Research Statement

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I am an applied microeconomist specialized in urban economics, industrial organization, and applications of big-data and machine learning in economics. My research is focused on agents' decision making and welfare distribution in peer-to-peer markets. Unlike traditional markets, these are highly dynamic markets where small heterogeneous players with limited capacities interact directly with each other. Studying these emerging markets requires both novel structural models and estimation methodologies. I will summarize my three working papers on welfare distribution and peer-to-peer markets.

"Welfare Estimation in Peer-to-Peer Markets with Heterogeneous Agents: The Case of Airbnb" (Job Market Paper)

In this paper, I explore distribution of welfare in the highly dynamic market of Airbnb with agent-level heterogeneity. I show that even though agents in less advantaged areas have lower opportunity cost of providing accommodation services, they cannot benefit from having access to Airbnb market as much as those in more advantaged areas. This disproportionate distribution of welfare is demand driven and is because people are willing to pay more for accommodations in upper-class neighborhoods. The lower demand for disadvantaged neighborhoods shifts the potential surplus away from these areas. This is a source of inequality in income for those who live in lower income areas. On the other hand, I show evidences for higher benefit of low-income suppliers within the same neighborhoods and suggest better local distribution of welfare toward more disadvantaged suppliers.

Studying welfare distribution among differentiated agents in the Airbnb, in specific, and in peer-to-peer markets, in general, is challenging because these markets are highly dynamic, size of the market changes frequently, and agents are small with limited capacities. To address these challenges, I build on the estimation method of Bayer, Ferreira, and McMillen (2007) and extend the existing methods to estimate agent-level and time varying consumer and producer surplus in the Airbnb market.

I apply a multinomial logit model to estimate a time varying individual-level demand

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for each listing in the Airbnb market. In my model, I allow for repeated sales and a panel-data estimation of supply and demand. Supplier have limited capacities, so I also allow for variations in the choice sets of guests in addition to possibility of having vacant unit in the market. Moreover, to estimate the producer surplus, I have a producer side with strategic pricing in my estimations. These extensions are essential for agent-level estimation of surplus in the Airbnb market. More broadly, my estimation method provides a framework for demand estimation in highly dynamic markets with small and differentiated agents that have limited capacities. My estimation method is applicable for studying housing sales and rental market or with some modifications for other peer-to-peer markets.

"Introducing a Micro-Founded Index of Consumption Welfare: A Big-Data Approach"

In this paper, I explore the heterogeneity in welfare from consumption based on a novel micro-founded index of welfare. I use Albania's 2012 Living Standard Measurement Survey to estimate the index in two steps. In the first step, I apply machine learning to find a non-parametric relation between households' consumption, and a large set of living conditions and characteristics indicators. In the second step, using the first step estimations, I find the distribution of households' marginal willingness to pay for each living condition indicator, and estimate an index of welfare based on the model of Bajari and Benkard (2005). I show that the index is highly correlated with households' consumption expenditures, but unlike consumption as a naive measure of welfare, it accounts for the existing heterogeneity among their living conditions and preferences. Finally, this paper studies the geographical, cross gender, and age distributions of the welfare index and compares the estimations with the distribution of consumption expenditure.

"Evaluating Regulations in Peer-to-Peer Markets: A Synthetic Control Approach to Study Santa Monica Ban on Airbnb"

Together with Peter Christensen, I focus on the effect of a ban on Airbnb in Santa Monica as a natural experiment. I apply "elastic net synthetic control" as a recently developed causal machine learning method. Synthetic control method provides a framework to generate a treatment group for each potential outcome using a pool of control groups, and study multiple outcomes in the market. I study the effect of the ban that targets entire-home, Airbnb rentals on incumbent listings' revenue, pricing behavior, and local competition in the market. I show that the ban significantly dropped the number of entry and increases the market power of those who remained in the market. Studying welfare effect of the ban is the next step in this paper.