

The demand for advertising on television: What guides firms' decisions and how their choices change during highly rated telecasts

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The Demand for Advertising on Television

What Guides Firms' Decisions and How Their Choices Change During Highly Rated Telecasts

Peter Stein

Adviser: Professor Julie Holland Mortimer

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Statement of the Topic

My thesis primarily revolves around advertiser demand, examining the choices of individual advertisers and trying to determine why they make the choice to advertise during a given telecast. To begin, I try to determine patterns both within and across industries, looking at firms' choices and trying to figure out the underlying mechanics guiding their decision to advertise. After gaining some initial insight, I will conduct my research within the context of telecasts that typically have very high ratings, like the Super Bowl, World Series, or Grammy Awards. My hope is to compare the advertising decisions of firms within an industry both during highly rated telecasts and during "control" telecasts with average ratings in order to draw some insights into the demand for advertising on TV.

Objectives

I hope to complete this research across multiple industries that advertise a reasonable amount, so I've geared my time towards studying the Quick Serve Restaurant, Candy, and Wireless Telecommunications industries. In addition to analyzing each industry on its own, I do inter-industry comparisons as well, looking at whether firms tend to break from their advertising patterns during a highly rated telecast in some industries more than others. The Stata output below displays how much each of my chosen industries advertises compared to others in the market. The three I've chosen are three of the biggest television advertisers.

ad_category_name	Freq.	Percent	Cum.
Chocolate Candy Bars	74,893	0.79	16.60
Cigars & Tobacco	2	0.00	16.60
Citrus Fruits	2,292	0.02	16.63
Non-Bar Chocolate Candy	74,444	0.78	55.29
Non-Carbonated Soft Drinks Corporate ..	225	0.00	55.29
Non-Carbonated Soft Drinks General Pr..	35	0.00	55.29
Non-Carbonated Soft Drinks NEC	5	0.00	55.29
Non-Carbonated Waters	2,596	0.03	55.32
Non-Chocolate Candy	41,907	0.44	55.76
Quick Serve Restaurants	241,519	2.54	88.10
Radio Networks & Syndicated Programmers	769	0.01	88.11
Radio Networks & Syndicated Programme..	4	0.00	88.11
Radio Stations	5	0.00	88.11
Railroad (Incl Industrial Dev)	4,716	0.05	88.16
Railroad Travel	4	0.00	88.16
Wireless Telecom Providers	127,816	1.34	99.68
Women's Scents, Fragrances, Perfumes	7,788	0.08	99.76
Womens Electric Shavers & Hair Removers	5	0.00	99.76
Womens Health Care Products	11,214	0.12	99.87

This table displays statistics from the first 6 months of 2012

Methods

New Programs

```

[steinp@leides ~/datafiles] module load stata
[steinp@leides ~/datafiles] stata-mp

(R)
-----
STATA 14.1 Copyright 1985-2015 StataCorp LP
Statistics/Data Analysis
4985 Lakeway Drive
College Station, Texas 77845 USA
800-STATA-9C http://www.stata.com
979-696-4600 stata@stata.com
979-696-4601 (fax)

MP - Parallel Edition

4-user 24-core Stata network perpetual license:
Serial number: 501406200673
Licensed to: Boston College
Chestnut Hill
    
```

Using Stata in Linux

```

\documentclass[10pt,oneside,11pt]{article} % Use "amsart" instead of "article" for AMSLaTeX
format
\usepackage{geometry} % See geometry.pdf to learn the layout options. There are lots
\geometry{letterpaper} % ... or a4paper or a5paper or ...
\geometry{landscape} % Activate for rotated page geometry
\usepackage{float} % Activate to begin paragraphs with an empty line rather than an
indent
\usepackage{graphicx} % Use pdf, png, jpg, or eps/eps with \pdfinclude; use eps in DVI
mode
\usepackage{amsmath} % LaTeX will automatically convert eps -> pdf in pdfLaTeX
\usepackage{amsfonts}
\usepackage{amssymb}
\usepackage{setspace}
\usepackage{fontspec}
\setfont{serif}
\setfont{sans-serif}
\setfont{monospace}
\begin{document}
\title{The Demand for Advertising on Television}
\author{Peter Stein}
\date{March 31, 2016}
\begin{table}
\caption{Table 1: Taco Bell Rival Effects Regression}
\end{table}
\begin{table}
\caption{Table 3: Quick Serve Restaurant Crossover Data (NBA)}
\end{table}
\end{document}
    
```

Using Latex to Word Process

The Demand for Advertising on Television

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Peter Stein

March 31, 2016

Finished Product

Revising my Strategy: From Regressions to Tables

Table 1: Taco Bell Rival Effects Regression

Variable	Coefficient
McDonald's	-.035 (.056)
Sonic	.171 (.061)
Subway	.179 (.059)
Constant	.679 (.062)
Number of Observations	R-Squared
171	.081
	Adjusted R-Squared
	.065

Table 3: Quick Serve Restaurant Crossover Data (NBA)

Restaurants	Ad %	BK	DQ	KFC	LJS	McD	Son	Sub	TB	Wendy's
Burger King	.455	1	.098	.438	.008	.371	.363	.531	.816	.195
Dairy Queen	.067	.658	1	.368	.053	.184	.368	.421	.737	.237
KFC	.520	.382	.048	1	.003	.324	.338	.515	.850	.211
Long John Silver's	.025	.143	.143	.071	1	.857	0	.143	.214	.429
McDonald's	.373	.452	.033	.452	.057	1	.281	.519	.786	.257
Sonic	.307	.538	.081	.572	0	.341	1	.457	.908	.168
Subway	.513	.471	.055	.522	.007	.377	.273	1	.889	.284
Taco Bell	.810	.458	.061	.546	.007	.362	.344	.564	1	.226
Wendy's	.226	.394	.071	.488	.047	.425	.228	.646	.811	1

Results

Top Genres in the Wireless Industry

TOP GENRES	AT&T	Boost Mobile	Consumer Cellular Wireless
Genre 1	Reality (36,078)	Comedy (1,741)	Documentary (1,814)
Genre 2	Movies (25,127)	Movies (1,479)	News (1,094)
Genre 3	Sports (20,652)	Reality (1,427)	Reality (733)
Total Telecasts	160,422	8,253	6,421
TOP GENRES	Sprint	Straight Talk Wireless	T-Mobile
Genre 1	Sports (12,639)	Reality (4,154)	Reality (13,356)
Genre 2	Reality (8,822)	Movies (3,278)	Movies (10,023)
Genre 3	Movies (3,638)	Documentary (1,658)	Sports (9,058)
Total Telecasts	44,597	15,665	60,835
TOP GENRES	Verizon	Virgin Mobile	
Genre 1	Sports (25,190)	Reality (3,886)	
Genre 2	Reality (18,620)	Music (2,241)	
Genre 3	Movies (13,110)	Movies (1,894)	
Total Telecasts	109,049	12,962	

Looking at Genre and Day Part: two tables that illustrate some key summary statistics in the Wireless Telecommunications and Quick Serve Restaurant industries

Overall Wireless Top Genres
Reality (91,225)
Sports (69,467)
Movies (62,325)
Comedy (43,108)
Documentary (35,049)
Drama (30,466)
News (29,140)
Music (29,022)
Total: 438,203

Overall Quick Serve Top Day Parts
Daytime (150,757)
Prime (142,189)
Late Fringe (93,612)
Early Morning (83,122)
Early News (81,655)
Early Fringe (78,178)
Total: 756,372

Top Day Parts in the Quick Serve Restaurant Industry

TOP DAY PARTS	Burger King	Dairy Queen	KFC
Day Part 1	Prime (21,767)	Daytime (9,678)	Daytime (13,474)
Day Part 2	Daytime (21,085)	Late Fringe (5,379)	Prime (10,427)
Day Part 3	Late Fringe (14,778)	Early Fringe (5,189)	Early Fringe (6,573)
Total Telecasts	113,631	40,994	57,880
TOP DAY PARTS	Long John Silver's	McDonald's	Sonic
Day Part 1	Daytime (12,967)	Daytime (22,068)	Prime (9,469)
Day Part 2	Prime (9,004)	Prime (18,177)	Daytime (8,304)
Day Part 3	Early Morning (5,817)	Early Morning (16,031)	Late Fringe (6,510)
Total Telecasts	49,820	101,909	49,296
TOP DAY PARTS	Subway	Taco Bell	Wendy's
Day Part 1	Daytime (24,463)	Prime (16,280)	Daytime (18,934)
Day Part 2	Prime (23,635)	Daytime (11,274)	Prime (15,497)
Day Part 3	Late Fringe (16,536)	Late Fringe (10,395)	Late Fringe (11,609)
Total Telecasts	123,482	73,931	83,633

The largest firms do not vary much at all from the industry trends when it comes to selecting genres and day parts. Small firms, on the other hand, will distinguish themselves from the larger firms when it comes to top genres and day parts. Logically, such behavior makes sense, as the smaller firms may try to find a new market segment in which to compete so as to avoid being overshadowed by larger competitors.

Data

I have been provided the data I need for my research from professor Julie Holland Mortimer. Professor Mortimer obtained an extremely extensive dataset that covers every advertisement run on every network for three years straight (2011-2013). The data contain all the relevant information about the ad, about when and where the ad ran, as well as a plethora of demographic information, covering 105 demographic cuts with 4 variables each. A successfully merged copy of all telecasts, demographics, and ads from the first 6 months of 2012 has over 9 million observations and over 450 variables. These figures don't even count the ~7 million unmatched telecast observations.

Sample Observation

unique_id	program_name	progra~o	series~o
10_21898282_1685242800000	2013 NBA Playoffs	21898282	2211891
episode_title	Eastern Conference Finals: Game 3: Miami Heat at Indiana Pacers from Bankers Life ..		
episode~o	year	tv_rat~g	genre
ECF3	2013	NONE	Sports
network~o	network~o	national_airdate~e	
10	10	20130527 03:00:00	
runtim~es	original_airdate	type	hh_rat~g
150	20130526 00:00:00		
averag~e	actual~e	share	
453663.4	32956.17	.0177797	
num_hhs	num_stbs	num_a~bs	num_ho~s
933230.3	955575.1	111659	1134159
num_a~rs	averag~d	eventtime	
82390.42	.2951501	1.685e+12	
dem~x376	dem~g376	dem~e376	dem~s376
1.007632	.0040038	50793.78	12686477
dem~x377	dem~g377	dem~e377	dem~s377
	.9357886	.0037183	44954.71
Gap in the Observation for 420 demographic variables...			
de~g5070	de~e5070	de~s5070	de~x5071
.004479	114127.9	25480486	1.069009
de~g5071	de~e5071	de~s5071	de~x5071
.004277	95507.83	22720265	
_merge1	ad_national_air~e	ad_pod~n	ad_rat~g
matched (3)	20130527 03:28:01	3.47	.0047364
ad_ave~e	ad_no		
540777.6	519916		
ad_name	ad_id	ad_parent_name	ad_advertiser_name
The Craving is Back	12489442	Yum Brands Inc	Taco Bell Restaurant
ad_bran~e	ad_product_name	runtim~ds	ad_category_name
Taco Bell	Taco Bell Beefy Crunch Burrito	30	Quick Serve Restaurants
_merge2	matched (3)		
_merge3	bad_ra~g	bad_in~x	bad_avg
matched (3)	0	0	0
bad_hhue	daypar~3	daypar~1	
0	03	3	
daypar~2	daypart		
03:00	overnight		

Next Steps and Acknowledgments

Moving forward from here, I will begin work on looking at how ratings affect ad placement and build a decision tree to start predicting how these firms should react given a particular highly rated telecast. Following that, I will create a dataset of highly rated programs and see if I can identify whether or not firms are breaking out of pre-established patterns.

I would like to sincerely thank Professor Julie Holland Mortimer for her help and support throughout this process. Without her, none of this research would have been possible.

I would also like to thank Professor Bob Murphy for encouraging me to write a thesis last fall. Thanks for the support!