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Evidence Based Pathology

Pathologists should probably forget about kappa. Percent agreement, diagnostic specificity and related metrics provide more clinically applicable measures of interobserver variability



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ABSTRACT

Kappa statistics have been widely used in the pathology literature to compare interobserver diagnostic variability (IOV) among different pathologists but there has been limited discussion about the clinical significance of kappa scores. Five representative and recent pathology papers were queried using clinically relevant specific questions to learn how IOV was evaluated and how the clinical applicability of results was interpreted. The papers supported our anecdotal impression that pathologists usually assess IOV using Cohen's or Fleiss' kappa statistics and interpret the results using some variation of the scale proposed by Landis and Koch. The papers did not cite or propose specific guidelines to comment on the clinical applicability of results. The solutions proposed to decrease IOV included the development of better diagnostic criteria and additional educational efforts, but the possibility that the entities themselves represented a continuum of morphologic findings rather than distinct diagnostic categories was not considered in any of the studies.

A dataset from a previous study of IOV reported by Thunnissen et al. was recalculated to estimate percent agreement among 19 international lung pathologists for the diagnosis of 74 challenging lung neuroendocrine neoplasms. Kappa scores and diagnostic sensitivity, specificity, positive and negative predictive values were calculated using the majority consensus diagnosis for each case as the gold reference diagnosis for that case. Diagnostic specificity estimates among multiple pathologists were > 90%, although kappa scores were considerably more variable. We explain why kappa scores are of limited clinical applicability in pathology and propose the use of positive and negative percent agreement and diagnostic specificity against a gold reference diagnosis to evaluate IOV among two and multiple raters, respectively.

1. Introduction

The results and conclusions of many studies in Pathology are considered clinically valid only after significant differences in prognosis, response to treatment and/or other dependent variables have been demonstrated using appropriate statistical tests [1]. The reliability of these conclusions depends on whether the independent variables being studied, such as diagnostic categories, growth patterns, immunophenotypes, and other features can be assessed in a consistent and reproducible manner. However, multiple studies have demonstrated considerable interobserver variability (IOV) and sometimes even intraobserver variability, that reflect the subjective interpretation of microscopic features and other diagnostic or prognostic variables [2-16]. This problem raises questions about the clinical validity and applicability of conclusions drawn from studies showing prognostic differences among entities that can be diagnosed variably by different pathologists [17-20]. For example, a study evaluating the effect of IOV in the differential diagnosis between usual interstitial pneumonia (UIP) and nonspecific interstitial pneumonia (NSIP) showed that changing diagnostic labels in a manner that simulated IOV in as few as 10% of the cases changed the statistical significance of the prognostic differences estimated in the selected literature [21].

Percent agreement and kappa are the metrics most commonly used in pathology to assess interrater agreement in the interpretation of diagnoses, immunohistochemical results and other test results by two or more observers. Positive and negative percent agreement are the

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simplest metrics designed to test for interrater reliability, but they do not consider the prevalence of the entities being rated and/or the possibility that certain agreements can be the result of chance [22]. For example, if the prevalence of tumor A is twice the prevalence of tumor B, some observers are more likely to favor a tumor A diagnosis. Selected investigators have suggested the use of 80% as the minimum interrater agreement level that is considered acceptable for most studies [23].

Jacob Cohen recognized in 1960 that percent agreement is a somewhat unreliable tool by which to assess interrater reliability in psychology and introduced the concept of kappa coefficient to measure the proportion of interrater agreement beyond chance in the interpretation of qualitative, categorical or nominal observations by two observers [24]. However, statisticians such as de Vet et al., have proposed that positive and negative percent agreement are better metrics than Cohen's kappa to compare diagnoses rendered by two raters [22]. Various other coefficients such as Fleiss' kappa (for 3 or more raters), tetrachoric (for dichotomous data and 2 raters), Pearson R, Spearman, Rho, Krippendorff's alpha and other correlation coefficients were proposed to evaluate interrater agreements among 2 or more observers, depending on the particular situation [2,4-6,9,11,13,23,25,26]. Fleiss' kappa, frequently used in the pathology literature, has been applied to IOV studies that compare the observations made by three or more observers [23,25-28].

Landis and Koch proposed a qualitative scale for the interpretation of kappa coefficients in 1977. Their scale has six levels in which kappa coefficients that are < 0 interpreted as no agreement, 0–0.20 as slight agreement, 0.21–0.40 as fair agreement, 0.41–0.60 as moderate agreement, 0.61–80 as substantial agreement and 0.81–1.0 as almost perfect agreement [9,11,13,18-21,27,29-33]. The use of this scale has been controversial in the psychometric literature and some authors have suggested that conclusions supported by kappa values < 0.67 should be discounted, that conclusions supported by kappa values ranging between 0.67 and 0.80 should be only tentatively accepted and that only conclusions supported by kappa values > 0.80 should be considered as definitive [27].

The scale proposed by Landis and Koch has been widely used in pathology without sufficient discussion as to its applicability for the interpretation of IOV in pathology. For example, is UIP a different clinico-pathologic entity than NSIP that can be diagnosed correctly by some pathologists and incorrectly by others, or are both part of a single clinico-pathologic continuum, in which case each pathologist offers an opinion in a situation where no one is certain about who has issued an accurate diagnosis? How should patients and their clinicians factor this uncertainty into their treatment decisions? Should clinico-pathologic entities that can only be diagnosed with fair, moderate or even substantial agreement be recognized as distinct from each other based on prognostic differences shown in retrospective observational studies or should clinico-pathologic entities only be accepted as distinct if most pathologists can diagnose them consistently and with accuracy similar to that expected for other laboratory tests?

We reviewed the IOV results from a few studies recently published in the lung cancer pathology literature to assess how the authors evaluated kappa values in the conclusions. Metrics commonly used to assess test accuracy in laboratory medicine were also applied to the dataset from one study to evaluate whether the study conclusions were supported when these metrics were applied.

2. Materials and methods

Using a previously described evidence-based approach, the specific questions listed in Table I were formulated to ascertain how pathologists currently assess and interpret IOV in selected problematic areas [34-36]. Four questions were designed to gather an anecdotal impression about the methods being used to assess IOV, the scale being used to interpret the results as clinically relevant, the minimum quantitative levels being used to conclude that IOV would not pose

Table I

Questions on the evaluation and interpretation of interobserver agreement levels in the recent pulmonary pathology literature.

Which method was used to evaluate interobserver agreement levels? What scale, if any, was used to interpret the results?

- Did the study define the minimum quantitative or qualitative level used to conclude that IOV would not pose a significant clinical problem in the interpretation of results?
- What solutions were offered to improve agreement levels?
- In instances where agreement levels were considered as problematic did the authors conclude that pathologists' perceptions or the definitions of the dependent variables (e.g. classification of tumor into subtypes, characteristics of immunostains, criteria being used to evaluate for stromal invasion) were the root cause of disagreements?
- In instances where agreement levels were less than optimal, did the study discuss whether the possibility that conclusions previously reported in the literature were or might be biased by the use of dependent variables that were not well defined?

significant clinical problems and the solutions being offered to decrease IOV. Two additional queries were formulated to gain insight into whether pathologists considered themselves or the criteria they were using to define the dependent variables responsible for the IOV in instances where agreement was suboptimal and to investigate whether pathologists suggested that the conclusions previously reported were or might have been biased by the use of dependent variables that could not be identified with acceptable accuracy by different pathologists [37,38]. Answers to these questions were collected from five arbitrarily selected papers that evaluated IOV in a variety of problematic issues in pulmonary cancer pathology [39-43]. These studies investigated reproducibility in each of the following: differential diagnosis of small cell carcinoma, scoring programmed cell death ligand-1 (PDL-1), differential diagnosis between multiple primary lung adenocarcinomas and intrapulmonary metastases, classification of small lung adenocarcinomas into adenocarcinoma in-situ, minimally invasive adenocarcinoma and invasive carcinoma and evaluation of risk of malignancy on cytology specimens.

Metrics commonly applied to assess accuracy in laboratory tests were applied to a dataset from 74 cases from a previous study by Thunnissen et al. [39] using kappa statistics to determine whether the use of immunohistochemistry improved the diagnosis of small cell lung cancer. Briefly, in their study whole slide digital images (WSI) selected from difficult cases had been circulated among 19 pulmonary pathologists from China, Japan, United States, Australia, Argentina, Italy, United Kingdom, and Germany in a manner that closely resembled actual clinical practice [39]. Each participant was provided with 3 opportunities to render "individual diagnostic opinions", as they studied each tumor stained with hematoxylin and eosin (H&E) and two successive sets of immunostains per case with observer selected or all available immunostains, respectively ("first", "second" and "third" level "individual diagnostic opinions"). Agreement levels were estimated with kappa statistics and the results interpreted using the Landis and Koch scale [39]. For the current illustration we categorized the "third level" "individual diagnostic opinions" diagnosing cases as small cell carcinoma, large cell neuroendocrine carcinoma, and typical or atypical carcinoid into true and false positive and true and false negative results, using the procedure shown in Table II and the majority consensus of the "third level individual diagnostic opinions" for each case as the gold standard diagnosis for that case. Sensitivity, specificity, positive predictive values (PPV) and negative predictive values (NPV) were calculated using MedCalc software (Ostend, Belgium). "Individual diagnostic opinions" diagnosing cases as non-Hodgkin lymphoma, basaloid squamous cell carcinoma or small round cell sarcoma were excluded because there were fewer than 3 cases in each of these categories.

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Step 5: Calculate for each of the 4 diagnostic possibilitie	Step 5: Calculate for each of the 4 diagnostic possibilities the percentages of TP, FP, TN and FN and use these data to calculate sensitivity, specificity, positive predictive value and negative predictive value	tive predictive valu
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1	Typical carcinoid	Typical carci

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Procedure to categorize the diagnostic opinions rendered for each case by multiple observers into true and false positive and true and false negative test results.

Table II

Typical carcinoid Typical arcinoid Typical carcinoid Typical carcinoid	Large cell neuroendocrine carcinoma	TN TV FP TN
	Small cell carcinoma	TN NT NT NT
	Atypical carcinoid	NT AFI NT NT
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	Majority diagnosis	Typical carcinoid Typical carcinoid Typical carcinoid Typical carcinoid
	Diagnosis selected by participant	Typical carcinoid Atypical carcinoid Large cell neuroendocrine carcinoma Small cell carcinoma
	Case number 1	

3. Results

IOV was evaluated with Fleiss' kappa in 3 of the 5 papers reviewed. One other paper used Cohen's kappa while the other did not specify the statistical method selected to estimate kappa scores. Three of the 5 papers interpreted the results using the Landis-Koch scale and one modified the scale classifying kappa scores of 0.61-0.80 as good, as initially reported by Cohen, rather than as "substantial". The fifth paper used a modification of the scale proposed by Cohen. None of the five papers explicitly indicated or proposed a minimum kappa score that would reasonably exclude the possibility that IOV could bias results in a clinically significant manner. One of the papers considered kappa scores < 0.40 as outliers while the others did not propose a particular kappa score to determine whether the level of IOV would be acceptable in clinical practice. The root cause of suboptimal IOV was attributed to overlapping diagnostic features that required the development of better criteria in all 5 studies. Two of the studies also opined that interpretation of current criteria by pathologists was part of the problem. Solutions proposed to decrease IOV included the use of immunohistochemistry, molecular studies and machine learning and development of more explicit microscopic features to distinguish the categories under investigation. None of the papers considered the possibility that the problem did not reside with the diagnostic criteria or pathologists but with the fact that the entities being differentiated are not entirely different from each other and that IOV could be resolved by combining them into fewer categories that could be recognized with greater accuracy.

Table III shows the percent agreement between "individual diagnostic opinions" and majority consensus diagnoses calculated from the data reported by Thunnissen et al. [20]. The table also shows the kappa scores reported in that study. Table IV shows the results of sensitivity, specificity, and positive and negative predictive value calculations using the majority consensus diagnoses as ground truth. Although the kappa scores in Table III are "Fair" to "Good" kappa scores, all four neuroendocrine neoplasms were diagnosed with > 90% specificity by the study participants (Table IV).

4. Discussion

While our review of only five studies from the pathology literature certainly does not represent a comprehensive evaluation of the methodologies being used to evaluate IOV and interpret the clinical implications, the results are consistent with our anecdotal impression about this literature. Most studies have evaluated IOV with kappa statistics and although they generally report % agreement, they use kappa scores interpreted with the Landis-Koch scale or minor modifications of this scale. However, there appears to be no consensus about how to interpret kappa scores in the context of clinical practice. None of the studies we sampled cited the existence of an expert opinion or evidencebased rule that could be used to determine which or even whether a particular level of kappa score, such as substantial, moderate or good would indicate that pathologists can establish particular differential diagnosis with an acceptable level of accuracy to ensure that patients are diagnosed in a reproducible manner. Interestingly, based on our analysis of the data from the study by Thunnissen et al., it would appear there is no need for such a rule for the diagnosis of lung neuroendocrine neoplasms. Indeed, diagnostic specificity was > 90% for the four neoplasms, in spite of kappa scores that were quite variable, suggesting that kappa statistics are of little clinical value when comparing the opinion of multiple raters against a "golden reference" diagnosis. We elected not to evaluate how best to compare diagnostic opinions among two raters by calculating the positive and negative percent agreement for each pairwise comparison of the raters, as estimates of median, minimum and maximum agreements for each diagnosis are not appropriate to compare the different raters to a gold reference diagnosis. The problem of how best to compare diagnosis by two raters was discussed in 2013

Table III

Fleisher kappa values and % agreement levels with majority consensus diagnoses.

Diagnosis $(n = number of observations)$	Kappa score	Interpretation of kappa score**	% Agreement***
Small cell lung carcinoma ($n = 589$)	0.60	Moderate	68.9
Typical carcinoid $(n = 342)$	0.74	Good	84.5
Large cell neuroendocrine carcinoma ($n = 247$)	0.49	Moderate	73.3
Atypical carcinoid ($n = 133$)	0.30	Fair	54.9

* From Thunnissen et al. [39].

** Kappa values were interpreted using the standard Landis and Koch [29].

*** Majority diagnoses were used as "gold standard" in this calculation.

Table IV

Metrics to estimate the accuracy of diagnoses using majority consensus diagnoses as the gold standard for "true positives".

Diagnosis	Sensitivity	Specificity	Positive predictive value	Negative predictive value
Small cell lung carcinoma	68.9%	92.8%	87.3%	80.6%
Typical carcinoid	84.5%	96.7%	89.2%	95.1%
Large cell neuroendocrine carcinoma	73.3%	92.9%	68.8%	94.2%
Atypical carcinoid	54.9%	93.0%	45.1%	95.2%

by de Vet et al. [22]. The study concluded that clinicians are justified in their unhappiness about using Cohen's kappa for comparisons of diagnoses among two raters and favored the use of positive and negative percent agreement for such estimates.

Analysis of the data collected for the study of Thunnissen et al. highlights the problems related to the identification of a "gold reference" to use as true and false positive and negative results when calculating sensitivity, specificity, NPV and PPV. Thunnissen et al. did not record the diagnoses rendered by the pathologists who provided the cases for investigation. As the "gold reference" or "ground truth" for each case we elected to use the majority consensus diagnoses among the study participants, although this method does not offer assurance that the majority was correct. Indeed, while almost all participants agreed on the diagnosis in certain cases, other cases were more problematic as evidenced by only a slight majority concurring with the "gold reference". Future studies comparing the IOV between multiple observers and the diagnoses rendered by the pathologists who submitted the cases or by experts could provide a better study design to evaluate IOV among multiple raters against an accurate "gold reference" diagnoses.

Recent studies published in 2020 also illustrate the lack of guidelines about how to evaluate the accuracy of immunohistochemistry interpretations by different pathologists. For example, Thunnissen et al. and Huang et al. compared agreement rates among pathologists evaluating programmed death-ligand 1 expression bv immunohistochemistry and ROS-1 by fluorescence in situ hybridization using positive and negative % agreement, while Williams et al. elected to analyze agreement in scores using intraclass correlation coefficients and concordance in patient's classification using Fleiss' kappa [40,44,45]. Although the application of different statistical methods is correct, this variability complicates the comparison of results across different studies comparing similar problems

There has been limited discussion about the effect of variability and uncertainty in diagnostic classifications, prognostic models, evaluation of the results of molecular studies and clinical trials. As recently reviewed by McHugh and other authors from the United States Food and Drug Administration (FDA), misclassifications by as little as 5% can be sufficient to significantly invalidate estimates of specificity, sensitivity and area under receiver operating curves [46]. Other studies have shown the effect of misclassifications on prognosis and other health related prediction models, but to our knowledge there are no evidence or expert opinion guidelines on how to control for this problem in future studies proposing new pathologic entities based on prognostic differences or evaluating the utility of new therapeutic options, by diagnosis [47,48]. In summary, our review of literature shows that kappa statistics have limited clinical applicability in pathology and suggests the need for guidelines that would help standardize the evaluation of IOV in a manner that would facilitate comparison of different studies and performance of meta-analysis. We concur with de Vet el al's conclusions that positive and negative agreement levels are the preferred metrics for evaluation of IOV among two raters and propose that estimates of diagnostic specificity, sensitivity and positive and negative predictive values against a "ground truth" are most useful to evaluate IOV among multiple raters.

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