

Consumer Willingness to Pay for Vehicle Characteristics: What Do We Know?

David Greene,* Anushah Hossein,** Gloria Helfand,*** Robert Beach**

*University of Tennessee

**RTI International

***U.S. Environmental Protection Agency

Why study willingness to pay (WTP) for vehicle attributes?

- Related to consumer vehicle choice modeling:
 - To use such a model to estimate the effects of policy on vehicle demand, one would want to know whether the model does a reasonable job of capturing responses.
 - If models using different data or estimation methods produce similar values for WTP, then the models may have found common consumer behavioral responses
 - If models produce different estimates, then how do we know that these models are consistently modeling or predicting behavior?
- The value of knowing a value:
 - As more fuel-saving technologies are used, there may be engineering tradeoffs (or complementarities) between fuel savings & other vehicle characteristics.
 - If we identify such tradeoffs/complementarities, having WTP values would allow us to monetize the changes in values.

WTP for Vehicle Attributes

- Many researchers have estimated demand for vehicles or their characteristics
 - Typically, in discrete choice models
 - Some in hedonic models
- Different kinds of data
 - Market-level data
 - Individual revealed preference data
 - Stated preference data
- These researchers have not necessarily reported WTP values implied by their analyses
 - Having these values would facilitate comparisons across studies

Estimating WTP from existing literature

- Goal: to estimate WTP values for vehicle attributes out of as many studies as possible
 - To have available estimates of these values
 - To see whether values are reasonably consistent across studies
- We decided to focus on US-based studies, 1995-present
 - Older & foreign studies are not as likely to be relevant
- Final sample of 52 papers

Population statistics

Paper count	52
Observation count	786
Unique attribute count	146
Literature type	
Peer-reviewed	45 (86.5%)
Grey	7 (13.5%)
Data type	
Revealed preference (RP) survey	19.3%
Stated preference (SP) survey	39.3%
Market data	29.0%
Other (Joint RP-SP, literature summaries)	12.4%
Model type	
Hedonic demand	9.6%
Multinomial logit (MNL)	29.4%
Nested multinomial logit (NMNL)	13.1%
Mixed logit (MXL)	30.2%
Berry-Levinsohn-Pakes (BLP)	7.7%
Other	11.0%

Bottom line: Lots of variability

- Results vary pretty widely, not only across studies, but within studies
- Modeling results appear sensitive to a number of factors, potentially including:
 - Sources of underlying data
 - Modeling methods
 - Included & omitted variables
 - Functional form
- They suggest a lack of robustness in the measurement of these WTP values.
 - Which raises the question of the robustness of the underlying models &/or parameters

Calculating Willingness to Pay (WTP)

- For discrete choice models, we calculated it as
 - - Marginal Utility (MU) of the Attribute/MU of price
 - This is not strictly correct, because it is the ratio of two random variables
 - It is the first-order approximation in a Taylor Series expansion
 - A second-order approximation requires knowing the covariance matrix of estimates, rarely reported in publications
 - Some back-of-the-envelope calculations suggest the bias is small when coefficients are statistically significant; for non-mixed logit, the bias shrinks for more correlated coefficients
- For hedonic models, we calculated it as
 - Derivative of price with respect to the attribute
 - Also an imperfect estimate, because it is not strictly a demand-side estimate

Sources of Variability/Uncertainty

- Different studies produce different estimates
 - Sometimes one study produces multiple estimates
 - By analyzing the data multiple ways
 - Our reporting of the mean, standard deviation, and median of these estimates focuses on these central estimates
 - “Raw” results include all WTP estimates
 - “Trimmed” results drop outliers
- Each of the individual estimates of WTP has a range around it
 - We use +/- one standard error for the attribute
 - Not accounting for variation in $dU/dPrice$
 - Some variation is due to variation in the population
 - When price interacts with income, we use 25th & 75th percentiles of income distribution
 - In random effects modeling we use +/- 1 standard deviation of the attribute variable
- And there are different measures of attributes
 - E.g., fuel economy may be \$/mile, miles/\$, \$/year, gallons/mile, miles/gallon

Example: WTP for horsepower

- 6 studies use this metric, producing 11 estimates (\$/hp, 2015\$)

Study	Low WTP	Central WTP	High WTP
Beresteanu & Li	-49,864.08	0.00	49,864.08
Beresteanu & Li	-55092.47	0.00	55092.47
Fifer & Bunn	31.42	39.41	47.40
Greene & Duleep	*	13.84	*
Greene & Duleep	*	13.81	*
Klier & Linn	-967.30	9.18	985.65
Klier & Linn	-109.09	1.24	111.58
Klier & Linn	-360.17	8.31	376.79
Klier & Linn	-117.03	1.33	119.70
McCarthy	297.02	355.01	412.99
Skeros & Raichur	143.23	147.99	152.74
	Raw Mean	\$53.65	
	Standard Deviation	\$103.73	
	Median	\$9.18	

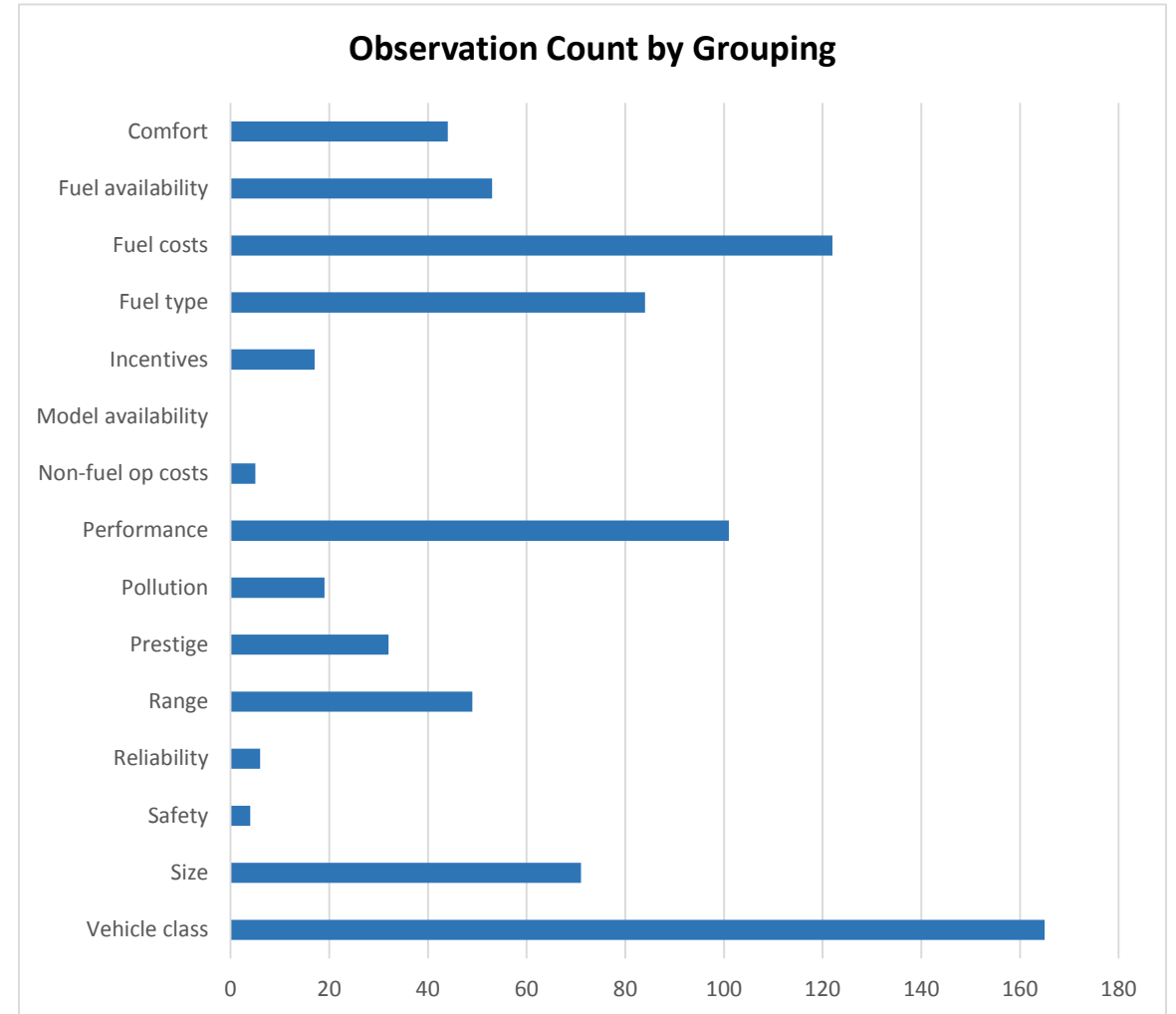
- Each estimate can have a big range
- 1 study can produce different estimates
- Across studies, even more variation

A trimmed mean might drop, e.g., the 2 estimates > \$100 as outliers; trimmed mean = \$9.68. If you drop 4 estimates (> \$100, = 0), trimmed mean = \$12.45

* Not enough information in the paper to calculate low & high values

Vehicle Characteristics

- The analysis identified 786 estimates of 146 unique attributes
- We categorized the 146 attributes into these 15 groups
 - E.g., Performance includes
 - 0-30 time
 - 0-60 time
 - Horsepower
 - Horsepower/weight
 - Top speed
- Where possible, we converted them to common units



Findings

- Following are findings on attributes with at least 6 observations
 - Fuel economy
 - Performance
 - Fuel type
 - Alternative fuel vehicle (AFV) range
 - Size
 - Comfort attributes
- Enough observations to observe variability
- Focus is on trimmed means – excluding outliers
 - This is intended to provide the best opportunity to find a robust central estimate
- Coefficient of variation as a measure of variation
 - Here, it is the dispersion of central values from the studies, not the variation around any one central value.
 - If the high and low WTP values are taken into consideration, variation will be much higher

Fuel cost: Five measures of fuel economy

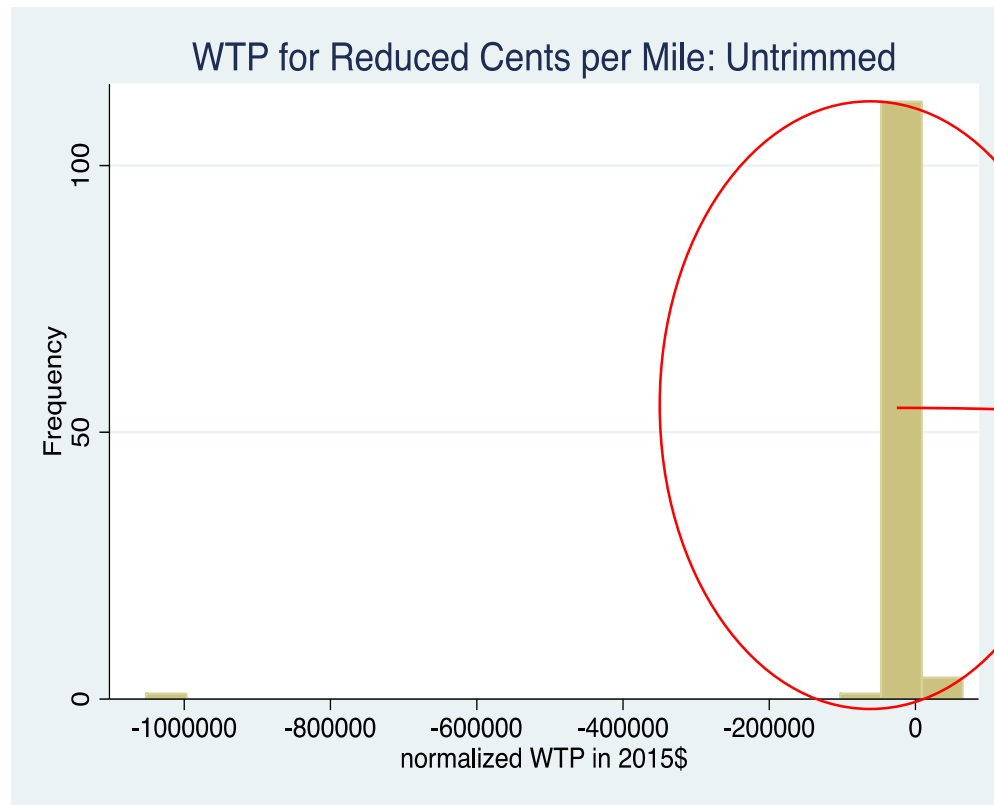
Measure	Present value comparison*	Trimmed Mean	Median	Coefficient of variation**
Reduce \$1/year	\$9/year	\$26	\$6	1.8 – 4.3
MPG	\$450	\$536	\$433	1.4 – 1.4
(10) Mi/\$	\$1800	-\$3,150	-\$2,837	0.8 – 1.4
Reduce 0.01 Gal/mi	\$2900	\$2,666	\$2,569	1.4 – 67.1
Reduce \$0.01/mi	\$1200	\$972	\$1012	2.4 – 2.6
Combined GPM and \$/mi	\$1200	\$640	\$769	11.6 – 14.9

*An order-of-magnitude calculation of the present value of a one-unit reduction in the measure, for comparison purposes.

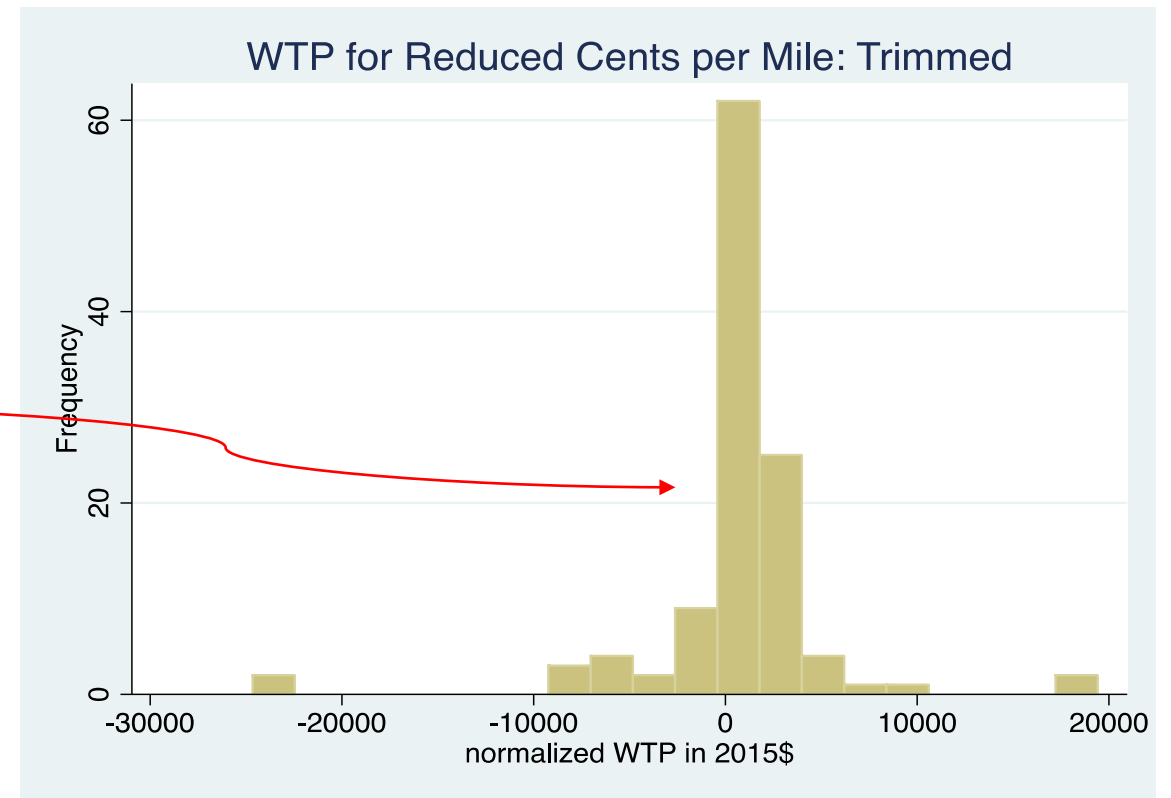
**The smaller value represents the coefficient of variation for the trimmed mean; the larger value is the coefficient for the set of estimates including outliers.

- Different studies use different measures, which makes comparisons difficult
- Best estimates of means are not necessarily close to “present value comparison”
- Mean & median are often different, meaning that the distribution of estimates is skewed
- Quite high variation around the estimates

Graphical portrayal of variation in the central estimates of \$0.01/mile decrease in fuel cost (2015\$)



Untrimmed

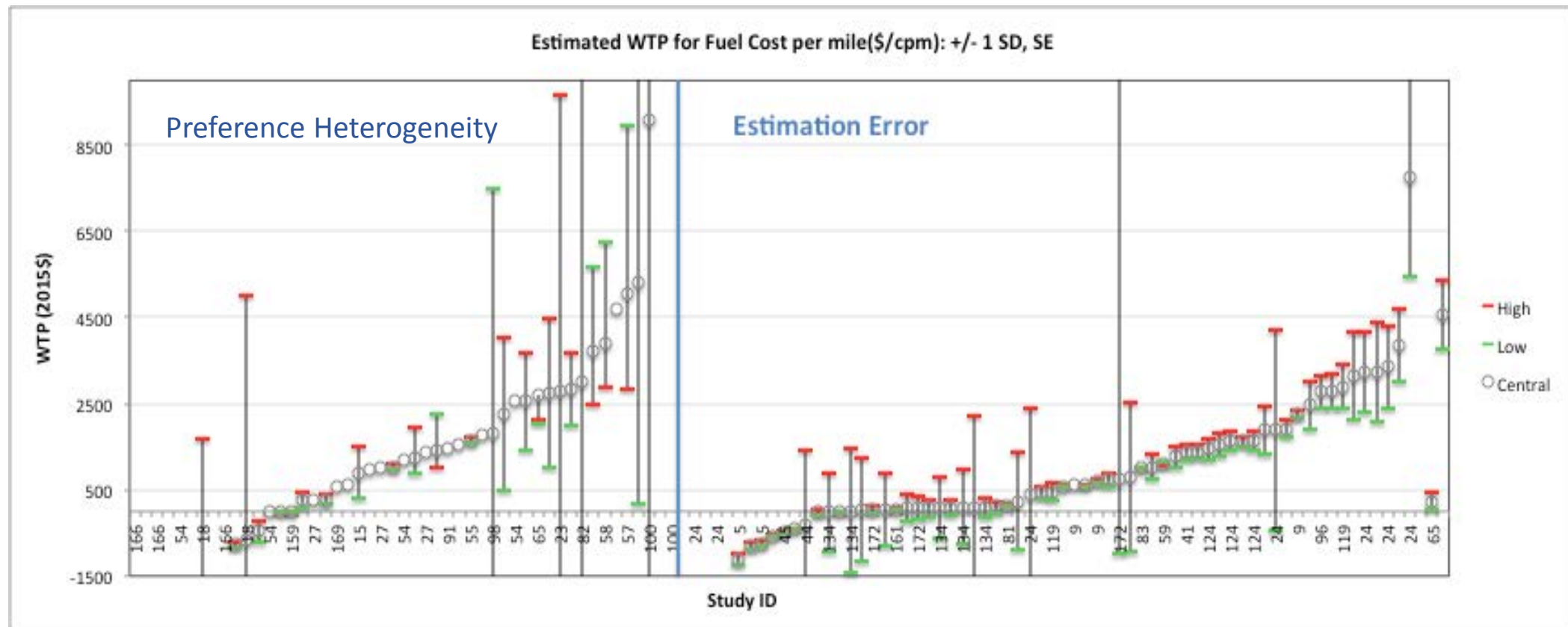


Trimmed

Different Scales on Axes!

Variation in WTP for \$0.01/mi decrease in fuel cost

- In some studies, variation is due to preference heterogeneity via random coefficients
- In some studies, variation is due to the standard errors about each estimate

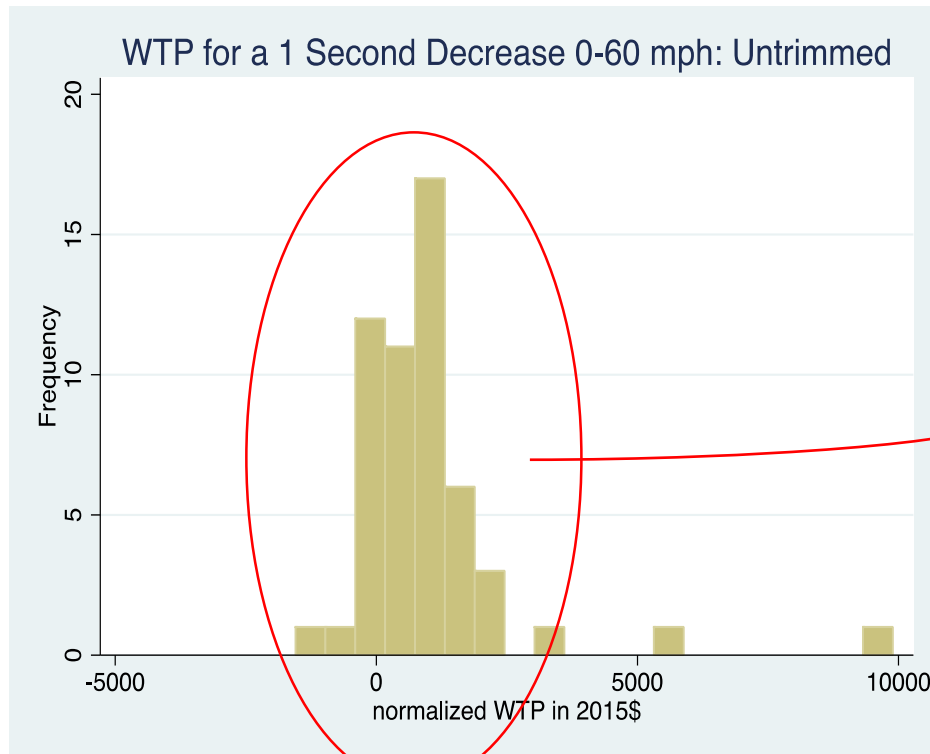


Performance: 5 measures

Measure	Trimmed Mean	Median	Coefficient of variation
Reduce 0-30 time (\$/sec)	\$1756	\$1916	1.1
Reduce 0-to-60 time (\$/sec)	\$1096	\$1183	0.6
Horsepower (\$/hp)	\$13	\$10	1 – 2
Horsepower/weight [\$/ (0.01hp/lbs)]	\$1334	\$346	1.6–1.9
Top speed (\$/mph)	\$100	\$75	0.6

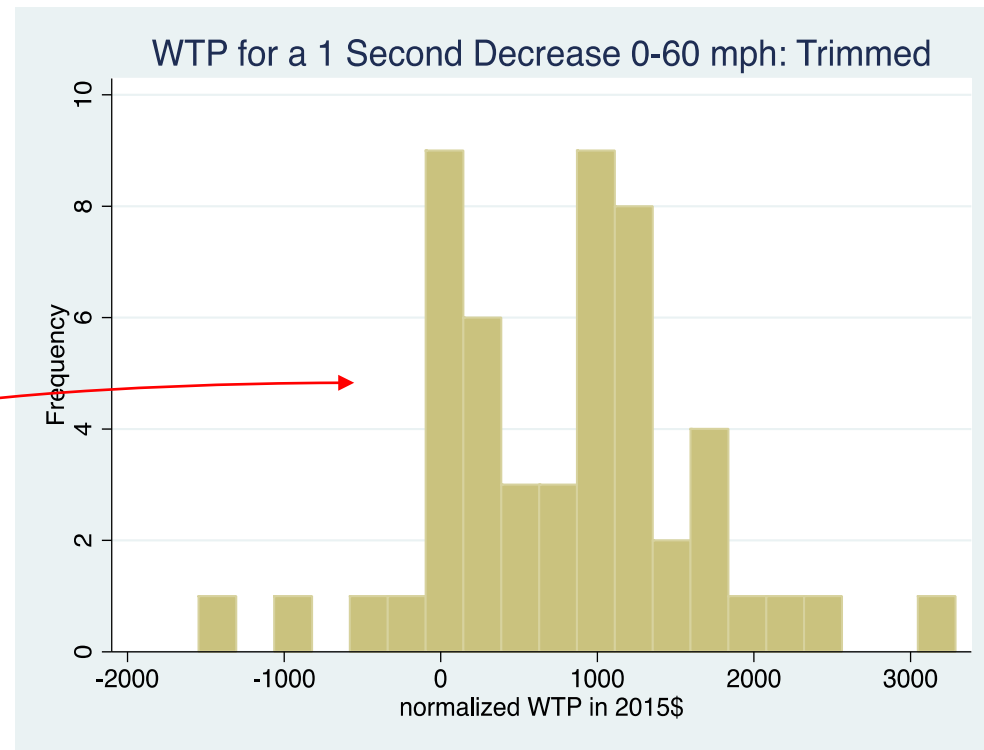
Performance: central tendencies, normalized to 0-to-60

- Combining 0-30 time, 0-60 time, hp/wt., via conversion factors



Untrimmed

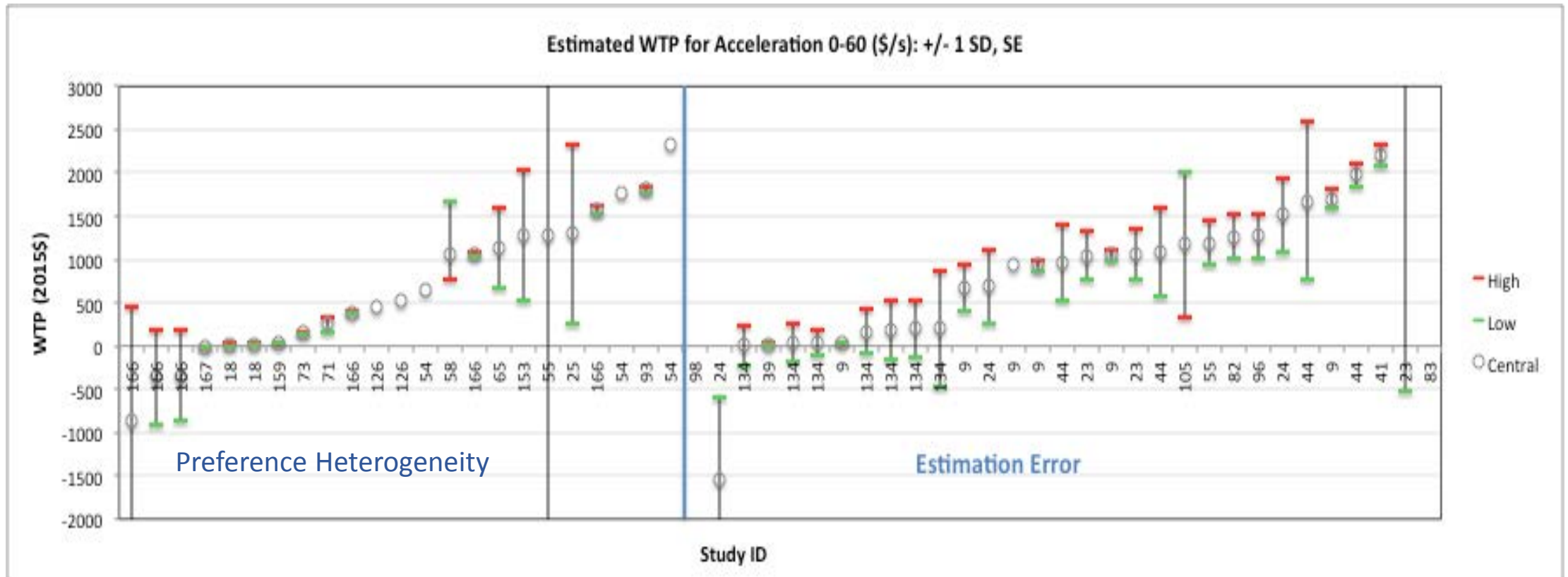
Different Scales on Axes!



Trimmed

Variation in Value for a One Second Decrease in 0-60 Time

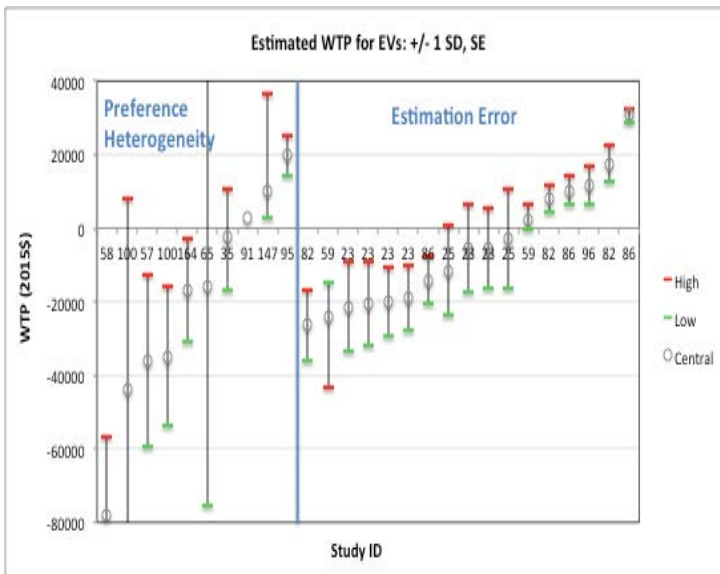
- In some studies, variation is due to the uncertainty about each estimate
- In some studies, variation is due to population variation – e.g., in income, via random coefficients



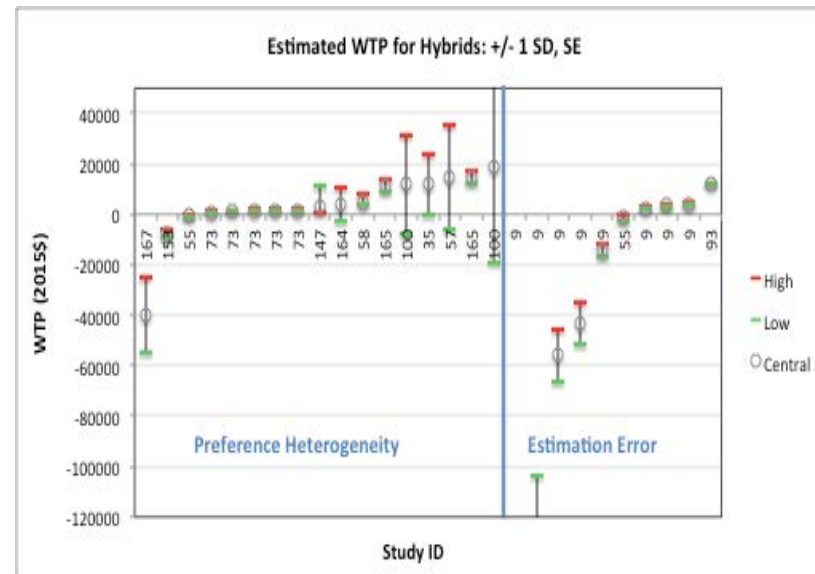
Fuel type

Measure	Trimmed Mean	Median	Coefficient of variation
EVs	-\$10,526	-\$11,392	2.2
Hybrids	-\$1,437	+\$2,375	3.5-12.9
Natural gas	-\$5,620	+\$4,620	4.2

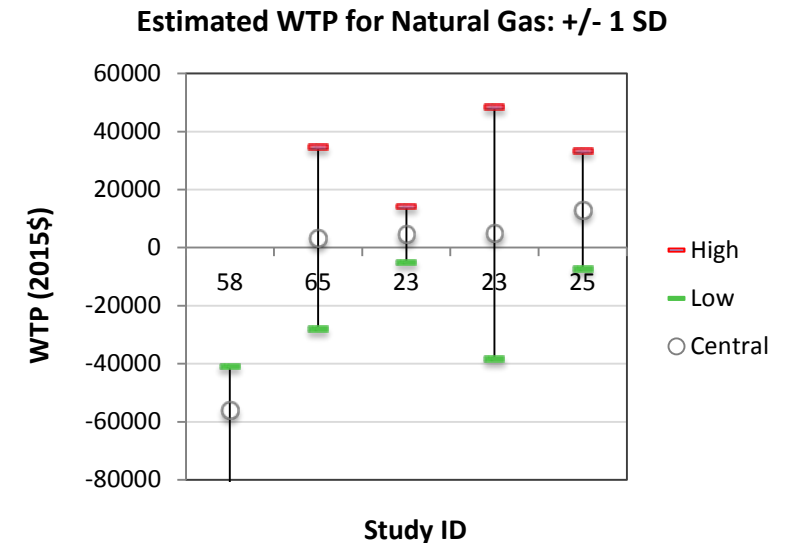
EVs



HEVs



NGVs



Alternative Fuel Vehicle Range

Measure	Trimmed Mean	Median	Coefficient of variation
Range	\$7	\$60	19.1

Size

Measure	Trimmed Mean	Median	Coefficient of variation
Footprint	\$3404	\$2283	1.3-4.0
Luggage space	\$1445	\$1100	0.9-2.3
Weight	\$6	\$1	1.4-2.0

Comfort attributes

Measure	Trimmed Mean	Median	Coefficient of variation
Automatic transmission (vs. manual)	\$1760	\$1522	2.1
All-wheel drive	\$32,031	\$26,779	0.6
Air conditioning	\$1085	\$4177	2.8
Shoulder room	\$1085	\$592	1.3

Summary

- This study contributes estimates of WTP for a variety of vehicle characteristics from a number of studies
- Results vary pretty widely, not only across studies, but within studies
- Modeling results appear sensitive to a number of factors, e.g.,
 - Sources of underlying data
 - Modeling methods
 - Included & omitted variables
 - Functional form
 - Measure
- They suggest a lack of robustness in the measurement of these WTP values.
 - Which raises the question of robustness in the underlying models &/or parameters

Implications

- Consumer choice modeling
 - It suggests that current models, and modeling approaches, can produce quite different results due to what might seem like minor changes
 - It would be helpful if researchers calculated & presented these values themselves
 - To facilitate comparisons
 - Or, they could provide sufficient information for others to calculate these values
- Estimating opportunity costs or ancillary benefits of changes in vehicle attributes, especially those other than fuel economy
 - It's not clear that these estimates are informative about these characteristics
 - There may be reasons to doubt that these studies are even estimating what they claim to estimate
 - E.g., is willingness to pay for fuel economy partly capturing effects of size and quality, since high fuel economy was historically associated with smaller, lower-quality cars?
 - Pre-footprint-based standards
 - It's possible that deeper digging might produce more sensible results
 - E.g., perhaps there's a trend in the value of performance over time, that this analysis won't recognize

Appendix

All values are in 2015\$

					Raw		Trimmed			
Grouping	Attribute	N	Units	Outliers	Mean	SD	Mean	SD	Median	Skew
Comfort	Auto-transmission	9	0/1	1	1760	3669	823	2518	1111	0.74
	Front wheel drive	6	0/1	0	-32031	18031	-32031	18031	-26779	1.20
	Air conditioning	13	0/1	0	3521	9544	3521	9544	4177	0.84
	Shoulder room	12	\$/inch	1	1085	1394	705	479	546	1.29
Fuel costs	Cost per mile	58	\$/cpm	2	-1251	3441	-1291	1194	-1147	1.13
	Cost per year	13	\$/(\$/yr)	1	-67	156	-26	50	-6	4.47
	Gallons per mile	20	\$/0.01gpm	4	14354	76395	-7972	18740	-580	13.74
	Miles per dollar	8	\$/((10mi)/\$)	1	-20181	27869	-11542	14477	-4216	2.74
	Miles per gallon	10	\$/mpg	1	365	659	174	281	64	2.70
Fuel type	Electric vehicle	24	0/1	1	-16515	21283	-13851	17191	-16837	0.82
	Hybrid	28	0/1	2	-11727	44322	-852	18441	2796	-0.30
	Natural gas	7	0/1	2	-5620	23691	6187	3851	5006	1.24
Performance	Acceleration (0-30)	11	\$/s	0	-1756	1886	-1756	1886	-1916	0.92
	Acceleration (0-60)	8	\$/s	0	-1096	627	-1096	627	-1183	0.93
	Horsepower	11	\$/hp	4	54	109	13	13	10	1.32
	HP/weight	29	0.01hp/lbs	1	1861	3523	1334	2126	346	3.85
	Top speed	9	\$/mph	0	100	58	100	58	75	1.33
AFV Range	Range	23	\$/mi	2	89	41	97	32	98	1.00
Size	Footprint	17	\$/ft^2	1	43401	163103	3856	4442	3273	1.18
	Luggage space	12	\$/ft^3	1	4209	9655	1445	1310	1100	1.31
	Weight	19	\$/lb	1	10	20	6	8	1	11.14

When Raw Mean differs strongly from Trimmed Mean, outliers matter.

When Trimmed Mean differs from Median, the distribution of estimates is skewed.

Median may be better than mean for a skewed distribution.

Large standard deviations indicate variation in estimates.

Studies included

- Allcott, Hunt, and Nathan Wozny. "Gasoline Prices, Fuel Economy, and the Energy Paradox." *Review of Economics and Statistics* 96, no. 5 (2014): 779-95. doi:10.1162/rest_a_00419.
- Axsen, Jonn, Dean C. Mountain, and Mark Jaccard. "Combining Stated and Revealed Choice Research to Simulate the Neighbor Effect: The Case of Hybrid-electric Vehicles." *Resource and Energy Economics* 31, no. 3 (2009): 221-38. doi:10.1016/j.reseneeco.2009.02.001.
- Beresteanu, Arie, and Shanjun Li. "Gasoline Prices, Government Support, And The Demand For Hybrid Vehicles In The United States*." *International Economic Review* 52, no. 1 (2011): 161-82. doi:10.1111/j.1468-2354.2010.00623.x.
- Berry, Steven, James Levinsohn, and Ariel Pakes. "Automobile Prices in Market Equilibrium." *Econometrica* 63, no. 4 (1995): 841. doi:10.2307/2171802.
- Brownstone, David, and Kenneth Train. "Forecasting New Product Penetration with Flexible Substitution Patterns." *Journal of Econometrics* 89, no. 1-2 (1998): 109-29. doi:10.1016/s0304-4076(98)00057-8.
- Brownstone, David, David S. Bunch, and Kenneth Train. "Joint Mixed Logit Models of Stated and Revealed Preferences for Alternative-fuel Vehicles." *Transportation Research Part B: Methodological* 34, no. 5 (2000): 315-38. doi:10.1016/s0191-2615(99)00031-4.
- Brownstone, David, David S. Bunch, Thomas F. Golob, and Weiping Ren. "A Transactions Choice Model for Forecasting Demand for Alternative-fuel Vehicles." *Research in Transportation Economics* 4 (1996): 87-129. doi:10.1016/s0739-8859(96)80007-2.
- Busse, Meghan R., Christopher R. Knittel, and Florian Zettelmeyer. "Are Consumers Myopic? Evidence from New and Used Car Purchases." *American Economic Review* 103, no. 1 (2013): 220-56. doi:10.1257/aer.103.1.220.
- Dasgupta, Srabana, S. Siddarth, and Jorge Silva-Risso. "To Lease or to Buy? A Structural Model of a Consumer's Vehicle and Contract Choice Decisions." *Journal of Marketing Research* 44, no. 3 (2007): 490-502. doi:10.1509/jmkr.44.3.490.
- Daziano, Ricardo A. "Conditional-logit Bayes Estimators for Consumer Valuation of Electric Vehicle Driving Range." *Resource and Energy Economics* 35, no. 3 (2013): 429-50. doi:10.1016/j.reseneeco.2013.05.001.
- Dreyfus, Mark K., and W. Kip Viscusi. "Rates of Time Preference and Consumer Valuations of Automobile Safety and Fuel Efficiency." *The Journal of Law and Economics* 38, no. 1 (1995): 79-105. doi:10.1086/467326.
- Espey, Molly, and Santosh Nair. "Automobile Fuel Economy: What Is It Worth?" *Contemporary Economic Policy* 23, no. 3 (2005): 317-23. doi:10.1093/cep/byi024.
- Fan, Qin, and Jonathan Rubin. "Two-Stage Hedonic Price Model for Light-Duty Vehicles." *Transportation Research Record: Journal of the Transportation Research Board* 2157 (2010): 119-28. doi:10.3141/2157-15.
- Feng, Ye, Don Fullerton, and Li Gan. "Vehicle Choices, Miles Driven, and Pollution Policies." *Journal of Regulatory Economics* 44, no. 1 (2013): 4-29. doi:10.1007/s11149-013-9221-z.
- Fifer, D.P.C. and N.P. Bunn. "Assessing Consumer Valuation of Fuel Economy in Auto Markets", Honors Thesis, Department of Economics, Duke University, Durham, NC (2009).
- Frischknecht, Bart D., Katie Whitefoot, and Panos Y. Papalambros. "On the Suitability of Econometric Demand Models in Design for Market Systems." *Journal of Mechanical Design J. Mech. Des.* 132, no. 12 (2010): 121007. doi:10.1115/1.4002941.
- Gallagher, Kelly Sims, and Erich Muehlegger. "Giving Green to Get Green? Incentives and Consumer Adoption of Hybrid Vehicle Technology." *Journal of Environmental Economics and Management* 61, no. 1 (2011): 1-15. doi:10.1016/j.jeem.2010.05.004.

Studies included, continued

- Goldberg, Pinelopi Koujianou. "Product Differentiation and Oligopoly in International Markets: The Case of the U.S. Automobile Industry." *Econometrica* 63, no. 4 (1995): 891. doi:10.2307/2171803.
- Gramlich, J. Gas Prices and Endogenous Produce Selection in the U.S. Automobile Industry, manuscript, Department of Economics, Yale University, New Haven, Connecticut, November, 20, 2008.
- Greene, D. L. "TAFV Alternative Fuels and Vehicles Choice Model Documentation." *Department of Energy*, 2001. doi:10.2172/814556.
- Greene, D.I. "Future Potential of Hybrid and Diesel Powertrains in the U.S. Light-duty Vehicle Market." *Oak Ridge National Lab*, 2004. doi:10.2172/885725.
- Haaf, C. Grace, Jeremy J. Michalek, W. Ross Morrow, and Yimin Liu. "Sensitivity of Vehicle Market Share Predictions to Discrete Choice Model Specification." *Journal of Mechanical Design J. Mech. Des.* 136, no. 12 (2014): 121402. doi:10.1115/1.4028282.
- Helveston, John Paul, Yimin Liu, Elea McDonnell Feit, Erica Fuchs, Erica Klampfl, and Jeremy J. Michalek. "Will Subsidies Drive Electric Vehicle Adoption? Measuring Consumer Preferences in the U.S. and China." *Transportation Research Part A: Policy and Practice* 73 (2015): 96-112. doi:10.1016/j.tra.2015.01.002.
- Hess, Stephane, Kenneth E. Train, and John W. Polak. "On the Use of a Modified Latin Hypercube Sampling (MLHS) Method in the Estimation of a Mixed Logit Model for Vehicle Choice." *Transportation Research Part B: Methodological* 40, no. 2 (2006): 147-63. doi:10.1016/j.trb.2004.10.005.
- Hess, Stephane, Mark Fowler, Thomas Adler, and Aniss Bahreinian. "A Joint Model for Vehicle Type and Fuel Type Choice: Evidence from a Cross-nested Logit Study." *Transportation* 39, no. 3 (2011): 593-625. doi:10.1007/s11116-011-9366-5.
- Hidrue, Michael K., George R. Parsons, Willett Kempton, and Meryl P. Gardner. "Willingness to Pay for Electric Vehicles and Their Attributes." *Resource and Energy Economics* 33, no. 3 (2011): 686-705. doi:10.1016/j.reseneeco.2011.02.002.
- Kavalec, Chris. "Vehicle Choice in an Aging Population: Some Insights from a Stated Preference Survey for California." *EJ The Energy Journal* 20, no. 3 (1999). doi:10.5547/issn0195-6574-ej-vol20-no3-5.
- Klier, Thomas, and Joshua Linn. "New-vehicle Characteristics and the Cost of the Corporate Average Fuel Economy Standard." *The RAND Journal of Economics* 43, no. 1 (2012): 186-213. doi:10.1111/j.1756-2171.2012.00162.x.
- Lave, Charles A., and Kenneth Train. "A Disaggregate Model of Auto-type Choice." *Transportation Research Part A: General* 13, no. 1 (1979): 1-9. doi:10.1016/0191-2607(79)90081-5.
- Liu, Yangwen, Jean-Michel Tremblay, and Cinzia Cirillo. "An Integrated Model for Discrete and Continuous Decisions with Application to Vehicle Ownership, Type and Usage Choices." *Transportation Research Part A: Policy and Practice* 69 (2014): 315-28. doi:10.1016/j.tra.2014.09.001.
- Liu, Yizao. "Household Demand and Willingness to Pay for Hybrid Vehicles." *Energy Economics* 44 (2014): 191-97. doi:10.1016/j.eneco.2014.03.027.
- McFadden, Daniel, and Kenneth Train. "Mixed MNL Models for Discrete Response." *Journal of Applied Econometrics* 15, no. 5 (2000): 447-70. doi:10.1002/1099-1255(200009/10)15:53.3.co;2-t.
- Mccarthy, Patrick S. "Market Price and Income Elasticities of New Vehicle Demands." *The Review of Economics and Statistics* 78, no. 3 (1996): 543. doi:10.2307/2109802.
- Mccarthy, Patrick S., and Richard S. Tay. "New Vehicle Consumption and Fuel Efficiency: A Nested Logit Approach." *Transportation Research Part E: Logistics and Transportation Review* 34, no. 1 (1998): 39-51. doi:10.1016/s1366-5545(97)00042-2.
- Mcmanus, Walter. "The Link Between Gasoline Prices and Vehicle Sales." *Business Economics* 42, no. 1 (2007): 53-60. doi:10.2145/20070106.

Studies included, continued

- Musti, Sashank, and Kara M. Kockelman. "Evolution of the Household Vehicle Fleet: Anticipating Fleet Composition, PHEV Adoption and GHG Emissions in Austin, Texas." *Transportation Research Part A: Policy and Practice* 45, no. 8 (2011): 707-20. doi:10.1016/j.tra.2011.04.011.
- Nixon, H., Saphores, J.D.. "Understanding household preferences for alternatives-fuel vehicle technologies." Report 10-11. Mineta Transportation Institute, San Jose, CA, 2011.
- Parsons, George R., Michael K. Hidrue, Willett Kempton, and Meryl P. Gardner. "Willingness to Pay for Vehicle-to-grid (V2G) Electric Vehicles and Their Contract Terms." *Energy Economics* 42 (2014): 313-24. doi:10.1016/j.eneco.2013.12.018.
- Petrin, Amil. "Quantifying the Benefits of New Products: The Case of the Minivan." *Journal of Political Economy* 110, no. 4 (2002): 705-29. doi:10.1086/340779.
- Sallee, James M., Sarah E. West, and Wei Fan. "Do Consumers Recognize the Value of Fuel Economy? Evidence from Used Car Prices and Gasoline Price Fluctuations." *Journal of Public Economics* 135 (2016): 61-73. doi:10.1016/j.jpubeco.2016.01.003.
- Segal, Robin. "Forecasting the Market for Electric Vehicles in California Using Conjoint Analysis." *The Energy Journal* 16, no. 3 (1995). doi:10.5547/issn0195-6574-ej-vol16-no3-4.
- Sexton, Steven E., and Alison L. Sexton. "Conspicuous Conservation: The Prius Halo and Willingness to Pay for Environmental Bona Fides." *Journal of Environmental Economics and Management* 67, no. 3 (2014): 303-17. doi:10.1016/j.jeem.2013.11.004.
- Shiau, Ching-Shin Norman, Jeremy J. Michalek, and Chris T. Hendrickson. "A Structural Analysis of Vehicle Design Responses to Corporate Average Fuel Economy Policy." *Transportation Research Part A: Policy and Practice* 43, no. 9-10 (2009): 814-28. doi:10.1016/j.tra.2009.08.002.
- Skeros, S., Raichur, V. "PRISM 2.0: Mixed Logit Consumer Vehicle Choice Modeling Using Revealed Preference Data." Electric Power Research Institute, 2013.
- Tanaka, Makoto, Takanori Ida, Kayo Murakami, and Lee Friedman. "Consumers' Willingness to Pay for Alternative Fuel Vehicles: A Comparative Discrete Choice Analysis between the US and Japan." *Transportation Research Part A: Policy and Practice* 70 (2014): 194-209. doi:10.1016/j.tra.2014.10.019.
- Tompkins, Melanie, David Bunch, Danilo Santini, Mark Bradley, Anant Vyas, and David Poyer. "Determinants of Alternative Fuel Vehicle Choice in the Continental United States." *Transportation Research Record: Journal of the Transportation Research Board* 1641 (1998): 130-38. doi:10.3141/1641-16.
- Train, Kenneth, and Garrett Sonnier. "Mixed Logit with Bounded Distributions of Correlated Partworths." *Applications of Simulation Methods in Environmental and Resource Economics The Economics of Non-Market Goods and Resources*: 117-34. doi:10.1007/1-4020-3684-1_7.
- Train, Kenneth, and Melvyn Weeks. "Discrete Choice Models in Preference Space and Willingness-to-Pay Space." *Applications of Simulation Methods in Environmental and Resource Economics The Economics of Non-Market Goods and Resources*: 1-16. doi:10.1007/1-4020-3684-1_1.
- Train, Kenneth E., and Clifford Winston. "Vehicle Choice Behavior And The Declining Market Share Of U.s. Automakers*." *International Economic Review* 48, no. 4 (2007): 1469-496. doi:10.1111/j.1468-2354.2007.00471.x.
- Walls, Margaret A. "Valuing the Characteristics of Natural Gas Vehicles: An Implicit Markets Approach." *The Review of Economics and Statistics* 78, no. 2 (1996): 266. doi:10.2307/2109928.
- Whitefoot, K., M. Fowlie, and S. Skeros. "Product Design Response to Industrial Policy: Evaluating Fuel Economy Standards Using an Engineering Model of Endogenous Product Design." Working Paper, University of Michigan (2011).
- Zhang, Ting, Sonja Gensler, and Rosanna Garcia. "A Study of the Diffusion of Alternative Fuel Vehicles: An Agent-Based Modeling Approach*." *Journal of Product Innovation Management* 28, no. 2 (2011): 152-68. doi:10.1111/j.1540-5885.2011.00789.x.