

Optimizing Research Payoff, 1: A Simple Way to Increase Research Efficiency

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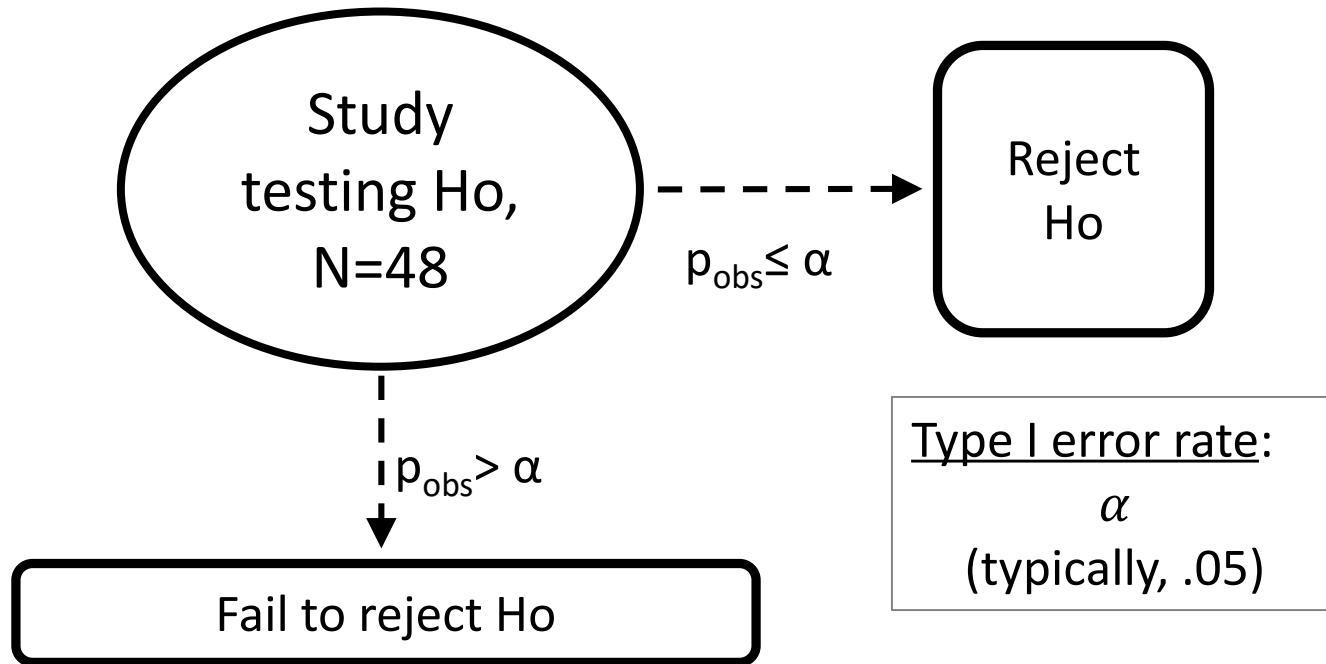
Avoidable waste in the production and reporting of research evidence. Chalmers & Glasziou (2009)

Biomedical research: increasing value, reducing waste. Macleod et al. (2014): “85% of research investment—equating to \$200 billion of the investment in 2010—is wasted.”

How can we do better?

1. A statistical model for assessing the research payoffs of traditional Ho testing.
2. An alternative “replication-based” strategy for Ho testing that often produces higher payoffs.

Traditional strategy



Payoffs for Traditional Ho Testing

Research Scenario Parameters:

4800 Ss, N=48 → 100 studies

2-sample *t*-tests

$\alpha=0.05$, 1-tailed

Base rate of false Ho's = .1

90 studies:
Ho is true

5%

95%

10 studies:
Ho is false
Effect size=0.5
→ power=0.525

52.5%

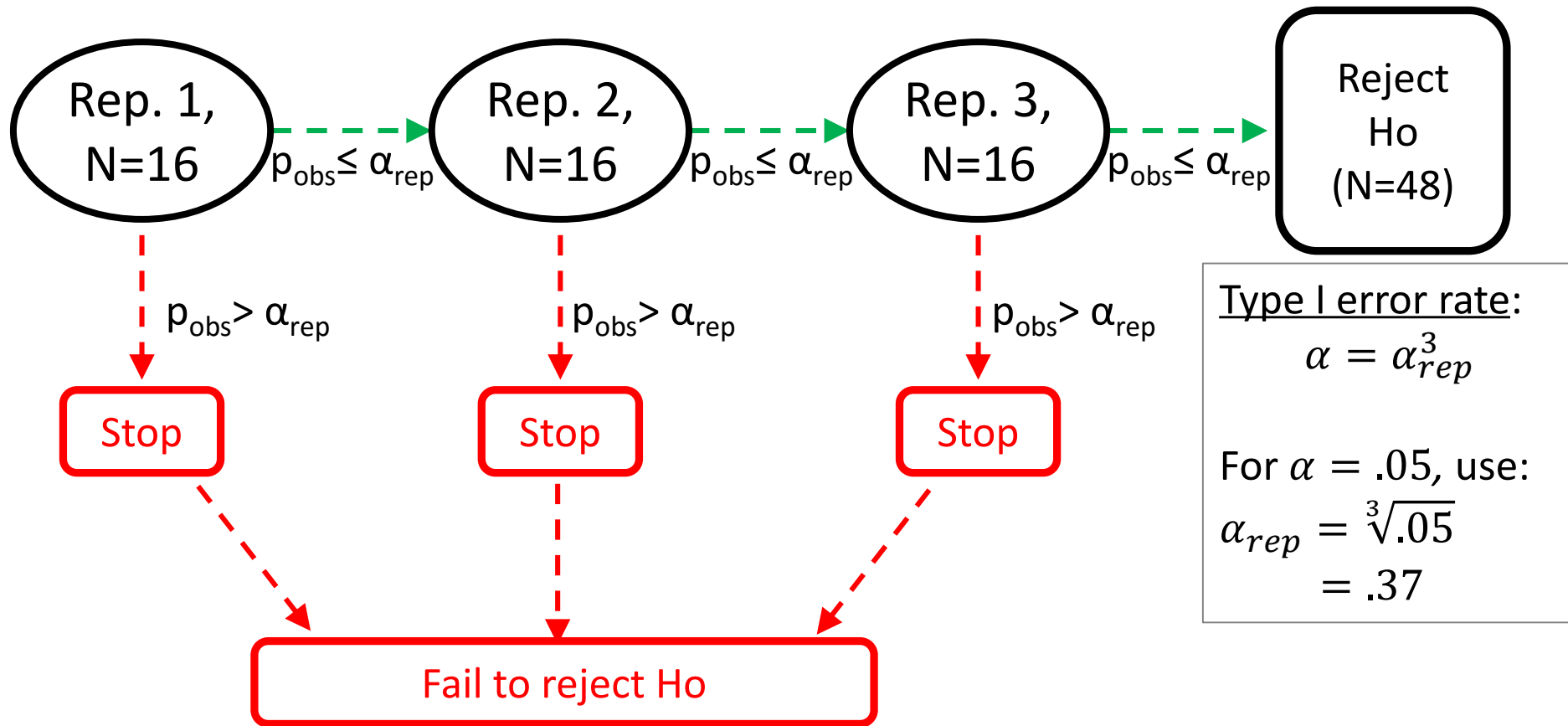
47.5%

Outcome	N_i	Payoff Pay_i	$N_i \times Pay_i$
False positive	4.5	-2.0	-9.00
True negative	85.5	0.1	8.55

True positive	5.25	1.0	5.25
False negative	4.75	-0.2	-0.95

Total payoff = 3.85

Replication-based strategy



Payoffs for Replication Strategy

Research Scenario:

4800 Ss, $N=16/32/48 \rightarrow \sim 189$ studies

2-sample t -tests

$\alpha=0.05$, 1-tailed

Base rate of false H_0 's = .1

~ 170 studies:
 H_0 is true

~ 19 studies:
 H_0 is false
Effect size=0.5
 \rightarrow power=0.415

Traditional power=0.525

Traditional
studies = 100

5%

95%

41.5%

58.5%

Outcome	N_i	Payoff Pay_i	$N_i \times Pay_i$
False positive	8.5	-2.0	-17.00
True negative	161.9	0.1	16.19

True positive	7.86	1.0	5.25
False negative	11.08	-0.2	-0.95

Total payoff = 4.79

Traditional payoff = 3.85

What About Other Research Scenarios?

We compared expected payoffs under all 2,592 possible combinations of these parameter values:

Parameter:

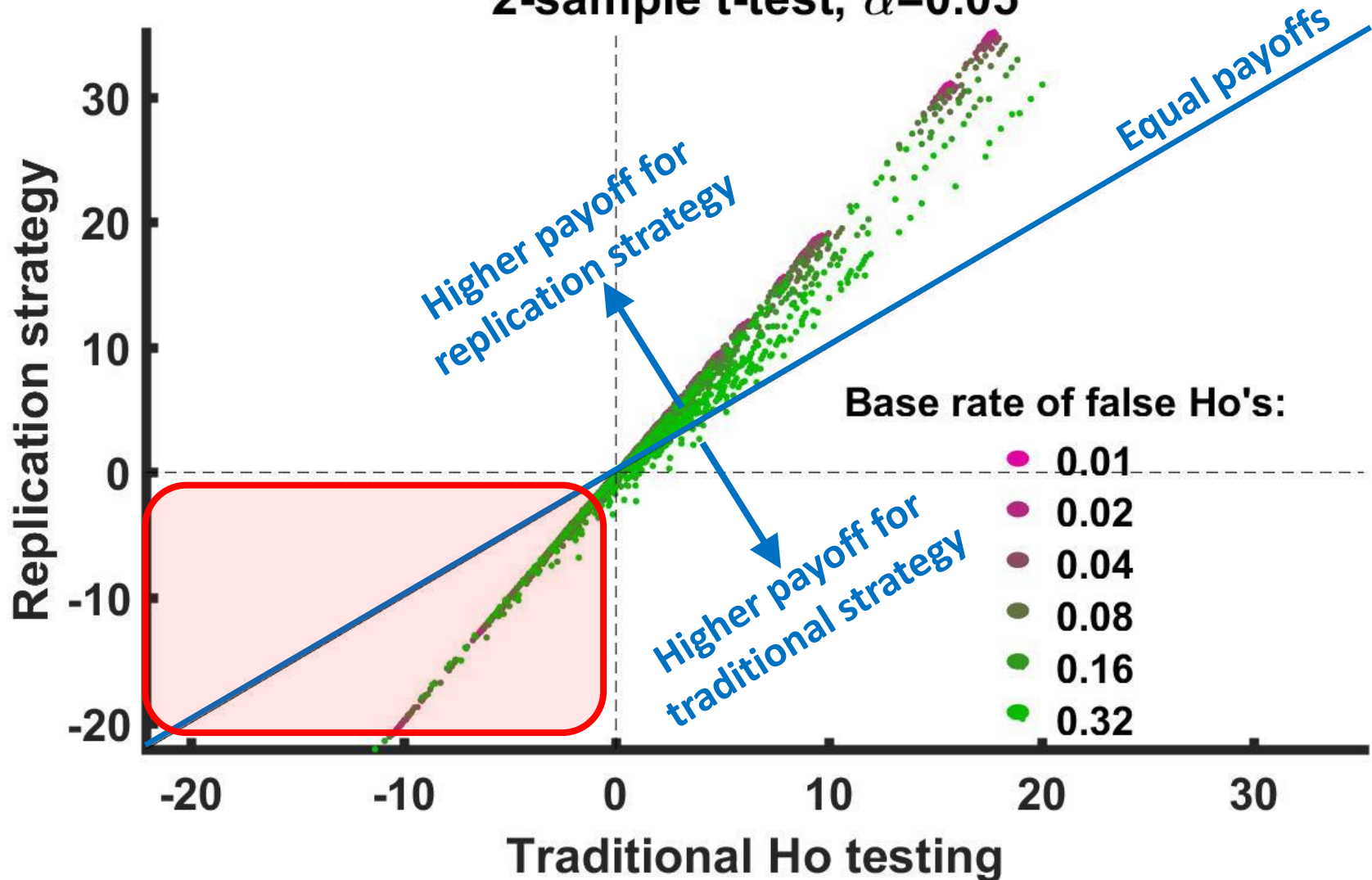
- base rate of false Ho's
- effect size
- sample size
- outcome payoffs:
 - true positive
 - true negative
 - false positive
 - false negative

Parameter values checked:

- 0.01, 0.02, 0.04, 0.08, 0.16, 0.32
- 0.2, 0.5, 0.8
- 24, 48, 96
- 1
- 0, 0.1, 0.2, 0.5
- -1, -2, -5
- 0, -0.1, -0.2, -0.5

Expected Payoffs

2-sample t-test, $\alpha=0.05$



Conclusions

Replication-based strategy looks promising:

- Simple extension of existing H_0 testing methods
- Increases payoffs in many research scenarios

Limitations:

- Single main H_0 to be rejected
- Works best when base rate of false H_0 's is low

Extensions: Are there even better strategies?

We *must* have a quantitative research payoff model to decide

Thanks for your attention. Questions?

Next talk: Which α level works best with the traditional strategy?