

# Bingkai Wang

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## Education

Ph.D in Biostatistics, Johns Hopkins University, 2021 (expected)  
M.S. in Mathematics, Fudan University, 2016

## Professional Experience

Research Assistant, Department of Biostatistics, Johns Hopkins University, 2016-present.  
Advisor: Michael Rosenblum and Brian Caffo.  
Research Assistant, School of Mathematics, Fudan University, 2014-2016.  
Advisor: Shuqin Zhang.  
Research Assistant, State Key Laboratory of Institute of Brain Science, 2014-2016.  
Advisor: Jiayi Zhang

## Honors and Awards

Cersi Scholarship, National Institutes of Health and Johns Hopkins University, 2017.

## Publications

### Peer-reviewed Journal Articles

1. Paniz Charkhchi, **Bingkai Wang**, Brian Caffo and David M. Yousem. (2018) Bias in Neuroradiology Peer Review: Impact of a “Ding” on “Dinging” Others. *American Journal of Neuroradiology*, December 2018.

### Preprints

1. Yi Zhao, **Bingkai Wang**, Stewart Mostofsky, Brian Caffo, Xi Luo. Covariate Assisted Principal Regression for Covariance Matrix Outcomes. *bioRxiv*, doi: <https://doi.org/10.1101/425033>.
2. **Wang, Bingkai**; Ogburn, Elizabeth; and Rosenblum, Michael, "Analysis of Covariance (ANCOVA) in Randomized Trials: More Precision, Less Conditional Bias, and Valid Confidence Intervals, Without Model Assumptions" (October 2018). *Johns Hopkins University, Dept. of Biostatistics Working Papers. Working Paper 292.*  
<https://biostats.bepress.com/jhubiostat/paper292>

## Teaching

Guest Lecture, Statistical Theory I, 2017

Teaching Assistant, Statistical Theory I-IV, 2017-2018

Guest Lecture, Advanced Data Science II, 2018

Teaching Assistant, Advanced Data Science I-II, 2018

## Professional Activities

### *Presentations*

Benefits of adjustment for baseline variables in randomized trials. *JSM, Baltimore, USA, July 2017.*

Clarifying How Adjustment for Prognostic Baseline Variables Leads to More Precision and Less Bias in Randomized Trials. *ENAR, Atlanta, USA, March 2018*