

Next Steps for the Colorado Risk-Limiting Audit (CORLA) Program

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Abstract

COLORADO PASSED A LAW MANDATING RISK-LIMITING TABULATION AUDITS (RLAs) STARTING IN 2018. ELECTIONS ARE CONDUCTED AT THE LEVEL OF JURISDICTIONS, WITHOUT STANDARDIZATION ACROSS JURISDICTIONS, SO SOME COUNTIES USE LEGACY VOTING MACHINES. 98.2% OF ACTIVE COLORADO VOTERS ARE REGISTERED IN COUNTIES THAT USE MACHINES THAT PRODUCE A CAST VOTE RECORD (CVR), ENABLING EFFICIENT BALLOT-COMPARISON AUDITS. THE REMAINING 1.8% OF VOTERS ARE IN COUNTIES THAT USE MACHINES THAT DO NOT PRODUCE A CVR. THESE COUNTIES MUST USE LESS EFFICIENT BALLOT-POLLING AUDITS. Colorado conducted RLAs across the state in 2017, including both ballot-level comparison audits and ballot-polling audits. Those audits only covered contests restricted to a single county; methods to efficiently audit contests that cross county boundaries and combine ballot polling and ballot-level comparisons have not been available.

Colorado's current audit software (RLATool) needs to be improved to audit these contests that cross county lines and to audit small contests efficiently.

This paper addresses these needs. It presents extremely simple but inefficient methods, more efficient methods that combine ballot polling and ballot-level comparisons using stratified samples, and methods

that combine ballot-level comparison and variable-size batch comparison audits in a way that does not require stratified sampling.

We conclude with some recommendations, and illustrate our recommended method using examples that compare them to existing approaches. Exemplar open-source code and interactive Jupyter notebooks are provided that implement the methods and allow further exploration.

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1 Introduction

A risk-limiting audit (RLA) of an election is a procedure that has a known, pre-specified minimum chance of correcting the electoral outcome if the outcome is incorrect—that is, if the reported outcome differs from the outcome that a full manual tabulation of the votes would find. RLAs require a durable, voter-verifiable record of voter intent, such as paper ballots, and they assume that this audit trail is sufficiently complete and accurate that a full hand tally would show the true electoral outcome. That assumption is not automatically satisfied: a *compliance audit* (Stark and Wagner, 2012) is required.

Risk-limiting audits are generally incremental: they examine more ballots, or batches of ballots, until either (i) there is strong statistical evidence that a full hand tabulation would confirm the outcome, or (ii) the audit has led to a full hand tabulation, the result of which should become the official result.

RLAs have been piloted in California, Colorado, and Ohio, and a test of RLA procedures has been conducted in Arizona. RLA bills are being drafted or are already under consideration in California, Virginia, Washington, and other states. A number of laws have either allowed or mandated risk-limiting audits, including California AB 2023 (Saldaña), SB 360 (Padilla), and AB 44 (Mullin); Rhode Island SB 413A and HB 5704A; and Colorado Revised Statutes (CRS) 1-7-515.

CRS 1-7-515 required Colorado to implement risk-limiting audits beginning in 2017. (There are provisions to allow the Secretary of State to exempt some counties.) The first set of coordinated risk-limiting election audits

across the state took place in Colorado in November, 2017.¹ Those audits only covered contests restricted to a single county, so counties could conduct audits independently. In future statewide elections and contests that cross county lines, counties will need to coordinate.

Colorado’s “uniform voting system” program² led many Colorado counties to purchase (or to plan to purchase) voting systems that are auditable at the ballot level: those systems export cast vote records (CVRs) for individual ballots in a manner that allows the corresponding paper ballot to be identified, and conversely, make it possible to find the CVR corresponding to any particular paper ballot. We call counties that have such systems “CVR” counties. It is estimated that by June, 2018, 98.2% of active Colorado voters will be in CVR counties. CVR counties can perform “ballot-level comparison audits,” (Lindeman and Stark, 2012) which are currently the most efficient approach to risk-limiting audits in that they require examining fewer ballots than other methods do, when the outcome of the contest under audit is in fact correct.

Other counties (“legacy” or “no-CVR” counties) have systems that do not allow auditors to check how the system interpreted voter intent for individual ballots. Their election results can still be audited, provided their voting systems create a voter-verifiable paper trail (*e.g.*, voter-marked paper ballots) that is conserved to ensure that it remains accurate and intact, and organized well enough to permit ballots to be selected at random. Pilot audits in California suggest that the most efficient way to audit such systems is by “ballot-polling” (Lindeman et al., 2012; Lindeman and Stark, 2012) (in contrast to “batch-level comparisons,” for example).

There is currently no literature on how to perform risk-limiting audits of contests that include CVR counties and no-CVR counties by combining ballot polling and ballot-level comparisons. Existing methods would either require all counties to use the lowest common denominator, ballot-polling (which does not take advantage of the CVRs, and thus is expected to require more auditing than a method that does take advantage of the CVRs), or would require no-CVR counties to perform batch-level comparisons, which were found in California to be (generally) less efficient than ballot-polling

¹See <https://www.sos.state.co.us/pubs/elections/RLA/2017RLABackground.html>

²<https://www.sos.state.co.us/pubs/elections/VotingSystems/UniformVotingSystem.html>

audits.³

The open-source audit software used for Colorado’s 2017 audits, RLATool (<https://github.com/FreeAndFair/ColoradoRLA/>), needs to be improved to audit contests that cross county lines and to audit small contests efficiently.

First, the current version (1.1.0) of RLATool needs to be modified to recognize and group together contests that cross jurisdictional boundaries; currently, it treats every contest as if it were entirely contained in a single county. Margins and risk limits apply to entire contests, not to the portion of a contest included in a county. RLATool also does not allow the user to select the sample size, nor does it directly allow an unstratified random sample to be drawn across counties.

Second, to audit a contest that includes voters in “legacy” counties (counties with voting systems that cannot export cast vote records) and voters in counties with newer systems, new statistical methods are needed to preserve the efficiency of ballot-level comparison audits that the newer systems afford.

Third, auditing contests that appear only on a subset of ballots can be made much more efficient if the sample can be drawn from just those ballots that contain the contest. While allowing samples to be restricted to ballots reported to contain a particular contest is not essential in the short run, it will be necessary eventually to make it feasible to audit smaller contests.

This document focuses on near-term requirements for risk-limiting audits in Colorado. Section 2 presents a number of crude approaches that could be implemented easily but might require examining substantially more ballots. Section 3 presents an approach based on comparison audits with different batch sizes. This approach is statistically elegant and relatively efficient, but might require changing how counties handle their ballots. Section 4 presents our recommended approach, which combines ballot-level comparisons in counties that can perform them with ballot-polling in the no-CVR counties. This approach is generalizable to other stratified audits, where ballots are divided into two independent “strata,” audited independently using appropriate methods, and combined into an audit of the overall contest. All the approaches require new software, including at least minor modifications to RLATool. We provide example software implementing the risk calculations for our recommended approach as a Python Jupyter notebook.⁴

³See Rivest (2018) for a different (Bayesian) approach to auditing contests that include both CVR counties and no-CVR counties.

⁴See <https://github.com/pbstark/CORLA18>.

Section 5 describes how audit efficiency could be improved in CVR counties by combining CVR data with data from Colorado’s voter registration system, SCORE.⁵ Sections 6 and 7 explain the recommended modifications to ballot-level comparison and ballot-polling audits, respectively. Section 8 summarizes our recommendations and considerations for implementation.

1.1 Priorities for Colorado audits

FOR THIS PARTICULAR PAPER FOCUSED ON STRATIFICATION, IS THE ORDER OF PARAGRAPHS HERE RIGHT? Auditing efficiency is controlled in part by how well the audit can limit the sample to ballots that contain the contests under audit. Some contests are on (essentially) every ballot, for instance the governor’s race. Others, such as mayoral contests, may appear on only a small fraction of ballots cast in a county. Partisan primaries—even for statewide office—are somewhere in between, because in general no single party’s primary appears on every ballot cast in the state. Thus, either we accept reduced efficiency for the sake of simplicity by continuing to sample ballots uniformly from within counties (or collections of counties), or we develop a way to focus the auditing on the ballots that contain the contest. The latter requires external information, e.g., from SCORE, as discussed below.

Moreover, party primaries for statewide offices (and perhaps other contests) will include CVR counties and no-CVR counties, so we need a method to audit across both kinds of voting technology.

This report addresses both issues, providing options for effectively auditing heterogeneous voting technology, varying in efficiency, complexity, and on whom any additional audit burden falls.

2 Crude (and unpleasant) approaches

Here and generally throughout the paper, we discuss auditing a single contest at a time, although the same sample can be used to audit more than one contest and there are ways of combining audits of different contests into a single process (Stark, 2009b, 2010). We use terminology drawn from a number of papers; the key reference is Lindeman and Stark (2012). An *overstatement error* is an error that caused the margin between *any* reported winner and

⁵SCORE is Colorado’s voter registration system, which also tracks who voted. See <https://www.sos.state.co.us/pubs/elections/SCORE/SCOREhome.html>.

any reported loser to appear larger than it really was. An *understatement error* is an error that caused the margin between *every* reported winner and *every* reported loser to appear to be smaller than it really was.

2.1 Hand count the legacy counties

The simplest approach to combining legacy counties with CVR counties is to require every legacy county to do a full hand count of the primaries, and to conduct a ballot-level comparison audit in CVR counties, based on contest margins adjusted for the results of the manual tallies in the CVR counties. For instance, imagine a contest with two candidates, reported winner w and reported loser ℓ . Suppose the total number of reported votes for candidate w is V_w and the total for candidate ℓ is V_ℓ , so that $V_w > V_\ell$, since w is the reported winner. Suppose that a full manual tally of the votes in the legacy counties shows V'_w votes for w and V'_ℓ votes for ℓ . Suppose that a total of N ballots were cast in the CVR counties. Then the *diluted margin* for the comparison audit in the CVR counties is defined to be $[(V_w - V'_w) - (V_\ell - V'_\ell)]/N$. Requiring a full hand count in the legacy counties has obvious disadvantages, except perhaps in very close contests where ballot polling is not efficient. (But it does have the advantage of not forcing CVR counties to do additional auditing to compensate for the legacy counties.)

2.2 Subtract error bounds for the legacy counties from vote totals

If ballot accounting and SCORE data can provide good upper bounds on the number of ballots cast in each contest in legacy counties, there are simple upper bounds on the total possible overstatement error each legacy county could contribute to the overall contest results; those can be subtracted from the overall margin (as in the previous subsection) and the remainder of the contests can be audited in CVR counties against the adjusted margins. For instance, consider a primary that appears on N ballots in a legacy counties. Suppose that in legacy counties, the overall, statewide contest winner, w , is reported to have received V'_w votes, and some loser, ℓ , is reported to have received V'_ℓ votes. (Note that V'_ℓ could be greater than V'_w : w is not necessarily the reported winner in the legacy counties.) Then the most overstatement error that the county could possibly have in determining whether w in fact beat ℓ is if every reported undervote, invalid vote, or vote for a different

candidate, t , had in fact been a vote for ℓ (producing a 1-vote overstatement), and every vote reported for w was in fact a vote for ℓ (producing a 2-vote overstatement). The reduction in the margin that would produce is $N - V'_w - V'_\ell + 2V'_w = N + V'_w - V'_\ell$ votes.

Whereas the previous approach places the auditing burden created by obsolescent equipment entirely on the legacy counties, this approach places it entirely upon the CVR counties. Also, in a close contest, it could require a full hand count in every county that might not otherwise be necessary.

2.3 Treat legacy counties as if every ballot selected from them for audit has a two-vote overstatement

A third simple-but-pessimistic approach is to sample uniformly from all counties as if one were performing a ballot-level comparison audit everywhere, but to treat any ballot selected from a legacy county as a two-vote overstatement. This approach has the same disadvantages as the previous approach.

3 Variable batch sizes

Another approach is to perform a comparison audit across all counties, but to use batches consisting of more than one ballot (batch-level comparisons) in legacy counties and batches consisting of a single ballot (ballot-level comparisons) in CVR counties.⁶ This requires that the no-CVR counties report vote subtotals for physically identifiable batches. If a county’s voting system can only report subtotals by precinct but the county does not sort paper ballots by precinct, this approach might require revising how the county handles its paper; we understand that this is the case in many Colorado counties.

That said, many California counties that do not sort vote-by-mail (VBM) ballots by precinct conduct the statutory 1% audits by manually retrieving the ballots for just those precincts selected for audit from whatever physical batches they happen to be in: the situation is identical to that in Colorado.

Another solution is the “Boulder-style” batch-level audit,⁷ which requires generating vote subtotals after each physical batch is scanned, and exporting

⁶For majority and plurality elections, including those in which voters can select more than one candidate, audits can be based on overstatement and understatement errors at the level of batches.

⁷See <http://bcn.boulder.co.us/~neal/elections/boulder-audit-10-11/>.

those subtotals in machine-readable form. That in turn may require using extra memory cards, repeatedly initializing and deleting tabulation databases, or other measures that add complexity and opportunity for human error.

While those two approaches are laborious, they would provide a viable short-term solution, especially combined with information from SCORE to check that the reported batch-level results contain the correct number of ballots for each contest under audit. Moreover, it does not unduly increase the workload in CVR counties to compensate for legacy equipment.

This kind of variable-batch-size comparison audit approach would require modifying or augmenting RLATool in several ways:

1. The CVR reporting tool would need to be modified to allow no-CVR counties to report batch-level results in a manner analogous to how CVR counties report ballot-level results, or an external tool would need to be provided.
2. The sampling algorithm would have to allow sampling batches—and sampling them with unequal probability, because efficient batch-level audits involve sampling batches with probability proportional to a bound on the possible overstatement error in the batch. It would also need to calculate the appropriate sampling probability for each batch (of whatever size). Again, this could be accommodated using an external tool to draw the sample from legacy counties.
3. The risk calculations would need to be modified. This, too, could be done with external software, with suitable provisions for capturing audit data from RLATool or directly from legacy counties.

None of these changes is enormous; the mathematics and statistics are already worked out in published papers, and there is exemplar code for calculating the batch-level error bounds, drawing the samples with probability proportional to an error bound, and calculating the attained risk from the sample results. Indeed, this is the method that was used in several of California’s pilot audits, including the audit in Orange County. A derivation of a method for comparison audits with variable batch sizes is given below in section 6.

4 Stratified “hybrid” audits

Other approaches involve *stratification*: partitioning the cast ballots into non-overlapping groups and sampling independently from those groups. One could stratify by county, but in general it is simpler and more efficient statistically (i.e., results in auditing fewer ballots) to minimize the number of strata. We consider methods that use two strata: CVR counties and no-CVR counties. Collectively, the ballots cast in CVR counties comprise one stratum and the ballots cast in legacy counties comprise a second stratum; every ballot cast in the contest is in exactly one of the two strata. We assume that the samples are drawn from the two strata independently.

As explained below, these stratified “hybrid” audits require the specification of some additional parameters: λ_1 for dividing the tolerable overstatement error up, and the strata risk limits $\{\alpha_s\}$.

4.1 Partitioning the total permissible overstatement into strata

The simplest approach to stratification involves partitioning the risk limit and the tolerable overstatement error of the tabulation into two pieces, one for the (pooled) CVR counties and one for the (pooled) no-CVR counties. Let $V_{w\ell} > 0$ denote the contest-wide margin (in votes) of reported winner w over reported loser ℓ . Let $V_{w\ell,s}$ denote the margin (in votes) of reported winner w over reported loser ℓ in stratum s . Note that $V_{w\ell,s}$ might be negative in one stratum. Let $A_{w\ell}$ denote the margin (in votes) of reported winner w over reported loser ℓ that a full hand count of the entire contest would show, that is, the *actual* margin rather than the *reported* margin. Reported winner w really beat reported loser ℓ if and only if $A_{w\ell} > 0$. Define $A_{w\ell,s}$ to be the actual margin (in votes) of w over ℓ in stratum s ; this too may be negative.

Let $\omega_{w\ell,s} \equiv V_{w\ell,s} - A_{w\ell,s}$ be the *overstatement* of the margin of w over ℓ in stratum s . Reported winner w really beat reported loser ℓ if and only if $\omega_{w\ell} \equiv \omega_{w\ell,1} + \omega_{w\ell,2} < V_{w\ell}$.

Pick $\lambda_1 \in \mathbb{R}$ and define $\lambda_2 = 1 - \lambda_1$. These values partition the total tolerable overstatement between the two strata: If $\omega_{w\ell,1} < \lambda_1 V_{w\ell}$ and $\omega_{w\ell,2} < \lambda_2 V_{w\ell}$, candidate w really received more votes than candidate ℓ . Some (λ_1, λ_2) pairs can be ruled out *a priori*, because (for instance) $\omega_{w\ell,s} \in [-2N_s, 2N_s]$, where N_s is the number of ballots cast in stratum s . There are other simple, sharper bounds, sketched below.

The choice of λ_1 (which determines the tolerable overstatement in each stratum), the strata risk limits $\{\alpha_s\}$, and details of the audit procedures affect the workload and the overall risk limit. (See section 4.1.1 and section 8.)

For ballot-level comparison audits, auditing to ensure that $\omega_{w\ell,s} < \lambda_s V_{w\ell}$ is discussed in section 6. It is a minor modification of the method embodied in RLATool.

For ballot-polling audits, auditing to ensure that $\omega_{w\ell,s} < \lambda_s V_{w\ell}$ is discussed in section 7. Note that this requires a more substantial modification of the standard ballot-polling calculations, because the standard calculations consider only the fraction of ballots with a vote for either w or ℓ that contain a vote for w , while we need to make an inference about the difference between the number of votes for w and the number of votes for ℓ . This introduces an additional unknown nuisance parameter, the number of ballots with votes for either w or ℓ .

4.1.1 Combining stratum-level risk limits

We audit to test the two hypotheses $\{\omega_{w\ell,s} \geq \lambda_s V_{w\ell}\}_{s=1}^2$, independently for the two strata. If we reject *both* hypotheses, we conclude that the contest outcome is correct; otherwise, we manually re-tabulate the contest in one or both strata, depending on the audit rules. Those rules matter: the two audits might need to be conducted to smaller risk limits individually than the desired risk limit for the contest as a whole.

Recall that the samples are drawn independently from the two strata. Pick $\alpha_1, \alpha_2 \in (0, \alpha)$. (Below we discuss the choice further.) We audit each stratum s to test the hypothesis $\omega_{w\ell,s} \geq \lambda_s V_{w\ell}$ (the overstatement exceeds the tolerable overstatement) at risk limit α_s , as if it were its own election. The audits can be conducted at the same time or sequentially; there is no coordination between the audits unless one of them leads to a full hand count but the other does not: see below.

How do these two stratum-level “risk limits” α_1 and α_2 determine the overall risk that the audit will not correct the outcome if the outcome is wrong? The overall risk depends on the rule for what we do if the audit in one stratum leads to a full manual tally of that stratum.

Here are the possibilities. Bear in mind that for the outcome to be wrong, at least one stratum must have a net overstatement greater its tolerable overstatement: That is, if $\omega_{w\ell,1} + \omega_{w\ell,2} \geq V_{w\ell}$, then $\omega_{w\ell,1} \geq \lambda_1 V_{w\ell}$ or $\omega_{w\ell,2} \geq \lambda_2 V_{w\ell}$, or both. If the tolerable overstatement is exceeded in only one stratum,

h , then the chance that the stratum will be fully hand counted is at least $1 - \alpha_h \geq 1 - \alpha$.

If both $\omega_{w\ell,1} \geq \lambda_1 V_{w\ell}$ and $\omega_{w\ell,2} \geq \lambda_2 V_{w\ell}$, then the chance both are completely tabulated by hand is at least $(1 - \alpha_1)(1 - \alpha_2)$, since the audit samples in the two strata are independent.

What should we do if the audit leads to a full tally in one stratum, h , that reveals that indeed its tolerable overstatement has been exceeded, but the other audit has not led to a full tabulation, because it has not started, because it is still underway, or because it terminated without a full hand tally? We consider two options. The simpler is to automatically require a full hand count of the other stratum. If the audit uses this rule, then we can take $\alpha_1 = \alpha_2 = \alpha$, and the procedure will have risk limit α . However, this rule creates the possibility of requiring a full hand count in circumstances where it may seem substantively superfluous. For instance, one can imagine an audit of a statewide contest in which the tolerable overstatement in no-CVR counties is exceeded, yet the outcome still could be verified without a full hand count in the CVR counties.

The second approach is to adjust the tolerable overstatement in the other stratum in light of the known manual tally $A_{w\ell,h}$ in the stratum h that has been fully hand tallied: we will test against the threshold $V_{w\ell} - A_{w\ell,h} \equiv \lambda'_t V_{w\ell}$, rather than the original value $\lambda_t V_{w\ell}$. (Because the overstatement in stratum h exceeded the tolerable overstatement, the updated tolerable overstatement in stratum t will be smaller than the original value.) Then to reject the new null hypothesis in stratum t is to conclude that the overall outcome is correct.

If and when the hypothesis in stratum t changes, the audit in that stratum might be able to stop on the basis of the data already observed; it might need to continue; or—if it had stopped based on the original threshold $\lambda_t V_{w\ell}$ —it might need to examine more ballots, possibly continuing to a full hand tally.

We will now show in detail that this rule allows the contest to be audited at risk limit α by selecting values of α_1 and α_2 that sum to a bit more than α : specifically, such that $(1 - \alpha_1)(1 - \alpha_2) < 1 - \alpha$. For instance, suppose we want the overall risk limit to be 5%. If we use a risk limit of 4% in the no-CVR stratum and a risk limit of 1.04% in the CVR stratum, the overall risk limit is not larger than $1 - (1 - \alpha_1)(1 - \alpha_2) \equiv 1 - 0.96 \times 0.9896 < 0.05$.

The statistical wrinkle is that adjusting for the manual tally in the hand-counted stratum h changes the hypothesis being tested in the other stratum t in a way that is itself random: whether the original null $\omega_{w\ell,s} \geq \lambda_t V_{w\ell}$ is tested or the new null $\omega_{w\ell,s} \geq \lambda'_t V_{w\ell}$ is tested depends on what the sample

reveals in stratum h . If the hypothesis does change, there is only one value possible for λ'_t —which depends on the reported margin $V_{w\ell}$ and the count $A_{w\ell,h}$ in stratum h —but λ'_t is unknown until $A_{w\ell,h}$ is known.

We assume that before any data are collected, the audit specifies two families of tests: for each stratum s , a family of level- α_s tests of the null hypothesis that the overstatement in the stratum is greater than or equal to c , for all feasible values of c . That is,

$$\Pr\{\text{reject hypothesis that } \omega_{w\ell,s} \geq c_s \mid \omega_{w\ell,s} \geq c_s\} \leq \alpha_s, \quad (1)$$

for $s = 1, 2$, and all feasible c_s . Moreover, we insist that the test depend on data only from ballots selected from its stratum. Because the samples in the two strata are independent, for all feasible pairs c_1, c_2 ,

$$\begin{aligned} & \Pr\{\text{reject neither hypothesis } \omega_{w\ell,s} \geq c_s, \ s = 1, 2 \mid \omega_{w\ell,s} \geq c_s \text{ for both } s = 1, 2\} \\ &= \prod_{s=1}^2 1 - \Pr\{\text{reject hypothesis that } \omega_{w\ell,s} \geq c_s \mid \omega_{w\ell,s} \geq c_s\} \\ &\geq (1 - \alpha_1)(1 - \alpha_2). \end{aligned} \quad (2)$$

What is the chance that the audit leads to a full hand tabulation if the outcome is incorrect? One way the audit can lead to a full hand tally is if it leads to a full count in one stratum, the null hypothesis in the other stratum is changed, and the audit in the second stratum then proceeds to a full manual tally. (There are other ways the audit can lead to a full hand tally, for instance, if neither null hypothesis is rejected, but this is one way.)

If the outcome is wrong, there is at least one stratum in which the overstatement $\omega_{w\ell,s}$ exceeds the threshold $\lambda_s V_{w\ell}$. Let h be one such stratum. Then the chance the audit in stratum h leads to a full manual tally in that stratum is at least $(1 - \alpha_h)$. If the audit leads to a full manual tally in stratum h and the overall outcome is wrong, then the (new) null hypothesis in the other stratum, t , must be true. If we started to audit that new hypothesis *ab initio*, the chance that we would reject it would be at most α_t , so the chance the audit would lead to a full hand count of stratum t is at least $1 - \alpha_t$. The question is whether “changing hypotheses” could make that chance smaller. The inequality 2 shows that it cannot: for any feasible pair of overstatements, $c = (c_1, c_2)$, if $\omega_{w\ell,1} \geq c_1$ and $\omega_{w\ell,2} \geq c_2$, the chance that neither the hypothesis $\omega_{w\ell,1} \geq c_1$ nor the hypothesis $\omega_{w\ell,2} \geq c_2$ will be rejected is at least $(1 - \alpha_1)(1 - \alpha_2)$.

And therefore, for this procedure, the chance that there will be a full hand count in both strata is at least $(1 - \alpha_1)(1 - \alpha_2)$ if the outcome is incorrect, even if the probability were zero that both of the original audits would proceed to a full hand count. The overall risk limit is thus not larger than $1 - (1 - \alpha_1)(1 - \alpha_2)$.

4.2 Constraining the total overstatement across strata

A more statistically efficient approach to ensuring that the overstatement error in the two strata does not exceed the margin is to try to constrain the *sum* of the overstatement errors in the two strata, rather than constrain the pieces separately. The null hypothesis $\omega_{w\ell,1} + \omega_{w\ell,2} \geq V_{w\ell}$ is true if and only if there exists *some* values of λ_1 and λ_2 such that $\omega_{w\ell,s} \geq \lambda_s V_{w\ell}$, $s = 1, 2$.⁸ Thus, fixing λ_1 and λ_2 at single values and requiring that we reject both stratum-level null hypotheses will be inefficient: there are many ways that the total overstatement could be less than $V_{w\ell}$ (i.e., the alternative hypothesis is true) without having the overstatement $\omega_{w\ell,s}$ in stratum s less than $\lambda_s V_{w\ell}$, $s = 1, 2$.

To that end, imagine *all* values ways of partitioning the error. If, for all (λ_1, λ_2) pairs, we can reject the hypothesis that the overstatement error in stratum 1 is greater than or equal to $\lambda_1 V_{w\ell}$ *and* the overstatement error in stratum 2 is greater than or equal to $\lambda_2 V_{w\ell}$, then we can conclude that the outcome is correct. This is more efficient because it only requires rejecting one of the two stratum-wise null hypotheses, for all possible (λ_1, λ_2) pairs, rather than rejecting *both* null hypotheses for a particular pair.

To test the conjunction hypothesis (i.e., that both of those null hypotheses are true), we use Fisher's combining function. Let $p_s(\lambda_s)$ be the p -value of the hypothesis $\omega_{w\ell,s} \geq \lambda_s V_{w\ell}$. If the null hypothesis that $\omega_{w\ell,1} \geq \lambda_1 V_{w\ell}$ and $\omega_{w\ell,2} \geq \lambda_2 V_{w\ell}$ is true, then the combination

$$\chi(\lambda_1, \lambda_2) = -2 \sum_{s=1}^2 \ln p_s(\lambda_s) \quad (3)$$

has a probability distribution that is dominated by the chi-square distribution with 4 degrees of freedom.⁹ Fisher's combined statistic will be small

⁸ Namely, letting $\lambda_1 = \frac{\omega_{w\ell,1}}{\omega_{w\ell,1} + \omega_{w\ell,2}}$ satisfies both inequalities.

⁹ If the two tests had continuously distributed p -values, the distribution would be exactly

when both null hypotheses are true and will be large when at least one null hypothesis is not true.

Hence, if, for all λ_1 and $\lambda_2 = 1 - \lambda_1$, the combined statistic $\chi(\lambda_1, \lambda_2)$ is greater than the $1 - \alpha$ quantile of the chi-square distribution with 4 degrees of freedom, the audit can stop.

This procedure involves maximizing Fisher’s combined statistic over all pairs (λ_1, λ_2) . The calculation of $p_s(\lambda)$ uses the procedures discussed in sections 6 and 7.

5 Sampling from subcollections

To audit contests that are contained on only a fraction of the ballots cast in one or more counties efficiently requires the ability to sample from just those ballots (or, at least, from a subset of all ballots that contains every such ballot). Because the CVRs cannot be entirely trusted (otherwise, the audit would be superfluous), we cannot rely on them to determine which ballots contain a given contest. However, if we have independent knowledge of the number of ballots that contain a given contest (e.g., from the SCORE system), then there are methods that allow the sample to be drawn from ballots whose CVRs contain the contest and still limit the risk rigorously. See Benaloh et al. (2011) and Bañuelos and Stark (2012) for details.

6 Batch comparison audits of a tolerable overstatement in votes

In this section we expand previous comparison auditing work (already embodied in RLATool) to handle two new requirements. The first allows the specification of the λ parameters discussed in section 4. The second handles batch-level auditing.

The first requirement requires that we consider auditing in a single stratum to test whether the overstatement of any margin (in votes) exceeds some fraction λ of the overall margin $V_{w\ell}$ between reported winner w and reported loser ℓ . If the stratum contains all the ballots cast in the contest, then for

chi-square with four degrees of freedom, but if either p -value has atoms when the null hypothesis is true, it is in general stochastically smaller. This follows from a coupling argument along the lines of Theorem 4.12.3 in Grimmett and Stirzaker (2001).

$\lambda = 1$, this would confirm the election outcome. For stratified audits, we might want to test other values of λ , as described above.

In Colorado, comparison audits have been ballot-level (i.e., batches consisting of a single ballot). This section also addresses the second requirement by deriving a method for batches of arbitrary size, which might be useful for Colorado to audit contests that include CVR counties and legacy counties. We keep the *a priori* error bounds tighter than the “super-simple” method (Stark, 2010). To keep the notation simpler, we consider only a single contest, but the MACRO test statistic (Stark, 2009b, 2010) automatically extends the result to auditing $C > 1$ contests simultaneously. The derivation is for plurality contests, including “vote-for- k ” plurality contests. Majority and super-majority contests are a minor modification (Stark, 2008).¹⁰

6.1 Notation

- \mathcal{W} : the set of reported winners of the contest
- \mathcal{L} : the set of reported losers of the contest
- N_s ballots were cast in all in the stratum. (The contest might not appear on all N_s ballots.)
- P “batches” of ballots are in stratum s . A batch contains one or more ballots. Every ballot in stratum s is in exactly one batch.
- n_p : number of ballots in batch p . $N_s = \sum_{p=1}^P n_p$.
- $v_{pi} \in \{0, 1\}$: the reported votes for candidate i in batch p
- $a_{pi} \in \{0, 1\}$: actual votes for candidate i in batch p . If the contest does not appear on any ballot in batch p , then $a_{pi} = 0$.
- $V_{w\ell,s} \equiv \sum_{p=1}^P (v_{pw} - v_{p\ell})$: Reported margin in stratum s of reported winner $w \in \mathcal{W}$ over reported loser $\ell \in \mathcal{L}$, in votes.
- $V_{w\ell}$: Overall reported margin of reported winner $w \in \mathcal{W}$ over reported loser $\ell \in \mathcal{L}$, in votes, for the entire contest (not just stratum s)

¹⁰So are some forms of preferential and approval voting, such as Borda count, and proportional representation contests, such as D’Hondt (Stark and Teague, 2014). Changes for IRV/STV are more complicated.

- V : smallest reported overall margin between any reported winner and reported loser: $V \equiv \min_{w \in \mathcal{W}, \ell \in \mathcal{L}} V_{w\ell}$
- $A_{w\ell, s} \equiv \sum_{p=1}^P (a_{pw} - a_{p\ell})$: actual margin in the stratum of reported winner $w \in \mathcal{W}$ over reported loser $\ell \in \mathcal{L}$, in votes
- $A_{w\ell}$: actual margin of reported winner $w \in \mathcal{W}$ over reported loser $\ell \in \mathcal{L}$, in votes, for the entire contest (not just in stratum s)

6.2 Reduction to maximum relative overstatement

If the contest is entirely contained in stratum s , then the reported winners of the contest are the actual winners if

$$\min_{w \in \mathcal{W}, \ell \in \mathcal{L}} A_{w\ell, s} > 0.$$

Here, we address the case that the contest may include a portion outside the stratum. To combine independent samples in different strata, it is convenient to be able to test whether the net overstatement error in a stratum exceeds a given threshold.

Instead of testing that condition directly, we will test a condition that is sufficient but not necessary for the inequality to hold, to get a computationally simple test that is still conservative (i.e., the risk is not larger than its nominal value).

For every winner, loser pair (w, ℓ) , we want to test whether the overstatement error exceeds some threshold, generally one tied to the reported margin between w and ℓ . For instance, for a simple stratified audit, we might take the threshold to be $\lambda_s V_{w\ell}$.

We want to test whether

$$\sum_{p=1}^P (v_{pw} - a_{pw} - v_{p\ell} + a_{p\ell}) / V_{w\ell} \geq \lambda_s.$$

The maximum of sums is not larger than the sum of the maxima; that is,

$$\max_{w \in \mathcal{W}, \ell \in \mathcal{L}} \sum_{p=1}^P (v_{pw} - a_{pw} - v_{p\ell} + a_{p\ell}) / V_{w\ell} \leq \sum_{p=1}^P \max_{w \in \mathcal{W}, \ell \in \mathcal{L}} (v_{pw} - a_{pw} - v_{p\ell} + a_{p\ell}) / V_{w\ell}.$$

Define

$$e_p \equiv \max_{w \in \mathcal{W}, \ell \in \mathcal{L}} (v_{pw} - a_{pw} - v_{p\ell} + a_{p\ell}) / V_{w\ell}.$$

Then no reported margin is overstated by a fraction λ_s or more if

$$E \equiv \sum_{p=1}^P e_p < \lambda_s.$$

Thus if we can reject the hypothesis $E \geq \lambda_s$, we can conclude that no pairwise margin was overstated by as much as a fraction λ_s .

Testing whether $E \geq \lambda_s$ would require a very large sample if we knew nothing at all about e_p without auditing batch p : a single large value of e_p could make E arbitrarily large. But there is an *a priori* upper bound for e_p . Whatever the reported votes v_{pi} are in batch p , we can find the potential values of the actual votes a_{pi} that would make the error e_p largest, because a_{pi} must be between 0 and n_p , the number of ballots in batch p :

$$\frac{v_{pw} - a_{pw} - v_{p\ell} + a_{p\ell}}{V_{w\ell}} \leq \frac{v_{pw} - 0 - v_{p\ell} + n_p}{V_{w\ell}}.$$

Hence,

$$e_p \leq \max_{w \in \mathcal{W}, \ell \in \mathcal{L}} \frac{v_{pw} - v_{p\ell} + n_p}{V_{w\ell}} \equiv u_p. \quad (4)$$

Knowing that $e_p \leq u_p$ might let us conclude reliably that $E < \lambda_s$ by examining only a small number of batches—depending on the values $\{u_p\}_{p=1}^P$ and on the values of $\{e_p\}$ for the audited batches.

To make inferences about E , it is helpful to work with the *taint* $t_p \equiv \frac{e_p}{u_p} \leq 1$. Define $U \equiv \sum_{p=1}^P u_p$. Suppose we draw batches at random with replacement, with probability u_p/U of drawing batch p in each draw, $p = 1, \dots, P$. (Since $u_p \geq 0$, these are all positive numbers, and they sum to 1, so they define a probability distribution on the P batches.)

Let T_j be the value of t_p for the batch p selected in the j th draw. Then $\{T_j\}_{j=1}^n$ are IID, $\mathbb{P}\{T_j \leq 1\} = 1$, and

$$\mathbb{E}T_1 = \sum_{p=1}^P \frac{u_p}{U} t_p = \frac{1}{U} \sum_{p=1}^P u_p \frac{e_p}{u_p} = \frac{1}{U} \sum_{p=1}^P e_p = E/U.$$

Thus $E = U\mathbb{E}T_1$.

So, if we have strong evidence that $\mathbb{E}T_1 < \lambda_s/U$, we have strong evidence that $E < \lambda_s$.

This approach can be simplified even further by noting that u_p has a simple upper bound that does not depend on v_{pi} . At worst, the reported result for batch p shows n_p votes for the “least-winning” apparent winner of the contest with the smallest margin, but a hand interpretation would show that all n_p ballots in the batch had votes for the runner-up in that contest. Since $V_{w\ell} \geq V$ and $0 \leq v_{pi} \leq n_p$,

$$u_p = \max_{w \in \mathcal{W}, \ell \in \mathcal{L}} \frac{v_{pw} - v_{p\ell} + n_p}{V_{w\ell}} \leq \max_{w \in \mathcal{W}, \ell \in \mathcal{L}} \frac{n_p - 0 + n_p}{V_{w\ell}} \leq \frac{2n_p}{V}.$$

Thus if we use $2n_p/V$ in lieu of u_p , we still get conservative results. (We also need to re-define U to be the sum of those upper bounds.) An intermediate, still conservative approach would be to use this upper bound for batches that consist of a single ballot, but use the sharper bound (4) when $n_p > 1$. Regardless, for the new definition of u_p and U , $\{T_j\}_{j=1}^n$ are IID, $\mathbb{P}\{T_j \leq 1\} = 1$, and

$$\mathbb{E}T_1 = \sum_{p=1}^P \frac{u_p}{U} t_p = \frac{1}{U} \sum_{p=1}^P u_p \frac{e_p}{u_p} = \frac{1}{U} \sum_{p=1}^P e_p = E/U.$$

So, if we have evidence that $\mathbb{E}T_1 < \lambda_s/U$, we have evidence that $E < \lambda_s$.

6.3 Testing $\mathbb{E}T_1 \geq \lambda_s/U$

To test whether $\mathbb{E}T_1 < \lambda_s/U$, there are a variety of methods available. One particularly “clean” sequential method is based on Wald’s Sequential Probability Ratio Test (SPRT) (Wald (1945)). Harold Kaplan pointed out this method on a website that no longer exists. A derivation of this “Kaplan-Wald” method is given in Stark and Teague (2014, Appendix A); to apply the method here, take $t = \lambda_s$ in their equation 18.

A different sequential method, the Kaplan-Markov method (also due to Harold Kaplan), is given in Stark (2009a).

7 Ballot-polling audits of a tolerable over-statement in votes

7.1 Conditional tri-hypergeometric test

We consider a single stratum s , containing N_s ballots. We will sample individual ballots without replacement from stratum s . Of the N_s ballots, $A_{w,s}$ have a vote for w but not for ℓ , $A_{\ell,s}$ have a vote for ℓ but not for w , and $A_{u,s} = N_s - A_{w,s} - A_{\ell,s}$ have votes for both w and ℓ or neither w nor ℓ , including undervotes and invalid ballots. We might draw a simple random sample of n ballots (n fixed ahead of time), or we might draw sequentially without replacement, so the sample size B could be random. For instance, the rule for determining B could depend on the data.¹¹

Regardless, we assume that, conditional on the attained sample size n , the ballots are a simple random sample of size n from the N_s ballots in the population. In the sample, B_w ballots contain a vote for w but not ℓ , with B_ℓ and B_u defined analogously. Then, conditional on $B = n$, the joint distribution of (B_w, B_ℓ, B_u) is tri-hypergeometric:

$$\mathbb{P}_{A_{w,s}, A_{\ell,s}}\{B_w = i, B_\ell = j | B = n\} = \frac{\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{j} \binom{N_s - A_{w,s} - A_{\ell,s}}{n - i - j}}{\binom{N_s}{n}}. \quad (5)$$

The test statistic will be the diluted sample margin, $D \equiv (B_w - B_\ell)/B$. This is the sample difference in the number of ballots for the winner and for the loser, divided by the total number of ballots in the sample. We want to test the compound hypothesis $A_{w,s} - A_{\ell,s} \leq c$. The value of c is inferred from the definition $\omega_{w\ell,s} \equiv V_{w\ell,s} - A_{w\ell,s} = V_{w,s} - V_{\ell,s} - (A_{w,s} - A_{\ell,s})$. Thus,

$$c = V_{w,s} - V_{\ell,s} - \omega_{w\ell,s} = V_{w\ell,s} - \lambda_s V_{w\ell}.$$

The alternative is the compound hypothesis $A_{w,s} - A_{\ell,s} > c$.¹² Hence, we will reject for large values of D . Conditional on $B = n$, the event $D = (B_w - B_\ell)/B = d$ is the event $B_w - B_\ell = nd$.

¹¹Sampling with replacement leads to simpler arithmetic, but is not as efficient.

¹²To use Wald's Sequential Probability Ratio Test, we might pick a simple alternative instead, e.g., $A_{w,s} = V_{w,s}$ and $A_{\ell,s} = V_{\ell,s}$, the reported values, provided $V_{w,s} - V_{\ell,s} > c$.

Suppose we observe $D = d$. The test will condition on the event $B = n$. (In contrast, the BRAVO ballot-polling method (Lindeman et al., 2012) conditions only on $B_w + B_\ell = m$.)

The p -value of the simple hypothesis that there are $A_{w,s}$ ballots with a vote for w but not for ℓ , $A_{\ell,s}$ ballots with a vote for ℓ but not for w , and $N - A_{w,s} - A_{\ell,s}$ ballots with votes for both w and ℓ or neither w nor ℓ (including undervotes and invalid ballots) is the sum of these probabilities for events when $B_w - B_\ell \geq nd$. Therefore,

$$\mathbb{P}_{A_{w,s}, A_{\ell,s}, N_s} \{D \geq d \mid B = n\} = \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} \frac{\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{j} \binom{N_s - A_{w,s} - A_{\ell,s}}{n-i-j}}{\binom{N_s}{n}}. \quad (6)$$

7.2 Conditional hypergeometric test

Another approach is to condition on both the events $B = n$ and $B_w + B_\ell = m$. We describe the hypothesis test here, but do not advocate for using it. We found that this approach was inefficient in some simulation experiments.

Given $B = n$, all samples of size n from the ballots are equally likely, by hypothesis. Hence, in particular, all samples of size n for which $B_w + B_\ell = m$ are equally likely. There are $\binom{A_{w,s} + A_{\ell,s}}{m} \binom{N_s - A_{w,s} - A_{\ell,s}}{n-m}$ such samples. Among these samples, B_w may take values $i = 0, 1, \dots, m$. For a fixed i , there are $\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{m-i} \binom{N_s - A_{w,s} - A_{\ell,s}}{n-m}$ samples with $B_w = i$ and $B_\ell = m - i$.

The factor $\binom{N_s - A_{w,s} - A_{\ell,s}}{n-m}$ counts the number of ways to sample $n - m$ of the remaining ballots. If we divide out this factor, we simply count the number of ways to sample ballots from the group of ballots for w or for ℓ . There are $\binom{A_{w,s} + A_{\ell,s}}{m}$ equally likely samples of size m from the ballots with either a vote for w or for ℓ , but not both, and of these samples, $\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{m-i}$ contain i ballots with a vote for w but not ℓ . Therefore, conditional on $B = n$ and $B_w + B_\ell = m$, the probability that $B_w = i$ is

$$\frac{\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{m-i}}{\binom{A_{w,s} + A_{\ell,s}}{m}}.$$

The p -value of the simple hypothesis that there are $A_{w,s}$ ballots with a vote for w but not for ℓ , $A_{\ell,s}$ ballots with a vote for ℓ but not for w , and $N - A_{w,s} - A_{\ell,s}$ ballots with votes for both w and ℓ or neither w nor ℓ

(including undervotes and invalid ballots) is the sum of these probabilities for events when $B_w - B_\ell \geq nd$. This event occurs for $B_w \geq \frac{m+nd}{2}$. Therefore,

$$\mathbb{P}_{A_{w,s}, A_{\ell,s}, N_s} \{D \geq d \mid B = n, B_w + B_\ell = m\} = \sum_{i=(m+nd)/2}^{\min\{m, A_{w,s}\}} \frac{\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{m-i}}{\binom{A_{w,s}+A_{\ell,s}}{m}}. \quad (7)$$

This conditional p -value is thus the tail probability of the hypergeometric distribution with parameters $A_{w,s}$ “good” items, $A_{\ell,s}$ “bad” items, and a sample of size m . This calculation is numerically stable and fast; tail probabilities of the hypergeometric distribution are available and well-tested in all standard statistics software.

7.3 Maximizing the p -value over the null set

The composite null hypothesis does not specify $A_{w,s}$ or $A_{\ell,s}$ separately, only that $A_{w,s} - A_{\ell,s} \leq c$ for some fixed, known c . The (conditional) p -value of this composite hypothesis for $D = d$ is the maximum p -value for all values $(A_{w,s}, A_{\ell,s})$ that are possible under the null hypothesis,

$$\max_{A_{w,s}, A_{\ell,s} \in \{0, 1, \dots, N\} : A_{w,s} - A_{\ell,s} \leq c, A_{w,s} + A_{\ell,s} \leq N_s} \sum_{\substack{(i,j) : i, j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} \frac{\binom{A_{w,s}}{i} \binom{A_{\ell,s}}{j} \binom{N_s - A_{w,s} - A_{\ell,s}}{n-i-j}}{\binom{N_s}{n}}, \quad (8)$$

wherever the summand is defined. (Equivalently, define $\binom{m}{k} \equiv 0$ if $k > m$, $k < 0$, or $m \leq 0$.)

7.3.1 Optimizing over the parameter c

The following result enables us to only test hypotheses along the boundary of the null set.

Theorem 1. *Assume that $n < A_{w,s} + A_{\ell,s}$. Suppose the composite null hypothesis is $N_w - N_\ell \leq c$. The p -value is maximized on the boundary of the null region, i.e. when $N_w - N_\ell = c$.*

Proof. Without loss of generality, let $c = 0$ and assume that $A_{w,s} = N_s - A_{w,s} - A_{\ell,s}$ is fixed. Let $N_{w\ell,s} \equiv A_{w,s} + A_{\ell,s}$ be the fixed, unknown number of

ballots for w or for ℓ in stratum s . The p -value p_0 for the simple hypothesis that $c = 0$ is

$$p_0 = \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} \frac{\binom{N_{w\ell,s}/2}{i} \binom{N_{w\ell,s}/2}{j} \binom{A_{u,s}}{n-i-j}}{\binom{N_s}{n}} = \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij}, \quad (9)$$

where T_{ij} is defined as the (i, j) term in the summand and $T_{ij} \equiv 0$ for pairs (i, j) that don't appear in the summation.

Assume that $c > 0$ is given. The p -value p_c for this simple hypothesis is

$$\begin{aligned} p_c &= \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} \frac{\binom{(N_{w\ell,s}+c)/2}{i} \binom{(N_{w\ell,s}-c)/2}{j} \binom{A_{u,s}}{n-i-j}}{\binom{N_s}{n}} \\ &= \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij} \frac{\frac{N_{w\ell,s}+c}{2} (\frac{N_{w\ell,s}+c}{2} - 1) \cdots (\frac{N_{w\ell,s}}{2} + 1) (\frac{N_{w\ell,s}-c}{2} - j) \cdots (\frac{N_{w\ell,s}}{2} - 1 - j)}{(\frac{N_{w\ell,s}+c}{2} - i) \cdots (\frac{N_{w\ell,s}}{2} + 1 - i) (\frac{N_{w\ell,s}-c}{2}) \cdots (\frac{N_{w\ell,s}}{2} - 1)}. \end{aligned}$$

Terms in the fraction can be simplified: choose the corresponding pairs in the numerator and denominator. Fractions of the form $\frac{\frac{N_{w\ell,s}}{2} + a}{\frac{N_{w\ell,s}}{2} + a - i}$ can be expressed as $1 + \frac{i}{\frac{N_{w\ell,s}}{2} + a - i}$. Fractions of the form $\frac{\frac{N_{w\ell,s}}{2} - a - j}{\frac{N_{w\ell,s}}{2} - a}$ can be expressed as $1 - \frac{j}{\frac{N_{w\ell,s}}{2} - a}$. Thus, the p -value can be written as

$$\begin{aligned}
p_c &= \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij} \prod_{a=1}^{c/2} \left(1 + \frac{i}{\frac{N_{w\ell,s}}{2} + a - i} \right) \left(1 - \frac{j}{\frac{N_{w\ell,s}}{2} - a} \right) \\
&> \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij} \left[\left(1 + \frac{i}{\frac{N_{w\ell,s}+c}{2} - i} \right) \left(1 - \frac{j}{\frac{N_{w\ell,s}}{2} + 1} \right) \right]^{c/2} \\
&= \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij} \left[1 + \frac{\frac{N_{w\ell,s}+c}{2}j + \frac{N_{w\ell,s}}{2}i + i}{\left(\frac{N_{w\ell,s}+c}{2} - i\right)\left(\frac{N_{w\ell,s}}{2} + 1\right)} \right]^{c/2} \\
&> \sum_{\substack{(i,j): i,j \geq 0 \\ i-j \geq nd \\ i+j \leq n}} T_{ij} \\
&= p_0
\end{aligned}$$

The last inequality follows from the fact that i and j are nonnegative, and that $i < \frac{N_{w\ell,s}+c}{2} - i$ is a possible outcome under the null hypothesis. \square

7.3.2 Optimizing over the parameter $A_{w,s}$

We have shown empirically (but do not prove) that this tail probability, as a function of $A_{w,s}$, has a unique maximum at one of the endpoints when $A_{w,s}$ is either as small or as large as possible, given N , c , and the observed sample values B_w and B_ℓ . If the empirical result is true, then finding the maximum is trivial; otherwise, it is a trivial one-dimensional optimization problem to compute the unconditional p -value.

7.4 Conditional testing

If the conditional tests are always conducted at significance level α or less, i.e., so that $\mathbb{P}\{\text{Type I error} | B = n\} \leq \alpha$, then the overall procedure has

significance level α or less:

$$\begin{aligned}\mathbb{P}\{\text{Type I error}\} &= \sum_{n=0}^N \{\text{Type I error} | B = n\} \mathbb{P}\{B = n\} \\ &\leq \sum_{n=0}^N \alpha \mathbb{P}\{B = n\} = \alpha.\end{aligned}\tag{10}$$

In particular, this implies that our conditional hypergeometric test will have the correct risk limit unconditionally.

8 Recommendations

We have outlined several methods Colorado might use to audit cross-jurisdictional contests that include CVR counties and no-CVR counties. We expect that stratified “hybrid” audits will be the most palatable, given the constraints on time for software development and the logistics of the audit itself, because the workflow for counties would be the same as it was in November, 2017.

What would change is the risk calculation “behind the scene,” including the algorithms used to decide when the audit can stop. Those algorithms could be implemented in software external to RLATool. The minimal modification to RLATool that would be required to conduct a hybrid audit is to allow the sample size from each county to be controlled externally, e.g. by uploading a parameter file once per round, rather than using a formula that is based on the margin within that county alone. The parameter file would be generated by external software that does the audit calculations described here based on the detailed audit progress and discrepancy data available from RLATools’ `rla_export` command.

To conduct a hybrid audit, one must choose two numbers in addition to the risk limit α :

- one stratum-wise risk limit, α_1 (the other, α_2 , is determined from α_1 and the overall risk limit, α)
- the tradeoff (allocation) of the tolerable overstatement between strata, λ_1 (the value of λ_2 is $1 - \lambda_1$)

Those parameters can be chosen essentially arbitrarily (provided $\alpha_1 \leq \alpha$) and the audit will still be risk-limiting; however, they can be optimized to reduce

the expected workload under various assumptions about tabulation errors in the two strata. (Software that can be used to run scenarios is available at <https://www.github.com/pbstark/CORLA18>; see below.)

In either stratum, increasing the risk limit or increasing the tolerable overstatement will decrease the required sample size from that stratum (assuming that the actual overstatement in that stratum is less than its allowable overstatement). The relative change in sample size as the risk limit changes scales similarly in the two strata, because the risk limit enters both ballot-level comparisons and ballot-polling the same way: as the logarithm. However, the relative change in sample sizes as the tolerable overstatement changes scales quite differently in the two strata: linearly in the ballot-level comparison stratum, but quadratically in the ballot-polling stratum. Hence, the workload is not as sensitive to how the risk limit is allocated across strata as it is to how the tolerable overstatement is allocated.

8.1 Software and examples

Examples of stratified hybrid audits are in Jupyter notebooks available at <https://www.github.com/pbstark/CORLA18>. The first two examples are contained in a single notebook, “hybrid-audit-example-1”. The first example is a hypothetical medium-sized election with a total of 110,000 votes and a diluted margin of 1.8%. 9.1% of the ballots come from no-CVR counties. The risk limit is 10%. If the audit in the CVR stratum found no errors and the allowable overstatement error was 30% of the margin, it would terminate after examining 1,213 ballots. In over 90% of 10,000 simulations, an audit of 250 ballots from the no-CVR stratum would have sufficed to confirm that the overstatement error in that stratum did not exceed its allocation, 70% of the margin. A sample of 450 ballots was sufficient to stop the audit in 99% of simulations. As always, λ_1 could be adjusted to rebalance the expected workload between strata, perhaps taking into account the expected workload for audits of countywide or intra-county contests, so as to minimize (or quite possibly eliminate) any additional burden imposed by the stratified audit.

If a CVR were available for all counties and we could have run a ballot-level comparison audit for the entire contest, rather than stratifying, an audit with risk limit 10% that found no errors would have concluded after examining just 263 ballots. The efficiency gained comes from two sources. First, ballot-level comparison audits are substantially more efficient than ballot-polling audits. If no CVR were available in any county and we had run a

ballot-polling audit for the entire contest, the expected sample size needed to confirm the audit with a risk limit of 10% would have been 13,988. The ballot-polling audit would require checking over 10% of ballots cast, an infeasible number in practice. Second, the hybrid audit requires dividing the margin and risk limit between two strata. This results in both strata using smaller risk limits. In order to keep the workload low, it is necessary to allocate a disproportionately high fraction of the margin to the no-CVR stratum; the CVR stratum must increase its workload to compensate.

Instead of fixing λ_1 and λ_2 , we could use Fisher’s method to combine the independent audits in each stratum over all values of (λ_1, λ_2) . In 93 of 100 simulations, an audit with samples of 400 ballots in the CVR stratum and 600 ballots in the no-CVR stratum and risk limit 10% would have sufficed to confirm that the total overstatement error did not exceed the overall margin. This combined method would require a total of 1,000 ballots to be examined, rather than 1,513 ballots using the stratified method with fixed allocation of allowable errors to each stratum. The workload is more than a pure ballot-comparison audit, but substantially less than a pure ballot-polling audit.

Another method discussed in Section 2.3 is to perform a ballot-level comparison audit statewide, but to treat any ballot sampled from the no-CVR county as showing a two-vote overstatement. In this example, this worst-case method would lead to a full hand count. However, the situation may be more optimistic for Colorado: if only 1.2% of ballots came from the no-CVR stratum and the overall margin were in fact 10,000 votes, then this method would require checking 430 ballots.

The second example is a hypothetical large statewide election with a total of 2 million ballots and a diluted margin of nearly 20%. The risk limit is 5%. If the audit in the CVR stratum found no errors and the allowable overstatement error was 10% of the margin, it would terminate after examining 50 ballots. In over 90% of 10,000 simulations, an audit of 50 ballots from the no-CVR stratum would have sufficed to confirm that the overstatement error in that stratum did not exceed its allocation, 90% of the margin. A sample of 100 ballots was sufficient to stop the audit in 99% of simulations. If a CVR were available for all counties and we could have run a ballot-level comparison audit for the entire contest, rather than stratifying, an audit with risk limit 5% that found no errors would have concluded after examining just 31 ballots. Using Fisher’s method to combine the audits in each stratum would reduce the workload in the no-CVR stratum: In 100% of 10,000 simulations, an audit with samples of 50 ballots in the CVR stratum and 20 ballots in the

no-CVR stratum would have sufficed to confirm that the total overstatement error did not exceed the overall margin at the 5% risk limit.

A second notebook, “hybrid-audit-example-2,” illustrates the workflow for conducting a hybrid audit of this kind. The example election has a total of 2 million ballots. The reported margin is just over 1%, but in reality the vote totals for the reported winner and reported loser are identical in both strata. The risk limit is 5%. The example illustrates three scenarios. In the first scenario, the audit in the CVR stratum escalates to a full hand count and the allowable overstatement in the no-CVR stratum must be adjusted. Using the new allowable overstatement in the no-CVR stratum makes it impossible to terminate the audit, even for samples as large as 5% of the ballots. In the second scenario, the audit in the no-CVR stratum terminates with a sample of 500 ballots. However, the audit in the CVR stratum will still lead to a full hand count and the audit in the no-CVR stratum must be redone using the adjusted allowable overstatement, putting us back in the first scenario. In the third scenario, we use Fisher’s method to combine the audits in the CVR stratum and no-CVR stratum. The sample sizes in each stratum are large, but the maximum Fisher’s combined p -value is over 20%, so the audit cannot be terminated. In each case, the audit leads to a full recount of all the ballots.

These notebooks can be modified and run with different contest sizes, margins, risk limits, and allocations of allowable error, in order to estimate the workload of different scenarios.

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