

Optimal Pre-Analysis Plans: Statistical Decisions Subject to Implementability

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Introduction

- Trial registration and pre-analysis plans (PAPs) have become a standard requirement for experimental research.
 - For clinical studies in medicine starting in the 1990s.
 - For experimental research in economics more recently.
- Standard justification: Guarantee validity of inference.
 - P-hacking, specification searching, and selective publication distort inference.
 - Tying researchers' hands prevents selective reporting.
 - Christensen and Miguel (2018); Miguel (2021).
- The widespread adoption of PAPs has not gone uncontested, however.
 - Coffman and Niederle (2015); Olken (2015); Duflo et al. (2020).

Open questions

1. Why do we need a commitment device?
Standard decision theory has no time inconsistency!
2. How should the structure of PAPs look like?
How can we derive optimal PAPs?

Key insight:

- Single-agent decision-theory cannot make sense of these debates.
- We need to consider multiple agents, conflicts of interest, and asymmetric information.

Introduction

Setup

Motivating example: Normal testing

Implementable decision functions

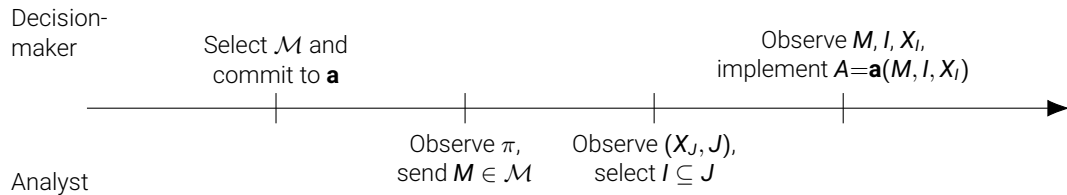
Hypothesis testing

Conclusion and outlook

Setup: Notation

- Two parties, decision-maker and analyst.
- Message \mathbf{M} (“pre-analysis plan”) sent from analyst to decision-maker.
- Data $\mathbf{X} = (X_1, \dots, X_n) \sim \mathbf{P}_\theta$.
 - Unknown parameter $\theta \in \Theta$.
- Index sets:
 - $\mathbf{K} = \{1, \dots, n\}$ fixed, finite, commonly known.
 - $\mathbf{J} \subset \mathbf{K}$ subset of data available to the analyst, privately known.
 - $\mathbf{I} \subset \mathbf{J}$ subset of available data reported to the decision-maker.
- Decision $\mathbf{A} \in \mathcal{A} \subseteq \mathbb{R}$.

Setup: Timeline



Discussion

- The analyst can withhold information, but they cannot lie.
- The components of X might represent different
 - hypothesis tests,
 - estimates,
 - subgroups,
 - outcome variables, etc.
- Possible model interpretations:
 1. Drug approval (pharma company vs. FDA).
 2. Hypothesis testing (researcher vs. reader).
 3. Publication decision (researcher vs. journal).

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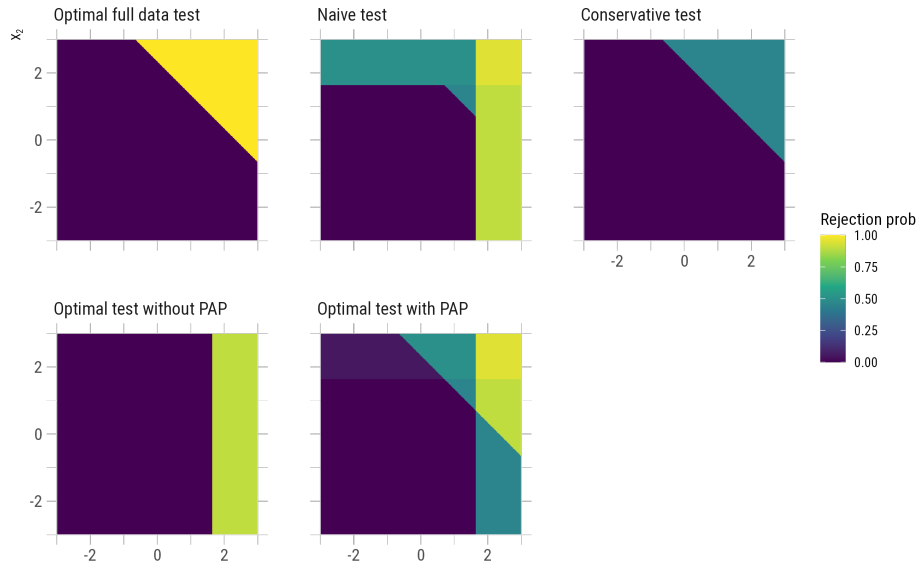
Motivating example: Normal testing

- $X_1, X_2 \sim N(\theta, 1)$.
- Prior of the decision-maker : $(J_1, J_2) \sim \text{Ber}(\eta_1) \times \text{Ber}(\eta_2)$.
- The analyst knows J .
- Null hypothesis $H_0 : \theta \leq 0$.
- The analyst selectively reports, to get a rejection of the null.

Compare 5 testing rules

0. The optimal full data test (infeasible).
1. The naive test (ignores selective reporting).
2. The conservative test (worst-case assumptions about unreported \mathbf{X}_t).
3. The optimal implementable test without a PAP.
4. The optimal implementable test with a PAP.

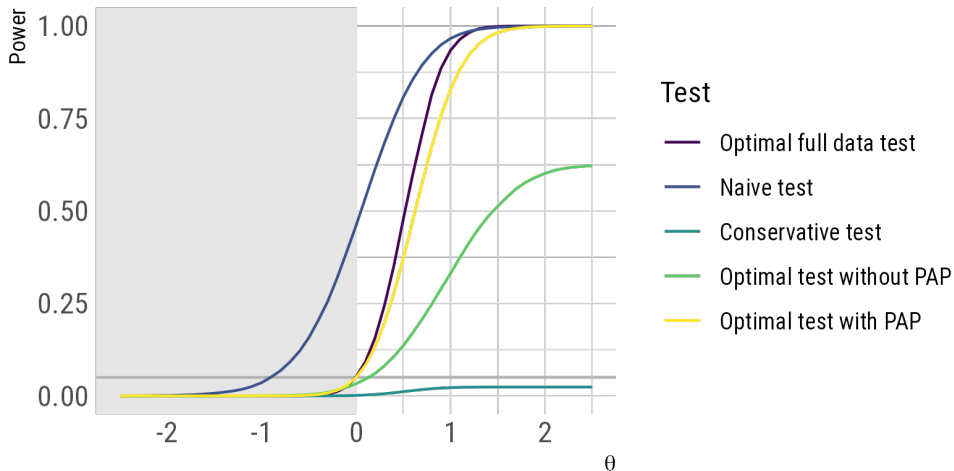
Rejection probabilities for different testing rules



x_1

Degrees of freedom $n = 10$

Power curves for different testing rules



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Implementable decision functions

- A **reduced form decision function** maps the full data into a decision **a**:

$$\bar{\mathbf{a}}(\pi, X_J, J)$$

- A reduced form decision function $\bar{\mathbf{a}}$ is **implementable**
 - if there exist a decision function **a**
 - with best responses $\mathbf{m}^*, \mathbf{i}^*$
 - such that

$$\bar{\mathbf{a}}(\pi, X_J, J) = \mathbf{a}(M^*, X_{I^*}, I^*).$$

- **Assumption:**

The analyst is an expected utility maximizer with utility

$$v(A)$$

for a (strictly) monotonically increasing function v .

Implementability without PAPs

Lemma

*If no pre-analysis messages \mathbf{M} are allowed,
a reduced-form decision function $\bar{\mathbf{a}}(\pi, \mathbf{X}_J, \mathbf{J})$ is implementable iff*

- 1. $\bar{\mathbf{a}}$ does not depend on π , and*
- 2. $\bar{\mathbf{a}}$ is **monotonic** in \mathbf{J} ,*

$$\bar{\mathbf{a}}(\mathbf{X}_I, \mathbf{I}) \leq \bar{\mathbf{a}}(\mathbf{X}_J, \mathbf{J})$$

for almost all \mathbf{X}, \mathbf{J} and all $\mathbf{I} \subseteq \mathbf{J}$.

Implementability with PAPs

Theorem

A reduced-form decision function $\bar{\mathbf{a}}$ is implementable iff both of the following conditions hold:

1. **Truthful PAP**

For almost all π and all π' ,

$$E[v(\bar{\mathbf{a}}(\pi', X_J, J)) | \pi] \leq E[v(\bar{\mathbf{a}}(\pi, X_J, J)) | \pi].$$

2. **Monotonicity**

For almost all π , X , J , and all $I \subseteq J$

$$\bar{\mathbf{a}}(\pi, X_I, I) \leq \bar{\mathbf{a}}(\pi, X_J, J)$$

Revelation and delegation

Lemma

A reduced-form decision rule $\bar{\mathbf{a}}$ can be implemented iff:

1. **Implementation by truthful revelation**

It can be implemented with a decision rule \mathbf{a} for which

$$\mathbf{a}(\pi, X_J, J) = \bar{\mathbf{a}}(\pi, X_J, J).$$

2. **Implementation by delegation**

It can be implemented with a decision rule \mathbf{a} for which

$$\mathbf{a}(b, X_J, J) = b(X_J, J),$$

where b is restricted to lie in some set \mathcal{B} .

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Hypothesis testing

- Null hypothesis $\theta \in \Theta_0$.
- Rejection probability $\mathbf{a} \in [0, 1]$.

\Rightarrow w.l.o.g. $\mathbf{v}(\mathbf{a}) = \mathbf{a}$.

- Size control at level $\alpha \in (0, 1)$:

$$\sup_{\theta \in \Theta_0} \sup_{\pi} \mathbb{E}[\bar{\mathbf{a}}(\pi, X_J, J) | \theta, \pi] \leq \alpha.$$

- Expected power:

$$\mathbb{E}[\bar{\mathbf{a}}(\pi, X_J, J)].$$

Implementing the optimal test by delegation

Theorem

- *The test with maximal expected power*
- *subject to implementability and size control*
- *can be implemented by requiring the analyst to communicate a full-data test t which satisfies, for all $\theta \in \Theta_0$,*

$$E[t(X)|\theta] \leq \alpha$$

- *and then implementing the test*

$$b(X_I, I) = \min_{X'; X'_I = X_I} t(X').$$

The analyst's problem as a linear program

$$\max_b \int b(X_J, J) dP_\pi(X, J), \quad \text{(Interim expected power)}$$

$$\text{s.t. } \int b(X, K) dP_{\theta_0}(X) \leq \alpha, \quad \text{(Size control)}$$

$$b(X_J, J) \in [0, 1] \quad \forall J, X, \quad \text{(Support)}$$

$$b(X_J, J) \leq b(X, K) \quad \forall J, X. \quad \text{(Monotonicity)}$$

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- Conflicts of interest, private information.
⇒ Not all decision rules are implementable.
- Mechanism design: Optimal implementable rules.
- Statistical reporting: Partial verifiability.
 1. No lying about reported statistics.
 2. Private information about which statistics were available.
- Pre-analysis plans:
 - No role in single-agent decision-theory.
 - But increase the set of implementable rules in multi-agent settings.
- We characterize
 1. implementable rules,
 2. optimal implementable hypothesis tests,

Thank you!