

The Dynamics of Automobile Expenditures

Adam Copeland

NY Federal Reserve

September 17, 2010

The views expressed herein are our own and not necessarily those of the Federal Reserve Bank of New York, or the Federal Reserve System

- A central question in durable goods analysis: how much do consumers time their purchase decisions?
- I focus on two issues arising with temporal substitution in the automobile market
 - 1 The degree of temporal substitution and its main patterns.
 - 2 Are dynamic pricing strategies important?
- I analyze the automobile market b/c
 - 1 Industry wisdom says consumers temporally substitute
 - 2 Seem to be large gains to temporally substituting
- With motor vehicles, have a unique window on timing of purchases
 - 1 detailed monthly price and sales vehicle data
 - 2 coordinated model-year cycle for motor vehicle
 - 3 demographic data

Overview of the Paper

- 1 Analyze how consumers time their new vehicle purchase decisions
- 2 Focus on decisions within the model year
 - choice set constant, but prices vary (decline)
 - clear tradeoff of enjoying vehicle now vs. waiting for price decline
- 3 Consumers modeled as an optimal stopping problem
- 4 Estimate parameters in the consumer's indirect utility function
- 5 Results:
 - 1 Consumers are price sensitive and willing to time purchases
 - 2 Entry/exit and temporal substitution are dominant channels of substitution. Cross-sectional substitution is relatively small.
- 6 Dynamic pricing strategies matter
- 7 Circumstantial evidence of price discrimination

- 1 Durable goods - infrequently purchased
 - (closest) Gowrisankaran and Rysman (2007) - DVD players
 - Carranza (2003, 2006), Song and Chintagunta (2003), Gordon (2006) and Nair (2005)
- 2 Non-durable goods - frequently purchased (grocery retail data)
 - impact of high/low pricing and other optimal pricing schemes,
 - Slade (1998), Aguirregabiria (1999), Pesendorfer (2002), Erdem et al (2003), Hendel and Nevo (2006).
- 3 Automobile pricing: incentives, dealer inventories, price cues (Zettelmeyer et al (2003,2006,2007)).
- 4 Macroeconomic issues
 - aggregate inventory behavior: Hall (2000) and Attanasio (2000)
 - GDP volatility: Ramey and Vine (2007)

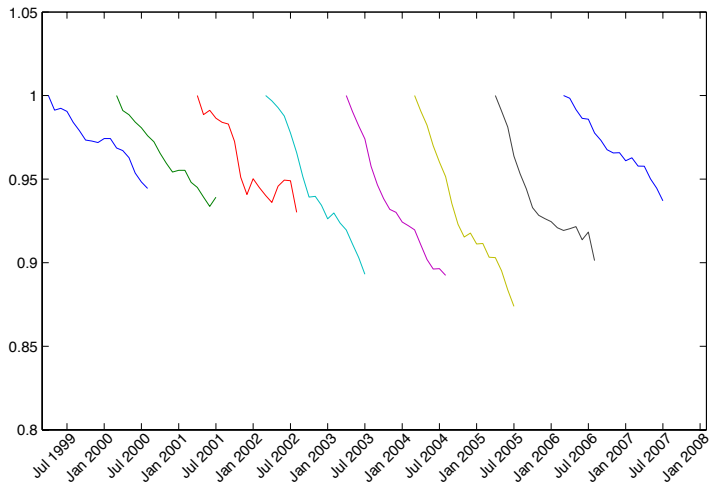
Talk outline

- 1 Data
- 2 Model
- 3 Results

- 1 First: Power Information Database (PIN), from JDPA
 - Collected from dealerships, 15-20% of all U.S. retail sales.
 - Monthly observations by model/model-year from 1999 to 2008
 - Observe price, cash rebate, and financial details
 - Observe model-year distribution of sales by model
- 2 Second: Wards Communications - US monthly sales by model.
- 3 Combine both data sets to get:
 - Monthly series of sales by model & model-year,
 - Monthly series of real market prices by model & model-year,

- 1 Prices fall over the model year at 9% annual rate
- 2 Mean income of new vehicle purchasers falls over model year
- 3 Model-level sales are hump-shaped
- 4 Aggregate sales volatile

Price indexes by model year: 2000 - 2007

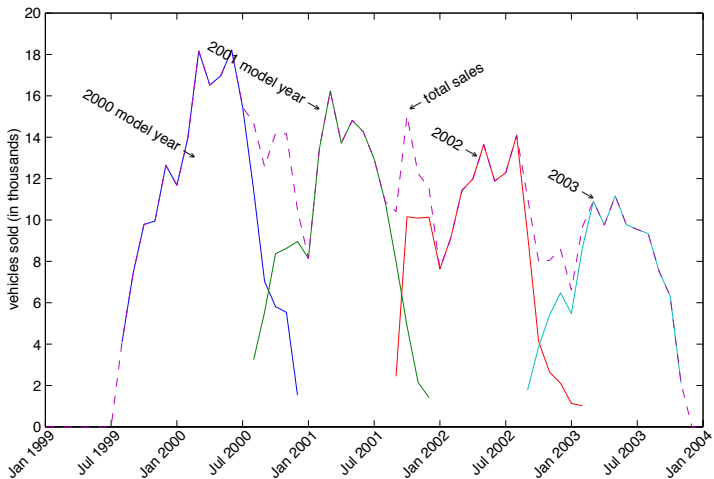


Data from Aizcorbe-Bridgman-Nalewaik 2007

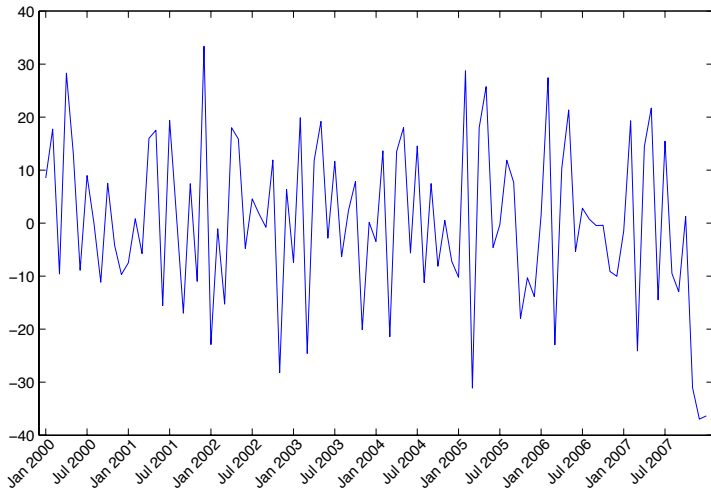
Quarter	Income (thousands of dollars) Data
1	74.973
2	73.075
3	71.460
4	68.603

Average Income of New Vehicle Purchasers

Pickup sales by model year



Percent Change in Aggregate Sales



Reduced form analysis of prices and sales

- Is there positive correlation between p_{t-1} and s_t ?
- First stage: detrend $\log(p)$, $\log(s)$ by model
- Second stage: for $n = \{1, 2, 3, \dots, 10\}$

$$\hat{s}_{jt} - \hat{s}_{j,t-n} = \varsigma + v(\hat{p}_{jt} - \hat{p}_{j,t-1})$$

periods of comparison	price coefficient	obs.
t,t-1	-0.434 (0.066)	33,566
t,t-2	-0.735 (0.078)	31,851
t,t-3	-1.087 (0.088)	30,135
⋮		
t,t-9	-0.567 (0.126)	19,878
t,t-10	-0.370 (0.133)	18,185

Model: Optimal stopping problem within Model Year

- All consumers show up at the beginning of the model year.
- Every month, each decides whether to buy a new vehicle or wait. If buy, then leave the market.
- In last month of the model year, waiting means not buying a new vehicle that model year.
- Remainder of income goes towards alternative consumption good.
- Aggregate across consumers to get predicted market shares
- Employ discrete-choice framework (BLP-style), includes
 - Consumer heterogeneity
 - Product differentiation
 - Constant choice set, changing prices
 - Perfect foresight on prices

Dynamic model

- if consumer i purchases vehicle j at time t , gets a net flow utility of

$$\delta_{ijt}^f - \beta^{T-t} \eta_i p_{jt} + \epsilon_{ijt} \quad (1)$$

where only δ_{ijt}^f persists over time.

- Let $\delta_{ijt} = \sum_{s=t}^{\infty} \beta^{s-t} \delta_{ijt}^f$, then indirect utility is

$$u_{ijt} = \delta_{ijt} - \beta^{T-t} \eta_i p_{jt} + \epsilon_{ijt}. \quad (2)$$

→ fits into the random coefficients discrete choice framework.

- Substituting in for δ ,

$$u_{ijt} = X_{jt} \gamma + \sum_{s=1}^{11} 1_{C_t=s} \zeta_s + \xi_{jt} + \sum_{k=1}^K \sigma_k \nu_{ik} X_{jkt} - \beta^{T-t} \frac{\eta}{y_i} p_{jt} + \epsilon_{ijt}, \quad (3)$$

- Price discounted as if paid in last period of the model year
- Separates consumption of alternative good from timing of the vehicle purchase (perfect smoothing assumption)

Dynamic Model – value functions

- Value function in last period

$$V_{iT} = \max \left\{ \pi_{iT}, \max_{j \in J} \{ u_{ijt} \} \right\}. \quad (4)$$

where $\pi_{iT} = \delta_{i0T}^f + \epsilon_{i0T}$ is utility flow from outside option (δ_{i0t}^f is set to zero)

- Value function for $t \in \{1, \dots, T - 1\}$

$$V_{it} = \max \left\{ \pi_{it} + \beta E[V_{i,t+1}], \max_{j \in J} \{ u_{ijt} \} \right\}. \quad (5)$$

- Expectations only over taste-for-variety term
- Perfect foresight: choice set, characteristics, and prices

Moment conditions & instruments

- GMM approach with 2 sets of moments:
 - ① regular BLP moments: $E[\xi|X] = 0$
 - ② average income of purchasers by quarter
- Problem with correlation between ξ and price
 - Firms and consumers observe ξ . Higher ξ likely implies higher price, violating moment condition
 - Usual instruments are competing vehicle characteristics.
 - Rely on vehicles' fuel efficiency and real gas price.
- Use real gas price as well as sum of fuel efficiencies across all other vehicles, all other manufacturers' vehicles, and all other vehicles produced by the same manufacturer.

Identification & Measurement

- Because of parametric assumptions on indirect utility, technically the model is identified.
 - worry about multicollinearity of observed characteristics
- What's the variation in the data driving the estimates?
 - 1 Across model-years, have changes in choice set with corresponding changes in sales
 - 2 Within a model-year, have changes in instruments and corresponding changes in sales
 - 3 Income moments are essential to pin-down distaste-for-price parameter
- caveat: is there a unique solution to the double loop iteration?

Used variation on the estimation strategy detailed in Gowrisankaran & Rsyman (2007).

- Parameters are precisely estimated
- Estimate heterogeneity in consumer tastes
- Positive trend in indirect utility over model year (consistent with inventories generating sales)
- Temporal substitution is a large force

Parameter		Estimate	Std Error
Mean	Constant	-4.594	(0.324)
	Acceleration	2.727	(0.441)
	Height	2.174	(0.594)
	Size	2.129	(0.729)
	Miles per Dollar	-2.906	(0.467)
Variance	Constant	3.746	(0.145)
	Acceleration	3.016	(0.391)
	Height	0.410	(0.224)
	Size	1.494	(0.127)
Distaste-for-price	η	4.565	(0.365)
Loss Criterion		155.50	
Observations		21,073	

Parameter		Estimate	Std Error
Month	1	-0.728	(0.050)
Dummies	2	-0.294	(0.046)
	3	-0.221	(0.051)
	4	-0.037	(0.055)
	5	-0.206	(0.052)
	6	-0.015	(0.047)
	7	0.255	(0.037)
	8	0.252	(0.028)
	9	0.334	(0.024)
	10	0.256	(0.023)
	11	0.122	(0.021)

Quarter	Data	Dynamic Model
1	\$74,973	\$73,702
2	\$73,075	\$72,438
3	\$71,460	\$71,163
4	\$68,603	\$70,002

Table: Average Income of New Vehicle Purchasers

Table: Own Price Elasticities for the 2002 Model Year (absolute value)

Market Segment	Dynamic Model Quartiles		
	1/4	1/2	3/4
CUV	1.40	1.51	2.27
Large car	1.50	1.59	1.67
Luxury car	2.05	2.41	2.94
Midsize car	1.25	1.41	1.50
Pickup	1.23	1.60	1.77
Compact car	0.87	0.94	1.07
SUV	1.63	1.90	2.30
Van	1.41	1.47	1.61
All	1.35	1.62	2.20

Table: Within Market Segment Cross-Price Elasticities for the 2002 Model Year

Market Segment	Dynamic Model		
	Min	Mean	Max
CUV	0	0.00017	0.00129
Large car	0	0.00030	0.00123
Luxury car	0	0.00004	0.00287
Midsize car	0	0.00028	0.00302
Pickup	0	0.00093	0.00726
Compact car	0	0.00016	0.00171
SUV	0	0.00038	0.00437
Van	0	0.00024	0.00207

note: Statistics are computed only over vehicles within the same market segment.

Three counterfactuals

- 1 A temporary price discount on GM pickups, holding all else constant
- 2 A temporary price discount on GM pickups, allowing for price responses
- 3 All firms collude to not offer incentives.

Table: \$500 Incentive for GM Pickups in month 6

Month	Chrysler	Ford	GM	All others
Unit sales: difference between data and counterfactual				
6	-32	-49	3,332	-76
7	-30	-47	-63	-69
8	-29	-45	-64	-66
9	-31	-48	-62	-72
10	-30	-48	-72	-70
11	-28	-45	-76	-71
12	-31	-51	-73	-79
Total	-210	-332	2,923	-503

→ 1,877 consumers entered the market

Estimating price responses

- Assume GM is the price leader for pickups
- d_j denotes price incentives on pickup j , d_{GM} is the average price incentive on GM pickups and m is a monthly trend
- inv_j are inventories of vehicle j , included to capture aggregate shocks
- Estimate regression

$$d_{jt} = \rho m_t + \omega inv_{jt} + \sum_{j=1}^J I_{v_t=j} \alpha_j + \sum_{k=1}^5 I_{C_t=k} (\kappa_k^0 d_{GM,t} + \kappa_k^1 d_{GM,t-1} + \kappa_k^2 d_{GM,t-2}) + \varepsilon_{jt} \quad (6)$$

- α are model fixed effects, $(\kappa^0, \kappa^1, \kappa^2)$ estimated for each company

Table: Predicted Incentive Responses to a \$500 GM Incentive in Month 6

	Month			
	6	7	8	9
GM	500	0	0	0
Chrysler	20	237	213	0
Ford	96	259	79	0
All others	50	20	52	0

Table: \$500 Price Decline for GM Pickups in month 6, with price responses

Month	Chrysler	Ford	GM	All others
Unit sales: difference between data and counterfactual				
6	-17	378	3,237	-91
7	603	1,539	-156	-139
8	572	360	-148	-61
9	-72	-110	-139	-168
10	-70	-109	-161	-162
11	-65	-103	-167	-165
12	-71	-115	-161	-182
Total	881	1,839	2,304	-969

→ 4,055 consumers entered the market

Table: Comparing flows with and without the price response

	Chrysler	Ford	GM	All others
<i>Units</i>				
no p.r.	-210	-332	2,923	-503
p.r.	881	1,839	2,304	-969
<i>Revenue (\$ thousands)</i>				
no p.r.	-4,889	-8,104	19,767	-12,721
p.r.	2,517	4,517	4,495	-28,351

Table: Average Cash Rebate over the 2002 Model Year

Months within the Automotive Model Year											
1	2	3	4	5	6	7	8	9	10	11	12
312	234	251	345	654	805	757	805	861	938	827	846

Table: 2002 Model Year Sales and Revenue with and without Incentives

Month	Unit Sales (thousands)				Revenue (\$ million)			
	CF	Data	diff.	cdiff.	CF	Data	diff.	cdiff.
1	514	505	9	9	13,254	12,902	352	352
2	1,135	1,117	18	27	29,168	28,379	790	1,142
3	1,112	1,094	19	46	29,113	28,339	774	1,915
4	1,186	1,175	10	56	31,159	30,421	738	2,653
5	1,020	1,038	-18	38	26,242	25,868	374	3,027
6	1,195	1,242	-46	-8	30,608	30,403	204	3,231
7	1,390	1,436	-45	-54	35,498	35,145	352	3,583
8	1,324	1,374	-49	-103	33,582	33,287	295	3,878
9	1,374	1,429	-55	-158	34,566	34,302	265	4,143
10	1,372	1,441	-69	-227	34,340	34,233	107	4,250
11	1,356	1,410	-54	-280	34,322	34,140	182	4,432
12	1,413	1,466	-54	-334	35,810	35,606	204	4,636

- With motor vehicles, have an excellent opportunity to measure temporal substitution
- Find consumers are price sensitive and willing to substitute temporally across many months
- Temporal substitution and exit/entry are the main channels of substitution
- B/c of temporal substitution, accounting for dynamic pricing strategies is important. For example revenue gains are vastly overestimated (5 times in my example).
- Circumstantial evidence that automakers take advantage of consumers' ability to temporally substitute and price discriminate.