-9.5) in mild intermittent asthmatics compared to 2.2% (0.2-4.3) in non-349 asthmatics.⁴² Children with and without follow-up information were comparable 350 (Table E2), although we cannot exclude that selection bias has affected the 351 composition of the final model. Finally, we interpreted missing values in potential 352 predictor variables as an absence of the respective risk factor, which may also have 353

affected the results. However, the number of missing values did not exceed 5.8% in
any of the potential predictor variables.

356 *Meaning of the study*

Our model was robust and results changed little with modifications of the inclusion criteria and outcomes. In fact, the performance improved (AUC 0.89 vs. 0.78) when we predicted moderately severe asthma, rather than any asthma. After internal validation, the AUC of main model and tool were similar to the ones before validation, suggesting that there was little over-fitting.

362 Our tool used only information on symptoms that can be gathered in a simple patient's history. Despite that, it had a similar or better predictive performance than 363 previous tools including more complex measurements.^{8, 11, 13-15} This suggests that a 364 detailed description of presented symptoms might predict later asthma equally well 365 as more invasive methods, including blood eosinophilia or skin prick tests.^{8, 11, 13-15} 366 Seven of 10 predictors (including the 5 strongest) describe the symptoms: frequency 367 of attacks, activity disturbance, shortness of breath, triggers (wheeze apart from 368 colds, exercise, aeroallergens) and eczema. This is consistent with the old 369 knowledge that frequent wheeze strongly predicts asthma persistence, ^{10, 43} and with 370 our previous report, showing that frequency of wheeze predicted asthma nearly as 371 well as the complicated API rule.¹⁹ In our tool, adding more symptoms (in addition to 372 wheeze frequency) improved the performance (AUC after internal validation 0.74 for 373 the tool vs. 0.57 for wheeze frequency only; Table E6). This shows that more 374 detailed assessment of symptoms in pre-school children improves prediction of later 375 asthma. 376

377 Future research

To further evaluate the predictive performance of the proposed tool and assess its generalizability to other populations, external validation in independent samples is necessary.³⁴ We therefore encourage the application and validation of this tool in ongoing epidemiological studies and clinical care (particularly primary care). Some earlier prediction models^{8, 9, 13} performed similarly in external populations, but their performance remained modest.^{15, 19, 37}

Compared to other prediction rules, our tool includes detailed description of symptom severity and pattern. This raises the possibility that further refinement in the

386 description of preschool wheeze phenotype might improve precision of prediction of

³⁸⁷ later asthma. Additional gains might be made by detailed assessment of age-related

388 changes, physiological measurements (lung function, bronchial

389 hyperresponsiveness, exhaled nitric oxide, atopy), environmental, socioeconomic

and genetic risk factors.¹⁷ All this could, however, compromise the tool's simplicity.

391 Conclusions

This tool represents a simple, low-cost and non-invasive method to predict the risk for later asthma in symptomatic preschool children, which is ready to be tested in other populations.

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Table I. Characteristics of the study population (N=1226) at baseline, by development of asthma 5 years later*

		5 yrs later: Asthma (n=345)		5 yrs later: No Asthma (n=881)		
		n	(%)	n	(%)	p-value†
Demographic and perinatal	data					
Male		224	(64.9)	454	(51.5)	< 0.001
Age (years):	1	85	(24.6)	251	(28.5)	0.388
	2	204	(59.1)		. ,	
	3	56	(16.2)	132	(15.0)	
Gestational age <37 weeks		35	(10.1)		(5.6)	0.006
Birth weight <2500 g		41	(11.9)	68	(7.7)	0.025
Wheeze-related symptoms:						
Current wheeze		272	(78.8)	425	(48.2)	< 0.001
Wheeze without colds		127	(36.8)	95	(10.8)	< 0.001
Frequency of attacks:	0	81	(23.5)	476	(54.0)	< 0.001
- •	1-3	111	(32.2)		(31.9)	
	4-12	117	· /		(12.0)	
	>12		(10.4)		(12.0) (2.0)	
Activity disturbance:	no	141	(40.9)			< 0.001
notivity distance.	little		(37.4)		· · · ·	0.001
	moderate	57	(16.5)		· /	
	a lot	18	(5.2)	8	(0.9)	
Shortness of breath:	never	129	(37.4)		. ,	< 0.001
	sometimes		(48.1)		. ,	
	always	50	(14.5)		. ,	
Exercise-related wheeze/coug	•	196	(56.8)		. ,	< 0.001
Aeroallergen-related wheeze/		52	(15.1)	37	(4.2)	< 0.001
Other symptoms‡						
Cough without colds		233	(67.5)	536	(60.8)	0.030
Duration of colds (weeks):	<1	75	(07.3) (21.7)		. ,	0.194
(1-2		(57.4)		· /	
	>2		(20.9)		. ,	
Nasal symptoms			(53.9)			< 0.001
Eczema (ever)			(55.1)		· · ·	< 0.001
Parental history						
Wheeze, asthma or						
bronchitis:	none	142	(41.2)	499	(56.6)	< 0.001
	father		(19.7)		. ,	-
	mother		(24.6)		. ,	
	both		(14.5)		. ,	
Hay fever:	none		(44.1)			0.001
2	father		(16.2)			
	mother		(27.0)			
	both		(12.8)		. ,	

* This table includes all predictors that were selected for the main model

† Fisher's exact test

‡ During the last 12 months

§ Wheeze or cough with running, playing, laughing or crying

Table II. Important factors for prediction of asthma at school age in symptomatic preschool children (selected by penalized logistic regression)

		OR§	Regression coefficient (RC)	Simpli- fied RC*	Order of inclusion
			Main model	Tool	
Demographic and p	erinatal data				
Male		1.48	0.394	1	9
Age: >1 year		1.19	0.171	1	16
Gestational age <37 v	weeks	1.11	0.108		18
Birthweight <2500g		1.17	0.154		17
Wheeze-related sym	ptoms†				
Current wheeze		1.18	0.163		13
Wheeze without cold	S	1.40	0.337	1	3
Frequency of attacks: Activity	>3	1.65	0.500	2	2
disturbance:	any	1.28	0.243	1	4
disturbance:	moderate or a lot	1.16	0.144		7
	a lot sometimes or	1.63	0.491	1	13
Shortness of breath:	always	1.98	0.684	2	1
	always	1.56	0.442	1	6
Exercise-related whee	eze/cough‡	1.26	0.233	1	5
Aeroallergen-related	wheeze/cough	1.22	0.198	1	10
Other symptoms†					
Cough without colds		1.09	0.086		18
Duration of colds: at	least 1week	0.97	-0.031		22
Nasal symptoms		1.17	0.157		12
Eczema (ever)		1.52	0.420	1	7
Parental history					
Wheeze, asthma or					
bronchitis:	mother or father	1.23	0.203	1	10
	both parents	1.26	0.235	1	13
Hay fever:	mother or father	1.03	0.025		21
	both parents	1.12	0.110		18
Number of binary p	redictors	22	22	13	22
Number of variables		17	17	10	17

* RC of the main model multiplied by 3 and rounded to the nearest integer

(simplification approach where the number of variables was substantially reduced without relevant decrease in predictive performance)

† During the last 12 months

[‡] Wheeze or cough with running, playing, laughing or crying

§ Confidence intervals for the ORs are not provided because OR estimates result from penalized logistic regression which is primarily a method for variable selection rather than for statistical inference. Estimates are deliberately biased toward null with the benefit of reducing their variance and improving overall prediction. Confidence intervals are misleading in this context.

523

524 Table III. Comparison of four asthma prediction tools for preschool children

	Leicester (present study) (Leicestershire Respiratory Cohort Studies)	Tucson (API)⁸* Tucson Children's Respiratory Study	IoWBC¹³ Isle of Wight Birth Cohort	PIAMA⁹ Prevention and Incidence of Asthma and Mite Allergy
N (included in analysis) Inclusion criteria	1226	776	336	2054
Age (y)	1-3	2-3	4	1-4
Symptoms	Health care visit due to respiratory problems plus at least one of the following symptoms in the past 12 months: Wheeze, cough without colds, cough at night	Entire cohort (including a majority of children without symptoms)	Wheeze at ages 1,2 and 4 yrs	Wheeze or cough at night without colds (or both) in the past 12 months
Outcome definition				
Age (y)	6-8	8	10	7-8
Prediction interval (y)	4-5	5	6	3-7
Criteria	Wheeze plus asthma medication (past 12 mo)	Doctor's diagnosis of asthma plus current wheeze, or more than 3 wheeze episodes (past 12 mo)	Current wheeze	At ages 7 and 8y: Current wheeze or prescription of inhaled corticosteroids or doctor's diagnosis of asthma (past 12 mo)
Outcome prevalence	28.1 %	13.7%	37.2%	11.7%
Predictor variables included in tool	Male sex, Age: >1y, wheeze without colds, frequent wheeze, activity disturbance, shortness of breath, exercise-related wheeze/cough†, aeroallergen-related wheeze/cough, eczema, parental asthma or wheeze bronchitis	Wheeze, frequent wheeze‡, wheeze without colds, eczema, parental asthma, blood eosinophilia, allergic rhinitis	Family history of asthma, recurrent chest infections (at 2yrs), skin prick test positivity (at 4yrs), nasal symptoms (at 1yr)	Male sex, post term delivery, wheeze/dyspnea without colds, frequent wheeze, eczema, respiratory infections, inhalation medication (parents), parental education
Method used to derive tool	Penalized logistic regression	The combination of predictors was chosen that yielded the highest PPV and specificity	Stepwise backward logistic regression	Stepwise backward logistic regression
Performance measures§	Score-cutoff: ≥ 5	Loose API	Score-cutoff: ≥ 3	Score-cutoff: ≥20
Youden index ³⁶	0.43	0.32	0.38	0.36
Sensitivity (%)	72	51	53	60
Specificity (%)	71	81	85	76
PPV (%)	49	29	68	23
NPV (%)	86	91	74	94

API, Asthma Predictive Index; PPV, positive predictive value; NPV, negative predictive value.

* To have a prediction interval comparable to the one in our tool, we focused here on the API for prediction at 8 yrs

† Wheeze or cough with running, playing, laughing or crying
‡ This variable is only part of the stringent API, but not of the loose API

525 526 527 528 529 \$ Reported for cut-off where sum of sensitivity and specificity pair was maximal. It is possible that a higher sum of sensitivity and specificity exists at a cut-off point that was not reported in the respective studies.

530 Figure legends

531

532 Fig 1. Wheeze, cough and health care visits in 1 to 3 year-old children.

Proportional Venn diagram for children aged 1 to 3 years, showing frequency of
health care visits due to wheeze or cough, current wheeze and chronic cough (cough
without colds or cough at night). The shaded grey represents our study population.

536

537 Fig 2. Calibration plot of main model (assessed in leave-one out cross-

validation). Children are grouped into deciles of their predicted probability. The

average predicted probability for later asthma among children within each decile is

540 plotted against the actual observed frequency (prevalence) of asthma in that group.

As a visual aid a smoothing technique (locally-weighted polynomial regression) was
applied to these data.

543 The straight line represents perfect calibration.

544

Fig 3. Asthma prediction tool. For any 1-3-year-old child seeking health care due
to wheeze or cough the applicable predictors are summed to a total score in the
upper part of the figure. The estimated probability of having asthma 5 years later is
given below for different total scores.

549

Fig 4. Receiver operating characteristic (ROC) curves for the main asthma
 prediction model and for the prediction tool.

552 The dots represent sensitivity and specificity for different cutoff-values of the 553 prediction tool.

554

1	A simple asthma prediction tool for pre-school children with wheeze or cough
2	
3	Anina M Pescatore, MSc, ¹ Cristian M Dogaru, MD, PhD, ¹ Lutz Duembgen, PhD ² ,
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26 **Details of statistical methods**

27 Development of the main prediction model

We used the R package glmnet to fit the penalized logistic regression. The 28 parameter alpha was set to 1 so that only a LASSO type penalty was included. This 29 tends to retain only the most influential predictors. The parameter lambda, which 30 determines the magnitude of the penalty was set to a value that maximized the area 31 under the receiver operating characteristic curve of resulting predictions in 10-fold 32 cross-validation¹. All potential predictors with more than 2 response categories were 33 34 coded as binary variables. If the original categories were ordered, these dichotomous variables represented all possible cut-off points separating lower from 35 higher categories. For instance, frequency of wheezing episodes in the past 12 36 months (0, 1-3, 4-12, >12) was coded into 3 binary variables indicating >0, >3, and 37 >12 episodes respectively. This procedure resulted in 38 binary variables entering 38 39 variable selection. Confidence intervals for the ORs are not provided because OR estimates result from 40 penalized logistic regression which is primarily a method for variable selection rather 41 42 than for statistical inference. Estimates are deliberately biased toward null with the

43 benefit of reducing their variance and improving overall prediction. Confidence

44 intervals are misleading in this context.

Data were prepared using Stata 11.0 and analysed using R version 2.12.2. We used
the R package ROCR to assess discrimination and the functions hosmerlem and
val.prob.ci to assess calibration².

48

49 Clinical prediction tool

50 To simplify our model to a practical tool, we considered three different approaches:

a) multiplying regression coefficients by factors 10, 5 and 3 and rounding them to the

nearest integer;²⁰ b) setting the penalty of the LASSO-penalized logistic regression
so that only a few important predictors (5 or 3) were retained, and c) considering a
model with frequency of wheeze as the only predictor.¹⁹ All these approaches aimed
to reduce the number of variables while maintaining a comparable predictive
performance.

In Table E7 the performance of these tools are compared with the main model in
sample (sample used for model development) and by internal validation (see below).
In a final step, we recalibrated the probabilities for later asthma of the preferred tool
by re-running a logistic regression of the outcome on simplified scores.

61

62 Internal validation

To assess the reliability of our result of model performance within our study sample 63 (i.e. to test its repeatability within our development sample) we tested our model in 64 leave-one-out cross-validation. The first step in this technique is to omit the first of 65 total n observations and to use the remaining n-1 observations from the entire study 66 sample to develop a new model. Using this new model, the probability for later 67 asthma is estimated for the one observation left out before. In total, this procedure is 68 repeated n times, each time omitting an observation that has not previously been left 69 out. In the end, internal validity of the model is tested based on these estimated 70 probabilities. 71

Because the purpose was to test the main model's predictive performance and not how the method performs (including variable selection), we chose leave-one-out cross-validation as an internal validation technique that aims to fit models which are very similar to the main model. Other approaches, such as bootstrapping, would result in fitting models that are less similar to the main model, and thus would have

77	tested the repeatability	of the method (vari	iable selection appro	bach and estimation of
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regression coefficients) rather than have validated the main model itself.

79

80

81 References

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- 83 E1. Friedman J, Hastie T, Tibshirani R. Regularization Paths for Generalized Linear
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Question			Total study population (N=1226)		5 yrs later: Asthma (N=345)		5 yrs later: No Asthma (N=881)			
number*			n	(%)	n	(%)	n	(%)	p-value [.]	
	Demographic and perinatal	data							•	
	Male		678	(55.3)	224	(64.9)	454	(51.5)	< 0.001	
	Age (years)	1	336	(27.4)	85	(24.6)	251	(28.5)	0.388	
		2	702	(57.3)		(59.1)		(56.5)		
		3	188	(15.3)	56	(16.2)	132	(15.0)		
	Gestational age <37 weeks			(6.9)	35	· /	49		0.006	
	Birth weight <2500 g			(8.9)	41	· · ·		(7.7)	0.025	
	South Asian ethnicity (versus	white)	316	(25.8)	78	(22.6)	238	(27.0)	0.127	
	Wheeze-related symptoms‡									
8	Current wheeze					(78.8)			< 0.001	
9	Wheeze without colds		222	(18.1)		· /	95	(10.8)	< 0.001	
10	Frequency of attacks:	0	557	· /	81	· /		(54.0)	< 0.001	
		1-3	392	· /	111	· /		· /		
		4-12	223	· /	117	. ,		(12.0)		
		>12		(4.4)		(10.4)		(2.0)		
11	Activity disturbance:	no		(64.4)	141	· /			< 0.001	
		little		· /		(37.4)		· /		
		moderate		(7.8)	57	(16.5)		(4.4)		
		a lot		(2.1)	18	(5.2)	8	(0.9)		
12	Shortness of breath:	never	797	(65.0)	129	· · · ·		(75.8)	< 0.001	
		sometimes	356	· /	166	· · · ·	190	· /		
12		always	73	(6.0)	50	· /	23		<0.001	
13	Sleep disturbance:	never	790					(72.9)	< 0.001	
		<1 >=1		(21.9)		(35.4)		(16.7)		
14	Exercise-related wheeze/coug	-		(13.6) (39.3)		(21.7) (56.8)		(10.4)	< 0.001	
14	Aeroallergen-related wheeze/		482 89	(7.3)		(30.8) (15.1)		(32.3)	< 0.001	
14 14	Food-related wheeze/cough	cougii		(7.5) (15.2)		(15.1) (15.7)			<0.001 0.791	
14	-		100	(13.2)	54	(15.7)	152	(15.0)	0.791	
15	Other symptoms:		7(0	((2,7))	222	((7,5))	526	((0,0))	0.020	
15	Cough without colds					(67.5)			0.030	
16 17	Cough at night	~1	631	(51.5)		(55.1)		(50.1)	0.127	
17	Frequency of colds:	<4 4-6				(29.3) (38.8)		· /	0.001	
		4-0 >6		(37.0) (25.9)		(38.8)		(37.1) (23.6)		
18	Duration of colds (weeks):	≥0 <1	278	(23.9) (22.7)	75	(31.9) (21.7)		(23.0) (23.0)	0.194	
10	Duration of colus (weeks).	<1 1-2	731	(22.7)	73 198	(21.7)		(23.0) (60.5)	0.174	
		>2	217	(17.7)	72	(37.4) (20.9)		(16.5)		
19	Ear infection(s):	0	599	(17.7) (48.9)	151	(43.8)		. ,	0.020	
17		1	351	(48.9) (28.6)	99	(43.8) (28.7)		(28.6)	0.020	
		>1		(22.5)	95	. ,		(20.0) (20.5)		
20	Nasal symptoms	· •				(27.3) (53.9)		· /	< 0.001	
20	Snoring					(77.4)			0.001	
	~			1, 1, 1, 0, 1					0.000	

Table E1. Characteristics of the study population (N=1226) at baseline by development of asthma 5 years later (all potential predictors considered in the analysis)

	Parental history Wheeze, asthma or		
23/24	bronchitis:	none	641 (52.3) 142 (41.2) 499 (56.6) <0.001
		father	204 (16.6) 68 (19.7) 136 (15.4)
		mother	267 (21.8) 85 (24.6) 182 (20.7)
		both	114 (9.3) 50 (14.5) 64 (7.3)
23/24	Hay fever:	none	626 (51.1) 152 (44.1) 474 (53.8) 0.001
		father	200 (16.3) 56 (16.2) 144 (16.3)
		mother	296 (24.1) 93 (27.0) 203 (23.0)
		both	104 (8.5) 44 (12.8) 60 (6.8)

* See Online Repository: Original questions used in questionnaires

† Fisher's exact test

‡ During the last 12 months

§ Wheeze or cough with running, playing, laughing or crying

Table E2. Characteristics of children at baseline, by availability of follow-up information (N=2444)

		Follow-up information available (N=1226)		infor ava	ow-up mation not ilable =1218)	
Demographic and residents	1 1 2 4 2	n	(%)	n	(%)	p-value*
Demographic and perinata	I data		/		/ ··	
Male		678	(55.3)	633	(52.0)	0.105
Gestational age <37 weeks		84	(6.9)	86	(7.1)	0.874
Birth weight <2500 g	•••	109	(8.9)	86	(7.1)	0.101
South Asian ethnicity (versu	s white)	316	(25.8)	386	(31.7)	0.001
Wheeze-related symptoms	÷					
Current wheeze		697	(56.9)	762	(62.6)	0.004
Wheeze without colds		222	(18.1)	272	(22.3)	0.010
Frequency of attacks:	0	557	(45.4)	482	(39.6)	0.012
	1-3	392	(32.0)	419	(34.4)	
	4-12	223	(18.2)	269	(22.1)	
	>12	54	(4.4)	48	(3.9)	
Activity disturbance:	no	790	(64.4)	725	(59.5)	0.044
	little	314	(25.6)	371	(30.5)	
	moderate	96	(7.8)	91	(7.5)	
	a lot	26	(2.1)	31	(2.5)	
Shortness of breath:	never	797	(65.0)	749	(61.5)	0.193
	sometimes	356	(29.0)	387	(31.8)	
	always	73	(6.0)	82	(6.7)	
Sleep disturbance:	never	790	(64.4)	728	(59.8)	0.059
	<1	269	(21.9)	304	(25.0)	
	>=1	167	(13.6)	186	(15.3)	
Exercise-related wheeze/cou	gh‡	482	(39.3)	531	(43.6)	0.033
Aeroallergen-related wheeze	/cough	89	(7.3)	104	(8.5)	0.261
Food-related wheeze/cough		186	(15.2)	196	(16.1)	0.540
Other symptoms†						
Cough without colds		769	(62.7)	798	(65.5)	0.152
Cough at night		631	(51.5)	612	(50.2)	0.571
Frequency of colds:	<4	447	(36.5)	420	(34.5)	0.498
	4-6	461	(37.6)	484	(39.7)	
	>6	318	(25.9)	314	(25.8)	
Duration of colds (weeks):	<1	278	(22.7)	268	(22.0)	0.897
	1-2	731	(59.6)	737	(60.5)	
	>2	217	(17.7)	213	(17.5)	
Ear infection(s):	0	599	(48.9)	613	(50.3)	0.481
- \-/-	1	351	(28.6)	322	(26.4)	

	>1	276	(22.5)	283	(23.2)	
Nasal symptoms		536	(43.7)	569	(46.7)	0.143
Snoring		880	(71.8)	877	(72.0)	0.928
Eczema (ever)		533	(43.5)	548	(45.0)	0.464
Parental history						
Wheeze, asthma or						
bronchitis:	none	641	(52.3)	647	(53.1)	0.581
	father	204	(16.6)	178	(14.6)	
	mother	267	(21.8)	276	(22.7)	
	both	114	(9.3)	117	(9.6)	
Hay fever:	none	626	(51.1)	646	(53.0)	0.702
	father	200	(16.3)	199	(16.3)	
	mother	296	(24.1)	271	(22.2)	
	both	104	(8.5)	102	(8.4)	

* Fisher's exact test

† During the last 12 months

‡ Wheeze or cough with running, playing, laughing or crying

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Table E3. Sensitivity analysis I: Testing performance of main asthma prediction model in alternative study populations

		e criteria ear-olds	C	Outcome definition 5 yrs later							
Study population	Health care visit and any wheeze or chronic cough	Health care visit and any wheeze		Any wheeze and asthma medication	Any wheeze and ever doctor-diagnosed asthma	>4 episodes of wheeze and inhaled corticosteroids	N Total	n Outcome	(%)	Brier (scaled)	AUC*
P0 (used for main model)	\checkmark			✓			1226	345	(28.1)	0.23	0.78
P1 P2		· ·		√ √			769 697	285 272	(37.1) (39.0)	0.21 0.22	0.77 0.77
P3	\checkmark	v		v	✓		1239	331	(39.0) (26.7)	0.22	0.77
P4	\checkmark					\checkmark	1053	71	(6.7)	-0.51†	0.87

Baseline and outcome criteria refer to the past 12 months, if not otherwise stated

*Area under receiver operating characteristic curve

[†] The negative scaled Brier score is due to the large difference in the prevalence of the outcome in P0 and P4. A simple recalibration without changing the score would lead to a scaled Brier score of 0.24

Table E4. Sensitivity analysis II: Testing performance of newly developed asthma prediction models based on alternative study populations

		e criteria ear-olds		e defini s later	tion						
Study population	Health care visit and any wheeze or chronic cough	Health care visit and any wheeze Any wheeze	Any wheeze and asthma medication	Any wheeze and ever doctor-diagnosed asthma	>4 episodes of wheeze and inhaled corticosteroids	No. of binary predictors in the model	N Total	n Outcome	(%)	Brier (scaled)	AUC*
P0 (used for main model)	\checkmark		✓			22	1226	345	(28.1)	0.23	0.78
P1		√	√			25	769	285	(37.1)	0.22	0.77
P2	,	\checkmark	✓	,		23	697	272	(39.0)	0.23	0.78
P3	v			\checkmark	,	26	1239	331	(26.7)	0.26	0.81
P4	\checkmark				\checkmark	20	1053	71	(6.7)	0.28	0.89

Baseline and outcome criteria refer to the past 12 months, if not otherwise stated

*Area under receiver operating characteristic curve

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TABLE E5. Selected predictors in sensitivity analysis II and corresponding ORs

		Main model*	New models (alternative populations			
			P1†	P2‡	P3§	P4I
		Odds Ratio				
D	• • • • • • •	(OR)	OR	OR	OR	OR
Demographic and per	rinatal data					
Male		1.48	1.43	1.49	1.68	1.00
Age (years)	≥2	1.19	1.53	1.51	1.28	1.00
	3	1.00	1.00	1.01	1.06	0.95
Gestational age <37 we	eeks	1.11	1.13	1.00	1.16	1.00
Birth weight <2500 g		1.17	1.18	1.28	1.34	1.00
South Asian ethnicity ((versus white)	1.00	1.00	1.00	1.00	0.53
Wheeze-related symp	toms¶					
Current wheeze		1.18	1.00	1.00	1.59	1.46
Wheeze without colds		1.40	1.55	1.45	1.42	2.11
Frequency of attacks	≥1	1.00	1.00	1.00	1.05	1.00
	>3	1.65	1.53	1.60	1.37	1.16
	>12	1.00	1.00	1.00	1.00	2.10
Activity disturbance	any	1.28	1.30	1.25	1.28	1.49
5	moderate or a lot	1.16	1.31	1.17	1.14	1.00
	a lot	1.63	1.94	1.87	1.81	2.18
Shortness of breath	sometimes or always	1.98	1.90	1.91	1.84	2.06
	always	1.56	1.40	1.41	2.10	2.70
Sleep disturbance	$\geq 1/\text{week}$	1.00	1.00	1.00	1.10	1.00
Sheep and and and a	>1/week	1.00	1.00	1.00	1.00	1.20
Exercise-related wheez		1.26	1.00	1.15	1.40	1.27
Aeroallergen-related w	•	1.20	1.05	1.04	1.33	1.00
Food-related wheeze/c	-	1.00	1.03	1.04	0.97	1.00
Other symptoms¶						
Cough without colds		1.09	1.10	1.07	1.16	1.37
Cough at night		1.00	1.10	1.13	1.06	1.00
	>3	1.00	1.12	1.00	1.00	1.00
Frequency of colds	>5 >6	1.00	0.97	1.00	1.00	1.00
Duration of colds	~0	1.00	0.97	1.00	1.00	1.00
(weeks)	≥1	0.97	0.89	0.90	0.80	1.00
(weeks)	>2	1.00	1.00	1.00	1.00	1.00
Ear infection(s)	≥ 1	1.00	1.13	1.00	1.00	1.00
La micetion(s)	>1	1.00	1.00	1.00	1.00	1.00
Nasal symptoms	× 1	1.17	1.14	1.13	1.18	1.14
Snoring		1.00	1.00	1.00	1.10	1.00
Eczema (ever)		1.52	1.42	1.50	1.39	1.62
Dorontal history						
Parental history		1.00	1 1 4	1.07	1 45	1.07
Wheeze or bronchitis	mother or father	1.23	1.14	1.06	1.45	1.07
	mother or both	1.00	1.00	1.00	1.00	1.00
	both parents	1.26	1.57	1.36	1.39	2.02
Hay fever	mother or father	1.03	1.00	1.00	1.00	1.09
	mother or both	1.00	1.05	1.01	1.00	1.00

both parents	1.12	1.28	1.37	1.41	1.34	

Baseline and outcome criteria refer to the past 12 months, if not otherwise stated * Inclusion criteria: 1-3 year-olds with health care visit plus either wheeze or cough without colds or cough at night;

Outcome: Wheeze plus asthma medication at age 6-8 yrs

† Inclusion criterion: 1-3 year-olds with wheeze; Outcome: same as in main model

‡ Inclusion criteria: 1-3 year-olds with health care visit plus wheeze; Outcome: same as in main model

§ Inclusion criteria: same as in main model; Outcome: Current wheeze plus doctor's diagnosis of asthma (ever) at age 6-8 yrs

Inclusion criteria: same as in main model; Outcome: >4 episodes of wheeze and using inhaled corticosteroids

¶ During the last 12 months

**Wheeze or cough with running, playing, laughing or crying

98 99 100

		No. of binary predictors in	Brier scor	e (scaled)	AU	C*
Simplification approach		the model	before val	after val¶	before val	after val¶
Main model Rounded model†	no simplification	22	0.23	0.20	0.78	0.76
	factor 10	20	0.23	0.19	0.78	0.75
	factor 5	19	0.23	0.21	0.78	0.77
Reduced model	factor 3††	13	0.22	0.16	0.78	0.74
	first five predictors only‡	5	0.14	0.13	0.75	0.64
	first three predictors only§	3	0.12	0.11	0.73	0.60
Frequent wheeze only**		3	0.13	0.12	0.70	0.57

Table E6. Predictive performance of simplified versions of the main asthma prediction model

* Area under receiver operating characteristics curve

*: Using simplified regression coefficients of the model (regression coefficients of main model multiplied by 10, by 5 or by 3, respectively, and rounded to the next integer)

‡ Shortness of breath due to wheeze, frequent wheeze episodes (>3), wheeze without colds, activity disturbance due to wheeze; exercise-related wheeze/cough

§ Shortness of breath due to wheeze, frequent wheeze episodes (>3), wheeze without colds

Before internal validation: assessment using same sample as used to develop the model

¶ After internal validation: assessment using leave-one-out crossvalidation

** A 4-level variable coded as 3 binary dummy variables; analysis using logistic regression without penalization †† Preferred model

Table E7 . Performance measures of the prediction tool for different
cutoff-values (calculated in sample used to develop the tool without
crossvalidation)

ciossvalluation)					
Score-cutoff	Sensitivity	Specificity	PPV	NPV	LR+	LR-
0	>0.99	< 0.01	0.28	NA	1.00	*
1	>0.99	0.02	0.29	0.95	1.02	0.12
2	0.96	0.14	0.30	0.89	1.11	0.30
3	0.91	0.37	0.36	0.92	1.45	0.23
4	0.79	0.57	0.42	0.87	1.84	0.37
5	0.72	0.71	0.49	0.86	2.47	0.40
6	0.62	0.80	0.55	0.84	3.18	0.47
7	0.52	0.88	0.62	0.82	4.19	0.55
8	0.42	0.92	0.68	0.80	5.53	0.63
9	0.33	0.96	0.77	0.79	8.32	0.70
10	0.22	0.98	0.79	0.76	9.36	0.80
11	0.13	0.99	0.80	0.74	10.45	0.88
12	0.06	>0.99	0.83	0.73	12.77	0.95
13	0.02	>0.99	0.89	0.72	20.43	0.98
14	0.01	>0.99	>0.99	0.72	*	0.99
15	< 0.01	>0.99	NA	0.72	*	>0.99

PPV, positive predictive value; *NPV*, negative predictive value; *LR*+, likelihood ratio positive; *LR*-, likelihood ratio negative

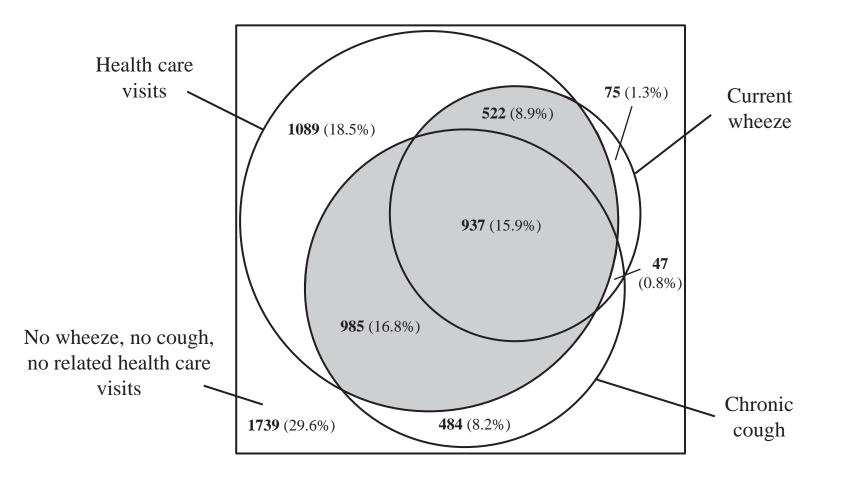
Sensitivity, Specificity, PPV, NPV: restricted to values between 0 and 1

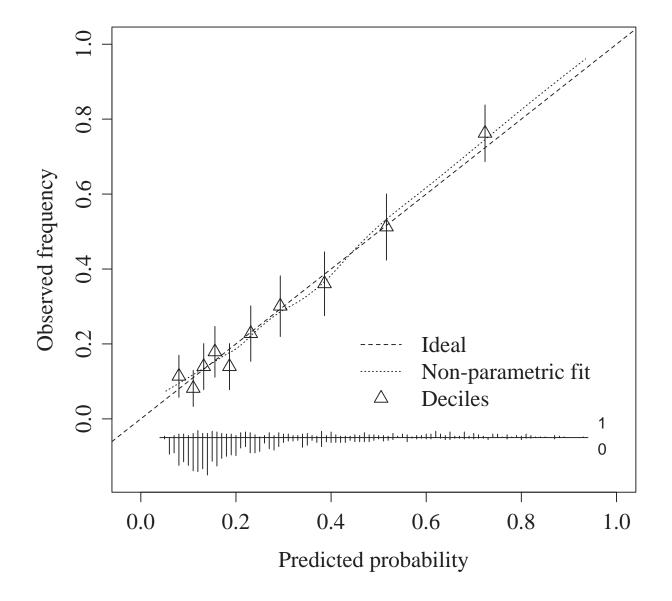
* Great uncertainty of estimate due to sensitivity and specificity close to 0 or 1

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Table E8. Comparison of percentage change in FEV₁ and MEF₅₀ after bronchodilator administration by questionnaire-based outcome definitions assessed at the same time

	administration by questionnaire-based outcome definitions assessed at the same time						
	Outcome criteria	Any as	thma	Moderately severe asthma			
		(current wheeze and asthma medication)		(>4 episodes of wheeze in the past 12 months			
				and inhaled corticosteroids)			
	Fulfilling	Yes	No	Yes	No		
	outcome criteria N _{FEV1}	111	340	30	389		
	Mean % change	5.46	2.59	9.10	2.76		
	in FEV_1 after	95%CI=[3.58,7.34]	95%CI=[1.96,3.21]	95%CI=[3.74,14.45]	95%CI=[2.15,3.38]		
	bronchodilator						
	administration						
	N _{MEF50}	109	334	29	382		
	Mean % change	16.66	10.65	18.60	11.21		
	in MEF ₅₀ after	95%CI=[12.80,20.53]	95%CI=[8.75,12.54]	95%CI=[9.75,27.46]	95%CI=[9.39,13.03]		
	bronchodilator						
102	administration	- 4 - m 1 - m					
103 104		atory volume in the 1st set $= -0.001$					
104 105	t-tests: any astnma:	$p_{\text{FEV1}} = <0.001; p_{\text{MEF50}} = 0.001$	JUS; moderately severe a	$p_{\text{FEV1}} < 0.001$; p_{ME}	EF50 = 0.039;		
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122	Fig E3. Origin	al questions used	as potential prec	lictive factors			
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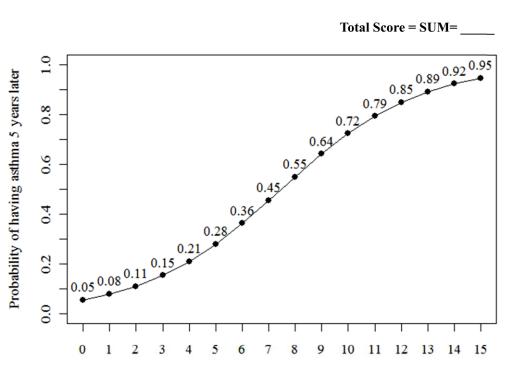




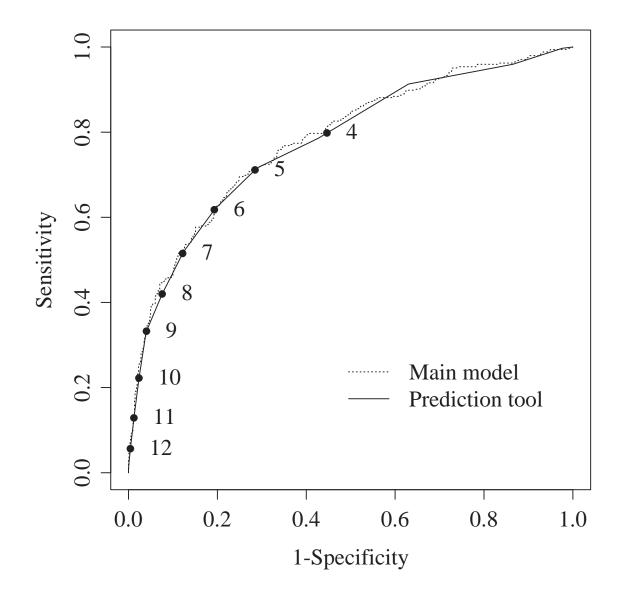
Asthma Prediction Tool

1.	What is the child's sex?	Female Male	
2.	How old is the child? (in years)	2	
3.	In the last 12 months, has the child had wheezing or whistling in the chest even without having a cold or flu?	No Yes	
4.	How many attacks of wheeze has the child had during the last 12 months?		
5.	In the last 12 months, how much did wheezing interfere with the child's daily activities?	No A little A lot	□ 0 □ 1 □ 2
6.	Do these wheezing attacks cause him/her to be short of breath? So	Never metimes Always	□ 0 □ 2 □ 3
7.	In the last 12 months, did exercise (playing, running) or emotions (laughing, crying or excitement) cause wheezing or coughing in the child?	No Yes	
8.	In the last 12 months, did contact with dust, grass, pets or other animals cause wheezing or coughing in the child?	No Yes	
9.	Has the child ever had eczema?	No Yes	
10.	Have the child's parents ever suffered from wheezing, asthma or bronchitis?	None Mother Father	









- 1. Has your child had wheezing or whistling in the chest in the last 12 months? yes no
- 2. Does your child usually have a cough apart from colds? yes

In the last 12 months, has your child had a dry cough at night, apart from a cough associated with a cold or a chest infection? yes no

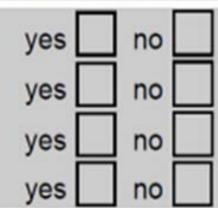
4. How often did your child see the GP for coughing or wheezing during the last 12 months? never once 2 - 3 times 4 - 6 times 7 or more times

5. In the last 12 months, has wheezing or asthma resulted in your child:

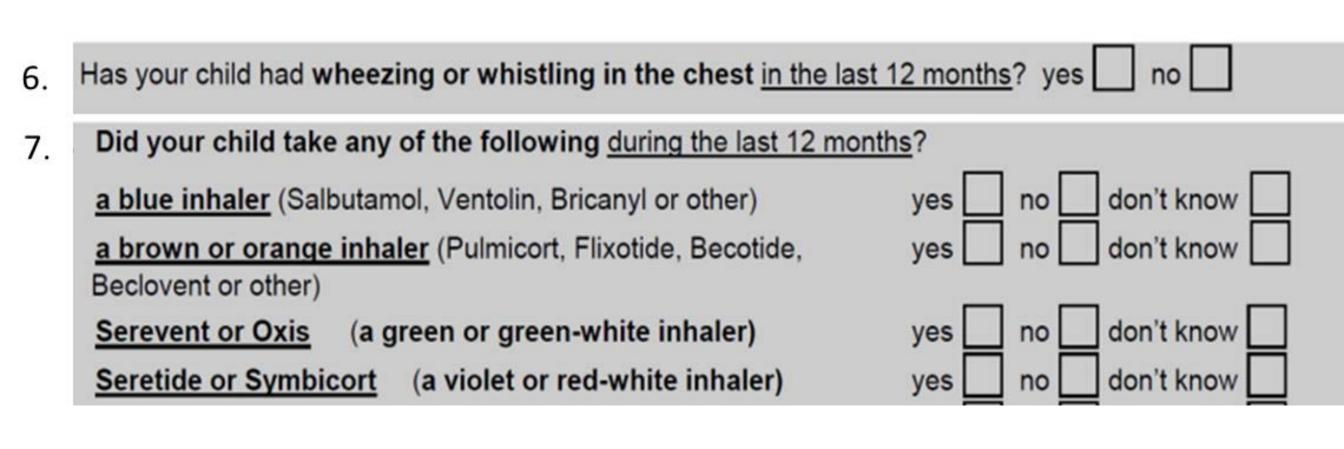
- being referred to a consultant in hospital
- being admitted to hospital

3.

- attending the casualty (A and E) department
- attending (or calling) the GP in an emergency



no



8.	Has your child had wheezing or whistling in the chest in the last 12 months? yes no
9.	In the last 12 months, has your child had wheezing or whistling in the chest even without having a cold or flu? yes no
10.	How many attacks of wheezing has your child had during the last 12 months? None 1 to 3 4 to 12 more than 12
11.	In the last 12 months, how much did wheezing interfere with your child's daily activities? not a all a little a moderate amount a lot
12.	Do these attacks cause him/her to be short of breath? yes, always yes, occasionally no, never
13.	In the last 12 months, how often, on average, has your child's sleep been disturbed due to wheezing? Never woken with wheezing less than one night per week one or more nights per week
14.	In the last 12 months did the following things cause wheezing in your child? • exercise (playing or running) yes no don't know • laughing, crying or excitement yes no don't know • contact with pets or other animals yes no don't know • food or drinks yes no don't know
15.	Does your child usually have a cough apart from colds? yes no
16.	. <u>In the last 12 months</u> , has your child had a dry cough at night , apart from a cough associated with a cold or a chest infection? yes no
17.	In the last 12 months, how many times has your child had a cold or flu? never 1 - 3 times 4 - 6 times 7 -10 times more than 10 times
18.	How long does a cold usually last in your child? less than 1 week 1 to 2 weeks 2 to 4 weeks more than 4 weeks
19.	In the past 12 months, has your child had ear infections? no, never yes, once yes, more than once
20.	In the past 12 months, has your child had a problem with sneezing, or a runny, or blocked nose when he/she did NOT have a cold or the flu? yes no
21.	. <u>Over the past 12 months</u> , has your child snored at night? yes no
22.	In the past 12 months, has your child had eczema? yes no
23.	 Has the child's father ever suffered from any of the following conditions? wheezing? yes no don't know asthma? yes no don't know bronchitis? yes no don't know hayfever? yes no don't know
24.	Has the child's mother ever suffered from any of the following conditions?
	wheezing? yes no don't know asthma? yes no don't know
	bronchitis? yes no don't know
	havfever? ves no don't know