Advertising and Demand for Addictive Goods: The Effects of E-Cigarette Advertising

Anna Tuchman

Northwestern University - Kellogg

FTC Microeconomics Conference November 2016

What Are E-Cigarettes? Why Study This Industry?



- 4 同 6 4 日 6 4 日 6

Research Questions

- 1. What is the effect of e-cigarette advertising on demand for cigarettes?
 - Direct: advertising spillovers?
 - Indirect: substitutes or complements?
- 2. What would be the impact of banning e-cigarette advertising?

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のなの

Ad Spillovers - Renormalization & Visual Smoking Cues



イロト (部) イモト (モ) 通言 ぶんで

Product Complementarities - Indoor Use





Overview of Empirical Analyses

1. Descriptive analysis

Market data: identification of ad effects Household data: addiction and substitution patterns

2. Structural demand model

Both market and household data

3. Counterfactual analysis: e-cigarette ad ban

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のなの

Data Description - Aggregate and Household-Level Data

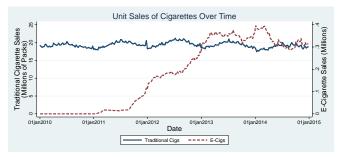
Nielsen Advertising Data

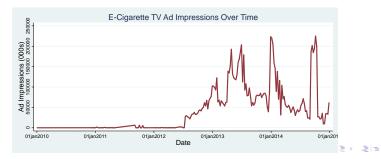
Weekly DMA-level TV ad impressions and GRPs 2009-2014

Nielsen Purchase Data

Weekly store sales volume and prices Daily household purchase panel 2010-2014

Data - Aggregate Trends in Sales and Advertising





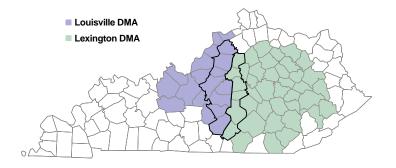
8/19

Data Description - E-Cigarette Shares by Brand

	Market Share	Ad Impression Share
Blu (Lorillard)	57.8%	74.1%
Vuse (RJ Reynolds)	1.1%	10.7%
NJOY	8.5%	8.4%
Fin	12.0%	4.2%
Other	20.6%	2.7%
Total	\$289,500,000	10,328,566,000

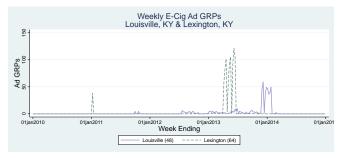
Identification of Advertising Effects - Border Discontinuities

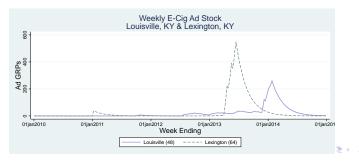
- Compare sales and advertising over time in counties on bordering DMAs (Shapiro, 2016)
- Control for different levels in demand Market FEs
- Control for regional demand shocks Border-Time FEs





Across Border Variation in Advertising Over Time





11/19

ъ

Market Level Border Counties Regression Results

$$Q_{mt} = \beta_m + \beta_{bt} + \phi \vec{A}_{mt} + \alpha \vec{p}_{mt} + \epsilon_{mt}$$

	E-Cig Cartridges		
E-Cigarette Ad GRPs	0.191***		
	(0.035)		
Smoking Cessation GRPs	-0.047***		
	(0.013)		
Price Controls	Y		
N Obs	52,236		
E-Cig Ad Elasticity	0.02		
$\%\Delta Q$ from 1 SD \uparrow A^e	4.86%		
*** p<0.01, ** p<0.05, * p<0.1			



Market Level Border Counties Regression Results

$$Q_{mt} = \beta_m + \beta_{bt} + \phi \vec{A}_{mt} + \alpha \vec{p}_{mt} + \epsilon_{mt}$$

	E-Cig Cartridges	Cigarette Packs			
E-Cigarette Ad GRPs	0.191***	-2.811***			
	(0.035)	(0.806)			
Smoking Cessation GRPs	-0.047***	-0.478			
	(0.013)	(0.315)			
Price Controls	Y	Y			
N Obs	52,236	52,236			
E-Cig Ad Elasticity	0.02	-0.004			
$\%\Delta Q$ from 1 SD \uparrow A^e	4.86%	-0.90%			
*** p<0.01	*** p<0.01, ** p<0.05, * p<0.1				



Household Data Description

- Observe all household purchases between 2010-2014
- Cigarettes, e-cigarettes, and smoking cessation products

Household Data Description

- Observe all household purchases between 2010-2014
- Cigarettes, e-cigarettes, and smoking cessation products
- 881 households buy e-cigarettes
- ▶ Mean HH: 3 packs cigarettes per week & 6 e-cig purchases
- Majority of HHs (83%) buy cigarettes before e-cigarettes

◆□ ▶ ◆□ ▶ ◆ □ ▶ ◆ □ ▶ ◆ □ ▶ ◆ □ ▶ ◆ □ ▶

Substitution and Addiction in Household Purchase Data

$$c_{it} = \alpha_i + \alpha_t + \gamma E_{it} + \beta C_{it} + \delta_1 P_{it} + \delta_2 G_{it} + \epsilon_{it}$$

	Cig Packs
E-Cig Cartridges in Prev 4 Weeks	-0.030***
	(0.008)
Cig Packs in Prev 4 Weeks	0.060***
	(0.007)
Nicotine Patches	Y
Nicotine Gum	Y
HH FE & Week FE	Y
N Obs	1,970,419
Mean DV	1.410
Effect $+1$ E-Cig as % of DV	-2.13%
*** p<0.01, ** p<0.05, *	p<0.1

Summary of Descriptive Results

- 1. E-cigarette advertising \uparrow e-cig demand and \downarrow cigarette demand
- 2. Household purchase patterns consistent with addiction
- 3. Traditional cigarettes and e-cigarettes are substitutes

Modeling Objectives and Challenges

1. Objectives

Leverage both individual and market-level data Identify advertising effects using border discontinuities Capture dynamic dependencies resulting from addiction Allow for heterogeneity in preferences

Simulate a counterfactual ban on e-cigarette advertising

Modeling Objectives and Challenges

1. Objectives

Leverage both individual and market-level data

Identify advertising effects using border discontinuities Capture dynamic dependencies resulting from addiction Allow for heterogeneity in preferences

Simulate a counterfactual ban on e-cigarette advertising

2. Challenges

Aggregation of individual-level model with state dependence and heterogeneity

Border discontinuity identification in structural model

▲□▶ ▲□▶ ▲□▶ ▲□▶ ヨ□ ののの

1. Household Demand

Addiction: t - 1 consumption c_{t-1} increases utility in t

$$u_{ijt} = \underbrace{\beta_j + \alpha p_{jt} + \phi A_t + \xi_{jt}}_{\delta_{jt}(\theta)} + \gamma c_{it-1} + \varepsilon_{ijt}$$

1. Household Demand

Addiction: t - 1 consumption c_{t-1} increases utility in t

$$u_{ijt} = \beta_j + \alpha p_{jt} + \phi A_t + \xi_{jt} + \gamma c_{it-1} + \varepsilon_{ijt}$$

$$\delta_{jt}(\theta)$$

2. Aggregate Demand

1. Household Demand

Addiction: t - 1 consumption c_{t-1} increases utility in t

$$u_{ijt} = \beta_{j} + \alpha p_{jt} + \phi A_t + \xi_{jt} + \gamma c_{it-1} + \varepsilon_{ijt}$$

$$\delta_{jt}(\theta)$$

2. Aggregate Demand

$$\sigma_{jt}(c_{t-1}| heta) = rac{e^{\delta_{jt}(heta)+\gamma c_{t-1}}}{1+\sum_k e^{\delta_{kt}(heta)+\gamma c_{t-1}}}$$

1. Household Demand

Addiction: t - 1 consumption c_{t-1} increases utility in t

$$u_{ijt} = \beta_{j} + \alpha p_{jt} + \phi A_t + \xi_{jt} + \gamma c_{it-1} + \varepsilon_{ijt}$$

$$\delta_{jt}(\theta)$$

2. Aggregate Demand

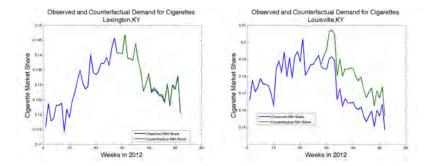
$$\sigma_{jt}(c_{t-1}| heta) = rac{e^{\delta_{jt}(heta)+\gamma c_{t-1}}}{1+\sum_k e^{\delta_{kt}(heta)+\gamma c_{t-1}}}$$

 Weight logit purchase probabilities for each consumption state by probability of the consumption state

$$egin{aligned} s_{jt} &= \mathbb{E}_{c_{t-1}}[\sigma_{jt}] \ &= \sigma_{jt}(c_{t-1} = 1| heta) imes extsf{Pr}(c_{t-1} = 1| heta) + \sigma_{jt}(c_{t-1} = 0| heta) imes extsf{Pr}(c_{t-1} = 0| heta) \ &= \sigma_{jt}(c_{t-1} = 1| heta) imes (1 - s_{0t-1}) \ &+ \sigma_{jt}(c_{t-1} = 0| heta) imes s_{0t-1} \end{aligned}$$

Counterfactual E-Cigarette Advertising Ban

- Impose a ban on e-cigarette advertising
- Simulate market demand using the counterfactual ad stock
- Median % Δ in cigarette market share: \uparrow 2.64%.



(日)

Contributions

1. Substantive

Advertising spillovers across categories First empirical analysis of e-cigarette ad effects at scale

Contributions

1. Substantive

Advertising spillovers across categories First empirical analysis of e-cigarette ad effects at scale E-cigarette advertising *decreases* demand for cigarettes Ban may have unintended consequences

Contributions

1. Substantive

Advertising spillovers across categories First empirical analysis of e-cigarette ad effects at scale E-cigarette advertising *decreases* demand for cigarettes Ban may have unintended consequences

2. Methodological

Aggregation of structural model with state dependence & unobserved heterogeneity

Identification of ad effects $\mathsf{w}/$ border discontinuities w/in nonlinear model

THANK YOU!

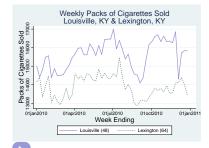
E-Cigarette Advertising Market Share by Media Type

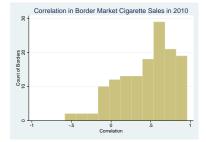
	Media Type	Dollar Share	Impression Share
National	Cable	73.4%	85.7%
	Network	3.0%	6.4%
	Syndicated	0.4%	0.5%
Local	Spot	23.1%	6.6%
	Network Clear Spot	0.0%	0.6%
	Syndicated Clear Spot	0.0%	0.2%
Total		\$54,185,012	10,328,566,000

Within and Across Market Variation in Ad Stock

	Ν	Min	Median	Mean	Max
Ave E-Cig GRP Stock	300	0.03	14.88	21.84	138.45
Ave Abs Δ E-Cig Ad Stock	150	0.70	21.25	27.89	139.54
Coeff Var E-Cig GRPs	300	1.64	3.56	4.01	10.00

Common Trends: Across Border Variation in Cig Sales





4/9

Identification Strategy - Challenges

- Identification relies on common trends assumption
- Problem if unobserved shock on one side of border correlated with sales and advertising.

County excise taxes increase County indoor smoking legislation tightens

Average Characteristics in Border and Non-Border Markets

	Border	Non-Border	
	Counties	Counties	<i>p</i> -value
% Female	50.14	50.16	0.764
% Population Under 18	22.22	22.74	0.000
% HS Diploma	83.31	85.16	0.000
% White	86.08	85.08	0.148
% Black	8.99	10.10	0.085
Per Capita Income	23,228	24,380	0.000
Population Density	169.4	502.1	0.001
N Counties	772	1,202	

Normalized Absolute Deviations in Demographics Across Bordering Markets

$$d_b^x = \frac{|x_{bi} - x_{bj}|}{\sigma_x}$$

	Ν	Min	Median	Mean	Max
% Female	150	0.00	0.57	0.87	5.59
% Population Under 18	150	0.00	0.58	0.79	3.59
% HS Diploma	150	0.01	0.46	0.61	3.88
% White	150	0.00	0.30	0.48	2.38
% Black	150	0	0.17	0.36	2.49
Per Capita Income	150	0.00	0.41	0.64	4.47
Population Density	150	0.00	0.17	0.48	4.81

Market Level Border Counties Regression Results

$$Q_{mt} = \beta_m + \beta_{bt} + \phi \vec{A}_{mt} + \alpha \vec{p}_{mt} + \epsilon_{mt}$$

	Patches	Gum
E-Cigarette Ad GRPs	-0.024	-1.546***
	(0.019)	(0.347)
Nicotine Patch Ad GRPs	-0.039*	1.062***
	(0.020)	(0.299)
Nicotine Gum Ad GRPs	-0.005	-0.310
	(0.015)	(0.217)
Price Controls	Y	Y
N Obs	37,077	37,077
E-Cig Ad Elasticity	-0.003	-0.006
$\%\Delta Q$ from 1 SD \uparrow A^e	-0.72%	-1.40%
*** p<0.01, ** p<	<0.05, * p<	<0.1

Assumptions on Error Term

$$Q_{mt} = \beta_m + \beta_{bt} + \phi \vec{A}_{mt} + \alpha \vec{p}_{mt} + \epsilon_{mt}$$

$$\uparrow \qquad \uparrow$$

$$\epsilon_{mt} = u_m + v_{bt} + \nu_{mt}$$

Assumptions

1.
$$Cov(\nu_{mt}, \nu_{mt-1}) = 0$$

No market-specific serial correlation after FEs Implied by common trends assumption

$$2. \quad Cov(\nu_{mt}, A_{mt}) = 0$$

Advertising not targeted based on demand in border markets

▲□▶ ▲□▶ ▲□▶ ▲□▶ □□ のQの