

Siddharth Prusty

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Education

Ph.D.	Duke University, Durham, NC	Expected 2026
Marketing (Quantitative)		
M.S.	Columbia University, New York City, NY	2020
Industrial Engineering and Operations Research		
B.Tech.	Indian Institute of Technology, Kanpur, India	2016
Electrical Engineering		

Research Interests

Substantive: Retail Media, Ad Auctions, Mechanism Design, Sustainability, Regulation.

Methodological: Structural Modeling, Analytical Modeling, Machine Learning, Statistics.

Publications and Working Papers

(Abstracts in Appendix) (*Authors Listed Alphabetically)

Siddharth Prusty, Carl F. Mela, and Hana Choi. “**Enhancing Position Auctions in Retail Media.**” Working Paper (*Job market paper*), 2025.

Wilfred Amaldoss* and Siddharth Prusty*. “**Sustainable Consumption: A Strategic Analysis.**” Forthcoming in *Marketing Science*, 2025.

Fengpei Li, Henry Lam, and Siddharth Prusty. “**Robust Importance Weighting for Covariate Shift.**” *Proceedings of the 23rd Conference on Artificial Intelligence and Statistics*, PMLR 108:352-362, 2020.

Selected Work in Progress

Wilfred Amaldoss* and Siddharth Prusty*. “**Regulating Sustainable Products: A Public Policy Perspective.**”

Conference Presentations

“Enhancing Position Auctions in Retail Media”

- ISMS Marketing Science Conference, University of Georgia (Washington, D.C.), 2025

"Sustainable Consumption: A Strategic Analysis"

- 18th Annual Bass Forms Conference, U.T. Dallas, 2024
- Marketing for Environmental Sustainability Conference, Stanford University, 2023
- ISMS Marketing Science Conference, University of Chicago Booth (Virtual), 2022

Awards, Honors, Service

AMA Sheth Doctoral Consortium, Ohio State University, 2025

Ad-Hoc Reviewer, Management Science, 2023

ISMS Doctoral Consortium Fellow, University of Chicago Booth (Virtual), 2022

Shardashish Interschool Fellowship, Columbia University, 2018

Honda Young Engineer and Scientist Award, Honda Foundation, 2014

Academic Excellence Award, Indian Institute of Technology, Kanpur, 2013

Teaching

Instructor, Statistical Programming Bootcamp

Duke University

- Developed and instructed week-long course on Python (for incoming Ph.D. students in Business Administration) 2023

Teaching Assistant

Duke University

- Value Creation in Martech (MBA, MQM), taught by Carl Mela 2021/23/24
- Strategy and Tactics of Pricing (MBA, EMBA, MQM), taught by Wilfred Amaldoss 2025
- Marketing Management (WEMBA), taught by Preyas Desai 2025
- Programming Analytics (MQM), taught by Allison Chaney 2023
- Strategic Brand Management (MBA, EMBA), taught by Tong Guo 2022/23
- Marketing Core (MMS), taught by Srinivas Tunuguntla 2022

Teaching Assistant

Columbia University

- Stochastic Models and Applications, taught by Ton Dieker 2020
- Introduction to Financial Engineering, taught by David Yao 2019/20

Graduate Coursework

Marketing & Business Administration

PhD level:

- | | |
|---|--------------------------|
| Quantitative Marketing Proseminar | Rick Staelin |
| Special Topics in Quantitative Marketing | Carl F. Mela |
| Structural Modeling in Marketing (<i>at UNC-CH</i>) | Sriram Venkataraman |
| Economic Models in Marketing (<i>at UNC-CH</i>) | Rajdeep Grewal |
| Special Topics in Quantitative Marketing | Allison Chaney, Tong Guo |

Economics

PhD level:

- | | |
|-------------------------|---------------------------------|
| Industrial Organization | Allan Collard-Wexler, Daniel Xu |
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Partial Identification: Theory and Applications in IO
Theoretical Industrial Organization (*at UNC-CH*)
Dynamic Discrete Choice
Causal Inference and Treatment Effects
Econometrics I
Microeconomics I

Allan Collard-Wexler, Adam Rosen
Gary Biglaiser
Peter Arcidiacono
Arnaud Maurel
Matt Masten
Philipp Sadowski, Curtis Taylor

Operations Research

PhD level: (*at Columbia University*)

Stochastic Modeling I
Stochastic Modeling II
Optimization I: Continuous Optimization
Optimization II: Discrete Optimization
Statistical Models for Simulations

Karl Sigman
Ton Dieker
Vineet Goyal
Yuri Faenza
Henry Lam

Statistics and Machine Learning

Master's level:

Introduction to Machine Learning
Optimization Methods in Machine Learning
Bayesian Models in Machine Learning
Time Series Analysis
Statistical Techniques in Data Mining

Cynthia Rudin
Satyen Kale
John Paisley
Amit Mitra
Amit Mitra

Professional Experience

IBM Research Labs , Research Intern, Armonk, NY	Summer 2020
American Express , Risk Management Analyst, Gurugram, India	2016-2018
iRageCapital Advisory Pvt. Ltd. , Quant Trading Intern, Mumbai, India	Summer 2015
Ecole Normale Supérieure , Software Intern (Research), Paris, France	Summer 2014
Avanti Fellows , Course Content Development Intern, Delhi, India	Summer 2013

Computer Languages and Skills

Programming languages: Python, R, Mathematica, Stata, MATLAB

Tools: LLMs (ChatGPT), LaTeX (Overleaf, Beamer), Microsoft (Word, PowerPoint, Excel)

References

Carl F. Mela

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Appendix: Abstracts

Siddharth Prusty, Carl F. Mela, and Hana Choi. “**Enhancing Position Auctions in Retail Media.**” Working Paper (*Job market paper*), 2025.

Retail media, a fast-growing channel for digital advertising, surpassed \$55 billion of ad spend in 2024. A common retail media format involves position auctions, in which advertisers bid for higher placements on a retailer's product listing page. Advertiser bids are combined with a retailer-set quality score to determine the allocation of sponsored slots and the resulting payments. Quality scores boost certain advertisers' positions and reduce their per-click price. Unlike search engine advertising, retail media position auctions can monetize sales commissions as well as clicks. This paper develops a quality score approach to effectively balance these monetization options.

To connect quality scores to retail revenues, the paper develops a structural model linking advertiser bids and revenues to the retailer's quality score choices coupled with a machine learning model of consumer behavior. These models are estimated using auction-advertiser level data from a quality score experiment conducted at a mid-size US based retail marketplace. Policy simulations show that a quality score approach that balances clicks and commissions improves retailer profits by 7% and advertiser surplus by 42% over click-based approaches typically used by retailers, leading to a win-win outcome for both.

Wilfred Amaldoss and Siddharth Prusty. “**Sustainable Consumption: A Strategic Analysis.**” Forthcoming in *Marketing Science*, 2025.

Consumers' growing concern for the environment has motivated firms to offer sustainable products in several categories. An exploratory survey shows that many consumers desire sustainable products and are willing to pay more for them, but some consumers dislike sustainable products and want to pay less for them. Using a theoretical model where firms are horizontally differentiated and two groups of consumers have divergent preference for sustainable products, we investigate the strategic implications of sustainable consumption. First, our analysis shows that when consumers' dislike for sustainable products is moderate, the price could increase as the dislike increases. Moreover, price could decrease if consumers' desire for sustainable products increases. Second, we find that competing firms' profits can decrease with consumers' desire for sustainability but increase with consumers' dislike for sustainability. Third, we clarify when and why enforcing minimal sustainability standards for products can backfire and reduce consumer surplus. Finally, we extend the model to capture additional facets of sustainable consumption, such as multi-product firms, sustainable luxury goods and political orientation of consumers, and tease out its counterintuitive implications for the firms supplying sustainable products.

Wilfred Amaldoss and Siddharth Prusty. “**Regulating Sustainable Products: A Public Policy Perspective.**”

At the core of sustainable consumption lies a public goods dilemma: while some consumers invest significant financial and personal resources in adopting sustainable practices, and some firms incur costs to develop eco-friendly products and services, a considerable proportion of both consumers and firms engage in free riding, reaping the environmental benefits of sustainability without contributing to its provision (Milinski et al., 2006). Moreover, sustainable consumer behavior frequently exhibits present bias, wherein short-term consumption benefits are weighted more heavily than the long-term environmental and social consequences of those choices (Trudel, 2019). Using a theoretical model that incorporates consumption externalities and behavioral biases in sustainable consumption, this paper investigates the strategic implications of regulatory instruments targeted towards consumers (e.g., price discounts) or firms (e.g., sustainability mandates). The theoretical analysis examines how policymakers can design regulatory instruments to incentivize firms and consumers to foster a sustainable future.

Fengpei Li, Henry Lam, and Siddharth Prusty. “**Robust Importance Weighting for Covariate Shift.**” *Proceedings of the Twenty Third International Conference on Artificial Intelligence and Statistics*, PMLR 108:352-362, 2020.

In many machine learning problems, the training and testing data follow different distributions and a particularly common situation is the covariate shift. To correct for sampling biases, most approaches, including the popular kernel mean matching (KMM), focus on estimating the importance weights between the two distributions. Reweighting-based methods, however, are exposed to high variance when the distributional discrepancy is large, and the weights are poorly estimated. On the other hand, the alternate approach of using nonparametric regression (NR) incurs high bias when the training size is limited. In this paper, we propose and analyze a new estimator that systematically integrates the residuals of NR with KMM reweighting, based on a control-variate perspective. The proposed estimator can be shown to either strictly outperform or match the best-known existing rates for both KMM and NR and thus is a robust combination of both estimators. Experiments show the estimator works well in practice.