

Jin Miao

CONTACT INFORMATION	Room 13.206 Naveen Jindal School of Management Richardson, TX 75080, USA	<i>Mobile:</i> 469-449-2160 <i>E-mail:</i> Jin.Miao@utdallas.edu <i>Website:</i> jinmiaomkt.github.io
EDUCATION	University of Texas at Dallas , Richardson, TX Ph.D. Quantitative Marketing	Aug 2020 - May 2026
	Columbia University , New York City, NY M.S. Marketing Science	Aug 2017 - May 2018
	Tsinghua University , Beijing, China B.A. Economics & B.S. Psychology	Aug 2013 - July 2017
RESEARCH INTERESTS	Substantive: Generative AI, Behavioral Economics, New Product Development Methodology: Generative Pre-trained Transformer (GPT), Deep Learning / AI, Game Theory	
WORKING PAPER	ProductGPT: A Generative Model of Consumer Decision Dynamics in Limited-Time Product Offerings Jin Miao, Fanglin Chen, Ying Xie	

Abstract: Managers frequently offer limited-time products across various industries, yet few studies examine consumers' dynamic decisions to guide managerial choices about re-issuing previous products or introducing new ones. We develop *ProductGPT*, a novel deep generative pre-trained Transformer designed to predict sequential purchase decisions in limited-time product settings. Three key features differentiate our *ProductGPT* from standard Transformer models. First, it uses triplet input tokens which simultaneously encapsulate decision history, the supply-side sequence of limited-time products, and the demand-side sequence of observed decision outcomes. This approach enables *ProductGPT* to capture meaningful product representations for all products in the assortment. Second, using the encoded long-range and short-range contexts, *ProductGPT* exclusively predicts consumer decisions. During the inference phase, we can leverage the model to generate consumer decision sequences under alternative limited-time product schemes, thus providing managers a decision tool to help design limited-time products. Third, *ProductGPT* incorporates product features as covariates, enabling managers to forecast sales of new limited-time products through feature-based product embeddings. We apply *ProductGPT* to a unique 20-month transaction dataset of video game loot boxes. Our model consistently outperforms several sophisticated benchmark models in sequential purchase predictions. We demonstrate its ability to forecast aggregate sales, discover behavioral patterns, and support product sequencing strategies.

Pricing of Services: An Analysis of the Impact of Availability Bias
Sanjay Jain, Jin Miao

Abstract: Firms often offer subscriptions for services such as extended warranties for automobiles or appliances, and subscription plans for golf or symphony. Before purchasing a subscription, consumers need to estimate the probability of needing the service in the future. Most of the prior research assumes that consumers form an unbiased estimate of their future needs. Empirical evidence, however, shows that consumers often make errors in predicting future needs. We draw on the literature on the availability bias (Tversky and Kahneman 1973) to model how consumers form subjective probabilities of needing the service. We develop a dynamic model in which a firm offers a menu of contracts to consumers, and the menu prices can change over time. We show that availability bias can lead firms to offering subscription plans with below marginal cost pricing for all consumers

at the time of service. Contrary to intuition, we find that availability bias can sometimes benefit consumers at the expense of firms. We also find that even perfect competition cannot eliminate the negative impact of availability bias on social welfare, and that consumer and social welfare can be negative even under perfect competition.

PUBLICATION

Designing Loot Boxes: Implications for Profits and Welfare

Jin Miao, Sanjay Jain

Marketing Science (2024) vol. 43, no. 6, pp. 1242–1259.

Abstract: A loot box is a probabilistic allocation of virtual products, the exact outcome of which is known to consumers only after purchase. Consumers sometimes purchase these goods multiple times until their preferred products are obtained. As loot boxes have been gaining enormous popularity in recent years, they are often criticized as exploitative and socially wasteful. In this study, we develop a stylized model to study the optimal design of loot boxes and its impact on profits and social welfare. We find that firms may assign asymmetric probabilities to *ex ante* symmetric products. Firms could use loot boxes to offer products at low prices to users who would not buy these products under the traditional pricing strategy. Loot boxes enable firms to earn higher profits due to better price discrimination and market expansion. Contrary to the widespread criticism of loot boxes as socially harmful, our analysis reveals that the loot box strategy can improve social welfare. Some platforms promise that consumers can obtain their preferred products with no more than a certain number of purchases. Contrary to conventional wisdom, our analysis reveals that such a strategy can increase firm's profits while reducing consumer welfare.

WORK IN PROGRESS

Design Rollover Policy in Subscription Economy

Jin Miao, Haokun Du, Sanjay Jain

Abstract: Service providers in subscription economy differ markedly in how they treat unused allowances – some offer no rollover, while others permit unlimited carryover. Many subscribers auto-renew their subscription plans without checking unused allowances and evaluating alternative plans. This paper investigates a firm's optimal rollover policy when consumers face uncertain usage and may forget to reevaluate their plans. We develop an infinite-horizon model in which the firm chooses a rollover cap and offers a menu of subscription plans to serve consumers who differ in their self-awareness of memory problems. We show that when consumers are naïve about their memory problems, the firm benefits from allowing rollover but chooses the least generous cap. Contrary to the conventional belief that consumers are better off with more generous rollover policies, we find that when consumers are partially sophisticated, the profit-maximizing firm may offer a more generous policy that ultimately leaves consumers worse off. In contrast, when consumers are fully sophisticated, rollover fails to enhance firm profits. From a policy perspective, mandatory auto-renewal reminders do not alter the rollover policy but improve consumer welfare.

INVITED TALKS

Tsinghua University	October 2025
Chinese University of Hong Kong, Shenzhen	October 2025
University of Texas, Austin	October 2025
Singapore Management University	September 2025
City University of Hong Kong	September 2025
Cambridge University	cannot attend
University of Miami	cannot attend
Peking University, HSBC Business School	cannot attend

TEACHING INTERESTS

Large Language Models in Marketing, AI-Driven Content Creation, Marketing Analytics
Digital Marketing, Pricing Analytics and Strategy, New Product Development

INSTRUCTOR	Principles of Marketing (BBA-Marketing) <i>solo instructor</i> (Class Size: 56, Teaching Evaluation: 5.0/5.0) <i>Outstanding Ph.D. Teacher Award</i>	Fall 2024
	Principles of Marketing (BBA-Marketing) <i>solo instructor</i> (Class Size: 48, Teaching Evaluation: 4.8/5.0)	Fall 2023
HONORS	Graduate Student Assembly Travel Award	Fall 2025
AWARDS	JSOM Outstanding Ph.D. Teacher Award	Fall 2025
SCHOLARSHIPS	ISMS Doctoral Consortium Fellow JSOM Ph.D. Student of the Year, Finalist ISMS Doctoral Dissertation Early-Stage Grant, Finalist Google Cloud Platform Credits Award AMA-Sheth Foundation Doctoral Consortium Fellow Betty and Gifford Johnson Travel Awards Center for Teaching and Learning (CTL) Fellow Academic Excellence Scholarship, Tsinghua University Baden-Württemberg-Stipendium, Mannheim University	Summer 2025 Spring 2025 Spring 2025 Spring 2025 Summer 2023 Summer 2023 Spring 2023 Fall 2016 Fall 2015
CONFERENCE PRESENTATION	INFORMS Annual Conference INFORMS Marketing Science Annual Conference BizAI Annual Conference INFORMS Marketing Science Annual Conference Production and Operations Management Conference	Atlanta GA, October 2025 Washington DC, June 2025 Richardson TX, March 2025 Miami FL, June 2023 Orlando FL, May 2023

Last Update: Nov 2025