

Identifying patients at risk for severe exacerbations of a validation of a multivariable prediction model

Thorax

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Citation Report

#	ARTICLE	IF	CITATIONS
1	The evidence on tiotropium bromide in asthma: from the rationale to the bedside. <i>Multidisciplinary Respiratory Medicine</i> , 2017, 12, 12.	0.6	21
2	The role of upper airway pathology as a co-morbidity in severe asthma. <i>Expert Review of Respiratory Medicine</i> , 2017, 11, 855-865.	1.0	42
3	Inflammatory and Comorbid Features of Patients with Severe Asthma and Frequent Exacerbations. <i>American Journal of Respiratory and Critical Care Medicine</i> , 2017, 195, 302-313.	2.5	346
4	Machine learning approaches to personalize early prediction of asthma exacerbations. <i>Annals of the New York Academy of Sciences</i> , 2017, 1387, 153-165.	1.8	138
5	Emerging Concepts in Evidence-Based Asthma Management. <i>Seminars in Respiratory and Critical Care Medicine</i> , 2018, 39, 082-90.	0.8	1
6	Exacerbations in Adults with Asthma: A Systematic Review and External Validation of Prediction Models. <i>Journal of Allergy and Clinical Immunology: in Practice</i> , 2018, 6, 1942-1952.e15.	2.0	49
7	Comorbid "treatable traits" in difficult asthma: Current evidence and clinical evaluation. <i>Allergy: European Journal of Allergy and Clinical Immunology</i> , 2018, 73, 1369-1382.	2.7	113
8	Performance of database-derived severe exacerbations and asthma control measures in asthma: responsiveness and predictive utility in a UK primary care database with linked questionnaire data. <i>Journal of Pragmatic and Observational Research</i> , 2018, Volume 9, 29-42.	1.1	18
9	Prospective predictors of exacerbation status in severe asthma over a 3-year follow-up. <i>Clinical and Experimental Allergy</i> , 2018, 48, 1137-1146.	1.4	48
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18	Development and validation of an asthma exacerbation prediction model using electronic health record (EHR) data. <i>Journal of Asthma</i> , 2020, 57, 1339-1346.	0.9	20

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20	The impact of comorbidities on severe asthma. <i>Current Opinion in Pulmonary Medicine</i> , 2020, 26, 47-55.	1.2	40
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32	Developing an ML pipeline for asthma and COPD: The case of a Dutch primary care service. <i>International Journal of Intelligent Systems</i> , 2021, 36, 6763-6790.	3.3	1
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40	Automatically Explaining Machine Learning Prediction Results on Asthma Hospital Visits in Patients With Asthma: Secondary Analysis. <i>JMIR Medical Informatics</i> , 2020, 8, e21965.	1.3	14
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57	Identifying asthma patients at high risk of exacerbation in a routine visit: A machine learning model. <i>Respiratory Medicine</i> , 2022, 198, 106866.	1.3	8
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63	Independent risk factors of asthma exacerbations: 3-year follow-up in a single-center prospective cohort study. <i>Annals of Translational Medicine</i> , 2022, 10, 1353-1353.	0.7	0
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