

A World Without Borders Revisited: The Impact of Online Sales Tax Collection on Shopping and Search

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Abstract

Since 2000, online shopping has grown dramatically as a share of total retail sales and, historically, it was an easy way to avoid sales taxes. Using household-level shopping data, sales tax data, and staggered online sales tax collection by Amazon, I identify price elasticities for households. I find household elasticities are -1.9 using changes in tax rates on Amazon, higher than previously found in the literature. This predicts a 13% decline in Amazon's revenues upon sales tax collection. I find that after Amazon collects sales taxes, households increase their spending on Amazon's taxed competitors, but not its untaxed competitors. Finally, I find no evidence that households change their overall search behavior or shift their spending offline. This is the first paper to determine whether households shift their online spending offline in response to sales taxes. Collecting sales taxes online will help local governments recapture lost tax revenue, but it is unlikely to shift customers back offline. As a result, enforcing sales tax collection online levels the playing field, but may not provide the local economic boost brick-and-mortar retailers are hoping for.

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1 Introduction

Online shopping has grown dramatically since 2000, reaching 11.4% of total retail sales in Q4 of 2019 (U.S. Census Bureau 2020). Reasons for this growth include lower costs (travel, time, etc.), higher convenience, and more variety (Thau 2013; Arnott 2016). Until 2018, traditional brick-and-mortar stores were at a structural disadvantage because they had to collect sales taxes while their online competitors did not. As a result, online retailers could offer consumers, on average, a 7% discount compared to brick-and-mortar retailers. This discount came strictly at the expense of state and local tax revenues, with losses ranging from \$8 to \$33 billion in 2018 (U.S. Supreme Court 2018).

In this paper, I estimate consumers’ price elasticity using price increases generated by sales taxes on online purchases. Due to pressure from state and local governments and aggressive fulfillment center expansion, online retailers (particularly Amazon) collect sales taxes in many states. Combining data on online shopping with local tax rates and Amazon’s tax collection behavior across states over time, I estimate how consumers’ online purchasing changes when sales taxes are collected online. Furthermore, I extend this analysis to include measures of online browsing as well as overall household spending (including offline expenditures) to estimate whether consumers’ search behavior or their composition of online and offline spending changes in response to online sales tax collection.

I combine online shopping data with offline sales tax rates to estimate whether areas with higher tax rates respond more strongly to Amazon’s sales tax collection. Previous research uses a variety of approaches to determine how sensitive consumers are to tax rates. Table 1 summarizes elasticity estimates from previous research. In the offline environment, research leverages cross-border variation in tax rates and estimates a wide range of elasticities from -30 to -0.2 (Asplund, Friberg, and Wilander 2007; Agarwal, Marwell, and McGranahan 2017; Davis 2011; Agarwal et al. 2017; Mikesell 1970). In the online environment, elasticity estimates range from -6 to 0 , but early efforts often used data from before 2001, i.e., before the mass adoption of the internet and before groundbreaking innovations like rating systems and free shipping were effectively implemented (Scanlan 2007; Ballard and Lee 2007; Alm and Melnik 2005; Goolsbee 2000). Recent work leverages detailed online shopping data, but often is limited to particular websites, product categories, or states (Einav et al. 2014; Anderson et al. 2010; Ellison and Ellison 2009; Hu and Tang 2014). My paper extends the work of Baugh, Ben-David, and Park (2018) and Houde, Newberry, and Seim (2017) by incorporating data on browsing and total household expenditures to present a fuller picture of how household behavior changes in response to Amazon’s sales tax collection.

Table 1: Estimated Elasticities

Elasticity Type	Paper	Estimate
Cross-border Price-Expenditure	Asplund, Friberg, and Wilander (2007) – Foreign price	0.2 to 0.5
	Asplund, Friberg, and Wilander (2007) – Domestic price	-0.2 to -1.3
	Agarwal, Marwell, and McGranahan (2017)	-2 to -30
	Davis (2011)	-2.2 to -3.6
	Agarwal, Chomsisengphet, Qian, and Xu (2017)	-2.3
	Mikesell (1970)	-6.3
Tax-Purchase	Scanlan (2007)	0.0
	Ballard and Lee (2007)	-0.2
	Alm and Melnik (2005)	-0.5
	Einav, Knoepfle, Levin, and Sundaresan (2014)	-1.8
	Goolsbee (2000)	-2.3
Tax-Quantity	Anderson, Fong, Simester, and Tucker (2010)	-1.9 to -2.9
	Ellison & Ellison (2009)	-6
Tax-Expenditure	Baugh, Ben-David, and Park (2018)	-1.2 to -1.4
	Houde, Newberry, and Seim (2017)	-1.3
	Hu and Tang (2014)	-3.75 to -4.5

Note: Early research focused on how taxes influenced the binary decision of whether or not to make an online purchase. Subsequent research has looked at how sales taxes affect actual online expenditures or quantities purchased. In order to delineate between these, I use "tax-purchase elasticity" to refer to the effect on the purchase decision while "tax-quantity elasticity" and "tax-expenditure elasticity" refer to the effect on online purchase quantities and expenditures, respectively.

I use a differences-in-differences approach to estimate a household’s price elasticity. The expansion of Amazon’s warehouse network and the passage of state laws requiring online sales tax collection generate variation in Amazon’s tax liability across states over time. As a result, I am able to examine how household behavior changes after Amazon begins collecting sales taxes. Furthermore, I leverage variation in tax rates between counties to better examine if a household’s responsiveness depends on the magnitude of their local tax rate.

I find that consumers reduce their Amazon spending by about 1.9% for each percentage point of sales tax applied online. Given an average sales tax rate of 7%, this elasticity translates into a 13% decline in retail spending on Amazon. Consumers also increase their spending on Amazon’s taxed competitors by 1%. This is one of the first papers to explicitly incorporate how consumer search behavior changes in response to tax changes. Even though consumers do shift their spending from Amazon to its competitors, I find no evidence that consumers’ browsing habits are significantly affected by Amazon’s sales tax collection. This is also one of the first papers to examine whether households shift their spending offline in response to online sales tax collection. I find no evidence that consumers make such a shift.

This paper is organized as follows. Section 2 describes the data. Section 3 analyzes how online spending responds to Amazon sales tax collection. Section 4 examines whether online browsing activity is affected by Amazon sales tax collection. Section 5 analyzes tax responsiveness across all consumer spending modes and Section 6 concludes.

2 Data Description

In this section, I describe the data used for my analysis and give a brief overview of their respective features.¹ comScore’s Web Behavior database provides information on household online shopping and browsing behavior. Nielsen’s Consumer Panel data provides information on household shopping and purchasing decisions. Finally, Tax Data Systems provides information on local sales tax rates.

2.1 comScore Web Behavior Database

I primarily use the comScore Web Behavior database, which contains the online browsing and transaction activity of households that opt-in to have their internet activity collected by comScore. The browsing data records how many minutes were spent and how many

¹Researcher’s own analyses derived based in part on data from The Nielsen Company (US), LLC and marketing databases provided through the Nielsen Datasets at the Kilts Center for Marketing Data Center at The University of Chicago Booth School of Business.

pages were viewed on each website. The transaction data records the website, product name, product category, price, quantity, and basket total (including shipping and taxes) of the purchase. The comScore data capture all online activity of a household and is not limited to particular goods or retailers, in contrast with previous research. I use this breadth of information to capture whether households substitute to other retailers and estimate how their aggregate online spending changes when sales taxes are collected online.

I restrict my sample to households that have complete demographic information and remove any purchases in categories which Amazon is a not competitor (e.g., no plane tickets, dating services, etc.) and focus on products that cost between \$1 and \$500 (in nominal dollars).² These filters reduce the original sample of about 576,000 households down to about 206,000 households and a total of about 2 million transactions.

Table 2: comScore Panel Summary Statistics

Statistic	Mean	Median	St. Dev.	Min	Max	Census (2016)
Household Size	3.04	3	1.40	1	6	2.53
Age	46.65	47	12.59	19	65	51.9
Income	59.01	62.50	31.66	7.50	100.00	59.04
Child Present	0.64	1	0.48	0	1	0.42
Hispanic	0.13	0	0.34	0	1	0.13
Sales Tax	0.07	0.07	0.02	0.00	0.11	-
N (Household-Years)	261,416					

Note: Age and income are reported in bins, so the midpoint of each bin is used. "Child Present" indicates whether a child is present in the household. Census data comes from "Historical Households Tables" and "Income and Poverty in the United States: 2016."

Table 2 shows the summary statistics of the panelists. The average household size is 3, average age is 47, and average income is \$59,000. About 64% of the households have a child and 13% are Hispanic. Households face an average sales tax rate of 7%. Overall, panelists are similar to the general population, but are slightly larger and more likely to have kids. Households rotate through comScore's panel with a median tenure of 12 months (the 25th percentile is 9 months and the 75th percentile is 12 months).

Table 3 shows the summary statistics for all 2 million transactions. The average real price of products (in December 2016 dollars) is \$39. 28% of products are purchased on Amazon, 42% from the website of a brick-and-mortar store and the remaining products are purchased from

²This restriction is primarily to screen out extreme prices that may be generated by the comScore monitoring software. For example, a \$20 item reduced to \$15 may mistakenly be recorded as \$2015 because of how the price information is captured.

Table 3: comScore Transaction Summary Statistics

Statistic	Mean	Median	St. Dev.	Min	Max
Real Product Price	39.09	20.82	57.06	1.00	608.90
Sales Tax	0.07	0.07	0.02	0.00	0.11
Amazon Purchase	0.28	0	0.45	0	1
Offline Amazon Competitor	0.42	0	0.49	0	1
Online Amazon Competitor	0.30	0	0.46	0	1
N	2,001,485				

Note: Prices are deflated to December 2016 price levels using the CPI. "Sales Tax" indicates average local sales tax rate, not the average sales tax paid on online transactions.

another online retailer.

Table 4: comScore Browsing Summary Statistics (Minutes)

Statistic	Mean	Median	St. Dev.	Min	Max
Total	152.50	51.7	388.33	0	10,196
Amazon	18.52	0	83.19	0	8,864
Untaxed Competitor	93.46	16	346.00	0	10,176
Taxed Competitor	40.51	8	100.53	0	9,810
N (Household-Months)	2,559,012				

Note: Using 2006–2016 comScore Web Behavior data, this table reports the distribution of monthly browsing durations, in minutes, on shopping websites.

Table 4 reports the summary statistics of browsing on shopping websites. The average household spends about two and a half hours per month on shopping websites. Browsing on Amazon’s untaxed competitors accounts for about one and a half hours, browsing on Amazon accounts for about 20 minutes and the remaining 40 minutes is on Amazon’s taxed competitors. There is a wide range of shopping behavior with many households spending no time on shopping websites in a given month and others spending up to five hours per day.

2.2 Nielsen Consumer Panel Data

I use the Nielsen Consumer Panel data from 2004–2016. This is a panel of about 178,000 unique households. I observe about 40,000 households each year from 2004–2006 and about 60,000 households each year from 2007–2016. Households scan all items that they purchase and then input information about quantities, prices, date of purchase, and store. Nielsen retains about 80% of its panel from year to year with the mean and median tenure of a household being four and three years, respectively.

Nielsen computes projection weights to ensure their sample is nationally representative. Weights are calculated to match population moments based on household size, income, age, race, ethnicity, education, occupation, and presence of children. All analyses use these projection weights unless otherwise stated. Table 5 presents descriptive statistics for households in the sample.

Table 5: Nielsen Consumer Panel Summary Statistics

Variable	Mean	SD	25th Pctile	75th Pctile	Census (2016)
Household income (\$000s)	56.53	31.41	27.5	85	59.04
Household size	2.55	1.45	1	3	2.53
Age	52.62	14.34	41.5	63	51.9
College Educated	0.38	0.48	0	1	0.37
Child present	0.33	0.47	0	1	0.42
Married	0.50	0.50	0	1	0.48
N (Household-Years)	637,493				
N (Households)	154,352				

Note: Data are weighted for national representativeness.

Unlike the comScore data, store identities are anonymized, but they are classified into broad categories such as “Grocery Stores,” “Electronics Store,” and “Online Shopping.” This categorization is enough to conduct a similar analysis as I do with the comScore data, with the caveat that the Nielsen data primarily focuses on basic household goods, so items like electronics and apparel will not be captured. For retailers with both online and offline presence, Nielsen classifies them separately. For example, if Firm X has both offline stores and a website, detergent purchased from a Company X store will be from a different “retailer” than detergent purchased from CompanyX.com. In the first case, the retailer would be classified as a “Discount Store” and in the second, the retailer is classified as “Online Shopping”.

Because of their addictive properties and laws regulating purchasing these products online, I exclude tobacco and alcohol products. I remove households with a student or military head of household as well as those with an annual income of less than \$5,000. Only about 2% of households are excluded and I use the remaining 154,000 households for my analysis. See Appendix A for further details of sample construction.

2.3 Additional Data Sources

I obtain state, county, and local sales tax rates from Tax Data Systems, now part of Thomson Reuters. These data contain monthly tax rates at the ZIP code level. I compile information

on state law changes and agreements with Amazon under which states began collecting taxes for online transactions. This information was gathered from a wide range of local, state, and national news sources. Prior to 2018, most states did not require online retailers to collect sales taxes. They have only been able to collect sales taxes from Amazon because of separate agreements or because Amazon has opened warehouses in their state.³ Before 2006, Amazon only collected sales taxes in Kansas, Kentucky, North Dakota, and Washington. By the end of 2016, Amazon collected sales taxes in an additional 25 states: Alabama, Arizona, California, Colorado, Connecticut, Florida, Georgia, Illinois, Indiana, Maryland, Massachusetts, Michigan, Minnesota, Nevada, New Jersey, New York, North Carolina, Ohio, Pennsylvania, South Carolina, Tennessee, Texas, Virginia, West Virginia, and Wisconsin as well.

Figure 1 shows the wide variation in sales tax rates across the United States in December 2016. About 80% of counties have combined state and local sales tax rates between 5.3% and 8.3%. However, sales tax rates range from 0% to over 10% in Louisiana.

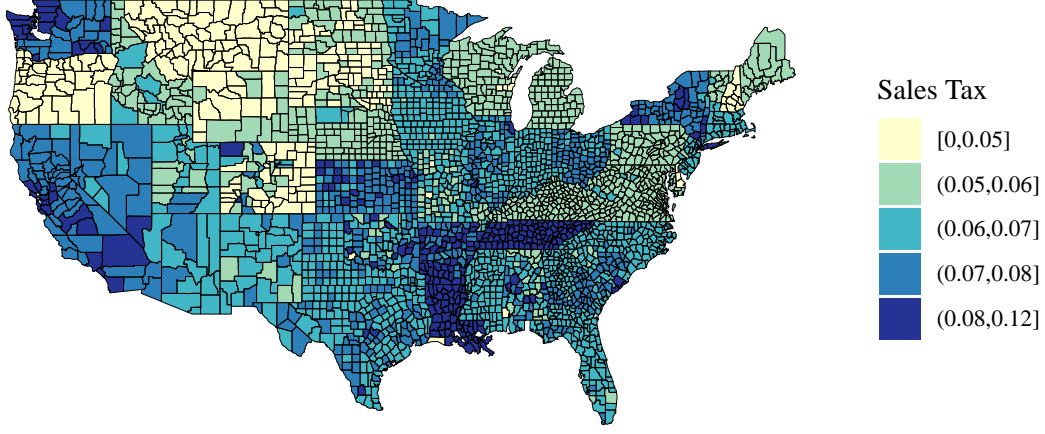
3 Amazon Sales Tax Collection and Online Spending

When shopping online, households can purchase from Amazon or one of its competitors. Amazon has two types of competitors: taxed and untaxed. Amazon's taxed competitors consist of traditional brick-and-mortar retailers, like Walmart and Target, who collect sales tax since they have physical locations across the country. Amazon's untaxed competitors are other online retailers, like Overstock.com or Etsy.com, which do not have physical locations across the country (generally just a headquarters location).

When Amazon begins collecting sales tax, consumers could respond in a variety of ways. First, they may not change their behavior and purchase on Amazon like usual (maybe not even notice the sales tax). Second, they could switch to one of Amazon's competitors. They could switch to an untaxed competitor if they value the tax savings or possibly a taxed competitor that offers a better selection or lower tax-inclusive prices. I examine each of these in turn to see how spending on Amazon changes and how spending at Amazon's taxed and untaxed competitors changes after Amazon begins collecting sales taxes.

³*National Bellas Hess v. Illinois* (1967) and *Quill Corp. v. North Dakota* (1992) ruled that retailers did not have to collect sales taxes in states where they did not have a physical presence. The Court held that tabulating tax liabilities for over 6,000 different tax jurisdictions would place an undue burden on many of these firms (Atkins 2005). The Supreme Court overturned these cases in *South Dakota v. Wayfair Inc.* (2018). Before 2018, consumers were supposed to self-report any unpaid taxes to the tax authorities, but compliance and enforcement were low, so these transactions were effectively tax-free (Manzi 2015).

Figure 1: Sales Tax Rates by County (Dec 2016)



Note: Using December 2016 Tax Data Systems data, this figure plots the spatial distribution of sales tax rates.

I use a differences-in-differences specification to identify the effect of Amazon's sales tax collection on a household's online spending:

$$Y_{ht} = \alpha + \beta AmazonCollect_{ht} * \tau_{ht} + \lambda_h + \lambda_t + \epsilon_{ht}, \quad (1)$$

where Y_{ht} measures real expenditures of household h in month t . $AmazonCollect_{ht}$ indicates whether Amazon collects sales tax (determined by month and state of residence of the household). τ_{ht} is the local sales tax rate.⁴ Household and time fixed effects are captured by λ_h and λ_t . β is the coefficient of interest measuring how spending changes after Amazon begins collecting sales taxes relative to what would have been expected had they not started collecting sales taxes. All standard errors are clustered at the state level because once Amazon collects sales taxes, it collects them across the whole state.

The policy of whether or not Amazon collects sales tax in a particular state or county is plausibly exogenous to the household spending decision. Often, it is prompted by the opening

⁴The local tax rate could also be included separately, but given the household fixed effect, this would only be identified off of changes in local tax rates which are relatively infrequent and when they do happen, are small.

of an Amazon warehouse in the state, but in a few cases, it is because of a change in state law. While there is a chance that these changes could be related to underlying economic fundamentals, Baugh, Ben-David, and Park (2018) show that sales tax collection by Amazon is not significantly related to state GDP growth, household income changes, or consumption declines. About 10% of households experience a change in Amazon's sales tax collection while they are in the sample.

Table 6: Online Spending Response to Amazon Sales Tax Collection

	Amazon	Untaxed Sites	Taxed Sites	Amazon	Untaxed Sites	Taxed Sites
	(1)	(2)	(3)	(4)	(5)	(6)
Collect	-0.422** (0.183)	-0.112 (0.150)	0.549** (0.249)			
Collect * Tax Rate				-6.173** (2.715)	-2.043 (2.165)	6.811* (3.545)
Household FE	Y	Y	Y	Y	Y	Y
Month-Year FE	Y	Y	Y	Y	Y	Y
Mean Spending	3.30	4.86	7.01	3.30	4.86	7.01
Mean Tax	0.068	0.068	0.068	0.068	0.068	0.068
Observations	5,076,040	5,076,040	5,076,040	5,076,040	5,076,040	5,076,040
Adjusted R ²	0.124	0.203	0.184	0.124	0.203	0.184

Note: *p<0.1; **p<0.05; ***p<0.01

Note: Using 2006–2016 comScore data on household-month online expenditures, this table reports the estimation results of Equation 1. "Collect" is a dummy variable indicating whether Amazon collected sales tax in a particular household-month. "Tax Rate" measures the local sales tax rate faced by a household in a particular month. All expenditures are real expenditures, deflated to December 2016 using the CPI. "Taxed Sites" refers to websites of retailers that have offline stores. "Untaxed Sites" are online-only retailers with no offline stores. Standard errors are clustered at the state level.

Table 6 reports the estimation results. Columns (1) – (3) only include an indicator for whether or not Amazon collects sales tax and then columns (4) – (6) allow for the response to vary with the sales tax rate. Column (1) demonstrates that Amazon spending decreases by an average of \$0.422 after Amazon collects sales taxes. Given that average monthly spending on Amazon is \$3.30 and the average sales tax rate is 6.8%, this equates to an elasticity of $\frac{-0.422/3.30}{0.068} = -1.88$. Columns (2) and (3) show that spending on Amazon's untaxed competitors does not significantly change while spending on its taxed competitors increases by \$0.549, implying a cross-elasticity of 1.14. Column (4) shows that the spending decreases on Amazon are stronger in areas with higher sales tax rates and the implied elasticity is a similar

-1.87.⁵ As before, there is no significant response on Amazon’s untaxed competitors and a marginally significant increase on Amazon’s taxed competitors. The spending increase on Amazon’s taxed competitors implies an elasticity of 0.97. Overall, after Amazon collects sales taxes, households reduce their Amazon spending and increase their spending on Amazon’s taxed competitors.

My estimated elasticity of -1.88 is higher (in magnitude) than similar estimates from Baugh, Ben-David, and Park (2018) and Houde, Newberry, and Seim (2017). My estimate differs from Houde, Newberry, and Seim (2017) for two reasons. First, while I use the same underlying data, I extend my sample for another three years through 2016, which doubles the number of states in which Amazon begins collecting sales tax from 12 to 25. Second, their analysis aggregates the data to the county-year level, while I aggregate the data to the household-month level. I get nearly identical estimates if I aggregate to the county-year level and limit my sample to 2006–2013. My estimate also differs from Baugh, Ben-David, and Park (2018) likely because of differences in the underlying data. First, my analysis spans 2006–2016 compared to 2011–2015, which adds an additional six states to my analysis.⁶ Restricting my analysis to 2011–2015 generates a slightly smaller elasticity of -1.66 , but this is still higher than the -1.2 to -1.4 estimated in Baugh, Ben-David, and Park (2018). The other possible contributor is the composition of our samples. The comScore data captures all online activity on a household’s computer and panelists are recruited to provide a representative measure of US internet users’ activity. On the other hand, the data used in Baugh, Ben-David, and Park (2018) is from an online account aggregator that likely targets younger, tech-savvy users interested in managing their finances effectively.⁷ These users are probably more likely to shop online and may be less likely to switch away from Amazon. This assertion is supported by comparing the average monthly spending on Amazon between the two samples. The average monthly Amazon spending of a comScore user is only \$3.30, but this increases to \$12.20 when restricting to only households that have made purchases on Amazon. In comparison, the average household in Baugh, Ben-David, and Park (2018) spends \$39. Overall, my estimate is higher than previous estimates because I incorporate more recent data and (arguably) a more representative sample of online shoppers.

⁵ $6.173 / 3.30 = -1.87$

⁶Baugh, Ben-David, and Park (2018) also restrict their analysis to households that spend more than \$200 on Amazon in 2011, but their Appendix B shows that removing this filter does not impact their estimate.

⁷One of the most popular financial aggregators, Mint.com, is reported to have a primarily young, male demographic. 71% of users were male and 64% were under 30 years of age back in 2008 (Perez 2008).

4 Amazon Sales Tax Collection and Online Browsing

The previous section shows that consumers are spending less on Amazon and more on Amazon’s taxed competitors. Do these changes in spending translate into changes in search behavior? I estimate Equation 1 with Y being minutes spent on Amazon or one of its competitors’ websites.

Table 7: Online Browsing Response to Amazon Sales Tax Collection

	Minutes Browsed					
	Amazon	Untaxed Sites	Taxed Sites	Amazon	Untaxed Sites	Taxed Sites
	(1)	(2)	(3)	(4)	(5)	(6)
Collect	0.076 (0.435)	0.387 (0.977)	0.512 (0.640)			
Collect * Tax Rate				0.369 (5.695)	6.317 (12.840)	7.160 (9.243)
Household FE	Y	Y	Y	Y	Y	Y
Month-Year FE	Y	Y	Y	Y	Y	Y
Mean Browsing (Min)	11.71	66.66	27.98	11.71	66.66	27.98
Mean Tax	0.068	0.068	0.068	0.068	0.068	0.068
Observations	5,076,040	5,076,040	5,076,040	5,076,040	5,076,040	5,076,040
Adjusted R ²	0.463	0.410	0.386	0.463	0.410	0.386

Note: *p<0.1; **p<0.05; ***p<0.01

Note: Using 2006–2016 comScore data on household-month online browsing, this table reports the estimation results of Equation 1. “Collect” is a dummy variable indicating whether Amazon collected sales tax in a particular household-month. “Tax Rate” measures the local sales tax rate faced by a household in a particular month. All browsing is in minutes. Standard errors are clustered at the state level.

Table 7 shows the results of this estimation. Our previous results indicate that households reduce their spending on Amazon only after sales tax is collected online. Because of this, we might expect that this reduced shopping activity would translate into reduced overall activity, measured in time spent on the website. Overall, I find no evidence that search on Amazon or its competitor websites is significantly affected by Amazon collecting sales tax. The lack of a significant browsing response may indicate that consumers are not changing their search behavior, but are simply switching their purchases away from Amazon since it no longer has a sales tax advantage.

Households’ relative unresponsiveness in search effort contrasts with the findings of Einav et al. (2014), which finds that when buyers realize sales taxes are added, they back out of

the transaction. This could be due to differences in user search between Amazon and eBay. On Amazon, the products are listed at a fixed price while on eBay, a share of items are sold at auction. In 2010 (the data used in Einav et al. (2014)), 40% of eBay sales were from auctions (eBay.com 2011). Given the risk of losing the auction, customers may be more likely to search on eBay relative to Amazon, where there is no risk of losing the purchase. Even if there is an effect that I cannot detect, it is likely to be small changes in browsing time, which could be generated by the extra effort needed to complete the purchase (e.g., the time needed to enter in address and credit card information).

Overall, households reduce their pre-tax expenditures on Amazon and shift to Amazon’s taxed competitors. In the next section, I examine whether households change their overall expenditures with a focus on whether their offline expenditures change in response to Amazon collecting sales tax.

5 Total Consumer Expenditure Analysis

The comScore data suggest that consumers reduce their Amazon expenditures when Amazon begins collecting sales tax, but they do not spend or browse significantly more on Amazon’s online competitors. The comScore analysis is limited to examining only online transactions and activity. Using Nielsen’s Consumer Panel Data, this section examines how households’ overall spending changes in response to Amazon’s sales tax collection. Households scan all of the items that they purchase for at-home consumption and input information about their shopping trip, including whether the purchase was from an online or offline retailer.⁸ Using these data, I can determine, for the common household items that are tracked, whether Amazon’s sales tax collection changes overall household expenditures and whether households are shifting spending offline in response.

To identify changes in consumer expenditures, I estimate Equation 1 where Y is the expenditures in either the online-only (likely untaxed) channel or offline channel. Since Amazon is most competitive in delivering less perishable, non-food products, I also separately examine whether online or offline non-food spending is affected. About 45% of households experience a change in Amazon’s sales tax collection during their tenure.

Table 8 shows that a range of household spending groups are unaffected by Amazon’s sales tax collection. Columns 1–3 show that there are no significant changes in total, online, or offline spending after Amazon begins collecting sales taxes. Even if there is an effect that I

⁸Retailers are anonymized in Nielsen, so I cannot identify whether an online purchase is made at a taxed or untaxed website as I could in the comScore data.

Table 8: Household Spending Response to Amazon Sales Tax Collection

	Real Spending					
	Total	Online	Offline	Online Non-Groc	Offline Non-Groc	
	(1)	(2)	(3)	(4)	(5)	
Collect	-0.682 (1.734)	0.036 (0.169)	-0.719 (1.772)	0.001 (0.095)	0.118 (0.458)	
Household FE	Y	Y	Y	Y	Y	
Month-Year FE	Y	Y	Y	Y	Y	
Mean Spending	316.39	4.7	311.69	2.93	94.74	
Mean Tax	0.068	0.068	0.068	0.068	0.068	
Observations	7,792,355	7,792,355	7,792,355	7,792,355	7,792,355	
Adjusted R ²	0.531	0.331	0.532	0.226	0.399	

Note: *p<0.1; **p<0.05; ***p<0.01

Note: Using 2006–2016 Nielsen Consumer Panel data, this table reports the estimates from regressing monthly spending on an indicator for Amazon sales tax collection (“Collect”) as well as household and month-year fixed effects and household demographics. Standard errors are clustered at the state level.

am not detecting, it is likely quite small. Part of this result is due to the fact that online shopping has low penetration into grocery and household non-durables (as indicated by online monthly spending averaging \$4). Even when analyzing only non-food items, where online shopping is most likely, there is no significant effect and the range of possible changes is quite small.

Because online shopping for household non-durables is relatively infrequent, the numerous months with no online expenditures may be masking changes in online purchases when they are made. To check this, I estimate the same regression, but only using months when purchases are made (i.e. conditional on making a purchase).

Table 9 reports the estimates conditional on making a purchase. Even after conditioning on making a purchase, online spending does not significantly change after Amazon collects sales taxes. Overall, there is no evidence of households shifting their spending offline in response to Amazon’s tax collection.

6 Future Research and Conclusion

Using data covering a broad range of online shopping activity, I find that consumers reduce their pre-tax spending on Amazon by about 1.9% for every percentage point of sales tax that

Table 9: Household Spending Response to Amazon Sales Tax Collection (Positive Purchases Only)

	Real Spending				
	Total	Online	Offline	Online Non-Groc	Offline Non-Groc
	(1)	(2)	(3)	(4)	(5)
Collect	-0.682 (1.734)	1.063 (1.106)	-0.699 (1.755)	-0.107 (0.974)	0.128 (0.458)
Household FE	Y	Y	Y	Y	Y
Month-Year FE	Y	Y	Y	Y	Y
Mean Spending	316.39	72.57	312.2	52.94	99.55
Mean Tax	0.068	0.068	0.068	0.068	0.068
Observations	7,792,355	482,702	7,782,003	408,811	7,464,450
Adjusted R ²	0.531	0.377	0.532	0.234	0.392

Note: *p<0.1; **p<0.05; ***p<0.01

Note: Using 2006–2016 Nielsen Consumer Panel data, this table reports the estimates from regressing monthly spending (conditional on have positive spending that month) on an indicator for Amazon sales tax collection ("Collect") as well as household and month-year fixed effects and household demographics. Standard errors are clustered at the state level.

Amazon collects. Furthermore, I find that households increase their spending on Amazon’s taxed competitors by 1% for each percentage point of sales tax Amazon collects. Even though households change their spending, they do not significantly change their search behavior on Amazon or its competitors. Finally, I find no evidence that households shift any of their spending offline after Amazon collects sales tax.

In light of the recent Supreme Court case, *South Dakota v. Wayfair* which increases state enforcement of sales tax collection online, state and local governments can expect a revenue boost because consumers are unlikely to shift their spending to untaxed channels. However, local policymakers and businesses will need to find other approaches if they want to encourage shoppers to move back offline. Online shopping is here to stay and more empirical work will be necessary to understand how traditional offline retailers can adapt to increased online shopping.

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A Data Appendix

A.1 comScore Web Behavior

This section provides more details about how I prepared the comScore data for analysis. All comScore data was obtained directly from Wharton Research Data Services (WRDS). I drop any households with incomplete demographic information. Additionally, I remove any households whose ZIP codes do not match with the Census Bureau’s 2010 Zip-to-County Relationship file. Table 10 shows that 35% of households remain based on these filters. The low retention rate is primarily a majority of households in the comScore data just browse the internet and make no online purchases while they are in the sample.

Table 10: comScore Sample

Step	HH
Starting HH:	586,420
Complete demographics:	585,867
Valid ZIPs:	576,457
Made Online Purchase:	206,435

Note: Table reports the number of households remaining in sample after each step of data cleaning.

For online transactions, I remove any transactions that are recorded for the same visit, same price, same time, and same product as a duplicate record. I also restrict my sample to products in categories that Amazon competes in, which excludes travel, dating, and financial products. Furthermore, I drop any transactions where the price is missing, less than \$1 or greater than \$500. Then, I remove any products that were sold by websites that do not feasibly compete with Amazon (e.g. daysinn.com or date.com).⁹ Table 11 shows that most transactions are omitted because they are duplicates or in non-Amazon competitive categories. The remaining portion of transactions are removed because the household did not make any Amazon purchase while they were in the sample. Overall, 41% of all transactions are made in Amazon competitive categories.

A.2 Nielsen Consumer Panel Data

This section provides more details about how I prepared the Nielsen Consumer Panel Data for analysis. I download the data directly from the University of Chicago Kilts Center for Marketing. I then remove any households with a military or student head of household or

⁹The full list of domains is available in the replication code.

Table 11: comScore Transactions

Step	Transactions
Starting Transactions:	4,934,867
Unduplicated Transactions:	3,956,424
Amazon Categories:	2,478,115
Invalid Prices:	2,269,680
Invalid Domains:	2,021,800

Note: Table reports the number of transactions remaining in sample after each step of data cleaning.

those that are making less than \$5,000 annually. Table 12 shows that only 2% of households are removed by these criteria.

Table 12: Homescan Sample

Step	HH
Starting HH:	158,004
Exclude military and students:	155,256
Exclude Households under 5k:	154,352

In the purchases data, I remove any alcohol and tobacco products as well as product categories that Nielsen has not consistently tracked over the full 2006–2016 period (i.e. “deferred modules”). I also exclude products that do not have Universal Product Codes (i.e. “magnet modules”). Finally, I remove any products with a recorded price of zero.

One final note for those familiar with Nielsen’s data products. Theoretically, I could use the Scanner data to determine whether an “online” retailer and another retailer share the same parent company (for the set of retailers that Nielsen track in their data). Unfortunately, none of the “Online Shopping” retailer codes are present in the Nielsen Scanner Data, so I cannot distinguish if the online retailer is the website of a traditional brick-and-mortar store (and thus taxed) or a stand-alone online store (and thus untaxed).