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A method to reduce variability in scoring anti-body mediated rejection in renal allografts: Implications for clinical trials.

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Abstract

Poor reproducibility in scoring antibody mediated rejection (ABMR) using the Banff criteria might limit the use of histology in clinical trials. We evaluated the reproducibility of Banff scoring of 67 biopsies by 6 renal pathologists at 3 institutions. Agreement by any two pathologists was poor: 44.8%–65.7% for glomerulitis, 44.8%–67.2% for peritubular capillaritis, and 53.7%–80.6% for chronic glomerulopathy (cg). All pathologists agreed on cg0 (n=20) and cg3 (n= 9) cases, however, many disagreed on scores of cg1 or cg2. The range for the incidence of composite diagnoses by individual pathologists was: 16.4%–22.4% for no ABMR; 17.9%–47.8% for active ABMR; and 35.8%–59.7% for cABMR. A “majority rules” approach was then tested in which the scores of 3 pathologists were used to reach an agreement. This increased consensus both for individual scores (ex. 67.2%–77.6% for cg) and for composite diagnoses (ex. 74.6%–86.6% cABMR). Modeling using these results showed that differences in individual scoring could affect the outcome assessment in a mock study of cABMR. We conclude that the Banff schema has high variability and a majority rules approach could be used to adjudicate differences between pathologists and reduce variability in scoring in clinical trials.

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Introduction

Improving renal allograft survival will almost certainly require the development of novel therapeutic agents that prevent chronic injury. To demonstrate the efficacy of these therapies, new surrogate endpoints for clinical trials are needed¹. Demonstrating a differential impact on renal allograft histology could be a very effective means of showing the efficacy of a drug. However, the poor reproducibility of the Banff scoring system among pathologists is a major drawback to its use as an endpoint in clinical trials^{2,3}.

The Banff 2017 Kidney Meeting Report recognized this problem and made recommendations on best practice for pathology endpoints in clinical trials including the development of an adjudication mechanism to address discordance between pathologists⁴. The report also presented revised criteria for the diagnosis of both active ABMR (aABMR) and chronic active ABMR (cABMR) and removed the terminology of “acute, active ABMR”. However, the diagnoses of aABMR and cABMR still rely heavily on the ability of a pathologist to identify and grade glomerulitis (g), peritubular capillaritis (ptc) and chronic glomerulopathy (cg). Thus, reproducibility remains an issue.

The goal of the current study was twofold. First, we aimed to compare the reproducibility of the main individual histologic scores (ptc, g, and cg) and the composite of these scores used for the aABMR and cABMR diagnoses based on the Banff classification. Second, we sought to test a process by which scores from multiple pathologists could be combined to yield a more consistent score, i.e. to provide an adjudication mechanism to address discordance between pathologists as directed by the Banff report. Our hypothesis for these studies was that the scores between the individual pathologists will not show close agreement, but that using a “majority rules” diagnosis from a group of 3 pathologists will improve the consistency of diagnoses and thus might provide a superior methodology to score biopsies in future clinical trials.

Materials and Methods

This study was approved by the Mayo Clinic Institutional Review Board (IRB 14–006319). The transplant database at Mayo Clinic in Rochester, MN was searched by the initial reviewing pathologist (LDC) to identify a subset of biopsies to adequately represent the different grades of peritubular capillaritis, glomerulitis, and chronic glomerulopathy (Banff ptc, g, and cg scores 0–3) for a total of 12 graded categories. A total of 67 biopsies obtained from years 2005–2013 were identified with at least 6 biopsies in each graded category. Forty eight biopsies had an original diagnosis of antibody-mediated rejection. Of those, 2 were mixed (one grade 1B and one grade 2A), 2 had concomitant IgA nephropathy and 1 had BK nephropathy.

Slide Selection:

Each biopsy case contained 10 slides with 3–4 histologic sections per slide, cut at 3–4 microns per section. These were stained for hematoxylin and eosin (H&E), periodic acid-Schiff (PAS), Masson trichrome, and Jones-methenamine silver, as per routine histologic preparation for clinical cases. From these 10 slides, 3 slides from each case (1 each H&E,

PAS, and silver stained) with the most representative tissue were chosen by the initial reviewing pathologist (LDC) and coded with a case number. These coded slides were shipped to each participating renal pathologist for interpretation. Biopsies were all scored without knowledge of preformed donor specific antibody (DSA) or C4d staining status, indication for biopsy, time post-transplant, or other clinical or laboratory parameters. In addition to the test cases, example cases with each of the graded categories as assessed by the initial pathologist (LDC) were reviewed prior to the test cases. Each pathologist was given written guidelines for scoring. At Mayo Clinic, C4d is performed by immunofluorescence on fresh-frozen tissue and therefore could not be repeated for this study (no frozen tissue existed). Therefore, the original C4d reading was used for this study.

Active and chronic antibody mediated rejection diagnosis:

In addition to assigning individual Banff scores, a diagnosis of no rejection, aABMR, or cABMR was given based on the aggregate of scores, purely based on strict Banff 2017 criteria⁴. Note that the diagnosis was not made by the pathologist, but only derived from the Banff scores. For these re-reads, pathologists were blinded to the C4d and DSA results and thus were not biased by these data. However, since one of our goals was to mimic central pathology re-reads in a clinical trial, the lack of central re-read for C4d actually mimics what happens in most trials. Thus we determined agreement on ABMR using the original C4d scoring done on 67 biopsies. Of these 50 (74.6%) were negative and 17 (25.4%) were positive. Active ABMR was assigned using the Banff 2017 criteria⁴ as: either 1) C4d = 1 and ptc + g score = 2 or 2) C4d > 1 and ptc > 0 or g > 0. A concomitant cg score > 0 resulted in the cABMR diagnosis. Electron microscopy and genomic testing was not routinely done and not used as criteria for aABMR or cABMR. The reproducibility of acute tubular injury was not assessed and therefore not considered when assigning the diagnosis.

Majority Rules Approach—For the majority rules approach we used a consensus score based on various combinations of 3 pathologist’s scores from a pool of 6 pathologists. Specifically, the 6 pathologists’ scores were combined in different ways resulting in 20 distinct groups (1+2+3, 1+3+4, etc.) of three pathologists. We then compared variability among the 20 groups by considering exclusive groups (10 total comparisons) to overestimation of agreement due to pathologist overlap. For individual Banff scores, the majority rules score was given by rounding the average Banff scores of the three pathologists that comprised the group. For the diagnosis, the majority rules diagnosis is simply a majority rules vote (equivalent to taking the rounded average of binary variables).

Mock Trial Simulation—We examined how a majority rules approach using three pathologists increases the probability that an assigned diagnosis actually approaches the “true diagnosis” and the correct patients would be enrolled in the trial. We assumed that as we increase the number of pathologists reading a biopsy, at some point a true consensus would emerge. First we considered the diagnosis of a consensus of 6 pathologists (4/6) as the “true diagnosis.” We then evaluated the frequency of a single pathologist versus 3 pathologists (using majority rules) reaching the “true diagnosis.” We used the diagnosis of cABMR vs no cABMR with the same reproducibility and frequency as described above. In order to demonstrate the impact of a lack of reliability, the mock trial process was repeated

over a grid of effect sizes as measured by the odds ratio from 0.15 to 0.5 by increments of 0.05.

Statistical Analysis:

Accuracy was assessed through the pairwise percent agreement. Reliability was assessed through Cohen's Kappa⁵. All comparisons carried out between groups of three pathologists only used exclusive groups of pathologists, (i.e. 1+2+3 vs 4+5+6, but not 1+2+3 vs 1+5+6). Kappa statistics were calculated using the 'psych' package and heatmaps were generated using the 'gplots' in R (Vienna, Austria).

In order to form a diagnosis tree, we hypothesized a 150 patient sample (to directly convert from percentages). Of this 150 patient sample, the 'true' diagnosis rate of cABMR vs. no cABMR was established by using the consensus of all 6 pathologists. The re-classification using a three pathologist combination or a single pathologist can then be calculated and directly converted to sensitivity, specificity, positive predictive value, and negative predictive value.

Subsequent results from a hypothetical clinical trial displayed as contingency tables with enrollment and endpoint classification variability were constructed with conditional probabilities of diagnosis based on latent classes⁶. Equivalently, the disagreement between any 2 pathologists (or 2 sets of 3 pathologists) means that the counts in a table could change – a patient may be recorded as no cABMR by one pathologist and cABMR by another pathologist. How often this occurs depends on the agreement between the pathologists. For example, consider a trial with 90% agreement and 75 patients on a treatment arm that resulted in 70 ABMR cases and 5 cases with no rejection. Then, a rescoring which results in $70 * 0.1 = 7$ patients switching from the 'ABMR' category to the 'no rejection' category would be contained within the variability of that diagnosis. Similarly, $5 * 0.1 = 0.5$ implies that 1 patient switching from the 'no rejection' category to the 'ABMR' category would also be contained within the variability.

1. Results

Patient Characteristics

Sixty-seven biopsies from 44 patients were studied (Table 1). 83.6% (56/67) of the biopsies were done for surveillance purposes at a median (IQR) time of 1.0 (0.6 – 2.1) years post-transplant. Of the subjects biopsied, the median (IQR) age at the time of transplantation was 47 (35–54) years, and the majority were female 68.2% (30/44), Caucasian 90.9% (40/44), and received their kidney from a living donor 95.5% (42/44). The leading cause of ESRD in this group was glomerulonephritis [43.2% (19/44)]. The majority of subjects [81.8% (36/44)] had a positive B-flow cytometric crossmatch at the time of transplantation. The majority of patients were on tacrolimus, mycophenolate mofetil, and prednisone for maintenance immunosuppression [95.4% (42/44)]. The majority of the allografts were still functioning at 2 and 5 years post-transplant (97.7% and 81.8%, respectively).

Reproducibility: Scoring agreement among individual pathologists

Agreement on individual Banff scores—We first examined the specific agreement for individual g, ptc and cg scores by performing pairwise comparisons among individual pathologists for each biopsy (ex. pathologist #1 vs #2, #1 vs #3, etc.) resulting in 6 pairwise comparisons. The specific agreement on scoring was relatively low. For the g score (glomerulitis), the exact agreement on the score among any two pathologists ranged from 44.8% to 65.7%; for ptc score (peritubular capillaritis), the range was 44.8% to 67.2%; and for cg (chronic glomerulopathy) the range was 53.7% to 80.6%. However, the vast majority of the g, ptc, or cg scores either matched or were ± 1 between pathologists (Table 2, Figure 1). For example, the number of cg scores with differences ≤ 2 ranged from 0 to 7 (0 to 10.5%). Thus, there was general agreement, but not exact agreement g, ptc and cg scoring. The average kappa statistics were 0.39, 0.38, and 0.48 for g, ptc, and cg scores, respectively.

Despite many discrepancies there was complete agreement regarding cg score among the 6 pathologists in 29 cases – 20 scored a cg0 and 9 scored a cg3. In contrast, there was never complete agreement regarding scores of cg1 or cg2 suggesting that assessing intermediate levels of cg had higher variability.

The effect of pathologists working at the same site was also investigated for possible confounding. By separating comparisons between pathologists across sites and within sites, we find that the percent agreement and kappa statistic within site is contained within the range of those across sites for all individual scores (data not shown). Therefore, pathologists working closely at the same site did not bias the mean and range of reported statistics in this study.

Agreement on the diagnosis of ABMR vs no ABMR—The agreement between single pathologists on a diagnosis of ABMR (either aABMR or cABMR) versus no rejection ranged from 86.6% to 97.0%. However, it was not always the same biopsies that led to disagreements in these pairwise comparisons. In the worst case scenario, 13.4% of biopsies would be classified differently just by being read by a different pathologist. In 77.6% of biopsies (52/67), all 6 pathologists agreed a diagnosis of either ABMR (aABMR or cABMR)/no ABMR including: Of the 52 with complete agreement, 6 were scored as no ABMR and 46 met criteria for ABMR (when combined with C4d data). In 15 biopsies (22%), at least one pathologist's composite score led to a different diagnosis (i.e. in 22% of cases, reading the same slide by a different pathologist would result in a different diagnosis).

Agreement on no ABMR, active ABMR, and chronic active ABMR—When the cg score is added to the ABMR composite score to determine presence or absence of cABMR, there are three possible agreement/disagreement groups: cABMR, aABMR and no ABMR. More choices resulted in less agreement. There was complete agreement among all 6 pathologists about the diagnoses (cABMR, aABMR and no ABMR) in only 43.3% of biopsies (29/67). All 6 pathologists agreed upon 6 cases of no ABMR, 6 cases of aABMR, and 17 cases of cABMR. However, in 56.7% (38/67) at least 1 pathologist's scores lead to a different composite diagnosis. The incidence of the three diagnoses as designated by individual pathologists ranged from 16.4% to 22.4% for no ABMR; 17.9% to 47.8% for

aABMR; and 35.8% to 59.7% for cAMBR. (Table 3). The kappa statistic for cABMR vs no cABMR (no ABMR or aABMR) ranged from 0.43 to 0.74.

Majority Rules approach

Table 3 shows the ranges of individual scores and diagnoses among individual pathologists and between the various majority rules groups of 3 pathologists. Using the majority rules approach, the agreement for g, ptc, and cg increased markedly to 72.2%, 70.4%, and 72.2%, respectively (Table 4). The kappa statistic also increased to 0.62, 0.60, and 0.57, for g, ptc, and cg respectively.

Table 4 shows the mean and range percent agreement and kappa statistics for Banff scores and diagnoses using a single pathologist or a ‘majority rules’ approach. As expected, the kappa statistic improved for all categories using majority rules. For the diagnosis of ABMR vs No Rejection, agreement improved from a range of 86.6% - 97.0% to a range of 91.0% - 97.0%. The “majority rules” kappa ranged from 0.71 to 0.91.

Similar to the diagnosis of aABMR, a majority rules approach involving 3 pathologist’s scores led to improved consistency in the various diagnoses. The various combinations of 3 pathologist’s scores led to the same diagnosis (cABMR vs aABMR or no rejection) in an average of 85.4% of cases and ranged from 80.6% to 89.6%. When considering the three-level diagnosis as no rejection vs. aABMR vs. cABMR, there was average agreement of 72.3% (range = 59.7% to 80.6%) using one pathologist and 80.0% (range = 74.6% to 86.6%) using a majority rules method. The incidence of the three diagnoses as designated by 3-pathologist majority rules ranged from 16.4% to 23.9% for no ABMR; 28.4% to 50.7% for aABMR; and 32.8% to 49.3% for cAMBR.

Implications of using a Majority Rules Approach in Clinical Trial Design: An Example

Inclusion criteria—Table 5 and Figure 2 show the mock trial and statistical variability of using one pathologist versus 3 in a majority rules approach. Calculations are based on the 6 pathologist consensus diagnosis. Note that 2 cases had a tie (3 vs 3) and these cases were excluded from subsequent calculations. When only one pathologist determined a positive cABMR diagnosis, a mean of 25.8% (100% minus 74.2%) of cases were classified differently than the consensus of 6 pathologists (true diagnosis). In contrast, when the majority rules approach was applied using 3 pathologists; on average, only 17.1% of cases were classified differently than the “true diagnosis.”

Clinical trial result—Clinical trial results can also be influenced by diagnostic reliability. To illustrate this, a mock clinical trial was designed with the following assumptions: 1) All patients enrolled had biopsy-proven cABMR [prevalence of the underlying disease was estimated using the average positive predictive value (number of true cABMR cases predicted correctly divided by the total predicted number of cABMR cases)], 2) The rate of treatment being successful was 45% in the control group and 68% in the treatment group consistent with reported treatment efficacies^{7,8}, and 3) the range of cABMR variability used corresponds to 79.5% for a single pathologist and 85.4% for a “majority rules” group based on known variability (Table 4). Using these assumptions we calculated the impact of

diagnostic variability on the trial results (Figure 3). In an ideal situation the reported odds ratio would be 0.390 ($p = 0.0049$), i.e. drug is effective in reversing ABMR Figure 3A. However, the results from a second pathologist reading the same slides could be an odds ratio of 0.530 ($p = 0.0757$) Figure 3B. If the majority rules approach (3 pathologists) applied, the clinical trial result would more likely match the result as that from the ideal situation [odds ratio of 0.482 ($p = 0.0370$)] and deem the drug effective as before Figure 3C. The impact of diagnostic variability can be generalized to other effect and sample sizes as shown in Figure 4.

Discussion

We found that the reproducibility of the individual Banff scores needed to diagnose ABMR (glomerulitis, peritubular capillaritis, and chronic glomerulopathy) was poor when evaluating the histologic interpretation of 67 biopsies by 6 pathologists from 3 transplant centers. The diagnosis of ABMR based on a composite of individual scores, had reduced variability but remained inadequate considering that reliable and accurate histologic diagnosis is essential for the design of effective clinical trials, as illustrated in our mock clinical trial design. . In only 43.3% of the cases did the composite of individual Banff scores lead to the same diagnosis for all pathologists. We devised a majority rules approach using 3 pathologists for ABMR diagnostic classification and found that it led to a more reliable and reproducible results than the use of a single pathologist. This alternative approach to histologic interpretation is consistent with the best practice recommendations for histologic diagnosis outlined in the Banff 2017 meeting report and would likely improve our ability to perform effective clinical trials in ABMR.

A potential limitation of this study was that most cases were purer forms of ABMR that lacked concurrent signs of cellular rejection. Thus, the reproducibility of the lesions related to ABMR in other types of patients (i.e those more conventional patients with non-adherence) might be limited.

It is well known that histologic interpretation is complex and the reproducibility of most Banff lesions is low⁹. Many studies have looked at the acute Banff scores with kappa values ranging from 0.19 to 0.50 for glomerulitis^{2,10-13}. Another study showed kappa values using two pathologists to score peritubular capillaritis were as low as 0.28 to 0.53^{14,15}. The inter-observer variability of transplant glomerulopathy was found to 0.14² although this was improved to values as high as 0.47 with later diagnosis criteria¹⁶. Similar poor inter-observer variability has been observed with other Banff lesions including: interstitial fibrosis (linearly weighted kappa = 0.04 – 0.15), peritubular capillary basement membrane multilaminations (kappa = 0.66 – 0.73), arteriolar hyalinosis (ICC = 0.06 – 0.38), and acute cellular rejection (kappa = 0.47 – 0.72)^{3,17-20}.

Emphasizing the importance of reproducibility of the histologic scores comprising the diagnosis of ABMR is not merely an academic argument. Currently, there are two Phase III clinical trials enrolling patients with biopsy-proven ABMR with the goal of either preventing eGFR decline/graft loss¹³ or preventing the progression of chronic glomerulopathy¹⁴. The distribution of diagnoses in the two trials with the same biopsies, but

a different central pathologist might be: 16.4% to 22.4% for no ABMR; 17.9% to 47.8% for aABMR; and 35.8% to 59.7% for cAMBR. This variability might have a major impact on determining the efficacy of a potential therapeutic agent, whether the drug effect is underestimated (false negative), or the effect cannot be reproduced in subsequent trials or clinical settings (false positive). Furthermore, our data show the progression of chronic glomerulopathy is subject to major variability when the damage is mild to moderate (Banff cg score 1/Banff cg 2). We suggest that a “majority rules” approach using the scores of 3 pathologists may be an appropriate compromise when using the Banff ABMR schema for clinical trials. In practice, a study could have a local read and a central read by a study pathologist. If there is disagreement regarding either criteria for inclusion or the primary endpoint, a third pathologist re-read would be needed to determine a “majority rules” score.

Although we found that the reproducibility of individual Banff scores was relatively low, we do not conclude that certain pathologists are performing poorly. On the contrary, we believe that the lack variability in pathologic interpretation results from the fact that currently pathologists are forced to provide a discrete score to a complex continuous process. Not only does the pathologist have to determine the presence or absence of a specific lesion (ie. glomerulitis), the pathologist must determine the severity of that score (1–3). For chronic glomerulopathy (cg), the pathologist must also determine the presence and score in the “most affected glomerulus not globally sclerosed.” Our data revealed that despite the lack of *exact* agreement, there was actually good *general* agreement among pathologists.

Going forward, reproducibility might be increased in other ways. First, use of images has been shown to improve inter-rater reproducibility^{12,21}. Next, a morphometric scoring system might decrease scoring variability. In this approach, a pathologist or a trained technician would annotate computer-scanned biopsies for each lesion (i.e. determine presence or absence). A computer program then would determine the extent to which the lesion is present and thus limit variability in this aspect of scoring. This approach might also allow for a continuous scoring system obtained through morphometric analysis of biopsies which should decrease problems associated with discretizing^{22,23}. For example, an interstitial fibrosis morphometric score correlations have been reported as high as 0.9636 although reproducibility was lower¹⁷. For cg, the fraction of the glomerular basement membrane with contour doubling may be a useful alternative. Such a system might be able to assess the progression of chronic glomerulopathy (Banff cg lesions) from mild (Banff cg score 1) to severe (Banff cg score 3) in serial biopsies.

The ideal ABMR histologic classification system should not only be highly reproducible, but also should have a high correlation with clinically-important outcomes such as graft function and graft survival. While validity, or the correlation of Banff scores with outcomes, is not guaranteed by a more reliable measure, any relationship with this measure is limited by the reliability. In clinical trials, histology might be used to assess response to therapy similar to oncology studies in which a histologic response is used. Essentially, this means validation of histology as a biomarker for ABMR. Whether or not the current Banff system is sufficient or requires further modification is unclear at the present. However, we believe that a major goal of the Banff working groups going forward should be to correlate histologic data with graft outcomes in order to validate future scoring systems. The use of

gene expression has been suggested to be a more accurate biomarker of ABMR than histology²⁴. Modeling histology, gene expression and clinical data might lead to a model that is truly predictive of outcomes and provides useful endpoints for future clinical trials.

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Abbreviations:

(aABMR)	Acute, active antibody mediated rejection
(ABMR)	Antibody mediated rejection
(cABMR)	Chronic, active antibody mediated rejection
(cg)	Chronic glomerulopathy
(CI)	Confidence interval
(ESRD)	End stage renal disease
(g)	Glomerulitis
(ptc)	Peritubular capillaritis
(sd)	Standard deviation

References

1. Stegall MD, Gaston RS, Cosio FG, Matas A. Through a glass darkly: seeking clarity in preventing late kidney transplant failure. *J Am Soc Nephrol.* 2015;26(1):20–29. [PubMed: 25097209]
2. Gough J, Rush D, Jeffery J, et al. Reproducibility of the Banff schema in reporting protocol biopsies of stable renal allografts. *Nephrol Dial Transplant.* 2002;17(6):1081–1084. [PubMed: 12032200]
3. Veronese FV, Manfro RC, Roman FR, et al. Reproducibility of the Banff classification in subclinical kidney transplant rejection. *Clin Transplant.* 2005;19(4):518–521. [PubMed: 16008598]
4. Haas M, Loupy A, Lefaucheur C, et al. The Banff 2017 Kidney Meeting Report: Revised diagnostic criteria for chronic active T cell-mediated rejection, antibody-mediated rejection, and prospects for integrative endpoints for next-generation clinical trials. *Am J Transplant.* 2018;18(2):293–307. [PubMed: 29243394]
5. Cohen J A Coefficient of Agreement for Nominal Scales. *Educational and Psychological Measurement.* 1960;20(1):37–46.
6. Lyles RH, Lin J. Sensitivity analysis for misclassification in logistic regression via likelihood methods and predictive value weighting. *Stat Med.* 2010;29(22):2297–2309. [PubMed: 20552681]
7. Kim MG, Kim YJ, Kwon HY, et al. Outcomes of combination therapy for chronic antibody-mediated rejection in renal transplantation. *Nephrology (Carlton).* 2013;18(12):820–826. [PubMed: 24033843]
8. Billing H, Rieger S, Susal C, et al. IVIG and rituximab for treatment of chronic antibody-mediated rejection: a prospective study in paediatric renal transplantation with a 2-year follow-up. *Transpl Int.* 2012;25(11):1165–1173. [PubMed: 22897111]

9. Mengel M, Sis B, Halloran PF. SWOT analysis of Banff: strengths, weaknesses, opportunities and threats of the international Banff consensus process and classification system for renal allograft pathology. *Am J Transplant.* 2007;7(10):2221–2226. [PubMed: 17848174]
10. Solez K, Hansen HE, Kornerup HJ, et al. Clinical validation and reproducibility of the Banff schema for renal allograft pathology. *Transplant Proc.* 1995;27(1):1009–1011. [PubMed: 7878779]
11. Furness PN, Taub N, Convergence of European Renal Transplant Pathology Assessment Procedures P. International variation in the interpretation of renal transplant biopsies: report of the CERTPAP Project. *Kidney Int.* 2001;60(5):1998–2012. [PubMed: 11703620]
12. Furness PN, Taub N, Assmann KJ, et al. International variation in histologic grading is large, and persistent feedback does not improve reproducibility. *Am J Surg Pathol.* 2003;27(6):805–810. [PubMed: 12766585]
13. Marcussen N, Olsen TS, Benediktsson H, Racusen L, Solez K. Reproducibility of the Banff classification of renal allograft pathology. Inter- and intraobserver variation. *Transplantation.* 1995;60(10):1083–1089. [PubMed: 7482712]
14. Gibson IW, Gwinner W, Brocker V, et al. Peritubular capillaritis in renal allografts: prevalence, scoring system, reproducibility and clinicopathological correlates. *Am J Transplant.* 2008;8(4):819–825. [PubMed: 18261174]
15. Batal I, De Serres SA, Mfarrej BG, et al. Glomerular inflammation correlates with endothelial injury and with IL-6 and IL-1beta secretion in the peripheral blood. *Transplantation.* 2014;97(10):1034–1042. [PubMed: 24406453]
16. Haas M, Sis B, Racusen LC, et al. Banff 2013 meeting report: inclusion of c4d-negative antibody-mediated rejection and antibody-associated arterial lesions. *Am J Transplant.* 2014;14(2):272–283. [PubMed: 24472190]
17. Farris AB, Chan S, Climenhaga J, et al. Banff fibrosis study: multicenter visual assessment and computerized analysis of interstitial fibrosis in kidney biopsies. *Am J Transplant.* 2014;14(4):897–907. [PubMed: 24712330]
18. Seron D, Moreso F, Fulladosa X, Hueso M, Carrera M, Grinyo JM. Reliability of chronic allograft nephropathy diagnosis in sequential protocol biopsies. *Kidney Int.* 2002;61(2):727–733. [PubMed: 11849416]
19. Liapis H, Gaut JP, Klein C, et al. Banff Histopathological Consensus Criteria for Preimplantation Kidney Biopsies. *Am J Transplant.* 2017;17(1):140–150. [PubMed: 27333454]
20. Liapis G, Singh HK, Derebail VK, Gasim AM, Kozlowski T, Nিকেleit V. Diagnostic significance of peritubular capillary basement membrane multilaminations in kidney allografts: old concepts revisited. *Transplantation.* 2012;94(6):620–629. [PubMed: 22936037]
21. Ozluk Y, Blanco PL, Mengel M, Solez K, Halloran PF, Sis B. Superiority of virtual microscopy versus light microscopy in transplantation pathology. *Clin Transplant.* 2012;26(2):336–344. [PubMed: 21955102]
22. Cox DR. Note on grouping. *Journal of American Statistical Association.* 1957;52:543–547.
23. Fedorov V, Mannino F, Zhang R. Consequences of dichotomization. *Pharm Stat.* 2009;8(1):50–61. [PubMed: 18389492]
24. Halloran PF, Reeve J, Akalin E, et al. Real Time Central Assessment of Kidney Transplant Indication Biopsies by Microarrays: The INTERCOMEX Study. *Am J Transplant.* 2017.

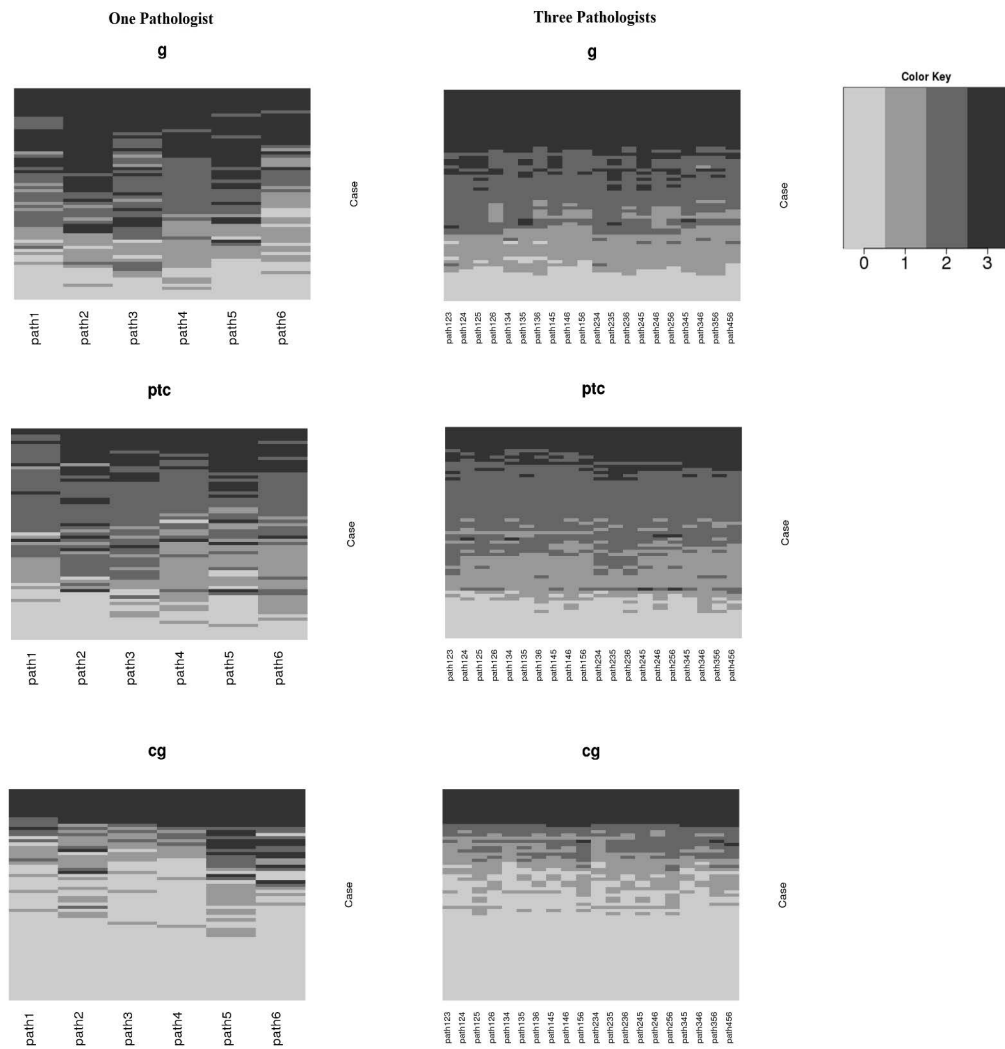


Figure 1. Heatmaps for Banff scores using one-pathologist versus a 3 pathologist ‘consensus’ score.

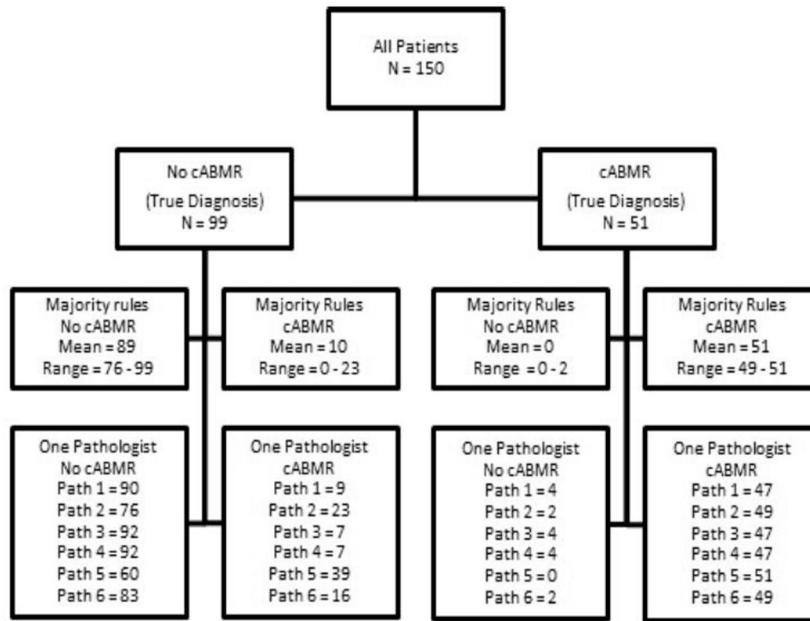


Figure 2. A flow diagram for diagnosis in a hypothetical clinical trial for cABMR involving 150 patients. The ground truth is given by the 6 pathologist consensus diagnosis and the range of diagnoses either in agreement or not is given for a three pathologist consensus or a single pathologist.

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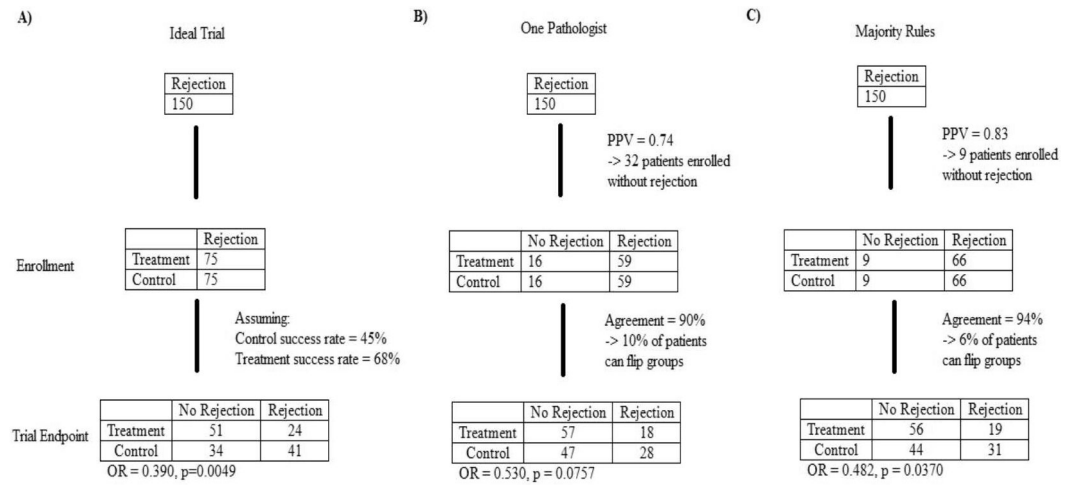


Figure 3. A mock trial involving 150 patients. In the ideal situation a pathologist would always and only identify positive cases as positive for cABMR. Given the PPV and diagnostic agreement for one pathologist and a three pathologist ‘majority rules’, the conclusion of the trial could be affected.

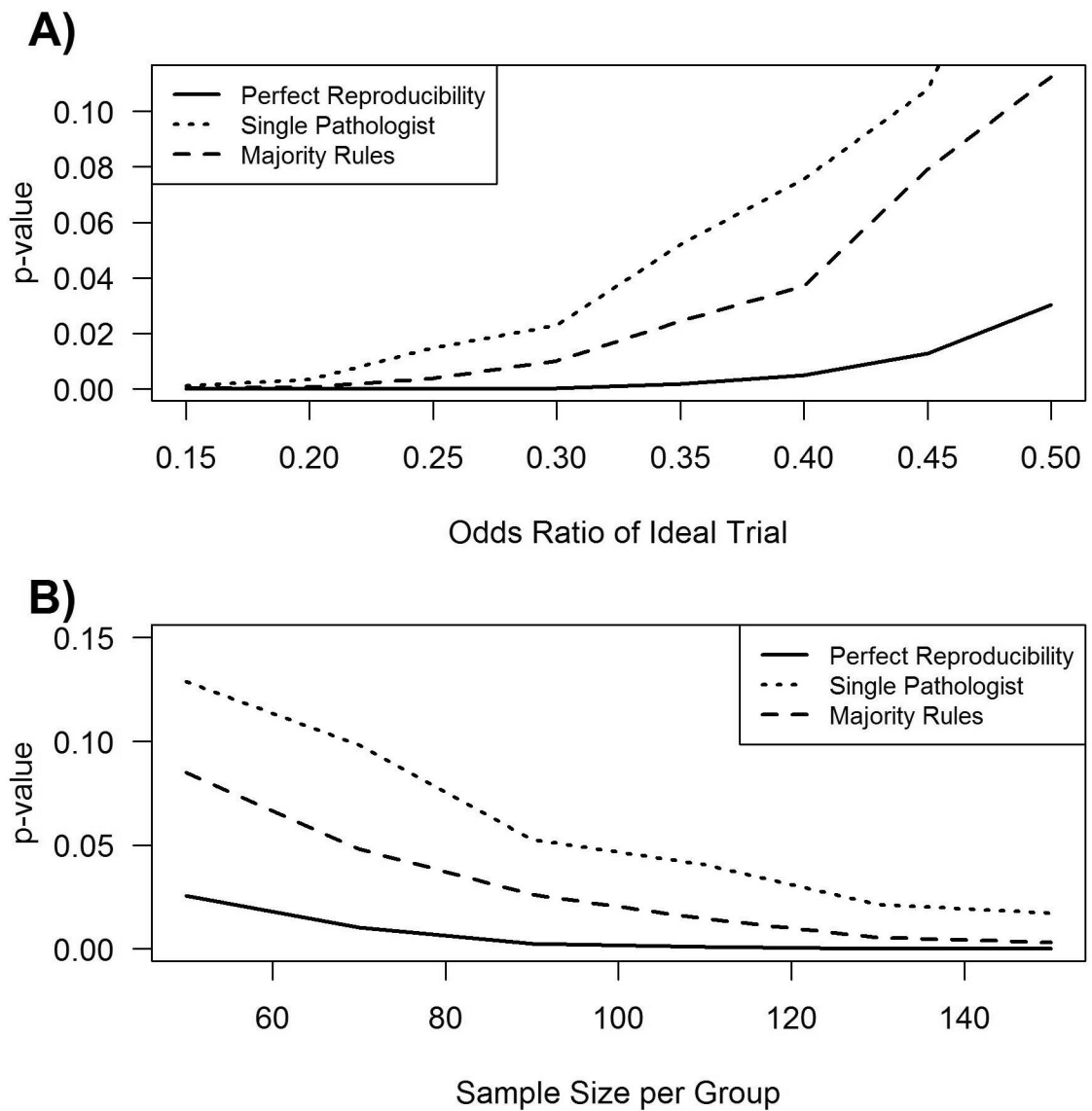


Figure 4. The effect of a lack of reproducibility versus A) effect size and B) sample size based on the above agreement and PPV. In order to produce lines, the mock trial was repeated for a variety of different effect sizes (odds ratios) and sample sizes.

Table 1.

Patient Characteristics

	All Patients (N=44)
Number of biopsies reviewed/patient n(%)	
1	28 (63.6%)
2	11 (25.0%)
3	3 (6.8%)
4	2 (4.5%)
Age at transplantation years median (IQR)	47(35-54)
Years from transplantation to biopsy median (IQR)	1.0 (0.6-2.1)
Female gender n(%)	30 (68.2%)
Race n(%)	
Caucasian	40 (90.9%)
Hispanic	2 (4.5%)
African American	1 (2.3%)
Pacific Islander	1 (2.3%)
Donor Type n(%)	
Living unrelated donor	25 (56.8%)
Living related donor	17 (38.6%)
Deceased donor	2 (4.5%)
Etiology of ESRD n(%)	
Glomerulonephritis	19(43.2%)
Cystic Kidney Disease	5(11.4%)
Diabetes Mellitus	5(11.4%)
Hypertension	4(9.1%)
Congenital Renal Disease	2(4.5%)
Other	7(15.9%)
Unknown	2(4.5%)
Positive B-flow crossmatch n(%)	36(81.8%)
Maintenance immunosuppression n(%)	
Tacrolimus, Mycophenolate mofetil, Prednisone	42(95.4%)
Other	2(4.5%)
Allograft status 2 years post transplant n(%)	
Active	43 (97.7%)
Failure	1 (2.3%)
Allograft status 5 years post transplant n(%)	
Active	36 (81.8%)
Failure	8 (18.2%)

Table 2.

Diagnostic Agreement among Pathologists Based on Individual Banff Scores. The cell values correspond to the count and percentage of the 67 total cases with no difference (zero), one, two, or three in Banff scores given by the two pathologists.

	First Pathologist	Second Pathologist	Magnitude of difference between first and second pathologist			
			zero	one	two	three
g	path1	path2	39 (58.21%)	22 (32.84%)	6 (8.96%)	0 (0%)
	path1	path3	30 (44.78%)	33 (49.25%)	4 (5.97%)	0 (0%)
	path1	path4	39 (58.21%)	27 (40.3%)	1 (1.49%)	0 (0%)
	path1	path5	41 (61.19%)	22 (32.84%)	3 (4.48%)	1 (1.49%)
	path1	path6	32 (47.76%)	27 (40.3%)	8 (11.94%)	0 (0%)
	path2	path3	33 (49.25%)	29 (43.28%)	5 (7.46%)	0 (0%)
	path2	path4	41 (61.19%)	23 (34.33%)	3 (4.48%)	0 (0%)
	path2	path5	44 (65.67%)	17 (25.37%)	6 (8.96%)	0 (0%)
	path2	path6	36 (53.73%)	18 (26.87%)	12 (17.91%)	1 (1.49%)
	path3	path4	37 (55.22%)	26 (38.81%)	4 (5.97%)	0 (0%)
	path3	path5	32 (47.76%)	27 (40.3%)	7 (10.45%)	1 (1.49%)
	path3	path6	34 (50.75%)	22 (32.84%)	9 (13.43%)	2 (2.99%)
	path4	path5	44 (65.67%)	20 (29.85%)	3 (4.48%)	0 (0%)
	path4	path6	38 (56.72%)	26 (38.81%)	3 (4.48%)	0 (0%)
	path5	path6	35 (52.24%)	20 (29.85%)	12 (17.91%)	0 (0%)
	cg	path1	path2	42 (62.69%)	21 (31.34%)	3 (4.48%)
path1		path3	54 (80.6%)	12 (17.91%)	1 (1.49%)	0 (0%)
path1		path4	52 (77.61%)	12 (17.91%)	3 (4.48%)	0 (0%)
path1		path5	37 (55.22%)	22 (32.84%)	6 (8.96%)	1 (1.49%)
path1		path6	44 (65.67%)	14 (20.9%)	6 (8.96%)	3 (4.48%)
path2		path3	46 (68.66%)	17 (25.37%)	3 (4.48%)	1 (1.49%)
path2		path4	49 (73.13%)	13 (19.4%)	4 (5.97%)	1 (1.49%)
path2		path5	39 (58.21%)	19 (28.36%)	5 (7.46%)	3 (4.48%)
path2		path6	46 (68.66%)	10 (14.93%)	8 (11.94%)	3 (4.48%)
path3		path4	53 (79.1%)	14 (20.9%)	0 (0%)	0 (0%)
path3		path5	35 (52.24%)	22 (32.84%)	8 (11.94%)	1 (1.49%)
path3		path6	46 (68.66%)	13 (19.4%)	6 (8.96%)	2 (2.99%)
path4		path5	39 (58.21%)	19 (28.36%)	7 (10.45%)	1 (1.49%)
path4		path6	48 (71.64%)	11 (16.42%)	6 (8.96%)	2 (2.99%)
path5		path6	46 (68.66%)	16 (23.88%)	1 (1.49%)	3 (4.48%)
Author Manuscript		path1	path2	34 (50.75%)	27 (40.3%)	5 (7.46%)
	path1	path3	37 (55.22%)	28 (41.79%)	2 (2.99%)	0 (0%)
	path1	path4	43 (64.18%)	21 (31.34%)	3 (4.48%)	0 (0%)
	path1	path5	30 (44.78%)	31 (46.27%)	5 (7.46%)	1 (1.49%)
	path1	path6	36 (53.73%)	25 (37.31%)	6 (8.96%)	0 (0%)
	path2	path3	43 (64.18%)	19 (28.36%)	4 (5.97%)	1 (1.49%)

		Magnitude of difference between first and second pathologist				
	First Pathologist	Second Pathologist	zero	one	two	three
ptc	path2	path4	36 (53.73%)	29 (43.28%)	2 (2.99%)	0 (0%)
	path2	path5	34 (50.75%)	26 (38.81%)	7 (10.45%)	0 (0%)
	path2	path6	35 (52.24%)	27 (40.3%)	5 (7.46%)	0 (0%)
	path3	path4	31 (46.27%)	33 (49.25%)	3 (4.48%)	0 (0%)
	path3	path5	34 (50.75%)	27 (40.3%)	5 (7.46%)	1 (1.49%)
	path3	path6	38 (56.72%)	25 (37.31%)	4 (5.97%)	0 (0%)
	path4	path5	34 (50.75%)	30 (44.78%)	2 (2.99%)	1 (1.49%)
	path4	path6	45 (67.16%)	20 (29.85%)	2 (2.99%)	0 (0%)
	path5	path6	38 (56.72%)	29 (43.28%)	0 (0%)	0 (0%)

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Table 3.

Range of diagnosis incidences using a single pathologist or a 'majority rules' approach.

Diagnosis	One Pathologist	Three Pathologists
No Rejection	16.4 - 22.4%	16.4 - 23.9%
aABMR	17.9 - 47.8%	28.4 - 50.7%
cABMR	35.8 - 59.7%	32.8 - 49.3%

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Table 4.

Mean and range percent agreement and kappa statistics for Banff scores and diagnoses using a single pathologist or a 'majority rules' approach.

Score	One Pathologist		Three Pathologist 'Majority Rules'	
	% Agreement [Mean (Range)]	Kappa [Mean (Range)]	Agreement [Mean (Range)]	Kappa [Mean (Range)]
g	55.2 (44.8, 65.7)	0.39 (0.24, 0.53)	72.2 (67.2, 79.1)	0.62 (0.55, 0.72)
ptc	54.5 (44.8, 67.2)	0.38 (0.27, 0.54)	70.4 (62.7, 80.6)	0.60 (0.49, 0.73)
cg	67.5 (53.7, 80.6)	0.48 (0.31, 0.65)	72.2 (67.2, 77.6)	0.57 (0.50, 0.64)
aABMR (Y/N)	90.2 (86.6, 97.0)	0.70 (0.53, 0.91)	94.0 (91.0, 97.0)	0.82 (0.71, 0.91)
cABMR (Y/N)	79.5 (70.1, 88.1)	0.59 (0.43, 0.74)	85.4 (80.6, 89.6)	0.70 (0.60, 0.78)
Diagnosis	72.3 (59.7, 80.6)	0.57 (0.39, 0.70)	80.0 (74.6, 86.6)	0.69 (0.61, 0.79)

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Table 5.

Sensitivity, specificity, positive predictive value, and negative predictive value each single pathologist. Average scores are given for single pathologists and three pathologist 'majority rules' for comparison as well.

	Sensitivity	Specificity	PPV	NPV
Three Path Average	100.00	89.4	82.9	100.0
One Path Average	94.8	83.0	74.2	96.9
Path 1	92.2	90.9	83.9	95.7
Path 2	96.1	76.8	68.1	97.4
Path 3	92.2	92.9	87.0	95.8
Path 4	92.2	92.9	87.0	95.8
Path 5	100.0	60.6	56.7	100.0
Path 6	96.1	83.8	75.4	97.6