Accuracy of Administrative Database Algorithms for Hospitalized Pneumonia in Adults: a Systematic Review

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BACKGROUND: Administrative data algorithms (ADAs) to identify pneumonia cases are commonly used in the analysis of pneumonia burden, trends, etiology, processes of care, outcomes, health care utilization, cost, and response to preventative and therapeutic interventions. However, without a good understanding of the validity of ADAs for pneumonia case identification, an adequate appreciation of this literature is difficult. We systematically reviewed the quality and accuracy of published ADAs to identify adult hospitalized pneumonia cases.

METHODS: We reviewed the Medline, EMBase, and Cochrane Central databases through May 2020. All studies describing ADAs for adult hospitalized pneumonia and at least one accuracy statistic were included. Investigators independently extracted information about the sampling frame, reference standard, ADA composition, and ADA accuracy.

RESULTS: Thirteen studies involving 24 ADAs were analyzed. Compliance with a 38-item study-quality assessment tool ranged from 17 to 29 (median, 23; interquartile range [IQR], 20 to 25). Study setting, design, and ADA composition varied extensively. Inclusion criteria of most studies selected for high-risk populations and/or increased pneumonia likelihood. Reference standards with explicit criteria (clinical, laboratorial, and/or radiographic) were used in only 4 ADAs. Only 2 ADAs were validated (one internally and one externally). ADA positive predictive values ranged from 35.0 to 96.5% (median, 84.8%; IQR, 65.3 to 89.1%). However, these values are exaggerated for an unselected patient population because pneumonia prevalences in the study cohorts were very high (median, 66%; IQR, 46 to 86%). ADA sensitivities ranged from 31.3 to 97.8% (median, 65.1%; IQR 52.5-72.4).

DISCUSSION: ADAs for identification of adult pneumonia hospitalizations are highly heterogeneous, poorly validated, and at risk for misclassification bias. Greater standardization in reporting ADA accuracy is required in studies using pneumonia ADA for case identification so that results can be properly interpreted.

KEY WORDS: pneumonia; clinical coding; data accuracy; systematic reviews.

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INTRODUCTION

Pneumonia is a leading cause of morbidity and mortality worldwide. In Canada and the USA, pneumonia is the 6th and 8th most commonly cited cause of death,^{1,2} respectively, and the most common infection requiring hospitalization.^{3–6} In 2013, the annual cost of hospital treatment of pneumonia in the USA approached \$9.5 billion.⁷

Given the public health importance of pneumonia, research is required regarding its burden, trends, etiology, processes of care, outcomes, health care utilization, and cost along with the effect of preventative or therapeutic interventions on these parameters. To accomplish such wide-ranging analyses, researchers and policymakers often rely on health administrative database algorithms (ADAs) to identify pneumonia cases for their studies.⁸⁻¹⁰ ADAs are combinations of diagnostic/ procedural codes along with patient/hospital demographic information that are used to identify cases of specific diseases using health administrative data. The use of ADAs is common in medical research in general, and in particular in pneumonia research.¹¹ However, most pneumonia analyses using ADAs, even when published in the most influential medical literature and aimed at guiding policy-making and clinical practice, do not address the accuracy ADAs for pneumonia case identification.¹²⁻¹⁵ This is important because it is only with a good understanding of the validity of ADAs for pneumonia case identification that the end-users of this literature (i.e., physicians, policy-makers, researches) can fully appreciate the relevance of these analyses for their practice, policy-making, or research-planning. To our knowledge, no systematic review of the literature has been done regarding the accuracy of ADAs for pneumonia. Thus, the objective of this study was to systematically identify and evaluate the quality of all published ADAs for adult hospitalized pneumonia and assess the overall performance of ADAs for identification of cases with this condition.

METHODS

Data Sources and Search Strategy

With the assistance of a health information specialist, we developed search strategies (Appendix 1) to interrogate three literature databases (Medline, 1946 to June 09, 2020; EMBase, 1947 to June 09, 2020; Cochrane Central, May, 2020) for studies that measured the accuracy of ADAs for the identification of adults assessed in the emergency room or admitted to hospital with pneumonia. Key search terms included "administrative data", "claims data", "coding", "billing data", "international classification of diseases", "ICD", "registry", and "medical records," among others. We did not set search restrictions on publication date, language, or completion status. The references of all selected studies were reviewed to identify additional relevant literature that might have been initially missed.

Study Selection and Data Abstraction

To be included in our review, studies had to be published in peer-reviewed journals and convey in their title or abstract that the accuracy of a pneumonia ADA (or a broader infectious diagnosis that could include pneumonia [e.g., "sepsis"]) was measured. The authors reviewed the relevance of all citations returned by our search strategy in duplicate by independently reviewing their titles and abstracts. Discrepancies were resolved by consensus.

Relevant articles were retrieved and reviewed in full-text also in duplicate and independently by the authors to determine final inclusion. To be included, studies had to describe at least one ADA for adult pneumonia requiring hospitalization and its accuracy. The description of the pneumonia ADA had to be detailed enough that it could be replicated by other researchers. We limited our analysis to adults because this is the most commonly studied demographic and because pediatric pneumonias can be very distinct clinically. We focused on cases requiring hospitalization because such cases are the most common pneumonia type studied with ADAs and their severity, presentation, and outcomes are generally distinct from cases assessed and treated in the community. We excluded ADAs that focused on etiology-typified pneumonias (i.e., pneumococcal pneumonia, atypical pneumonia, etc.) because the microbiological cause of most pneumonias is not usually determined with ADAs. We also excluded studies that focused upon hospital-acquired pneumonia because this disease is distinct from pneumonias acquired in the community by virtue of well-defined risk factors, etiologies, management, and outcomes.¹⁶ Finally, studies had to report at least one of the following accuracy statistics: positive predictive value (PPV), negative predictive value (NPV), sensitivity, specificity, or positive/negative likelihood ratio.

We used a structured abstraction form (available from researchers upon request) to independently abstract relevant data from the included studies. When more than one eligible pneumonia ADA was presented in a study, all were included in our assessment. Discrepancies were solved by consensus.

Analysis

The methodological and reporting quality of the studies was assessed using criteria from Benchimol et al.¹⁷ To simplify reporting, we summed for each study the number of criteria that were met (out of a maximum of 38). For each study, we extracted the total sample size (defined as the number of people to which the reference standard was applied) and the number of patients classified as having true pneumonia using the study's reference standard to estimate the study pneumonia prevalence. For each ADA, we abstracted the number of patients to which the ADA was applied, the number of these patients meeting criteria for the ADA, and the number of patients categorized as true pneumonia cases as per the study reference standard. With this information, we constructed $2 \times$ 2 contingency tables to calculate (when possible) the ADA's sensitivity, specificity, PPV, NPV, positive likelihood ratio (LR+), and negative likelihood ratio (LR-). Prevalence measurements were rounded to the unit whereas accuracy measurements were rounded to one decimal point.

RESULTS

After removal of duplicates, 896 citations were screened for study relevance (Fig. 1). Of these, 64 were selected for full-text review. Cross-referencing identified 4 additional articles of interest. In total, 13 studies were included in our final review and analysis.^{8,10,18–28}

Study Description

Studies varied extensively in their design and characteristics (Table 1). Seven studies (54%) were completed in the USA, 8,18,19,22,23,26,27 4 (31%) in Europe, 10,20,21,25 and 1 each in Australia²⁴ and Canada.²⁸

Five studies^{18,21–23,26} involved state- or nation-wide hospital systems whereas the rest involved oligo- or monoinstitutional systems. Six studies^{10,18-20,24,26} included cases that were solely from 2000 onwards whereas the rest included cases that partially $^{8,21-23}$ or fully 25,27,28 pre-dated that time. The inclusion criteria of 9 studies $^{10,18,20-24,26-28}$ (70%) resulted in populations with a higher risk of pneumonia by virtue of increased age,^{18,24} admission to the intensive care units,^{19,25} high-risk of acute respiratory distress syndrome,¹⁹ being coded with cancer²⁰ or rheumatoid arthritis,²³ being coded with pneumonia in the hospitalization of interest²² or in a pre-dated outpatient visit,²¹ or having hospitalizations of \geq 3-day duration.²⁶ In 5 of these studies,^{18,20,21,23,26} the inclusion criteria included the actual ADA whose accuracy was to be measured. Overall, only 1 study included a population that was not limited to patients with a pneumonia ADA or with higher risk for pneumonia.8



Figure 1 Flow chart of literature search and study selection.

Study samples were either all-inclusive or randomly generated and ranged in size from 23 to 10,748 (mean 2186, SD 950; median, 293). In the study having the most generalizable inclusion criteria (Aronsky et al.⁸), pneumonia prevalence was 2%. Otherwise, pneumonia prevalence in the other studies ranged from 5 to 100%. Overall, pneumonia prevalence in the 13 studies analyzed ranged from 2 to 100% (median, 66%; IQR, 46 to 86%).

Most of the reference standards used to define pneumonia cases relied upon physician notation of a clinical diagnosis or suspicion of pneumonia in the medical records. Only 4 studies (31%) used reference standards with explicit objective criteria (clinical, laboratory, and/or radiographic).^{10,18,20,26} Nine studies (69%) stated that the application of their reference standards was blinded^{8,10,19,20,23–25,27,28}; however, the inclusion criteria in 4 of these studies^{20,23,27,28} required that patients be coded for pneumonia or have a recorded clinical diagnosis of this condition, making effective blinding difficult.

Compliance with the 38 quality items proposed by Benchimol et al.¹⁷ varied from 17 to 29 (mean, 23; SD 9; median, 23) (Appendix 2).

Characteristics and Accuracy of Pneumonia ADAs

The 13 studies included in the review measured the accuracy of 24 pneumonia ADAs (Appendix 3). Sixteen ADAs (67%) used the International Classification of Disease (ICD)-9 coding system,^{8,10,18,19,22,23,25–28} 4 (17%) used ICD-10,^{20,24,25} 3 (13%) used the Diagnosis-Related Group (DRG) system,^{8,27} and 1 (4%) used the Read coding system (Table 2).²¹ Twentythree ADAs^{8,10,18–20,22–28} (96%) used codes assigned at the hospitalization of interest whereas the remaining ADA²¹ used codes from a community information system. ADA construction varied extensively, even if they used the same coding system (Appendix 3). For example, while most ICD-9-based ADAs used the codes 480.x-486.x for pneumonia identification, there was large variation among them with regard to criteria for the positioning of these codes in discharge diagnoses lists (e.g., any position vs. primary position only), the inclusion of or combination with other codes that could also represent pneumonia, and the addition of other criteria beyond the coding system of interest (e.g., radiographic criteria, minimum hospitalization length). Only 1 ADA (based on ICD-9 coding) was validated internally.¹⁹ Only 1 ADA (based on DRG coding) was validated externally (although the studies used distinct sampling frameworks and reference standards).^{8,27}

Pneumonia ADA accuracy varied extensively (Table 2). PPV was the most common accuracy statistic, being measured in 23 (96%) of the 24 ADAs. Values ranged from 35.0 to 96.5% (median 84.8%, IQR 65.3–89.1%). NPV was measured in 15 (63%) of the ADAs, with values ranging from 65.2 to 99.4% (median 98.0%, IQR 90.0–99.8%). Sensitivity was measured in 17 (71%) ADAs with values ranging from 31.3 to 97.8% (median 65.1%, IQR 52.5–72.4). Specificity was measured in 15 ADAs with values ranging from 64.5 to 99.8% (median 98.0%, IQR 91.2–99.8). The LR+ and LR– of the ADAs could be calculated in these latter studies and we used

Studies
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Table 1

Study (year)	Country	Sampling fran	Je		Sample			Reference	standard			Benchimol
		Scope	Time period	Inclusion criteria	Selection	N	Pneumonia prevalence*	Explicit criteria	Application	Blinded to code	Multiple reviewers	score (max. 38) [†]
Wiese (2018) ¹⁸	USA	State-wide health care	2008– 2012	Age > 50 years, MedicAid eligible and hospitalization	Random	340	96%	Yes	Retrospective	No	No	20
Ahmed (2014) ¹⁹	NSA	County-wide health care	2010	Admission to ICU and high risk of ARDS	All	1443 [‡]	8%	No	Prospective	Yes	Yes	26
Holland-Bill (2014) ²⁰	Denmark	Single center	2006– 2010	Coded with cancer and hospitalization coded with	Random	93	89%	Yes	Retrospective	Yes	No	22
Meropol (2012) ²¹	UK	Nation-wide database	1985 - 2006	Hospitalization within 30 days of an outpatient visit coded	Random	59	86%	No	Retrospective	No	No	17
Yu (2011) ²²	USA	State-wide health care	1997 - 2005	win pneumonia Hospitalization coded with pneumonia	All	3991	%09	No	Retrospective	No	No	27
Grijalva (2008) ²³	NSA	State-wide health care network	1995– 2004	MedicAid eligible, coded with rheumatoid arthritis and hospitalization coded with	Random	161	84%	No	Retrospective	Yes	Yes	20
Skull (2008) ²⁴	Australia	Two hospitals	2000– 2002	Age ≥65 years, hospitalization code ≥65 vears, hospitalization code with pneumonia (cases)	All	3343	46%	No	Retrospective	Yes	No	29
Gedeborg	Sweden	Single center	1994-	Admission to ICU	All	7615	5%	No	Prospective	Yes	No	24
van de Garde	Netherlands	Seven	2000- 2000-	Enrolled in pneumonia clinical	All	293	100%	Yes	Prospective	Yes	No	21
(2007) Schneeweiss (2007) ²⁶	NSA	State-wide health care	2001 - 2001 - 2004	Veteran, hospitalization for 23 days, hospitalization coded	All	23	70%	Yes	Retrospective	No	No	25
Aronsky	USA	Single center	1999–	Admission to ED	All	10,748	2%	No	Retrospective	Yes	Yes	25
Whittle Wittle	USA	Single center	2000 1989–	Hospitalization coded with	Random	144	65%	No	Retrospective	Yes	Yes	23
Marrie (1987) ²⁸	Canada	Single center	1984	Hospitalization with clinical diagnosis of pneumonia or coded with pneumonia	All	159	66%	No	Prospective	Yes	No	20
USA refers to Un quality items prop	ited States. IC vosed by Benc.	U refers to intensi himol et al. ¹⁷ (for	ive care un details of	it. ARDS refers to acute respiratory the performance of each study in ii	distress sym	drome. *A: ulity items,	s per the applica please see Appe	tion of the r_{i} and r_{i} the T_{i}	eference standard lidation sample f	. [†] Number o or this study	of positive item	s out of the 38

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Table

Study*†	Code system	Validation	Number of ca +	ses algorithm	Number of ca –	ses algorithm	Algorithm a	ccuracy					Probability of pneumonia [‡]	
			Pneumonia +	Pneumonia -	Pneumonia +	Pneumonia -	Sensitivity	Specificity	Add	NPV	LR+	LR -	Algorithm +	Algorithm -
Wiese ¹⁸	ICD-9-	No	328	12	0	0	,		96.5%					
Ahmed ¹⁹	ICD-9-	$\mathrm{Yes}^{\mathrm{II}}$	94	26	28	1295	77.0%	98.0%	78.3%	97.9%	38.5	0.2	49.7%	0.5%
Holland-Bill ²⁰	ICD-10-	No	83	10	0	0		ı	89.2%	I		ı		
Meropol ²¹ Yu ²²	Read ICD-9-	No No	51 325	8 33	0 189	0 407	- 63.2%	- 92.5%	$86.4\% \\ 90.8\%$	- 68.3 <i>%</i>	- 8.4	- 0.4	- 17.7%	$\frac{1}{1.0\%}$
	CM (a) ICD-9-	No	1214	150	651	870	65.1%	85.3%	89.0%	57.2%	4.4	0.4	10.1%	1.0%
Grijalva ²³	ICD-9-	No	135	26	0	0			83.9%	I	I	ı		
	ICD-9-	No	103	5	0	0			95.4%	I	I	ı		
	ICD-9-	No	32	21	0	0			60.4%	I	I			
Skull ²⁴	ICD-10-	No	1509	644	34	1156	97.8%	64.2%	70.1%	97.1%	2.7	0.03	6.5%	0.1%
Gedeborg ²⁵	ICD-9-	No	89	165	96	3831	48.1%	95.9%	35.0%	97.6%	11.7	0.5	23.1%	1.3%
	ICD-9-	No	40	62	88	3991	31.3%	98.5%	39.2%	97.8%	20.9	0.7	34.9%	1.8%
	ICD10	No	116	135	105	3078	52.5%	95.8%	46.2%	96.7%	12.5	0.5	24.3%	1.3%
	ICD10-	No	71	45	115	3203	38.2%	98.6%	61.0%	96.5%	27.3	0.6	41.2%	1.5%
van de Conde ¹⁰	ICD-9-	No	212	0	81	0	72.4%	ı	ı	ı	ı	ı	ı	ı
Schneeweiss ²⁶	ICD-9-	No	16	L	0	0		ı	69.6%	ı	ı			
Aronsky ⁸	ICD-9-	No	109	20	06	10,529	54.8%	99.8%	84.5%	99.2%	274.0	0.5	87.5%	1.3%
	ICD-9-	No	136	23	63	10,526	68.3%	99.8%	85.5%	99.4%	341.5	0.3	89.8%	0.8%
	ICD-9-	No	139	25	60	10,524	69.8%	99.8%	84.8%	99.4%	349.0	0.3	89.9%	0.8%
Whittle ²⁷	DRG (d) DRG (e) DRG (e)	No Yes [§]	89 124 79	13 21 7	110 75 15	10,536 10,528 43	44.7% 62.3% 84.0%	99.9% 99.8% 86.0%	87.3% 85.5% 91.9%	99.0% 99.3% 74.1%	447.0 311.5 6	0.6 0.4 0.2	92.0% 88.9% 13.3%	1.5% 1.0% 0.5%
ĉ	CM (a) DRG (b)	No	70	S - S	24	45	74.5%	90.0%	93.3%	65.2%	7.5	0.3	16.1%	0.8%
Marrie ²⁰	ICD-9- CM	No	73	54	32	0	69.5%		57.5%	I	ı			
*For detailed de: 2.5%. [¶] Data sho	scription of wn correspo	the algorithms, vads to the inte	please refer to A rnal validation a	Appendix 3. $^{\dagger}Lette$	ers in (), if preser udv. [§] This algori	ıt, corresponded ithm is considere	with the same of external vali	letters in Appe dation of DRC	ndix 3. [‡] As 3-based als	suming a zorithm of	true prevo [°] Whittle's	ilence o studv	f pneumonia in	population is

them to estimate the pneumonia probability in patients of a hypothetical population with a true prevalence of pneumonia of 2% (the prevalence measured in the study by Aronsky et al.,⁸ the most generalizable sampling frame of the included studies). In patients who were ADA positive, pneumonia probabilities ranged from 4.3 to 90.1% (median 19.3%; IQR 8.2–64.4%). In contrast, in patients who were pneumonia ADA negative, pneumonia was essentially excluded with disease probabilities ranging from 0 to 1.4% (median 0.8%; IQR 0.4–1.0%).

DISCUSSION

Our systematic review identified 24 published pneumonia ADAs. Despite having the same goal (i.e., to identify adults admitted to hospital with pneumonia), these ADAs varied extensively in their construction, the patient population to which they were applied, the prevalence of pneumonia in these populations, the rigor of the reference standards used to define pneumonia cases, their overall methodological quality, and their accuracy measurements. Of the 24 ADAs identified, only 1 was internally validated¹⁹ and only 1 was externally validated.^{8,27}

Our systematic review highlights important points that researchers, health professionals, and policy-makers should consider when studying pneumonia using ADAs and/or assessing literature that uses such strategy. First, the PPV and NPV should be used cautiously to assess the performance of these ADAs for the identification of pneumonia cases. PPV and NPV values are highly dependent on disease prevalence; therefore, the reported values are only generalizable to populations having similar disease prevalence.^{29,30} Many of the studies in our review included populations with clinical characteristics and/or sampling frames that notably increased pneumonia prevalence (Table 1). Thus, the PPVs and/or NPVs for the ADAs reported in these studies are not generalizable to less selected populations having a lower prevalence of pneumonia. In fact, only one study (Aronsky et al.)⁸ in our review involved a population that could be considered representative of unselected adults seeking acute medical care in the emergency department. To avoid biased statistics from varying disease prevalence, one should use sensitivity, specificity, and their corresponding LRs to evaluate ADA performance.^{29,30} Unfortunately, these statistics are much less commonly reported (Table 2). Second, the median sensitivity from the 17 ADAs in which it was measured was 65.1%, suggesting that analyses using ADAs likely miss a significant proportion of pneumonias cases. Third, in the absence of prospectively applied explicit to determine pneumonia status (used in only 4 of the studies included in this review),^{10,19,25,28} studies commonly used manual chart review to determine whether or not physicians noted a clinical impression of pneumonia as the reference standard. This strategy makes the reliability and reproducibility of these studies difficult because the suspicion

and diagnosis of pneumonia in clinical settings frequently relies on the subjective interpretation of the probability that unspecific symptomatology and/or clinical, laboratorial, and/ or imaging findings, together or in isolation, are explained by a pneumonic process. Moreover, clinicians may be under administrative pressure to document a diagnosis even when the degree of certainty for such diagnosis is suboptimal. Fourth, ADAs frequently involved criteria that were extraneous to the diagnostic codes of interest (Table 1), mostly to increase the probability of pneumonia. If other researchers use the cited diagnostic codes (Appendix 3) in isolation (i.e., without the other additional criteria), the originally reported accuracy statistics will not apply. *Finally*, only 1 pneumonia ADA⁸ was an external validation of a previously published pneumonia ADA.²⁷ Therefore, the generalizability of the other 23 ADAs is uncertain. The performance of most predictive models (which ADAs essentially are) deteriorates when they are applied to an external population. Such performance deterioration usually occurs when ADA developers overfit their original model to maximize its accuracy. Hence, external validation of ADAs is key to truly understand their accuracies and validity for external use.

Despite the above limitations of ADAs for pneumonia, administrative data remain a valuable resource for timely and practical analyses of this condition at large scale. However, to improve the interpretation of the validity and applicability of these analyses, several strategies should be applied. The most robust but also complex strategy would be the development of a contemporary, accurate, and reproducible pneumonia ADA that allows for the calculation of precise ADA likelihood ratios and that can then be used for the estimation of pneumonia probability in any other population in which pneumonia prevalence can be measured. Developing such an ADA should include unselected populations of hospitalized patients, involve multiple centers in broad jurisdictions, use explicit reference standard criteria to determine pneumonia status, and should be designed to accurately measure pneumonia prevalence in the study sampling frame. The resulting pneumonia ADA and its accuracy must then be validated in external populations. Ideally, this "reference" ADA should return a probability of pneumonia instead of the dichotomous outcome (pneumonia present/not present) as this would also allow for the use of statistical methods like bootstrap imputation to accommodate for misclassification bias in future studies using such ADA.^{31–33} Clearly, the development of such ADA would be logistically challenging and consume large amounts of time and resources. An alternative and more practical approach is that all pneumonia ADA studies report, at a minimum, the details of its ADA in enough detail that it can be replicated, disclose whether a reference standard was used, describe the details of the said reference standard, and report the pneumonia prevalence and ADA accuracy measured in a representative sample of their study population. Preferably, reference standards in such investigations should be based in objective clinical, radiographic, and/or laboratorial criteria. End-users of these reports can then use this information to more appropriately interpret the results of these analyses.

Limitations of our review include its restriction to articles published in peer-reviewed journals, articles whose objectives (as stated in their title or abstract) included the assessment of ADAs' accuracy, and articles that explicitly reported at least one accuracy metric of interest. These restrictions may have resulted in missed relevant information outside the peer-reviewed literature or articles that focused on accuracy of ADAs only tangentially. Also, as administrative database studies are not always accurately indexed in literature, relevant articles may not have been captured by our literature search. Information about the mode of data extraction (manual vs. automated) for each ADA was not consistently reported in the reviewed papers and we could not assess the potential impact of this variable on ADA accuracy (for example, manual extraction could have introduced further bias or error from the person or people carrying the data extraction). Most of the ADAs in our review were based on the ICD (9th and 10th versions) diagnosis coding system. Therefore, whether ADAs based on other coding systems show similar accuracy performance as ICD-based ADAs remains unclear. Also, our review did not account for differences in the source of administrative data (local hospital administration, governmental offices, insurance claims, etc.) which could also affect individual performances of ADAs. Given the remarkable heterogeneity in the reviewed literature (in ADA design, the coding systems used, the use and design of gold standards, and the accuracy metrics reported, among others), we could not perform any meaningful meta-analysis of the data. Nonetheless, such heterogeneity is a principal finding of our review and it stresses the need for more standardized reporting of the accuracy of ADAs in all analyses that use such strategy for pneumonia case identification.

CONCLUSION

Published studies of ADAs for identification of hospitalized pneumonia cases vary widely in their study setting, study design, ADA composition, use and nature of reference standards, and overall methodological quality. Most of the ADA accuracy statistics derived from these studies were obtained from analyses of populations selected for their high likelihood of pneumonia, and therefore their applicability to less selected populations is questionable. Moving forward, greater standardization in the reporting of ADA accuracy is required in studies that use pneumonia ADAs for case identification so that the end-users of this literature can properly interpret the results of these analyses. **Corresponding Author:** Vicente F. Corrales-Medina, MD, MSc; The Ottawa Hospital Civic Campus, Ottawa, Ontario, Canada (e-mail: vcorrales@toh.ca).

Authors' Contribution V.F.C-M and C.v.W equally participated in the study design, data gathering, performance of analyses, interpretation of results, and writing of the manuscript. Both authors approved the final version of the manuscript and take responsibility for its integrity and accuracy.

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Compliance with Ethical Standards:

Conflict of Interest: The authors declare that they do not have a conflict of interest.

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