

# Simulations of Ballot Polling Risk-Limiting Audits

Oliver Broadrick<sup>1</sup>   Sarah Morin<sup>1</sup>   Grant McClearn<sup>2</sup>  
Neal McBurnett   Poorvi L. Vora<sup>1</sup>   Filip Zagórski<sup>3,4</sup>

<sup>1</sup>Department of Computer Science, The George Washington University  
(odbroadrick@gmail.com)

<sup>2</sup>Department of Computer Science, Stanford University (grantmcc@stanford.edu)

<sup>3</sup>Wroclaw University of Science and Technology (filip.zagorski@gmail.com)

<sup>4</sup>Votifica

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# Outline

- ▶ Risk-Limiting Audits (RLAs)
  - ▶ BRAVO and MINERVA

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- ▶ Discussion and Future Work

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- ▶ Compliance and tabulation audits
- ▶ Risk-Limiting Audits
  - ▶ Given that the election outcome is incorrect, the probability with which the audit stops, declaring the outcome correct, is at most the risk limit,  $\alpha$ .

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  1. Election results announced
  2. In a public procedure, sample ballots at random and manually interpret them
  3. Compute a pre-specified error measure, the maximum risk, and compare to the risk limit
    - ▶ If smaller, stop the audit
    - ▶ Else, sample more (goto 2)

- ▶ Most commonly used ballot polling RLA



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- ▶ Is thus the most efficient RLA when the decision of whether to stop the audit is made after each ballot is drawn (ballot-by-ballot)
- ▶ In real audits, decisions are taken after many ballots are drawn (round-by-round)
- ▶ BRAVO can be implemented as:
  - ▶ Selection-Ordered (SO) BRAVO, where ballot selection order is retained, and the decisions are taken as though the audit were ballot-by-ballot
  - ▶ End-of-Round (EoR) BRAVO, where the decision using the BRAVO stopping rule is taken once, after the entire round of ballots is drawn

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  - ▶ Unknown how the audits compare for smaller stopping probability or for rounds after the first

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- ▶ Simulate RLAs for election results from the 2020 Presidential election (all margins above 0.05)
  - ▶  $10000 = 10^4$  trials assuming the underlying election was as announced
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- ▶ Stopping probabilities: 0.90 and 0.25

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An audit  $\mathcal{A}$  takes a sample of ballots  $X$  as input and gives as output either (1) *Correct*: the audit is complete, or (2) *Uncertain*: continue the audit.

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- ▶ The tie is the hardest incorrect outcome to detect
- ▶ Probability of stopping given a tie should be low
- ▶ Probability of stopping given a correctly announced outcome should be high for as few ballots as possible

# Experiments

## Definition (Maximum Risk)

The maximum risk  $R$  of audit  $\mathcal{A}$  with sample  $X \in \{0, 1\}^*$  drawn from the ballots is  $R(\mathcal{A}) = \Pr[\mathcal{A}(X) = \textit{Correct} \mid H_0]$ .

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## Definition (Risk-Limiting Audit ( $\alpha$ -RLA))

An audit  $\mathcal{A}$  is a Risk-Limiting Audit with risk limit  $\alpha$  iff  $R(\mathcal{A}) \leq \alpha$ .



# Experiments

## Definition (Stopping Probability)

The stopping probability  $S_j$  of an audit  $\mathcal{A}$  in round  $j$  is  $S_j(\mathcal{A}) =$

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## Definition (Conditional Stopping Probability)

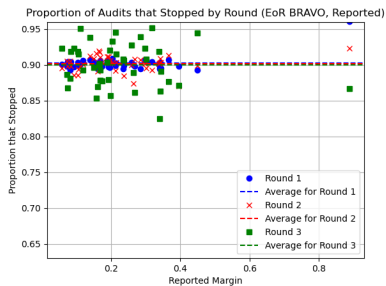
The conditional stopping probability of an audit  $\mathcal{A}$  in round  $j$  is

$$\chi_j(\mathcal{A}) =$$

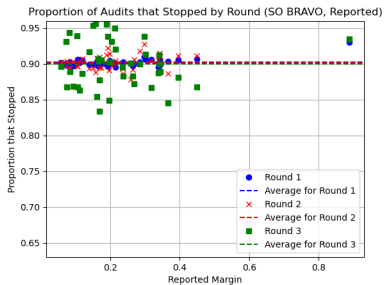
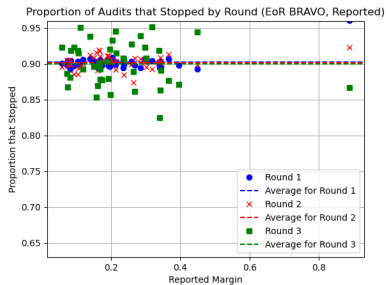
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Results: Stopping Probability ( $\chi_1 = 0.9$ )

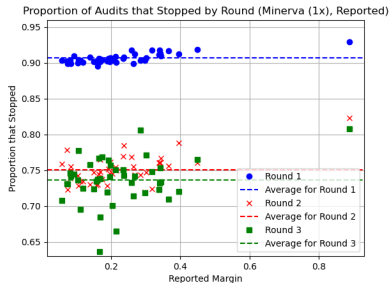
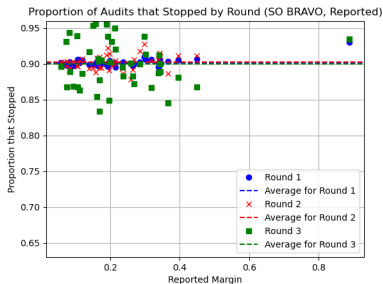
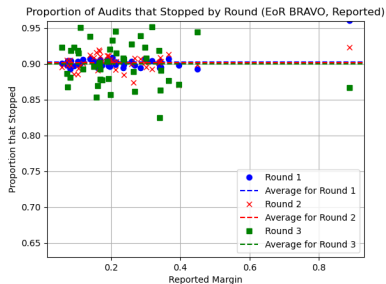
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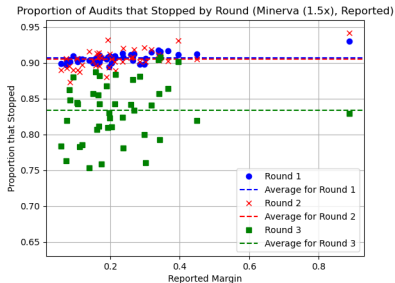
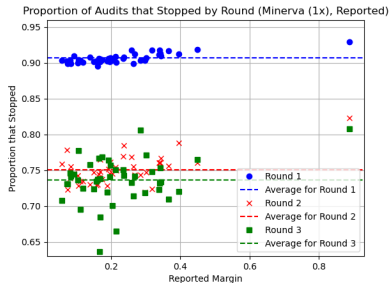
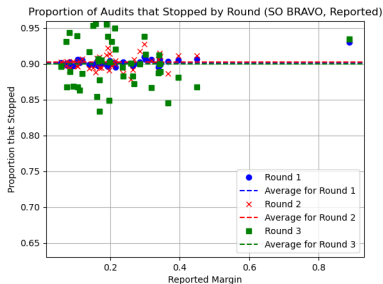
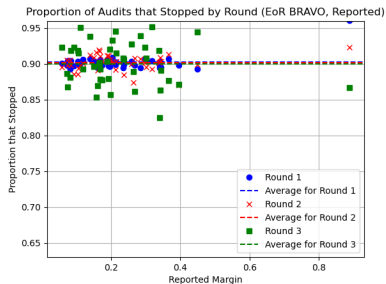
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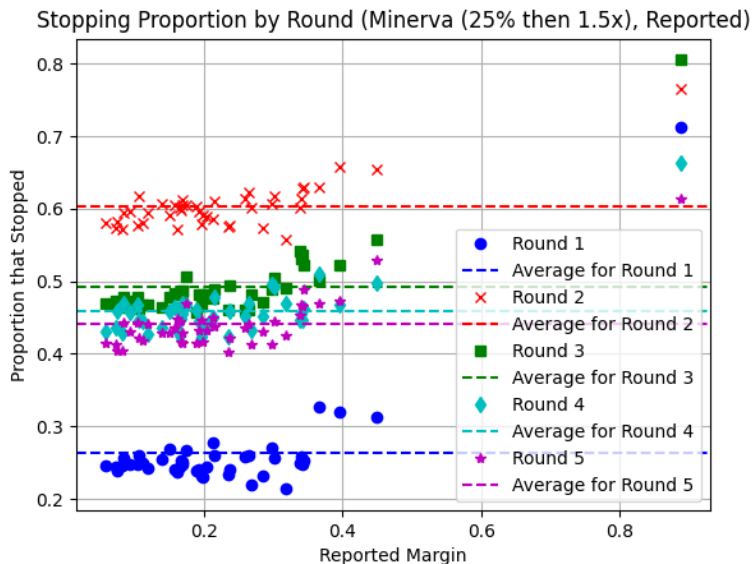


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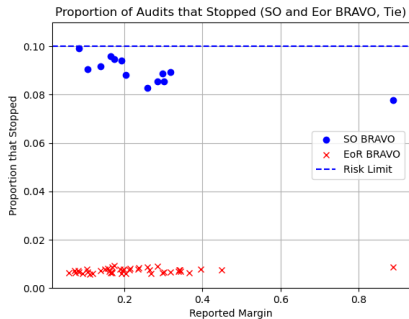




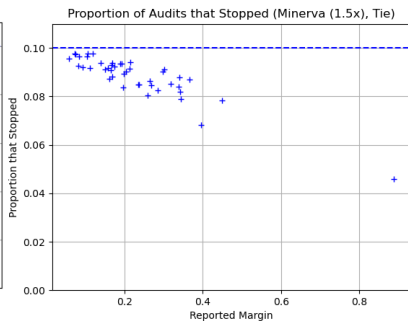
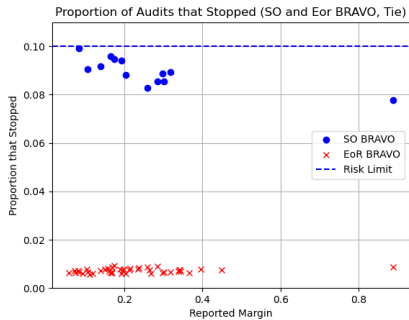
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# Results: Risk ( $\chi_1 = 0.9$ )



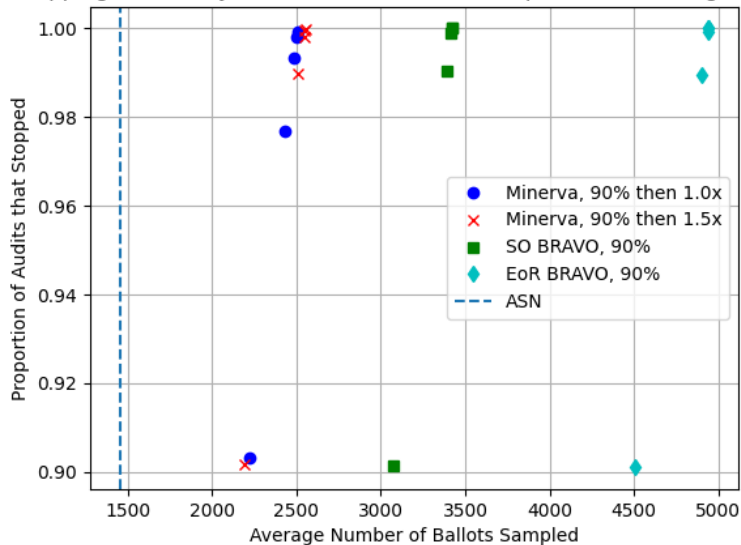
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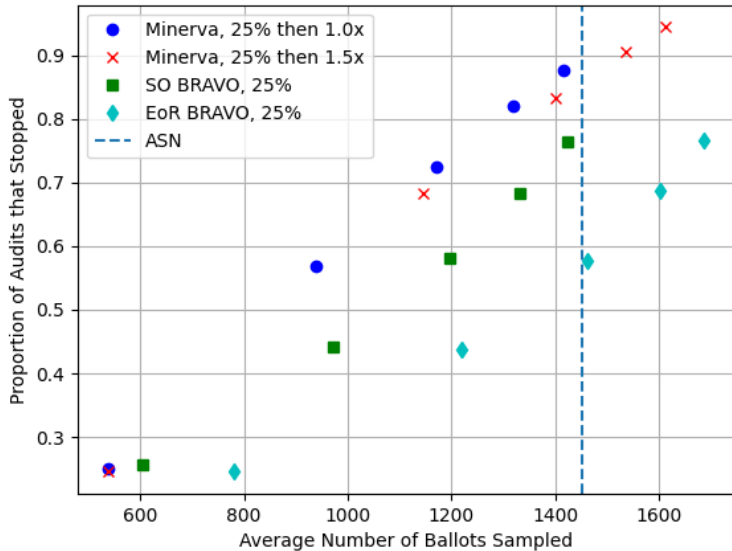
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Stopping Probability for Number of Ballots Sampled [Texas: margin 0.057]



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- ▶ EoR BRAVO:
  - ▶ for  $\chi_1 = 0.9$  requires twice as many as MINERVA
  - ▶ for  $\chi_1 = 0.25$  requires a fourth to a half more (depending on margin) than does MINERVA

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- ▶ with multiplying factor 1,  $\chi_2 \approx 0.75$  and  $\chi_3 \approx 0.74$
- ▶ with multiplying factor 1.5,  $\chi_2 \approx 0.91$  and  $\chi_3 \approx 0.83$

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## Future Work

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- ▶ Simulations with other underlying distributions

Thank you

odbroadrick@gmail.com