

# Industrial Reorganization: Learning about Patient Substitution Patterns from Natural Experiments

Devesh Raval   Ted Rosenbaum   Nathan Wilson

Federal Trade Commission

November 2015

## Disclaimer:

The opinions expressed here are those of the authors and not necessarily those of the Federal Trade Commission or any of its Commissioners.

# \Industrial Disorganization"

Many new empirical industrial organization studies forecast counterfactual outcomes [...], without a clear foundation in experience. [...] We'd expect such a judgment to be based on evidence showing that the simulation-based approach delivers reasonably accurate predictions.

{Angrist and Pinske (JEP 2010)

# Discrete Choice Demand - A Shaky Foundation?

Cornerstone of empirical IO

Models based on **untestable** assumptions

- | IIA
- | Limited differentiation among products

Little exogenous variation in choice sets

# Natural Disasters Randomly Change Choice Set

Hospitals destroyed/closed by natural disasters

Internal Validity: Areas immediately surrounding relatively unaffected

External Validity: Disasters in range of environments

# Hospitals Good Industry to Examine Demand Models

Hospitals are important

- | Hospital care more than 5 percent of GDP
- | Hospital demand models used to address a variety of questions

Rich patient-level data helpful for identification

Several critiques

- | Brennan and Guerin-Calvert, 2013; Doane, Froeb, and Van Horn, 2012; May, 2013

## Experimental Validation in Other Settings

Vending Machines: Conlon and Mortimer (2013)

School Choice: Pathak and Shi (2014)

Schooling and Fertility: Todd and Wolpin (2006)

# What you should take away from this talk

Heterogeneity in unobserved hospital quality is important

We provide guidance on which models to use

- | Use combination of semiparametric and parametric models

Widely used models lead to different policy conclusions

- | Model choice matters



# Overview

- 1 Disasters
- 2 Discrete Choice Models
- 3 Model Performance
- 4 Welfare
- 5 Conclusions

# Natural Disasters

Location	Month/Year	Severe Weather	Hospital(s) Closed
Northridge, CA	Jan-94	Earthquake	St. John's Hospital
Americus, GA	Mar-07	Tornado	Sumter Regional Hospital
New York, NY	Oct-12	Superstorm Sandy	NYU Langone Bellevue Hospital Center Coney Island Hospital
Moore, OK	May-13	Tornado	Moore Medical Center

# Tornado: Americus, GA

# Tornado: Moore, OK

# Superstorm Sandy: Manhattan, NY

# Superstorm Sandy: Brooklyn, NY

# Earthquake: Los Angeles, CA

# Primitives

Hospitals  $j = 0; 1; \dots; J$ , where 0 indexes the outside option.

Patients  $i = 1; \dots; N$



# Primitives

Hospitals  $j = 0; 1; \dots; J$ , where 0 indexes the outside option.

Patients  $i = 1; \dots; N$

Patient's choice of hospital is denoted  $h_i$

Patients choose hospital one time

# Focus is Prediction and Welfare

Probability of choosing



# Experimental Design

Estimate models on period before disaster

# Some Models Match Market Shares Well

Sumter

# Examine Relative Model Performance

Examine Percent Improvement in RMSE over Indicator Model

$$1 - \left( \frac{\text{RMSE}_{\text{Model}}}{\text{RMSE}_{\text{Indic}}} \right)$$

Three Sets of Predictions:

- | Aggregate Shares
- | Aggregate Diversion Ratios:  $\frac{y_{j;1} - y_{j;0}}{y_{\text{dest};0}}$
- | Individual Predictions

# Comparing Models: Aggregate Shares

# Comparing Models: Aggregate Diversion Ratio



# Comparing Models: Individual Predictions

# Optimal Model Combination

$$y_{ij} = \text{Semipar} \hat{y}_{ij}^{\text{Semipar}} + \dots + \text{CDS} \hat{y}_{ij}^{\text{CDS}} +$$

Weights non-negative, sum to one (Timmerman (2006))

Averaged across experiments

# Comparing Models: Individual Predictions

## Model Combination

# How Well Do Discrete Choice Models Perform?

Correlation Coefficient

# Robustness

Changing Choice Set Patients

Capacity Constraints

Case Mix

Medicare Insurance Only

Removing Destroyed Areas

Doctors

# Counterfactual Hospital Mergers

Hospitals with higher WTP have higher market power

- | Suppose hospital  $k$  and  $l$  are merging
- | Change in WTP

$$WTP_{(k;l)}(S) - WTP_k(S) - WTP_l(S)$$

approximates change in market power from merger

# Counterfactual Hospital Mergers

Hospitals with higher WTP have higher market power

- | Suppose hospital  $k$  and  $l$  are merging
- | Change in WTP

$$WTP_{(k;l)}(S) - WTP_k(S) - WTP_l(S)$$

approximates change in market power from merger

We examine all possible counterfactual mergers (95 in total)

# Meaningful Predicted Welfare Differences Across Models

SD / Mean



# Meaningful Predicted Welfare Differences Across Models

Correlation, RMSE and Percent Change in WTP

# Conclusion

In general, structural analysis and credible identification are complements.

# Conclusion

In general, structural analysis and credible identification are complements. [...] That this should not be an either-or proposition seems quite obvious to us.

{Nevo and Whinston (JEP 2010)}

Future Work: Examining Machine Learning Models