#### **Bayesian Post-Election Audits**

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1

#### Outline

**Post-Election Audits** 

**Bayesian Ballot-Polling** 

**Bayesian Comparison Audits** 

**Experimental Results** 

Lessons and Open Questions

### **Post-Election Audit Objectives**

By examining by hand sufficiently many randomly selected paper ballots:

Confirm to a high degree of confidence that the reported (scanner-based) outcome is correct or else that the actual (full hand-count) outcome is different.

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By examining by hand sufficiently many randomly selected paper ballots:

- Confirm to a high degree of confidence that the reported (scanner-based) outcome is correct or else that the actual (full hand-count) outcome is different.
- Convince the losers they really lost!

### Single-ballot Audits

- Sequential decision-making (Wald).
- Examine paper ballots one at a time, in random order.
- Determine actual type of each ballot (as opposed to its reported type).
- At each stage, decide whether to
  - Stop: Reported outcome looks **OK**.
  - Continue: more auditing needed.

(We assume that full hand count needed to overturn reported outcome.)

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Given what you've seen in the audit so far, what is the probability that each candidate would win if all ballots were examined?

Then you can stop audit if/when the reported winner has at least (say) 95% probability of winning.

## Actual ballot types (by hand): ???????????...

Probability A wins: 50.0% Probability B wins: 50.0%

## Actual ballot types (by hand): A??????????... Probability A wins: 75.0% Probability B wins: 25.0%

## Actual ballot types (by hand): A A ? ? ? ? ? ? ? ? ? ? ... Probability A wins: 87.5% Probability B wins: 12.5%

22

### Actual ballot types (by hand): A A B ? ? ? ? ? ? ? ? ...

Probability A wins: 68.8% Probability B wins: 31.2%

### Actual ballot types (by hand): A A B B ? ? ? ? ? ? ? ...

Probability A wins: 50.0% Probability B wins: 50.0%

### Actual ballot types (by hand): A A B B A ? ? ? ? ? ...

Probability A wins: 65.6% Probability B wins: 34.4%

### Actual ballot types (by hand): A A B B A A ? ? ? ? . . .

Probability A wins: 77.4% Probability B wins: 22.6%

### Actual ballot types (by hand): A A B B A A A ? ? ? . . .

Probability A wins: 85.6% Probability B wins: 14.4%

### Actual ballot types (by hand): A A B B A A A A ? ?...

Probability A wins: 91.0% Probability B wins: 9.0%

### Actual ballot types (by hand): A A B B A A A A A ? . . .

Probability A wins: 94.5% Probability B wins: 5.5%

### Actual ballot types (by hand): A A B B A A A A A A A ...

Probability A wins: 96.7% Probability B wins: 3.3%

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 $\rightarrow$  Stop auditing!  $\leftarrow$ 

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#### A ? A ? ?

Q: What is the probability that A won?

### Answer (Bayesian)

# To make Q well-posed, need a model (a prior) for the likelihood of different outcomes.

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#### 95%

If your error limit is 5%, stop auditing!

	tally	5:0	<b>4:1</b>	<mark>3:2</mark>	<b>2:3</b>	1:4	0:5
_	prior	1/6	1/6	1/6	1/6	1/6	1/6

tally	5: <mark>0</mark>	4:1	<b>3:2</b>	2:3	1:4	<b>0:5</b>
prior	1/6	1/6	1/6	1/6	1/6	1/6
likelihood(AA)	$\frac{5}{5} \cdot \frac{4}{4}$	$\frac{4}{5} \cdot \frac{3}{4}$	$\frac{3}{5} \cdot \frac{2}{4}$	$\frac{2}{5} \cdot \frac{1}{4}$	0	0

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product	<u>10</u> 60	$\frac{6}{60}$	$\frac{3}{60}$	1 60	0	0

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product	<u>10</u> 60	$\frac{6}{60}$	$\frac{3}{60}$	<u>1</u> 60	0	0
posterior	<u>10</u> 20	<u>6</u> 20	$\frac{3}{20}$	1 20	0	0

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	A wins 95%				vins 5	5%

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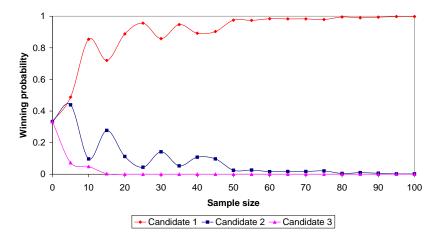
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Can sample faster using gamma variates (see paper).



#### Winning probabilities vs. sample size in a Bayes audit

### Arbitrary voting system

We note that a Bayes audit works for an *arbitrary voting system* as long as the number of ballot types is not too large; all you need is a way to compute the winner of a profile of ballots, and a way of sampling ballots. We have tested it on

- plurality
- IRV
- Borda
- Schulze

with good results.

#### Bayesian comparison audits

- Same idea, but have one urn for each reported type.
- Much more efficient!! (But needs way of matching paper ballots with their reported types.)

2011 Monterey Peninsula Water Mgt District Director

- Ballot-polling.
- Two candidates (plus write-ins).
- 2011 votes cast: 1353 for Lewis, 742 for Mancini (reported).
- Stark's ballot-polling audit with 10% risk limit examined:

89 ballots.

A Bayes ballot-polling audit with *ϵ* = 0.10 examines:

23 ballots on average 11 ballots (median)

#### 2011 Stanislaus Oakdale Measure O

- Comparison audit.
- Yes/No proposition.
- 3152 votes cast: 1728 Yes, 1392 No, 32 undervotes (reported).
- Stark's comparison audit with 10% risk limit examined:

49 ballots.

A Bayes ballot-polling audit with *ϵ* = 0.10 examines:

92 ballots (average) 39 ballots (median).

#### Discussion

- We conjecture that a Bayes audit is in fact "risk-limiting" (perhaps given some suitable assumptions or constant factors in parameterization). But this is just a conjecture.
- The Bayes audit admits the use of other priors, such as those a very partisan observer might have.
- The Bayes audit admits the use of *multiple* priors; only stopping when all auditors (with different priors) agree to do so.

Summary – Bayes Audit Advantages

- High efficiency (few ballots get audited).
- Small/controllable miscertification rates observed.
- Simple in structure / easy to implement.
- Handles ballot-polling audits, comparison audits, and many different voting systems.
- No MOV computation required to start.
- Admits flexible (multiple) choice(s) of prior.
- Can be stopped early with meaningful results.

Summary – Bayes Audit Disadvantages

- Only works (so far) for single-ballot audits.
- Unclear relationship to risk-limiting audits.
- Results depend on choice(s) for prior.
- Need program to compute winning probabilities.

# The End

#### For more info and code, contact authors or see:

http://people.csail.mit.edu/rivest/bayes/