# The Generic Holdout: Preventing False-Discoveries in Adaptive Data Science

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### Motivation

Classical science was non-adaptive:

- 1. Scientist fixes a [set of] hypothesis
- 2. Collects data to test the hypothesis.

Modern science is adaptive:

- 1. Scientist first collects data
- 2. Explores data to find plausible hypotheses
- 3. Tests hypotheses on the same data

#### Naively, "adaptive" science is NOT statistically valid:

- Hypotheses depend on the data used to test them.
- Hypothesis may be "overfit": appears true on data, but actually false.
- Leads to false-discoveries -- key factor in the "reproducibility crisis" in science.

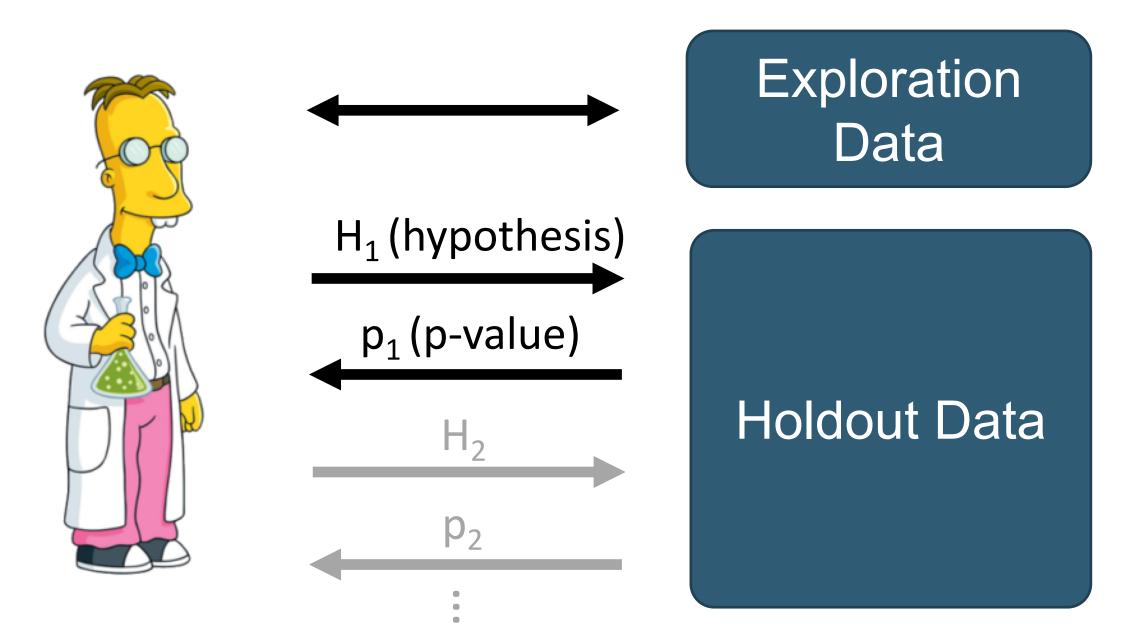
Goal: Provide a *statistically-sound* methodology for adaptive science – to prevent scientists from publishing false discoveries.

### Prior Approaches

Prior approaches: statistically-invalid, or sample-inefficient. Ex: "Naïve holdout"

Option 1: Collect new holdout set for each hypothesis (inefficient)

Option 2: "Reuse" same holdout set (invalid)



Method	Data Size (# samples)	Queries Possible
Naïve Holdout	n	O(n)
Reusable Holdout [Dwork et. al. '15]	n	$O(n^2)$
Our Method	n	$\exp(n)$

### Our Results

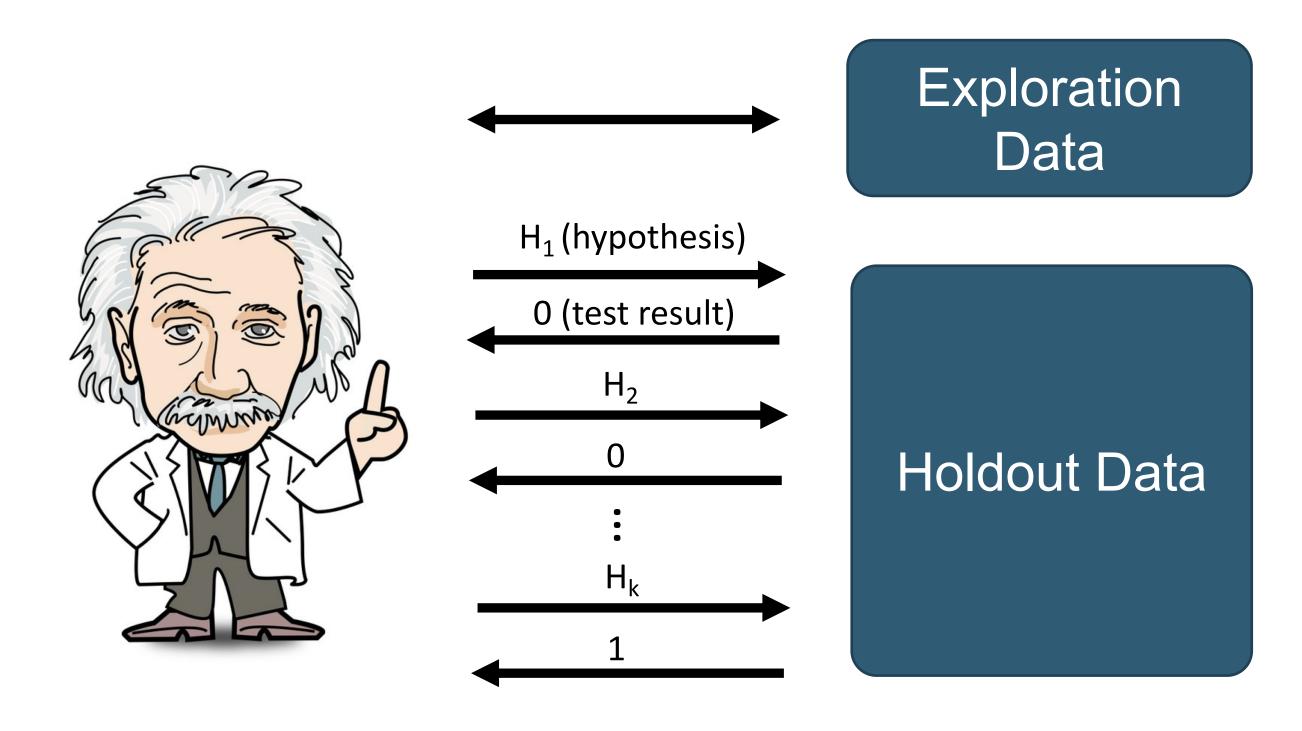
We propose a simple, statistically-sound, and sample-efficient framework for adaptive data science.

Our method allows the scientist to:

- 1. Explore part of the data to propose hypotheses
- 2. Adaptively propose & test hypotheses, based on previous hypothesis tests.
- 3. Test up to **exponentially-many** hypotheses in the size of the dataset, until discovering a true hypothesis (or several).
- 4. Bound the overall probability of a false-discovery

# The Proposal: Generic Holdout

Key Idea: Holdout set only revels a single bit – whether hypothesis test passed or not. (NOT "how well" it fits, etc)



- Leaks minimal information from holdout set → prevents overfitting.
- Scientist can test many false-hypotheses before finding a true one.

Theorem: Suppose a scientist interacting with the Generic Holdout generates a sequence of up to m adaptively-chosen hypotheses ( $H_1$ ,  $H_2$ , ...), and stops once t hypotheses are confirmed. If the false-positive probability of each hypothesis test  $H_i$  (on independent data) is bounded by p, then Pr[scientist accepts a false hypothesis] ≤  $m^t p$ 

### User Manual

How to use the Generic Holdout in your scientific process:

#### Assume we:

- Want to find a single true discovery
- Want to bound the probability of a false-discovery by p
- Will propose at most m hypotheses total (can be large).

#### Procedure:

1. Pick **n** large enough such that any hypothesis you pose can be tested with **n** iid samples, with false-positive probability  $\leq \frac{p}{m}$ .

E.g. usually requires  $n \sim \log(\frac{m}{p})$ 

- 2. Collect data (iid samples), and split it into a Holdout set of size **n**, and an Exploration set.
- 3. Use the Exploration Set as in your usual scientific process, to find plausible hypotheses.
- 4. Just before publishing a result, test the hypothesis against the holdout set, at false-positive level  $\frac{p}{m}$  (seeing only the binary result).
- 5. If the test failed, you are free to repeat Steps 3-4 until finding a true hypothesis.

This controls the probability of false-discoveries, regardless of the method used to generate hypotheses.

# Applications

- 1. Main Application: Preventing false discoveries for individual scientists/groups.
  - Alternative to pre-registration
- 2. Journal Application
  - Setting: Large public dataset collected once, many groups publish studies on it (eg, genomic data).
  - Proposal: Journals keep some of the data secret, as holdout. Use it to confirm every to-be-published study involving the public data.
  - Guarantee: Journal can confirm many true publications, before catching several false ones.

Full paper: https://arxiv.org/abs/1809.05596

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