

# Pretesting, Selective Inference

Master's Thesis Workshop 1

---

Yasuyuki MATSUMURA (Kyoto University)

Last Updated: April 13, 2025

<https://yasu0704xx.github.io>

# Yasuyuki Matsumura

Graduate School of Economics, Kyoto University

Home

Research

Japanese

## About Me



I am a second-year master's student in [Graduate School of Economics, Kyoto University](#).

Under the guidance of [Yoshihiko Nishiyama](#) and [Takahide Yanagi](#), my research focuses on Econometric Theory and Statistics.

Contact: matsumura.yasuyuki.w85 [at] kyoto-u.jp

Address: Yoshida Honmachi, Sakyo, Kyoto 606-8501, Japan

Other Links: [Linktree](#)

# Introduction

---

- In empirical studies, researchers often implement hypothesis tests before going through with their main analyses, which are used to show certain plausibility of identification conditions.
  - The Hausman test (OLS vs 2SLS)
  - The Durbin-Wu-Hausman test (RE vs FE)
  - The Sargan-Hansen J-test (GMM)
  - Balance test (RCT)
  - Pre-trend test (DiD)
  - Manipulation test (RD)
- These tests are classified as **falsification tests**, namely, **pretests**, whose null hypothesis is preferred to be accepted (i.e., failed to reject).

# Problems Arising with Pretest

- Recent literatures argue that researchers should not implement such pretests.
- **Orthogonality**: Specification tests provide no information about the existence of asymptotic bias.
- **Low statistical power**: Pretests have low statistical power.
- **Inconsistent estimates**: Conditional on the result of pretesting, the estimator might lose its consistency.
- **Low coverage rate**: Conditional on the result of pretesting, the coverage probability of  $1 - \alpha$  confidence interval might be smaller than  $1 - \alpha$ .

# Manipulation Test

- In (sharp) RD designs, researchers implement pretests to examine whether the identification condition

For each  $t = 0, 1$ ,

$\mathbb{E}[Y_i(t)|X_i = x]$  is continuous at  $x = c$

is actually fulfilled.

- **The fundamental problem:** Note that only one out of  $\mathbb{E}[Y_i(0)|X_i = x]$  and  $\mathbb{E}[Y_i(1)|X_i = x]$  can be observed. That is, the above condition cannot be directly examined.
- Instead, researchers often examine certain necessary conditions (**manipulation test**).

# Literature: Manipulation Tests

- Testing the continuity of the density of the assignment variable:
  - McCrary (2008) [24]
  - Otsu, Xu and Matsushita (2013) [25]
  - Cattaneo, Jansson and Ma (2020) [6]
- Testing the continuity of the conditional distribution of covariates:
  - Lee (2008) [19]
  - Canay and Kamat (2018) [5]
  - Fusejima, Ishihara and Sawada (2024) [10]
- Testing other necessary conditions of the identification conditions:
  - Arai, Hsu, Kitagawa, Mourifie and Wan (2022) [3] <sup>1</sup>

---

<sup>1</sup>Quoting lecture slides by T. Yanagi.

## Recent Research Interest (1)

- (Master thesis?) Point out that, in RD designs, the estimator of LATE may have asymptotic bias conditional on the result of manipulation test.
- Roth (2022) [27]
  - The event-study estimators have asymptotic bias conditional on the result of pre-trend test.
- Sueishi (2023) [28]
  - The Hausman test and the J test provide no information about the existence of asymptotic bias in the efficient estimator (i.e., **correct specification and valid estimation are different issues**). Note that its framework originates from Chen and Santos (2018) [7].



## Recent Research Interest (2)

- Construct a robust estimation/inference procedure in RD designs, which overcomes the problems of pretesting mentioned above.
- Guggenberger (2010a, 2010b) [11],[12]
  - On the impact of the Hausman test on subsequent inference.
- Berk, Brown, Buja, Zhang and Zhao (2013) [4]
  - On valid post-selection inference, which is robust even if wrong models are selected by AIC, BIC or Lasso etc.
- Lee, Sun, Sun and Taylor (2016) [20]
  - On post-selection inference with application to the Lasso.
- Dzemski, Okui and Wang (2025) [9]
  - On location characteristics of conditional confidence intervals concerning publication bias issues.
- Keywords: **Selective Inference**, Post-Model-Selection, ...

# Acknowledgement

- I wish to express my sincere gratitude to Yoshihiko Nishiyama<sup>2</sup>, Takahide Yanagi<sup>3</sup>, Naoya Sueishi<sup>4</sup>, Ryo Okui<sup>5</sup> and Hisatoshi Tanaka<sup>6</sup> for valuable supports and comments.

---

<sup>2</sup>Kyoto University (Supervisor)

<sup>3</sup>Kyoto University (Supervisor)

<sup>4</sup>Kobe University

<sup>5</sup>University of Tokyo

<sup>6</sup>Waseda University

# Survey

---

- メモ：
- Roth (2022) みたいなサーベイをやる.
- とりあえず直近の課題：Sharp RD の実証ペーパー集めてくる. Fusejima et al (2024) でめっちゃくちゃ集めてたから参考になりそう.

# Theoretical Analysis




---

# Practical Recommendations

---





## References





---




-  Andrews, D. W. and P. Guggenberger (2009a). Hybrid and size-corrected subsampling methods. *Econometrica* 77(3), 721-762.
-  Andrews, D. W. and P. Guggenberger (2009b). Incorrect asymptotic size of subsampling procedures based on post-consistent model selection estimators. *Journal of Econometrics* 152(1), 19-27.
-  Arai, Y., Y. C. Hsu, T. Kitagawa, I. Mourifie and Y. Wan (2022). Testing identifying assumptions in fuzzy regression discontinuity designs. *Quantitative Economics* 13, 1-28.











-  Berk, R., L. Brown, A. Buja, K. Zhang, and L. Zhao (2013). Valid post-selection inference. *Annals of Statistics* 41(2), 802-837.
-  Canay, I. A. and V. Kamat (2018). Approximate permutation tests and induced order statistics in the regression discontinuity design. *Review of Economic Studies* 85, 1577-1608.
-  Cattaneo, M. D., M. Jansson and X. Ma (2020). Simple local polynomial density estimators. *Journal of the American Statistical Association* 115(531), 1449-1455.
-  Chen, X. and A. Santos (2018). Overidentification in regular models. *Econometrica* 86(5), 1771-1817.




-  Doko Tchatoka, F. and W. Wang (2021). Size-corrected bootstrap test after pretesting for exogeneity with heteroskedastic or clustered data. SSRN Working Paper.
-  Dzemski, A., R. Okui and W. Wang (2025). Location characteristics of conditional selective confidence intervals via polyhedral methods. arXiv:2502.20917v1.
-  Fusejima, K., T. Ishihara and M. Sawada (2024). A unified diagnostic test for regression discontinuity designs. arXiv:2205.04345v4.
-  Guggenberger, P. (2010a). The impact of a hausman pretest on the asymptotic size of a hypothesis test. *Econometric Theory* 26(2), 369-382.

-  Guggenberger, P. (2010b). The impact of a hausman pretest on the size of a hypothesis test: The panel data case. *Journal of Econometrics* 156(2), 337-343.
-  Guggenberger, P. and G. Kumar (2012). On the size distortion of tests after an overidentifying restrictions pretest. *Journal of Applied Econometrics* 27(7), 1138-1160.
-  Judge, G. G. and M. E. Bock (1978). *The Statistical Implications of Pre-test and Stein-Rule Estimators in Econometrics*. North-Holland
-  Kabaila, P. (1995). The effect of model selection on confidence regions and prediction regions. *Econometric Theory* 11(3), 537-549.

-  Kitamura, Y., T. Otsu and K. Evdokimov (2013). Robustness, infinitesimal neighborhoods, and moment restrictions. *Econometrica* 81(3), 1185-1201.
-  Kivaranovic, D. and H. Leeb (2021). On the length of post-model-selection confidence intervals conditional on polyhedral constraints. *Journal of the American Statistical Association* 116(534), 845-857.
-  Kivaranovic, D. and H. Leeb (2024). A (tight) upper bound for the length of confidence intervals with conditional coverage. *Electronic Journal of Statistics* 18, 1677-1701.

-  Lee, D. S. (2008). Randomized experiments from non-random selection in US House elections. *Journal of Econometrics* 142(2), 675-697.
-  Lee, J. D., D. L. Sun, Y. Sun and J. E. Taylor (2016). Exact post-selection inference, with application to the lasso. *Annals of Statistics* 44(3), 907-927.
-  Leeb, H., and B. M. Pötscher (2005). Model selection and inference: Facts and fiction. *Econometric Theory* 21(1), 21-59.
-  Leeb, H., and B. M. Pötscher (2006). Can one estimate the conditional distribution of post-model-selection estimators? *Annals of Statistics* 34(5), 2554-2591.

-  Leeb, H., B. M. Pötscher and K. Ewald (2015). On various confidence intervals post-model-selection. *Statistical Science* 30(2), 216-227.
-  McCrary (2008). Manipulation of the running variable in the regression discontinuity design: A density test. *Journal of Econometrics* 142(2), 698-714.
-  Otsu, T., K. L. Xu, and Y. Matsushita (2013). Estimation and inference of discontinuity in density. *Journal of Business and Economic Statistics* 31(4), 507-524.
-  Pötscher, B. M. (1991). Effects of model selection on inference. *Econometric Theory* 7(2), 163-185.

-  Roth, J. (2022). Pretest with caution: Event-study estimates after testing for parallel trends. *American Economic Review: Insights* 4(3), 305-322.
-  Sueishi, N. (2023). A misuse of specification tests. arXiv:2211.11915v1.
-  Tibshirani, R. J., J. Taylor, R. Lockhart and R. Tibshirani (2016). Exact post-selection inference for sequential regression procedures. *Journal of the American Statistical Association* 111(514), 600-620.