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PUBLIC TRANSIT BUS PROCUREMENT:

The Role of Energy Prices, Regulation and Federal Subsidies

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ABSTRACT

The U.S. public transit system represents a multi-billion dollar industry that provides essential transit services to millions of urban residents. We study the market for new transit buses. Unlike private vehicles, the fuel economy of public buses has not improved during the last thirty years. Our empirical analysis of bus fleet turnover and capital investment suggests that transit agencies: (1) do not respond to energy prices; (2) respond to environmental regulations; (3) prefer purchasing buses from local manufacturers; (4) exhibit significant brand loyalty effects; (5) favor domestically produced buses when they have access to more federal funding.

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1. INTRODUCTION

In 2011, there were nearly 70,000 public transit buses operating in the United States. Public transit riders in the U.S. traveled 56 billion passenger miles. 38% of these miles were covered by public buses. Of the total of 21.4 billion passenger miles on public buses, transit systems in the three largest cities; New York, Los Angeles and Chicago, travelled about 4, 2 and 1 billion miles, respectively. Transit agencies spend about \$2.5 billion on new buses and \$3.5 billion to maintain the existing stock.¹ In 1991, the aggregate capital and operating expenditure on public transit equaled \$36.3 billion (in 2011 dollars) and this grew to \$55 billion by 2011. These facts highlight that the public transit bus fleet is a major urban capital stock.

Both private vehicles and public buses move people within cities and the vast majority of both types of vehicles use fossil fuels as their primary energy source. Figure 1 presents a thirty-year time trend in fuel economy measured as miles per gallon for three types of vehicles.

Figure 1: Miles per gallon for cars, light trucks and buses in use

¹ Source: 2013 Public Transportation Fact Book.

² Source: Transportation Energy Data Book, Oak Ridge National Laboratory, 2013. The estimates are based on fuel

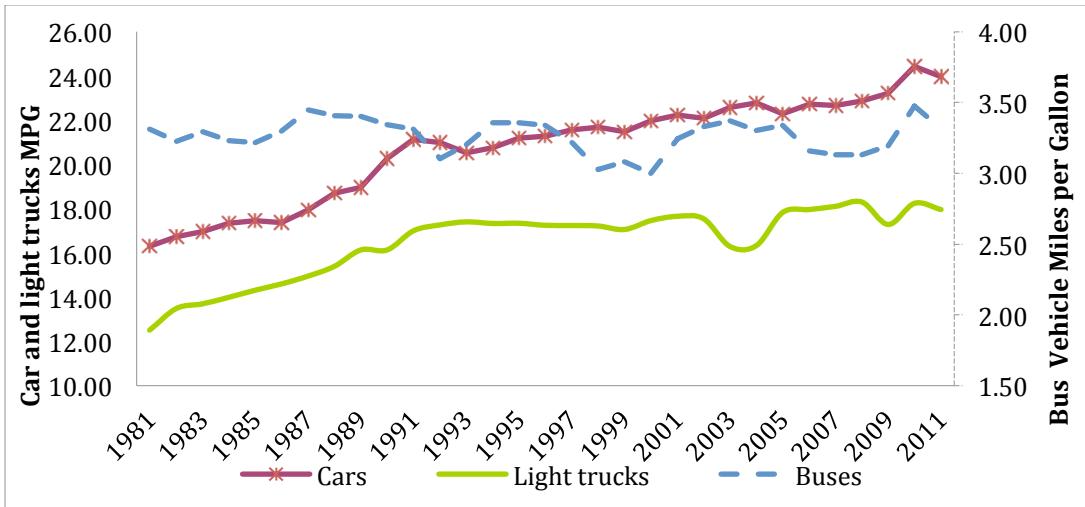


Figure 1 shows that the fuel economy of private vehicles (cars and light trucks) in use has improved by over 40% during the last thirty years but public transit bus fuel economy has been stagnant.² As we discuss in Section 2, the fleet fuel economy of buses does not appear to be responsive to fuel prices. Motivated by these findings, this paper examines transit agencies' decisions on fleet management and presents new evidence on the demand and supply of U.S. public transit buses. This is an important question for at least two reasons.

First, public transit agencies' bus purchase decisions represent distinctive market transactions featuring a non-profit (the public transit agency), who is a local monopoly in providing transit services, purchasing expensive durable capital from for-profit firms who compete in a differentiated goods market. Transit agencies often use "other people's money" (federal transfers) for capital investment. Therefore, this sector offers an opportunity to understand the revealed preference of public decision makers and we contrast their choices with those exhibited by private vehicle buyers (McFadden 1976).

² Source: Transportation Energy Data Book, Oak Ridge National Laboratory, 2013. The estimates are based on fuel consumption and vehicle miles by vehicle type. The figures for buses are miles per gasoline-equivalent gallon.

Second, we know of no recent economics research investigating bus procurement and fleet management. This research gap is notable because public buses play a central role in providing basic transportation services for a large share of the urban poor and constitute an important element of urban quality of life (Glaeser, Kahn and Rapaport 2008). Public buses also represent a viable substitute for private vehicle driving. If urban travelers substitute from using cars to riding the bus there could be significant impacts on reducing three key urban externalities: air pollution, greenhouse gas emissions from automobiles, and road congestion (Parry and Small 2005, 2009).

The market of public transit buses differs significantly from that of private vehicles. The private vehicle market features private firms selling to individuals and auto makers all over the world compete for U.S. consumers. This competition leads to a wide range of differentiated products in a rich attribute space for consumers. Consumers in this market respond to gas prices in choosing whether to scrap their existing vehicle and in choosing their new utility maximizing vehicle (Li, Timmins and von Haefen 2009). During times when gas prices are high, consumers seek out imports such as the Toyota Prius and for-profit sellers direct their product mix and innovation efforts to supply such vehicles. Such induced innovation shifts the attributes bundled into differentiated products at any point in time (Newell, Jaffe and Stavins 1999, Knittel 2012).

In the case of public buses, a large number of different buses are produced around the world and highly fuel-efficient buses are produced in China, Japan and South Korea. However, the U.S. public bus fleet is predominantly produced by domestic sellers who focus on the U.S. market and who are small in scale relative to major international bus makers. As public entities, U.S. transit agencies

must comply with the U.S. rules and regulations that we describe below.

We build a model of bus procurement and bus scrappage decisions for local transit agencies that yields testable hypotheses. A public transit agency has a very different optimization problem than private individuals or businesses as it trades off the costs of operating expenses such as vehicle maintenance; capital expenses; complying with Federal Clean Air Act regulation; and pleasing various constituents such as labor unions and funding agencies. Transit agencies are expected to guarantee a certain level of transit service to the cities they serve. In providing this service, they recognize that past investments in mechanic human capital, spare parts for specific buses and characteristics of their repair facilities all create an asset and human capital specificity that encourages them to concentrate their bus purchases on brands they have previously bought (Williamson 1988). In addition, public transit systems face mandatory “Buy America” requirements from the Federal government when availing themselves of Federal subsidies. By raising the relative “effective” price of internationally traded goods, the Buy American policy causes distortions in the allocation of resources in much the same way that a tariff does (Lowinger 1976).

We estimate a series of reduced form models of fleet inventory dynamics and new bus demand to document the role these various factors play. We examine how different agencies manage their bus inventory as they face various constraints such as the Buy America requirement and environmental regulations and recognize that maintenance costs are increasing in a vehicle’s age and cumulative mileage. Our results show: (1) transit agencies in non-attainment counties (counties designated by the EPA as having sub-standard air quality) for Ozone and PM2.5 tend to

scrap their diesel buses earlier than others and more likely to buy natural gas buses; (2) bus scrappage and purchase decisions do not respond to fuel prices; (3) transit agencies tend to buy buses of the same make and fuel type as what they already have, which we characterize as brand loyalty or lock-in effects; (4) transit agencies prefer purchasing buses from manufacturers whose plants are located in the same state; (5) the inflow of federal funding expedites the scrappage of old buses and increases purchases of domestically produced buses.

In Section 2, we present a detailed analysis of the bus fleet's fuel economy dynamics as a function of energy prices and we contrast these estimates with the private vehicle stock. This set of results yields a stark contrast between the investment choices pertaining to the public capital stock (buses) versus the private transportation capital stock (cars). Section 3 describes the bus procurement environment. Section 4 provides a model of transit authority decisions to help motivate our empirical model. In Section 5, we first discuss our data and present our empirical tests regarding bus procurement practices by transit agencies. We then discuss the implications of these findings and conclude.

2. FLEET FUEL ECONOMY OF U.S. PUBLIC TRANSIT BUS IS ENERGY PRICE INELASTIC

Ongoing research in environmental and energy economics has studied how the private vehicle fleet's composition and utilization responds to gas price dynamics. These studies have generally concluded that both new vehicle purchases and used vehicle scrappage respond to gasoline prices to various degrees (Li, Timmins, von Haefen 2009, Klier and Linn 2012, Knittel 2012, and Jacobsen and van Benthem 2013). At the intensive margin, households reduce driving where gasoline price

rises (Small and van Dender 2007, Gillingham 2013).

In this section, we examine the fuel economy trends for public buses. We focus on the comparison with cars instead of the private fleet of heavy-duty trucks such as Fed Ex or greyhound buses due to data limitations. The demand for Fed Ex and Greyhound reflects the need for cargo shipment and long-distance travel while transit buses and cars are used mostly for within-city passenger travel.³

We find very small and effectively zero energy price elasticity for public buses based on a panel data set from 1997 to 2011 from the National Transit Database (NTD). We generate the fuel economy of the bus fleet for each urbanized area (UZA) in a year based on total fuel consumption (gasoline-equivalent) and total vehicle miles travelled. This variable, gallons per mile (GPM), is “realized” fleet fuel economy determined by both the average fuel economy of all the buses and their individual utilizations.

Table 1 presents the regression results from six different specifications where the dependent variable is $\ln(\text{GPM})$ and the key explanatory variables are fuel prices. All the regressions include UZA fixed effects. All regressions include time trend and the last two only include observations with diesel accounting for more than 95% of the total energy consumption. Specifications 4 and 6 also include census division-specific time trend. In all specifications, gasoline and CNG prices do

³ The private fleet of heavy-duty trucks offers an alternative comparison group. The anecdotal evidence suggests that they respond to fuel price changes. For example, there is a recent trend of switching from diesel trucks to CNG trucks due to low natural gas prices. <http://online.wsj.com/news/articles/SB10001424052702304707604577422192910235090>

not affect the fuel economy of the bus fleet.⁴ In addition, the time trend variables are small in magnitude and not jointly statistically significant, consistent with the long-term trend depicted in Figure 1.

For-profit sellers of transit buses do not respond to rising gasoline prices by engaging in induced innovation. This finding stands in contrast with the fact that private vehicle suppliers respond to rising gasoline prices by offering more fuel-efficient vehicles (for long-run trends see Knittel 2012). In Appendix Table 2, we document the fact that there is a negative and statistically significant relationship between private vehicle's gallons per mile and gas prices. In Web Appendix 2, we explain our procedure for inferring how suppliers of new buses respond to gas prices. The regression results are reported in Appendix Table 3. There is no statistically significant association between gasoline prices and the supply of more fuel-efficient new buses.

To begin to understand why we observe these differences between the private transport capital stock and the public transport capital stock, it is relevant to note that the landscape of the U.S. bus market is dramatically different from the car market in that foreign producers are almost completely absent in the bus market while they represent more than half of the car market through both production in the U.S. and imports.

Table 2 lists the major bus manufacturers who sold buses in the U.S. between the years 1997 and 2011. Unlike the private vehicle market, buses produced by foreign bus makers account for less

⁴ Appendix Table 1 provides three additional regressions for different sub-samples and the finding of a zero energy price elasticities remains. The three subsamples focus on large UZAs, UZAs with fast population growth, and recession years, respectively.

than 1.5 percent of the market. No foreign bus makers have production facilities in the U.S. so all of the foreign buses are imported. In addition, there are no Asian exports of buses to the U.S.⁵

The bus fleets in Seoul and Tokyo are both more fuel efficient than in the U.S. The fleet fuel economy of buses in the U.S. was 3.54 miles per gallon (of gasoline-equivalent fuel) in 2011, compared with 4.74 in Tokyo that also operates a diesel-dominated fleet of about 1500 buses.⁶ In Seoul, the average fuel economy of 61 diesel buses was 5.05 and that of 7,469 CNG buses was 4.04 in 2011.⁷ While these durable buses are traded on the world market, U.S. transit agencies have chosen not to buy them. We will examine why this is the case.

3. THE BUS INDUSTRY AND TRANSIT BUS FLEET MANAGEMENT

We first discuss the industry background and then provide the institutional details that underlie transit agents' decisions of bus fleet management.

3.1 INDUSTRY BACKGROUND

In contrast with the passenger car market, foreign bus makers are by and large absent in the U.S. market. At the same time, U.S. bus makers are almost solely domestically oriented. While the Big Three (GM, Ford and Chrysler) have extensive network of sales and manufacturing plants in other countries, New Flyer, the largest bus maker in North America operates only in U.S. and Canada

⁵ U.S. transit buses are required to be wheelchair accessible. Such buses are produced and sold in Asia as well.

⁶ Tokyo Statistical Yearbook 2011: <http://www.toukei.metro.tokyo.jp/tnenkan/2011/tn11q3e004.htm>.

⁷ Based on a personal conversation with their personal at Seoul City Office.

while Gillig, the second largest, concentrates on the U.S. market.⁸ On the other hand, major international bus makers such as Daimler Buses (owner of Mercedes-Benz buses) and Volvo have a significant presence in other markets including Asia, Europe and South America. Volvo has sales in 85 countries and production facilities in Europe, North America, South America, and Asia. The largest bus maker in China, King Long, sells buses in about 80 countries and regions. Although their international sales have been mostly in the developing world, they are poised to have a larger presence in Europe.

U.S. bus makers are much smaller in scale. The top two bus makers in the U.S. (New Flyer and Gillig) sell about 1000-1500 buses each year in the U.S. where the total annual sales are 4000 to 5000 buses. Major international bus makers such as Daimler and Volvo are an order of magnitude larger than U.S. bus makers: Daimler sells 30,000 to 40,000 buses and chassis and Volvo over 10,000 on a yearly basis. The top two bus makers in Japan (Hino and Fuso) each sell over 2000 buses domestically in a year out of over 9000 total buses sold in Japan. The largest bus maker in China, King Long, sold 29,000 buses in 2012.

The U.S. bus production is concentrated in several states that are not known for automobile manufacturing. Among all the buses in service in 2011, over 24% of them were produced in California while over 23% were from Minnesota. Among the new buses produced in 2011, California produced 32% them and Minnesota 24%.⁹ Car manufacturing heavyweights such as Michigan, Ohio and Indiana only produced about 5% of them. These shares are based on where

⁸ Both GM and Ford used to produce transit buses but sold or closed their production plants. Ford still produces chassis for shuttle buses.

⁹ New Flyer has two assembly plants in Minnesota while Gillig is based in California.

the final assembly took place. Bus makers have very complex supply chains. For example, the fare system installed in New Flyer buses includes components produced by part suppliers in twelve states and the communication and the signal equipment from eighteen states.

3.2 BACKGROUND ON TRANSIT FLEET MANAGEMENT

The Agency's Objective

A transit agent needs to meet the local peak demand that is affected by many factors including demographics and population size (Bar-Yoset et al. 2013). The Federal Transportation Administration (FTA) requires that the metro agencies create a “Fleet Management Plan” that explains how the metro agency will meet peak demand, and be prepared for unexpected demand shocks. Such agencies usually have to supply a level of services that is 20% greater than peak demand. Surplus capacity allows the agency to engage in maintenance and rebuilding and preventative maintenance while still providing service to riders. In addition to this stated objective, agencies have additional goals including creating local public sector jobs, coping with political pressure to award contracts to assembly plants in the same state. In addition, there can be pressure from constituents and environmental regulators for the adoption of cleaner buses such as hybrid or electric buses.

THE BUDGET CONSTRAINT

Fleet management represents a set of interrelated decisions concerning the outflow of the old buses through scrappage and the inflow of new buses through procurement. In bus replacement decisions, the main tradeoff that the transit agencies face is between the upfront capital costs for new buses and the operating and maintenance costs for old buses. The availability of the federal

funding eases the tight budget constraint that transit agencies often face.¹⁰

The transit agency faces a budget constraint that varies over time depending on the inflow of various funding sources. Federal funding is an important source of revenue and this funding is provided to match local funds.¹¹ Federal funding for bus purchases mainly comes from two Federal Transit Administration (FTA) programs: the Urbanized Area Formula Program, and the transit capital investment program. The former is the primary federal funding source for capital costs and accounts for more than 80 percent of the federal funding. Funding is allocated based on legislative formulas: for urban areas with a population between 50,000 to 199,999, the formula is based on population and population density.¹² For larger urban areas, it is based on a combination of factors including bus vehicle miles, passenger miles, and population.

The second program, the transit capital investment program provides supplemental funds for three categories of capital projects including replacement of buses and bus facilities. Funds are allocated on a discretionary basis each year and are primarily intended to support one-time or periodic capital needs left unmet by federal formula funding or by local or State funding sources. While the first source of funding is allocated lump-sum to the transit agencies who then can decide on how to allocate them to different capital projects, the second source of funding is earmarked for designated projects in the funding application.

¹⁰ In its 2012 Annual Report, New Flyer reports: “The Company’s principal customers are municipal and other local transit authorities that rely on funding from various levels of government to purchase heavy-duty transit buses. There can be no assurance that this funding will continue to be available at current levels, on the same terms or at all. Eighty percent of the total eligible funding for purchases of new heavy-duty transit buses in the United States is provided by the federal government through allocations to the FTA.” (page 67 of 2012 Annual Report)

¹¹ The maximum level of subsidy is 80% of the capital expenditure on new buses. Rules for accessing federal funding require that a bus’s total life expectancy should be at least twelve years.

¹² See http://www.fta.dot.gov/grants/13093_3561.html

One important requirement for obtaining federal funding is to “Buy America” and requires: (1) the vehicles must undergo final assembly in the U.S., and (2) at least 60% of the vehicle's components, by cost, must be manufactured in the United States.¹³ The prospect of receiving substantial matching funds from the FTA would discourage transit agencies from purchasing foreign buses. The Buy America requirement constitutes an important entry barrier that foreign bus makers face to compete in the U.S. market.¹⁴ While foreign firms can setup a plant here (one of the five major bus makers was set up to assemble Hungarian buses), foreign firms are likely to lose their comparative advantage as they have to buy at least 60% of the materials from the U.S. and cannot rely on cheaper non-U.S. labor.

Federal transfers to transit agencies increase during recession years. This fact helps to explain why private car sales are highly pro-cyclical, but bus purchases are not. This is due to the fact that federal funding accounts for up to 80% percent of transit agencies' capital expenditure. Anderson (1983) argues that local public transit agencies are well aware of these subsidies as they pursue their goals of bureaucratic growth.

Table 3 highlights how federal funding for bus capital expenditure varies across years. The dependent variable in the four specifications measures the level or share of federal funding for capital expenditure. Controlling for transit agency fixed effects, the regression coefficients indicate that relative to the omitted category of 1992, in the recession years (especially starting in 2009),

¹³ The location of headquarters or company offices is not a factor in the FTA's Buy America's regulation.

¹⁴ The 2012 annual report New Flyers, a leading bus maker in North America explicitly mentions the Buy America as a significant entry barrier.

federal transfers for capital expenditure sharply increase. The regression coefficients suggest that during the Clinton Presidency that expenditure per-capita increased strongly and during George W. Bush's Presidency that expenditure fell.

The sharpest increase in expenditure occurs in response to the deep recession of 2008. The U.S. government uses fiscal policy to stimulate the U.S. economy in times of recession. New public bus purchases provide a specific example of how federal government expenditure translates into increased investment in the capital stock. For example the American Recovery and Reinvestment Act of 2009 provided \$6.9 billion dollars for new equipment for public transportation projects. Articles in the popular media have highlighted this point. The President of Gillig argues that the 2009 fiscal stimulus was crucial for propping up the demand during recessions and saving jobs in the industry.¹⁵

THE ROLE OF THE CLEAN AIR ACT ON BUS INVESTMENT

Environmental regulations are posited to play an important role in bus replacement and purchase decisions. There are three main types of buses in terms of fuel source: Diesel, CNG and hybrid buses. Diesel buses pollute more than other two types and are a significant contributor to local air pollution including particulate matter (PM) and ozone. Federal regulations from the Clean Air Act establish national ambient air quality standards for six pollutants including PM and ozone. State governments are responsible for meeting these standards. When counties are not in compliance

¹⁵ The President of Gillig, Denny Howard said "I've been with Gillig for 32 years, so I've been watching the transit industry for decades. ... if the stimulus hadn't come through, the market would have dropped 40% to 3,000 buses a year for last year, this year, and at least through next year. That would have devastated our employees. I've heard the talk about the stimulus not creating enough jobs, but you never read a statistic about the number of jobs *saved*. ... I'd call the Recovery Act a success." (source <http://fastlane.dot.gov/2010/04/american-busmaker-gillig-and-its-workers-a-stimulus-success-story.html#.UcRrYfmce9E>)

with the Clean Air Act, regulators will work with local governments to come up with steps to comply with the standards. Transit agencies are often under pressure to reduce pollution from diesel buses by replacing them with less-polluting buses such as newer buses or CNG and hybrid buses. States such as California have passed stringent standards to reduce PM10 emissions from buses.¹⁶ As these standards grew stricter in the late 1990s, agencies increasingly purchased CNG buses that perform better on particulates but have worse fuel economy than diesel buses. For example, Los Angeles has retired all of its diesel buses and replaced them with CNG buses.¹⁷

4. A MODEL OF TRANSIT BUS PROCUREMENT AND INVENTORY DYNAMICS

The scrappage of old buses and purchases of new buses can be integrated into a dynamic model of differentiated durable products demand. Each of the public transit systems at a point in time has a stock of buses it owns in inventory. It must choose whether to scrap them or not. In addition it chooses how many buses to buy and what types of buses to buy. The bus buyers know the price of each bus offered, the federal subsidies they can access, and the attributes bundled into each bus. The decisions are inherently dynamic due to the durability of buses. On the supply side, for-profit firms assemble buses with specific characteristics and sell them to the buyers. A resulting hedonic pricing gradient emerges and buyers sort and choose the “best bus” given their complex objective and their budget constraint defined over expenditures on labor, existing capital, new capital and energy.

¹⁶ For specifics about California’s metro bus rules for zero emissions buses see <http://www.arb.ca.gov/msprog/bus/zbus/zbus.htm>. California has long had the reputation as the nation’s leader on implementing and enforcing vehicle emissions regulations (Kahn 1996).

¹⁷ http://www.metro.net/news/simple_pr/Metro-receives-AQMDs-Clean-Air-Award-for-achieving/

The goal of this section is to shed light on the decision process of transit agencies as they decide their sequence of bus scrappage and procurement of new buses. We present a model of public transit's decision problem with respect to bus acquisitions. Before we proceed, we define our notation. Let

- S_t be the amount of bus services (measured in total miles) provided at time t
- L_t be the amount of labor employed at t
- E_t be the amount of energy consumed at t
- w_t be the wage of labor at t
- P_t be the price of energy at t
- i_t be the municipal bond interest rate at t
- I_t be the cost of new equipment/buses
- g_t be the percentage of I_t subsidized at t
- T_t the total tax revenue and income available to the transit authority at t

The agency wishes to produce services S_t and has an infinite lifetime objective function discounted by β . The state space representation of the objective function is:

$$\Phi_t = S_t + \beta \Phi_{t+1} \quad (1)$$

Define maintenance costs μ as a function of bus mileage M such that $\mu'(M) > 0$. Further define the level of services $S = S(L, A|V)$ where:

$$A = \begin{cases} a > 0, & \text{if } V(t) \neq V(t-1) \\ a = 0, & \text{if } V(t) = V(t-1), \end{cases}$$

denotes changes in the vehicle fleet. $V = \{V^*, V^{**}\}$, where V^* is the initial bus fleet and V^{**} is a new bus fleet such that the energy used by the fleet for any level of labor input L is: $\alpha(L, V^*) = \alpha(V^*)L > \alpha(V^{**})L = \alpha(L, V^{**})$.

Let $g_t = \{1 \text{ for all } V_t = V^*; \text{ and } 0 \text{ if no subsidy at } t; \text{ } g^{**} \text{ if subsidy at } t \text{ and } V_t = V^{**}\}$. Prob (subsidy at time t) = ρ_t with $\rho_t > \beta^k \rho_{t+k}$ for all $k > 0$. We first examine a world of constant prices and revenue to ease exposition: $I_t = I$, $w_t = w$, $i_t = i$, and $T_t = T$ for all t . We further assume that the expected change in energy prices follows: $P_t = P_1 > P_0$ for all $t > 0$.

A transit agent's problem at time t is:

$$\text{Max}_{\{L, V\}} E\{ \Phi_t \} = S_t(L_t, V_t | V_{t-1}) + \beta E\{ \Phi(V_{t+1}) \}, \text{ s.t.} \quad (2)$$

$$T = w L_t + P \alpha(V_t) L_t + i(1 - g_k) + A + \mu(M_t). \quad (3)$$

The budget constraint is static in nature. The left-hand side, T , includes revenue from fares and funding from local, state and federal governments. The right-hand side is total expenditure including maintenance and operating costs (labor and fuel costs) and spending on new capital projects. If $V_{t-1} = V^{**}$, then a maximum of (2) is achieved where:

$$L_t = \frac{T - i(1 - g_k) - \mu(M_t)}{w + P \alpha(V^{**})}. \quad (4)$$

where $k = t - i$ such that $V_t \neq V_{t-i}$ and $i > 1$. If however, $V_{t-1} = V^*$ then the transit agency must decide whether or not to upgrade the fleet. That decision requires the evaluation of the following two terms:

$$\Phi^{**} = \frac{\beta}{1-\beta} S_t \left(\frac{T-i(1-g_t)-\mu(M_t)}{w+P_t \alpha(V^{**})}, V^{**} \right) + S_t \left(\frac{T-i(1-g_t)-a-\mu(M_t)}{w+P_t \alpha(V^{**})}, V^{**} \right) \quad (5)$$

$$\Phi^* = S_t \left(\frac{T-\mu(M_t)}{w+P_t \alpha(V^*)}, V^* \right) + \beta \rho \Phi_{t+1}(\cdot | V_t = V^*, g^{**}) + \beta(1-\rho) \Phi_{t+1}(\cdot | V_t = V^*, 0). \quad (6)$$

If $\Phi^{**} > \Phi^*$, then $V_t = V^{**}$, else $V_t = V^*$.

Proposition 1: If $V_0 = V^*$, and at $t = 0$, $g=1$, then there exists a sequence of $\{\rho, g^*, \mu\}$ such that $\Phi^{**} < \Phi^*$, for all t.

This result implies that in the event government subsidies are sufficiently small, or the cost of changing the fleet size is sufficiently large relative to the energy savings, no purchase of new energy efficient buses will occur.

Proof: let $g^* = \varepsilon$, and $\mu = \gamma$ sufficiently small such that:

$$\Phi^{***} \approx S_t \left(\frac{T-i-a}{w+P_t \alpha(V^{**})}, V^{**} \right) / (1-\beta) + S_t \left(\frac{T-i-a}{w+P_t \alpha(V^{**})}, V^{**} \right) < \Phi^{**}$$

Let $\rho = \rho_t$ for all t . Then $\Phi^{***} < \Phi^{**} < \Phi^*$ for all t and there is a corner solution with no switching between V^* and V^{**} .

Proposition 2: If $V_t = V^*$, and no subsidy is offered at time 1, and the conditions of Proposition 1 do not hold, then there exists an $\{a, g^{**}, \rho, \beta\}$ such that $V_1 = V^*$, but $V_k = V^{**}$ for some k . The proof is provided in the Web Appendix 3. This Proposition shows that a variable subsidy, say as contracyclical macroeconomic policy, could trigger new bus purchases.

It is relevant to note that this model highlights a key option value component for delaying investment in new buses where the uncertainty is associated with the size of future federal government subsidies. This is particularly true if $\mu(M_t)$ increases rapidly enough as the existing fleet ages. If the probabilities $\{\rho_t\}$ change over time, then it would also be possible for a ρ_t sufficiently low to make the value of g^{**} negligible and the energy benefit $\alpha(V^{**})$ could, if high enough, overcome the adjustment cost A . The proof is presented in the Web Appendix.

To summarize the implications of the model we consider four discrete changes: government transfers g , gas prices P , the quality of the bus fleet μ , and the imposition of a “Buy America Mandate.”

- a. A change in government transfers: in equations (5) and (6), Φ^* is the value function for the no capital investment state is not a function of g and Φ^{**} is an increasing function of g . Since g , the level of government subsidies is unbounded, there exists a subsidy that would induce a fleet renewal decision at every point in time.
- b. A change in gas prices: a change in gas prices will enter Φ^* and Φ^{**} only through fuel usage $\alpha(V^*)L$ and $\alpha(V^{**})L$. Since $\alpha(V^*) > \alpha(V^{**})$ an increase in P will increase Φ^* more than Φ^{**} and make a fleet switch more likely. It should be noted however that as d. below argues, absent the “Buy America Mandate,” it would make a fleet switch (to foreign produced buses) even more likely.
- c. The quality of the bus fleet: as the fleet ages, maintenance costs increase. Therefore $\mu(M_t)$ by construction increases. For the fleet rollover decision this only affects Φ^*

by decreasing the resources available for bus services. Therefore an aging fleet makes a fleet swap decision more attractive.

d. The Buy America mandate: the mandate, can be analyzed in our model by considering a third alternative bus type V_f where $\alpha(V_f) < \alpha(V^{**})$ and $i_f < i$. In other words the buses are less expensive and they have better gas mileage. Then $\Phi^f < \Phi^{**} + (1 - \beta)^{-1}(ig_k)$. In other words, absent the subsidy, the bus company would choose the foreign manufacturer rather than the domestic manufacturer if a fleet change was indicated. The Buy America mandate is then binding when $\Phi^f > \Phi^{**}$ as the bus authority only receives the subsidy if they buy the domestic product. Thus the Buy America mandate acts as a tariff on foreign buses such that the fuel economy gained is outweighed by the net effective price to the purchaser.

5. EMPIRICAL ANALYSIS OF FLEET MANAGEMENT DECISIONS

In this section, we investigate the fleet management decisions of transit agencies and study various factors underlying these decisions. We first present our data. We then examine the outflow of the buses through scrappage and after that, the investment in new buses through procurement. This analysis allows us to study how the quantity and quality of public capital evolves over time.

5.1 DATA

The major data source for our analysis is the National Transit Database (NTD). This provides us with detailed transit data from transit agencies each year from 1997 to 2011. The NTD was established by Congress to gather and distribute information and statistics on the transit systems

in the country. Currently, over 660 transit agencies in urban areas, who are recipients or beneficiaries of federal transit grants, report to the NTD.

For each transit agency in each year, the NTD provides data on the total count of active buses, their model year distribution, total miles travelled in previous years by bus system/make/model year. For each bus, we also observe the bus manufacturer's name and fuel type. Table 4 present summary statistics of the bus inventory data. Since many of our explanatory variables are at the UZA level, we aggregate data from transit agencies to Urbanized Areas (UZA) which could contain multiple transit agencies. The unit of observation is a vintage-model by UZA-year. The average number of buses per vintage-model is 33 with the maximum being 763. The average scrappage rate is 6 percent and the average bus age is 8.27 with about 7 percent of them being new buses. Figure 2 presents the distribution of scrappage rate and miles traveled by age. Among all buses, 87 percent are diesel buses while 10 percent are compressed natural gas (CNG) buses. The vast majority of the buses (98 percent) are by domestic producers (or producers with plants in the U.S.). About 11 percent of the buses are produced in the same state as the buyer.

The NTD also provides total energy consumption across all of the buses in the fleet broken out into twelve categories such as gasoline, diesel, CNG, liquefied natural gas (LNG) etc. We use these energy consumption data and standard index weights to compute total gasoline-equivalent consumption for each bus system in each year.¹⁸ A separate NTD file provides information on annual operating expenditure and capital expenditure by public transit system/year. The NTD

¹⁸ Our data sources for these emissions factors are Knittel (2012) and the US Energy Information Administration - (<http://www.eia.gov/oiaf/1605/coefficients.html#tbl2>).

does not provide information on the purchase price of each bus and we could not find systematic price information anywhere, likely due to the fact that bus makers and transit agencies negotiate one to one on transactions. Instead, annual data are provided on total expenditure on the new capital stock and the count of new buses purchased of each make. While the NTD names the bus makes, it provides few details about the physical attributes of the buses except for its occupant capacity and length.

We also collect state and county demographic (state population and gross state product per capita) and environmental variables (county PM2.5 and Ozone nonattainment status).¹⁹ We collect gasoline prices at the city level from the Cost of Living Index database. Although diesel is the major fuel used by public buses, we use gasoline prices for our analysis below. We are not aware of comprehensive city-level diesel prices and state-level coverage from the Energy Information Administration is not complete as well. In addition, diesel and gasoline prices are highly correlated as both are derived from petroleum and subject to similar market forces: the correlation coefficient between gasoline price and diesel prices for nine states available from the Energy information administration (EIA) is 0.98 between 1997 and 2011. We collect annual state-level natural gas vehicle fuel prices from the EIA.

5.2 BUS TOP SCRAPPAGE

We examine how transit agencies make scrappage decisions taking into account various factors and constraints discussed above such as maintenance costs, fuel costs, federal transfers and

¹⁹ <http://epa.gov/airquality/greenbk/>

environmental regulations. Our main goal here is to understand the importance of these various factors in transit agencies' decisions and we do not attempt to estimate a dynamic model.

In contrast with the single-agent dynamic model estimated in Rust (1987), a formal dynamic stochastic model of transit agency decisions would be challenging to set up and estimate for several reasons.²⁰ First, the objective function of a transit agency is unlikely to be strictly profit optimization as such an agency may prioritize additional factors such as creating local employment and engaging in bureaucratic growth (Anderson 1983). In addition, agencies also face multiple constraints as we discussed above. Second, each transit agency owns a large number of buses with different vintage and cumulative miles and the decision should be made simultaneously over the whole fleet. Third, scrappage decisions and purchase decisions are jointly made. Purchase decisions themselves are multi-dimensional in that they are multiple bus makers and different fuel types as well as the fact they may buy from multiple makers (for different units) at the same time.

Denote j as vintage-model (e.g., 2000 Gillig Phantom) and c as UZA (city) and t as year. We estimate the following scrappage equation:

$$R_{jct} = \alpha_1 Age_{jct} + \alpha_2 Miles_{jct} + \alpha_3 NA_{ct} + \alpha_4 Fed_{ct} + X_{jct}\beta + \mu_t + \eta_{jc} + \varepsilon_{jct}, \quad (9)$$

where the dependent variable is the scrappage rate defined as the change in the number of buses of a vintage model in a fleet from year $t-1$ to year t divided by the number of buses in $t-1$. We use

²⁰ Rust (1987) studies the optimal timing of replacing a bus engine by the bus manager weighing the upfront replacement costs against future maintenance costs.

vehicle age and cumulative mileage to control for (expected) maintenance costs. Given that maintenance costs are likely to be a convex function of these variables as reflected in Figure 2, we include polynomials of age and mileage in the regression as well.

To examine if transit authorizes respond to environmental regulations by retiring their diesel buses earlier than otherwise, we include indicators of a county's (one-year lagged) non-attainment status in PM2.5 and Ozone. The dummy variable NA_{ct} is one for county c in year t if there is violation of federal standards in either PM2.5 or Ozone in year $t-1$. Regular diesel buses tend to emit more pollutants than CNG and hybrid buses and these pollutants can lead to high PM2.5 and Ozone concentrations. In addition, earlier vintages of diesel and CNG buses are more polluting than newer vintages due to improvement in emission-reduction technology and the tightening of emission standards over time.²¹ Since the regulations will mainly affect older buses in scrappage decisions if at all, we interact the non-attainment status dummy variable with a dummy variable for vehicles ten years old and above.

As we discussed above, federal funding accounts for a significant share of total capital expenditure of transit agencies and it often increases during economic recessions due to stimulus measures. Increased federal funding will relax the tight budget constraint many transit agencies face and expedite the scrappage of old buses. In one of the specifications we present below, we include Fed_{ct} , the per capita federal funding for capital expenditure (in logs).

²¹ According to the EPA, buses that meet EPA's 2010 standards emit 95 percent less pollution than pre-2007 vehicles and are 60 times cleaner than pre-1991 buses.

The regression also includes other control variables in X_{jct} such as vintages to reflect changes in production technology over time, gasoline prices to reflect changes in operating expenses, and variables that capture demographic shifts and income growth. We include population growth and area growth in the UZA to control for demand growth for bus services which may affect scrappage decisions.

We control for time-varying unobservables (μ_t) and UZA-vintage-model level (η_{jc}). In an OLS framework, we can control for both types of unobservables using fixed effects. However, given that the dependent variable ranges from zero to one, we use Tobit regressions to estimate the parameters. Due to the incidental parameters problem, we use random effects to control for unobservables η_{jc} in Tobit regressions. We add manufacturer dummies and state dummies in random effects Tobit regressions to control for unobservables at the make and state level that might be correlated with the explanatory variables.

Table 5 presents estimation results for five specifications. Given the small number of hybrid buses and the fact that 90% of them are five years old and younger, we focus on diesel and CNG buses in these regressions. The first column is from OLS where we include year fixed effects and UZA-vintage-model fixed effects as specified in equation (1). The second column presents results from a Tobit model where we include year fixed effects, manufacturer fixed effects and state fixed effects but do not control for UZA-vintage-model fixed effects. The third and fourth columns report random effects Tobit regressions which control for UZA-vintage-model unobservables using a

random effects specification. The third column does not include state fixed effects while the fourth does. The fifth column adds the federal funding variable which one might worry being endogenous as we noted above. We focus on the results from the fourth column, our preferred specification, noting the results are generally robust across specifications where unobservable factors are controlled in different ways.

Several findings emerge. First, bus age and cumulative miles are both highly statistically significant determinants of scrappage rates. Within the range of vehicle age (1 to 30) and cumulative miles (0 to 230,000 miles), the effects of age and miles on scrappage are both positive. For a 15-year old bus, the scrappage rate increases by 10 percentage points when it ages by one year. In terms of the effect of vehicle miles, the scrappage rate increases by 2 percentage points when the vehicle miles grow from 200,000 to 300,000.

Second, the positive coefficient estimate on county nonattainment status (for buses age ten and above) suggests that PM2.5 and Ozone regulations affect bus scrappage decisions. County non-attainment for either PM2.5 or Ozone leads to faster retirement of old buses. The magnitude of the effect is similar to the difference in scrappage rate between a 19-year old bus and a 20-year old bus. This finding suggests that transit agencies respond to federal regulations.

Third, gasoline prices do not affect scrappage decisions: the coefficient estimates on gasoline price variables are small in magnitude and not statistically significant. This finding is consistent with the results reported in Table 1 that gasoline prices do not affect fleet fuel economy of buses. But it

is in contrast with the finding for the scrappage of passenger cars (Li, Timmins and von Haefen 2009; Jacobsen and van Benthem 2013).²²

Lastly, the increase in federal funding is associated with earlier retirement of old buses as transit agencies take advantage of these funding to purchase new buses. We recognize the potential endogeneity of including this federal funding variable.²³ Adding the federal funding variable does not change the coefficient estimates on other variables in any meaningful way except for gross state product per capita.

5.3 BUS PROCUREMENT

In this section, we examine the factors underlying bus procurement decisions. We focus on two decision margins: how many to buy from different makes (choices over make), and how many to buy for each fuel type (choices over fuel type). The choice set of makes includes ten of the largest domestic makes, foreign make (lumping all foreign makes into one), and small makes (lumping all other makes into one). The choice set of fuel types includes three: diesel, natural gas, and hybrid. In practice, these two decisions are likely to be made jointly and can be potentially modeled

²² In the regressions, we use gasoline prices in the current year to capture expected gasoline prices following the literature, which amounts to a random walk assumption. Alternatively, we could use the gasoline prices in the next year following a perfect foresight assumption. But the qualitative conclusion does not change for this scrappage equation and the same holds for the purchase equations examined below.

²³ We argue that the correlation between this variable and the error term is likely to be small. First, as we discussed above, the majority of the federal funding comes from the formula program that allocates funds based on observed variables such as population from the census. To the extent that some of these variables affect scrappage decisions, they are controlled either directly such as the inclusion of an area's population growth or indirectly through fixed effects. Second, the other source of funding is discretionary and is allocated once applications from transit agencies are approved. This portion of funding might be more likely to be correlated with unobserved factors. Due to the endogeneity concerns, we only include the transfers variable in the last specification. We have tried using population growth and area growth as IVs for federal funding in the OLS framework (the first specification) but they prove to be only weakly correlated with federal funding.

simultaneously in a unified framework such as a discrete choice model in a product characteristics space as in Berry, Levinsohn and Pakes (1995) where make and fuel type will be characteristics of each choice (i.e., bus model). A technical challenge in estimating such a discrete choice problem is how to allow purchases of multiple units of each model as well as from multiple makes. Allowing for such possibilities would make the discrete choice problem an extremely high dimensional decision.²⁴ We examine the decision margins over bus make and fuel type separately. This allows us to look at different factors affecting the two margins in a practical and parsimonious way.²⁵ We do not jointly model the purchase decision and the utilization decision, unlike Dubin and McFadden (1984). As we discussed above, transit agencies are required to meet peak demand and in the short run, the changes in bus routes are small over time. We include population, income and state-specific time trend to control for changes in travel demand. The key results are robust to the exclusion of these variables.²⁶

²⁴ Studies such as Berry, Levinsohn and Pakes (1995, 2002) and Goldberg (1995) have formulated a demand system for cars based on a utility maximization framework over the product space. The multiple constraints and objectives discussed above make bus purchase decisions more complicated. In addition, the lack of data on bus characteristics and multi-unit demand make the framework infeasible in our context.

²⁵ As a robustness check on whether much is lost by modeling the make and fuel type decisions separately, we also estimate a multinomial logit model with random effects. The choice set includes all available combinations of 11 makes and three fuel types (such as a Gillig diesel bus). There are 20 choices in 1997 and 24 in 2011 due to the fact that not all bus makers produce hybrid buses and hybrid buses started to appear only after 2005. To generate multinomial outcomes (a vector with one element being one and the rest being zero), we assign one to the choice with the largest purchase and zero otherwise. Among 2280 purchase incidences in our data, two-thirds are single purchases in an UZA-year. While circumventing the problem of multiple choices, this simplification treats the other (smaller) purchase incidences to be non-purchase. Appendix Table 4 presents the estimation results for several different specifications. These results are qualitatively the same as those based on estimating separate equations.

²⁶ We note that the purchase decisions are not jointly modeled with the scrappage decisions. Although the correlation between these two types of decisions would be interesting to study and may affect the efficiency of the estimation, we choose not to model them jointly for parsimony and tractability, recognizing that ignoring it is unlikely to lead to bias in our estimates.

5.3.1 PURCHASE DECISIONS OVER BUS MAKE

To study choices over the demand for a bus make, we estimate the following equation.

$$S_{kct} = \alpha_1 L_{kc} + \alpha_2 F_{kt} * D_{kc} + \alpha_3 B_{kct} + X_{ct}\beta + \mu_t + \lambda_k + \eta_{ck} + \varepsilon_{kct}, \quad (10)$$

where the dependent variable, S_{kct} , is the share of new buses of make k among total new buses purchased by UZA c in year t . L_{kc} is a dummy variable being 1 for make k in UZA c if the bus manufacturer has production facility in the state where UZA c is. A positive coefficient on this variable would suggest local favoritism. We use local favoritism as a broad term and it may be due to many different reasons such as the concern for the employment in the state and better relationships between the transit agencies and bus makers due to closer proximity. To examine the effect of federal funding, we interact the share of federal funding F_{kt} with the domestic dummy variable D_{kc} which is one for domestically produced buses. Due to the potential concern regarding the endogeneity of the federal funding variable, we only include it in one of many specifications. B_{jkt} is the percent of buses from the same make among the existing buses, capturing the cost savings in spare parts stock when all buses are from the same make.

We include other control variables in X_{ct} such as the average age of the bus fleet and variables that capture a UZA's population change and income growth. μ_t captures time fixed effects while λ_k captures bus make fixed effects. η_{ck} controls for time-invariant UZA-make unobservables that capture special relationship between a UZA-make pair.

The estimation strategy is similar to our approach in estimating the scrappage equation. The dependent variable ranges from zero to one so we use Tobit regressions. In our preferred random effects Tobit model, we use random effects to capture UZA-make unobservables while including state fixed effects. The estimation results are presented in Table 6. The first column is from OLS where we include year fixed effects, make fixed effects and UZA fixed effects. Given that about 90 percent of the observations are left-censored at zero, we expect a large bias from OLS. The other three columns are all from Tobit regressions where the two-sided censoring is taken into account. The second column is from a Tobit where we have included year fixed effects, make fixed effects and state fixed effects but do not control for UZA-make unobservables. The third and fourth columns are from random effects Tobit regressions which control for UZA-make unobservables using random effects. The third column does not include state fixed effects while the fourth does. The last specification adds the interaction term between federal funding and domestic bus dummy. As expected given the heavy censoring of the data, the OLS coefficient estimates are much smaller than those from Tobit regressions. The estimates are generally the same across the four specifications using Tobit.

In Table 6, our preferred model's results are reported in column (4). These estimates yield several relevant findings. First, the positive and significant coefficient estimate on the local dummy suggests that transit agencies prefer buses produced within their own state. As we discussed above, this is likely due to considerations of the local employment effects and special relationships between transit agents and bus makers. The coefficient estimate of 0.457 on the local dummy is economically significant; the mean of the dependent variable (the share of each bus make) is 0.68 for 2,287 non-zero shares. Ideally, the identification of local favoritism (the coefficient on the

local dummy) should come from exogenous shocks that re-locate a manufacturing plant from one state to another. However, we do not observe plant relocation during our data period. Therefore, the local dummy variable is not time-varying for a given UZA-make pair.²⁷

Second, the coefficient estimate of 0.426 on the interaction between pro-union state and domestic bus dummy implies that pro-union states have a preference for domestic buses (relative to other states).²⁸ The estimation results (marginally insignificant) provide some evidence that decisions of transit agencies are subject to union influences. Protecting domestic employment is one key argument used for the Buy American mandate. Labor is a significant component of bus assembly as well as manufacturing at the part suppliers. According to the 2011 annual report of New Flyer, labor at the assembly plants accounts for 9% of the bus assembly while materials account for 71%. Paul Saubry, the CEO of New Flyer suggests that the bus assembly creates one to two jobs per bus while one bus is associated with four to five jobs for bus parts suppliers.

Third, transit agencies prefer to purchase buses of the same make they purchased before. A typical public transit agency has built up both physical capital and human capital that is bus brand specific. They have invested in customizing their maintenance facilities to work with a specific type of bus and their mechanics have been trained to work on such buses. In addition, they have

²⁷ We cannot include UZA-make fixed effects in the model to control for time-invariant unobservables (other than the local relationship). Instead, we use random effects at the UZA-make level in the Tobit model to control for the unobservables. The difference between columns (2) and (4) is that column 2 does not include random effects at the UZA-make level but column (4) does. The coefficient estimates across these two specifications are generally quite close and this gives us some confidence that the results are not driven by UZA-make unobservables (such as transit agencies in the same states like the size or color of the buses made in the state) that are not related to the broad definition of local favoritism.

²⁸ We define those states that are not Right to Work states to be “pro-union”.

invested in a spare parts inventory that is customized for specific buses.²⁹ These factors are likely to cause a lock-in effect such that bus buyers tend to buy from the same bus sellers. The bus companies also explicitly seek to build relationships with transit agency officials in order to generate repeat business.³⁰ Lastly, based on the final specification, transit agencies favor domestically produced buses especially when they have access to more federal funding. This is likely due to the Buy America mandate as we discussed above. We posit that this mandate could be an important reason behind the lack of international competition and high prices of American made buses.

We note that we have controlled for UZA-make random effects in the last three specifications in order to control for unobserved factors that contribute to the likelihood of a repeat purchase of the same make by a transit agency. The second specification does not control for them. Although the coefficient estimates on the percent of the old buses from the same make becomes smaller in the last three specifications as expected if unobservables exist, they are still significant and large in magnitude.³¹

5.3.2 PURCHASE DECISIONS OVER FUEL TYPE

We now turn to the decisions over the fuel type of the buses and we focus on three types: diesel,

²⁹ For example, the diesel bus has no spark ignition system but CNG vehicles do. The issue of fleet standardization often dictates purchase decisions of airlines. For example, Southwest airlines operate only Boeing 737s. As the largest operator of 737 worldwide, they are able to press Boeing to produce more fuel-efficient models to better meet their needs.

³⁰ In its 2012 Annual report, New Flyer reports; “Ongoing dialogue with existing customers, enhanced by a service or support relationship, assists in identifying opportunities for new business. Over the last five years, approximately 90% of buses produced by the Company were for customers who had purchased buses from the Company in the preceding five years.”

³¹ The (fleet-size) weighted average of the percent of old buses from the same make is 0.30.

CNG and hybrid. Buses of different fuel type differ in upfront purchases costs, fuel costs during operation and with respect to their emissions levels. A standard 40-foot transit bus costs approximately \$300,000 while a CNG bus costs about \$30,000 more. A hybrid bus costs between \$450,000 and \$550,000. Among the three types, hybrid buses have the best fuel efficiency (25-50% better than diesel buses) while CNG buses have the lowest (about 25% lower than diesel buses). In addition, CNG buses require expensive special re-fueling infrastructure and maintenance facilities compared to diesel buses, which are the traditional transit bus choices. In terms of environmental performance, hybrid buses rank the best while diesel buses the worst, especially diesel buses manufactured before 2007. Emissions standard for diesel engines in buses have tightened quite aggressively over time and the difference in emission levels between buses of different fuel type have narrowed.

This section examines how transit agencies consider these tradeoffs by estimating the following equation:

$$S_{fct} = NA_{ct}F_f\alpha + FP_{ct}F_f\gamma + X_{fct}\beta + \mu_t + \lambda_f + \eta_{cf} + \varepsilon_{fct}, \quad (11)$$

where the dependent variable, S_{fct} , is the share of new buses of fuel type f among total new buses purchased by UZA c in year t . NA_{ct} is equal to one if the jurisdiction is a non-attainment area in either PM2.5 or Ozone during the last year. F_f is a vector of three dummies variables corresponding to three fuel types. The interactions of non-attainment status and fuel type dummies examine if fuel type choices are affected by environmental regulations. FP_{ct} is defined as fuel (gasoline) prices and it interacts with fuel type dummies, F_f , if (expected) fuel expenses are taken into account in purchase decisions. X_{fct} is a vector including all other control variables.

Similar to equation (10), we control for UZA-invariant unobservables over time μ_t , fuel type fixed effects λ_f and time-invariant UZA-fuel type unobservables η_{cf} that capture special preferences a UZA may have for buses of fuel type f (e.g., a UZA may prefer CNG buses because they have easy access to natural gas pipelines or re-fueling stations).

The dependent variable ranges from zero to one so our estimation strategy follows those for equation (10) closely as well. We use a random effects Tobit regression as the preferred specification where we control for year fixed effects, fuel type fixed effects, state fixed effects, and random effects at the UZA-fuel type level. Table 7 presents regression results for five specifications. The first column is from OLS and the other four columns are from Tobit regressions. The second column does not control for UZA-fuel type unobservables while the last three columns do. Column (3) does not control for state fixed effect and columns (4) and (5) do. In column (5), we add the interaction term between federal funding and hybrid bus dummy.

The estimation results are very similar in columns (3) to (5) and we focus our discussion on those from columns (4). First, non-attainment status in PM2.5 or Ozone leads to less demand for diesel buses and more for CNG buses. The coefficient estimate on the interaction between non-attainment status and diesel bus is -0.576 while that on the interaction between non-attainment status and CNG bus is 0.579. Both are large in magnitude: the average share by fuel type is 0.89 among 1315 non-zero observations and it is 0.43 among all 3479 observations. The coefficient estimate on the interaction between non-attainment status and hybrid dummy is not statistically

significant which could partly be a reflection of the fact that hybrid buses account for only five percent of the new buses and their purchase decision could be subject to more idiosyncratic factors that we do not observe.

Second, fuel prices have no statistically significant effects on fuel type decisions although the signs of the coefficient estimates are intuitive except that on the interaction between gasoline price and CNG bus dummy. This result is consistent with the finding that scrappage decisions are not affected by fuel prices. Both of them confirm our motivating fact that fuel efficiency of the bus fleet is not responsive to fuel prices.

Third, based on the last specification, demand for hybrid buses is positively affected by federal funding. Hybrid buses are at least 50 percent more expensive than diesel counterparts and the upfront purchase costs could be a hurdle for their adoption especially under tight financial conditions. Additional capital funding relaxes the constraint and leads to more purchases of this type of bus.

6. DISCUSSION

The bus market can be characterized as a differentiated product oligopoly. Buses are differentiated among many dimensions such as floor height, fuel type, seating capacity and size. The bus market is more concentrated than the auto market with the top five makers accounting for over 75% of the U.S. market. In addition, as we documented, the fleet standardization is an important consideration in bus purchases. On average a transit agency purchases 52.5% of its

buses from its largest bus supplier and 24% of its buses from its second largest bus supplier based on bus purchased from 1997 to 2011. For comparison, the top two U.S. bus makers, Gillig commanded 25% of the market share while New Flyer 22% during this same period. These comparisons highlight the lock-in effect, which would increase the bargaining power of the bus suppliers. Union work rules may also re-enforce these lock-in effects as bus drivers may not want to learn how to drive many different types of buses. In addition, the lock-in effects could also reduce transit agencies demand responses to fuel prices because of limited varieties offered by each bus maker. For example, Gillig and New Flyer each only have two product lines currently available, with each product line offering two or three choices of length.

Federal funding is an important source of capital expenditure and can cover up to 80% of the capital costs of buses. The Buy America mandate combined with the federal funding essentially plays the role of a subsidy for purchasing domestic buses. This restriction on access to imports, similar to import quotas, could have important implications on the industry by reducing incentives for firms to innovate and to improve efficiency (Krueger 1974, Lenway et al. 1996). The domestic bus makers stand to gain at the expense of social welfare due to the misallocation of resources.³² The Buy American mandate requires 60 percent minimum domestic content and U.S. final assembly. So although foreign bus makers can set up bus assembly plants as automobile manufacturers do in the U.S., the requirements could greatly reduce their advantage for example

³² Lenway et al. (1996) find empirical evidence on that trade protectionism leads to more lobbying efforts among less productive and innovative firms and higher return to those firms in the American Steel industry. Based on data from www.opensecrets.org, none of the domestic bus makers are contributing to Political Action Committee campaigns. “Buy American” applies to broad purchases of using federal funding and domestic bus buyers are just small stakeholders in the broad context.

in lower labor, economies of scale, or material costs.³³ The lack of competition from abroad may cause inefficiency in production due to the lack of an incentive to innovate. Even if domestic bus buyers demand more fuel-efficient vehicles, if their total demand for such vehicles is low this may not provide sufficient revenue for domestic suppliers to cover the fixed costs of developing new varieties (Acemoglu and Linn 2004).

The subsidy on domestic buses and the lack of international competition imply that U.S. tax payers face a higher price for urban bus services and U.S. owners of the domestic firms that produce the buses gain some monopoly rents. Based on NTD data from 1997 to 2011, the average price for a U.S. public transit bus (in year 2011 dollars) was \$309,000 with the 10th percentile of the empirical distribution being \$104,000 and the 90th percentile at \$497,000. While it is difficult to construct a hedonic bus price regression where we control for key public transit bus characteristics including quality, our research suggests that measured in comparable units, buses in Tokyo and Seoul are half the price of U.S. buses and buses produced in China are even cheaper.³⁴ While cynics might question the quality of China's buses, it is notable that wealthy and well governed Singapore is importing buses from China.³⁵ In addition, different transit agencies

³³ North American Bus Industries (NABI) was originally setup in 1992 to assemble unfinished buses from a sister company in Hungary as a way to allow compliance with the Buy America mandate. In early 2013, NABI discontinued the practice and relegate all manufacturing and final assembly activities in the US. The company has since been acquired by New Flyer Industries.

³⁴ A regular 40-feet diesel bus costs about \$300,000 in the U.S. while a hybrid bus costs about \$450,000 to \$550,000. According to the 2011 annual report of New Flyer, the largest bus maker in North America, the average selling price per bus is about \$450,000 where the composition of buses sold are 44% diesel, 43% hybrid and 13% CNG buses. According to Volvo's 2011 annual report, Volvo sold 12,786 buses with a total revenue of 22.3 billion SEK, implying an average price of about \$272,000 per bus.

³⁵ Transportation costs of shipping international buses cannot explain why the United States imports cars but not buses. Transit buses weigh from 25,000 to 40,000 pounds while a passenger vehicle typically weighs from 2,000 to 5,000 pounds. Given the price of a typical transit bus is about 10 times of a passenger vehicle, the value per unit of weight is not necessarily smaller for buses.

may have different preferences or need for quality (such as air conditioning) and face different level of budget tightness, therefore, one should not simply rule out Chinese buses as preferred buses by all transit agencies.

High domestic costs for buses leads to fewer buses being purchased due to capital constraints and this hinders the economics of scale that is vital for determining the quality of public transit such as service frequency and coverage (Morhing 1972). Although transit agencies are required to meet demand, demand is endogenous to the level of service. Higher costs could lead to fewer buses purchased and a lower level of service, which would depress demand and hinder scale economies.

When the price premium of making energy efficient buses in the U.S. is sufficiently large, one can expect that a foreign manufacturer may open a U.S. factory. In 2013, BYD, a major producer of electric cars and buses in China, opened a manufacturing site in California and signed contracts with the cities of Los Angeles and Long Beach to sell over 100 electric buses to their respective public transit agencies. The electric buses sell at about \$800,000 per unit, more than twice as expensive as a diesel bus. In 2014, it was announced that this project was suspended because the federal government withdrew its grant money.³⁶

7. CONCLUSION

Public buses are a major component of the public transit system and represent a multi-billion dollar investment. Motivated by the findings that the fleet fuel economy of public buses has not

³⁶ http://www.greencarreports.com/news/1091186_byds-long-beach-ca-electric-bus-deal-on-indefinite-hold-now

improved over time and that it does not respond to fuel price, this paper has used several data sets to study the demand and supply for transit buses over the years from 1997 to 2011.

Our analysis suggests that the stagnancy in the fuel economy of U.S. bus fleet and its non-responsiveness to fuel prices reflect several features of the bus market. On the demand side, non-profit transit agencies purchase expensive buses relying extensively on federal subsidies. These subsidies are provided with “strings attached” known as the Buy American mandate. In addition, their purchase decisions are influenced by other considerations including environmental regulations, local employment, and commonality with their existing bus fleet. On the supply side, a handful of domestic bus makers account for the vast majority of the market share and imports are nearly non-existent. The domestic bus makers supply a small number of differentiated bus models. The lack of competition could retard incentives to develop more fuel-efficient buses. We argue that demand-side constraints dampen the effect of fuel prices on fleet fuel economy while the nature of the supply side re-enforces the non-responsiveness.

The Buy America mandate serves as an important entry barrier to the U.S. bus market. In the resulting hedonic equilibrium, U.S. public transit agencies pay more for buses than they would have if there had been free international trade in buses. Such bus sellers can raise the price of their buses above world prices because the binding Buy America subsidy means that bus buyers have very little incentive to consider cheaper international exports. The nature of a differentiated product oligopoly of the bus market exacerbates the positive effect of subsidies on bus prices. High

bus prices in turn imply that the quality of bus services (e.g., coverage across space and time) and the demand for public transit could be negatively affected.

Private vehicle usage contributes to local air pollution, greenhouse gas production and to the urban traffic congestion (Anderson 2013, Parry and Small 2005, 2009, Meyer, Kain and Wohl 1965). Due to economies of scale, buses could produce lower levels of local air pollution and greenhouse gas emissions per passenger mile. Improvements in the quality of public transit would reduce all three of those externalities because many private vehicle users would substitute to the cheaper public transit alternative. While buses offer social benefits over cars, the slow speed of urban buses and the waiting times lead many richer households to seek the convenience of private vehicles (Glaeser, Kahn and Rappaport 2008). If U.S. cities offered the quality of the public bus fleet that other cities such as Singapore offer, then voter opposition to introducing road pricing (and hence increasing urban traffic speeds) would be likely to decline (Harsman and Quigley 2010).

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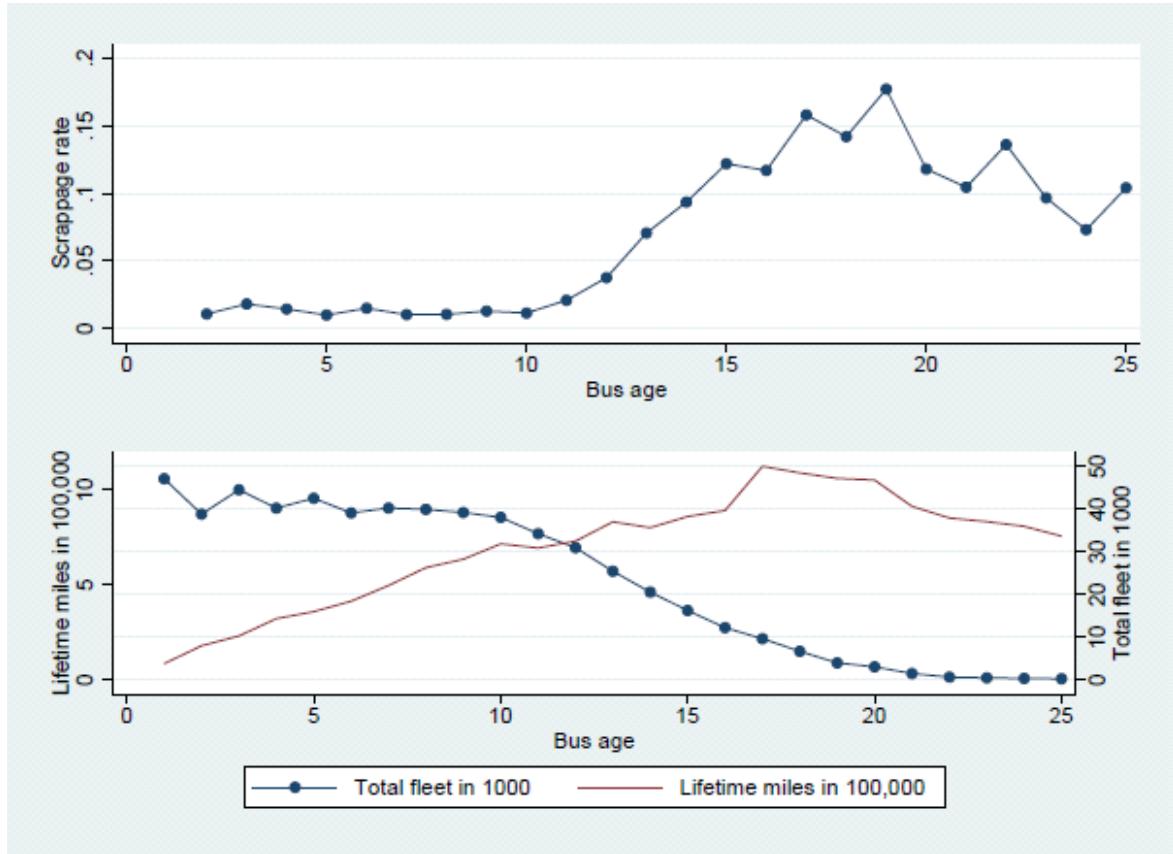
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Figure 2: Bus Scrappage Rate, Lifetime VMT and Fleet by Age



Notes: The plots are based on NTD data from 1997 to 2011. The top panel is the average (year-to-year) scrappage rate by bus age. The bottom panel shows the fleet size and average cumulative miles by age.

Table 1: Impacts of Fuel Prices on Gallons Per Mile (Total Fuel Consumption/Total VMT)

	(1) Ln(GPM)	(2) Ln(GPM)	(3) Ln(GPM)	(4) Ln(GPM)	(5) Ln(GPM)	(6) Ln(GPM)
Ln(gasoline price)	0.006 (0.015)	0.002 (0.015)	-0.002 (0.015)		-0.004 (0.013)	
Ln(CNG price)	0.005 (0.011)	0.007 (0.011)	0.008 (0.011)			
Ln(gas price:3-year average)				0.004 (0.032)		0.027 (0.028)
Ln(CNG price:3-year average)				0.003 (0.017)		
Ln(No. of passengers/mile)		0.044** (0.019)	0.045** (0.019)	0.045** (0.019)	0.022* (0.013)	0.021 (0.013)
Average bus age		0.012 (0.008)	0.013 (0.008)	0.013 (0.008)	0.016** (0.007)	0.016** (0.007)
Average bus age squared		-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)	-0.000 (0.000)
Time trend	-0.023 (0.015)	-0.017 (0.015)	-0.007 (0.032)	-0.003 (0.034)	-0.042 (0.031)	-0.030 (0.032)
Time trend squared	-0.006 (0.007)	-0.007 (0.007)	-0.006 (0.007)	-0.008 (0.009)	0.009 (0.006)	0.002 (0.008)
Constant	3.282*** (0.022)	3.108*** (0.050)	3.102*** (0.049)	3.108*** (0.053)	3.142*** (0.040)	3.133*** (0.041)
Observations	2748	2748	2748	2748	2190	2190

Notes: The dependent variable is Ln(gallons per mile). Gallons per mile is defined as total gasoline consumption/total vehicle miles. The unit of observation is a UZA-year. All regressions include UZA fixed effects. Regressions 3-6 also includes census division specific time trend. The robust standard errors are clustered at the UZA level. Standard errors are in parentheses. *: $p < 0.10$, **: $p < 0.05$, ***: $p < 0.01$.

Table 2: New Buses by Make and Type 1997 to 2011

Make	No. of new buses	Share (%)
Gillig Corporation	8,363	24.81
New Flyer of America	7,710	22.87
Orion Bus Industries Ltd.	3,772	11.19
NOVA Bus Corporation	2,740	8.13
Motor Coach Industries International	2,690	7.98
North American Bus Industries (NABI)	2,565	7.61
ElDorado National	1,444	4.28
Ford Motor Corporation	647	1.92
Neoplan - USA Corporation	426	1.26
Blue Bird Corporation	412	1.22
Small makes	2,439	7.24
Foreign	503	1.49
Total	33,711	100.00

Fuel type	No. of new buses	Share (%)
Diesel	26,363	78.20
CNG	3,051	9.05
Hybrid	1,736	5.15
Other	2,561	7.60
Total	33,711	100.00

Notes: Some of the bus makers merged during the data period but we keep them separate in this table. For example, NABI and Blue Bird merged together in 2006-2007. In 2013, New Flyer acquired NABI but the NABI brand is kept.

Table 3: Federal Funding for the Public Transit Capital Stock over the Business Cycle

	(1) Log(federal l funding)	(2) Federal funding share	(3) Log(federal funding per capita)	(4) Log(total funding per capita)
year=1993	0.001 (0.125)	-0.001 (0.013)	-0.001 (0.125)	-0.010 (0.112)
year=1994	-0.066 (0.124)	0.004 (0.013)	-0.061 (0.124)	-0.058 (0.115)
year=1995	0.184 (0.134)	-0.000 (0.014)	0.190 (0.134)	0.179 (0.126)
year=1996	0.351*** (0.128)	0.022 (0.015)	0.358*** (0.128)	0.315*** (0.118)
year=1997	0.706*** (0.126)	0.038*** (0.013)	0.712*** (0.126)	0.598*** (0.118)
year=1998	0.694*** (0.132)	0.041*** (0.013)	0.699*** (0.131)	0.582*** (0.124)
year=1999	0.681*** (0.135)	0.021 (0.015)	0.684*** (0.135)	0.611*** (0.121)
year=2000	0.651*** (0.124)	0.025* (0.015)	0.656*** (0.124)	0.573*** (0.116)
year=2001	0.839*** (0.127)	0.027* (0.015)	0.654*** (0.126)	0.584*** (0.116)
year=2002	0.798*** (0.132)	0.018 (0.017)	0.613*** (0.132)	0.597*** (0.118)
year=2003	0.816*** (0.122)	0.020 (0.016)	0.631*** (0.122)	0.569*** (0.111)
year=2004	0.845*** (0.126)	0.024 (0.016)	0.661*** (0.126)	0.589*** (0.116)
year=2005	0.830*** (0.131)	0.023 (0.017)	0.644*** (0.130)	0.568*** (0.118)
year=2006	0.712*** (0.125)	0.024 (0.017)	0.525*** (0.125)	0.481*** (0.115)
year=2007	0.736*** (0.126)	0.030* (0.016)	0.550*** (0.125)	0.489*** (0.118)
year=2008	0.598*** (0.129)	0.010 (0.017)	0.412*** (0.128)	0.401*** (0.120)
year=2009	1.104*** (0.131)	0.034** (0.017)	0.918*** (0.131)	0.883*** (0.117)
year=2010	1.537*** (0.124)	0.105*** (0.016)	1.351*** (0.124)	1.181*** (0.115)
year=2011	1.339*** (0.128)	0.093*** (0.017)	1.023*** (0.128)	0.873*** (0.113)
Constant	12.633*** (0.098)	0.690*** (0.011)	0.357*** (0.097)	0.834*** (0.089)
Observations	5206	5206	5206	5206

Notes: All regressions include UZA fixed effects. 1992 is the omitted category. Standard errors

in parenthesis are clustered at the UZA level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Table 4: Summary Statistics of the Bus Inventory Data

Variable	Mean	Std. Dev.	Min	Max
Number of buses	32.57	56.22	1.00	763.00
Annual scrappage rate	0.06	0.16	0.00	0.99
Bus age	8.27	5.29	1.00	25.00
New bus dummy	0.07	0.25	0.00	1.00
Miles per bus (100,000 miles)	5.03	5.74	0.00	49.95
Diesel bus	0.87	0.33	0.00	1.00
CNG bus	0.10	0.30	0.00	1.00
Hybrid bus	0.01	0.11	0.00	1.00
Other bus	0.01	0.11	0.00	1.00
Domestically produced	0.98	0.14	0.00	1.00
Locally produced	0.11	0.31	0.00	1.00
Pro-union state	0.69	0.46	0.00	1.00
Share of the same make	0.38	0.29	0.00	1.00
Share of the same fuel type	0.86	0.25	0.00	1.00
Federal funding per capita	12.73	13.41	0.00	97.02
Total funding per capita	27.18	35.44	0.00	344.40
PM2.5 nonattainment	0.16	0.35	0.00	1.00
Ozone nonattainment	0.31	0.45	0.00	1.00
Population (in 1000000)	2.90	4.49	0.05	18.35
Gasoline price (in 2011 \$)	2.42	0.71	1.27	4.05

Notes: The unit of observation is a vintage-model in a UZA in a year. The number of observations is 19,325.

Table 5: The Determinants of Public Transit Bus Scrappage

	(1) OLS	(2) Tobit	(3) R.E. Tobit	(4) R.E. Tobit	(5) R.E. Tobit
Average bus age	0.0454*** (0.0088)	0.1766*** (0.0237)	0.1769*** (0.0239)	0.1802*** (0.0238)	0.1791*** (0.0236)
Average age squared	-0.0001 (0.0003)	-0.0027*** (0.0006)	-0.0026*** (0.0006)	-0.0026*** (0.0006)	-0.0025*** (0.0006)
Lifetime mileage	-0.0118*** (0.0037)	0.0285*** (0.0041)	0.0220*** (0.0047)	0.0234*** (0.0047)	0.0228*** (0.0046)
Lifetime mileage squared	0.0002* (0.0001)	-0.0006*** (0.0002)	-0.0004** (0.0002)	-0.0005*** (0.0002)	-0.0005*** (0.0002)
PM or Ozone nonattainment*old bus	0.0176 (0.0110)	0.0874*** (0.0254)	0.0639** (0.0274)	0.0824*** (0.0285)	0.0744*** (0.0281)
Population growth*old bus	0.0712 (0.0522)	0.0239 (0.1550)	-0.1483 (0.1595)	0.1856 (0.1775)	0.0819 (0.1780)
Area growth*old bus	-0.0749* (0.0428)	-0.0453 (0.1275)	0.0296 (0.1346)	-0.1558 (0.1481)	0.0215 (0.1483)
Ln(gas price)*old diesel bus	0.0050 (0.0170)	-0.0269 (0.0570)	-0.0222 (0.0580)	-0.0094 (0.0584)	-0.0039 (0.0578)
Ln(gas price)*old CNG bus	-0.0095 (0.0084)	-0.0220 (0.0321)	-0.0182 (0.0333)	-0.0204 (0.0332)	-0.0228 (0.0327)
Ln(GSP per capita)*old bus	-0.0038** (0.0016)	-0.0029 (0.0056)	-0.0026 (0.0057)	-0.0052 (0.0057)	-0.0228*** (0.0060)
Ln(federal funding/pop)*old bus					0.0772*** (0.0087)
Constant	-0.0118*** (0.0037)	0.0285*** (0.0041)	0.0220*** (0.0047)	0.0234*** (0.0047)	0.0228*** (0.0046)
Year fixed effects	Yes	Yes	Yes	Yes	Yes
Manufacturer fixed effects	No	Yes	Yes	Yes	Yes
State fixed effects	No	Yes	No	Yes	Yes
UZA-vintage-model fixed effects	Yes	No	No	No	No
S.D. of random effects			0.1956*** (0.0109)	0.1702*** (0.0111)	0.1596*** (0.0112)
S.D. of error term		0.4260*** (0.0075)	0.3975*** (0.0073)	0.3976*** (0.0073)	0.3957*** (0.0073)

Notes: The dependent variable is the scrappage rate by vintage-model by UZA and year. The number of observations is 10,314. All the regressions include dummies variables for three vintages (pre-1980 buses, 1980 to 1990, 1990 to 2000) and CNG bus dummy. PM or Ozone nonattainment captures PM or Ozone attainment status of the counties in the UZA. Columns (3) and (5) use random effects to control for UZA-make unobservables. Standard errors are in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Table 6: The Determinants of New Bus Purchases

	(1) OLS	(2) Tobit	(3) R.E. Tobit	(4) R.E. Tobit	(5) R.E. Tobit
Local dummy	-0.016 (0.024)	0.264** (0.114)	0.503*** (0.182)	0.457** (0.194)	0.449** (0.193)
Pro-union state*Domestic bus	0.017 (0.011)	0.275 (0.192)	0.094 (0.079)	0.426 (0.277)	0.400 (0.275)
% of old buses from the same make	0.515*** (0.025)	3.378*** (0.138)	1.812*** (0.163)	1.844*** (0.164)	1.866*** (0.165)
Ln(federal funding/pop)*Domestic bus					0.008*** (0.003)
Constant	0.066*** (0.012)	-2.164*** (0.299)	-1.676*** (0.144)	-2.342*** (0.427)	-2.360*** (0.425)
Year fixed effects	Yes	Yes	Yes	Yes	Yes
Manufacturer fixed effects	Yes	Yes	Yes	Yes	Yes
State fixed effects	No	Yes	No	Yes	Yes
UZA fixed effects	Yes	No	No	No	No
S.D. of random effects			1.114*** (0.053)	1.105*** (0.053)	1.096*** (0.053)
S.D. of error term		1.757*** (0.046)	1.479*** (0.040)	1.481*** (0.040)	1.483*** (0.040)

Notes: The dependent variable is the share of new buses by make among all new buses purchased by a UZA in a year. The number of observations is 15,230. Columns (3) and (5) use random effects to control for UZA-make unobservables. Standard errors are in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Table 7: The Determinants of New Bus Fuel Type

	(1) OLS	(2) Tobit	(3) R.E. Tobit	(4) R.E. Tobit	(5) R.E. Tobit
PM or Ozone nonat.*diesel bus	-0.043 (0.030)	-0.355*** (0.136)	-0.535*** (0.172)	-0.576*** (0.183)	-0.551*** (0.181)
PM or Ozone nonat.*CNG bus	0.033 (0.028)	0.305** (0.152)	0.620*** (0.204)	0.579*** (0.215)	0.567*** (0.212)
PM or Ozone nonat.*hybrid bus	0.039 (0.040)	0.206 (0.214)	0.243 (0.262)	0.178 (0.278)	0.114 (0.274)
Ln(gas price)*diesel bus	-0.009 (0.184)	-0.587 (1.321)	-0.505 (0.784)	-0.520 (1.228)	-0.566 (1.232)
Ln(gas price)*CNG bus	0.082 (0.183)	0.536 (1.335)	-0.244 (0.815)	-0.254 (1.259)	-0.169 (1.263)
Ln(gas price)*hybrid bus	0.053 (0.225)	0.368 (1.666)	0.829 (1.205)	0.800 (1.588)	0.891 (1.597)
Ln(CNG price)*diesel bus	0.031 (0.023)	0.285* (0.170)	0.212 (0.162)	0.239 (0.167)	0.264 (0.169)
Ln(CNG price)*CNG bus	-0.021 (0.019)	-0.437** (0.201)	-0.257 (0.205)	-0.235 (0.211)	-0.267 (0.213)
Ln(CNG price)*hybrid bus	-0.009 (0.036)	0.242 (0.264)	0.252 (0.276)	0.335 (0.297)	0.256 (0.299)
% of buses of the same fuel type	0.899*** (0.037)	4.943*** (0.278)	3.467*** (0.353)	3.482*** (0.354)	3.570*** (0.351)
% democratic votes*hybrid bus	0.194 (0.261)	1.629 (1.833)	1.591 (2.109)	1.823 (2.377)	1.180 (2.335)
Ln(GSP)*hybrid bus	0.066 (0.118)	0.393 (0.938)	0.265 (1.076)	0.092 (1.238)	-0.262 (1.226)
Ln(federal funding/pop)*hybrid					0.406*** (0.126)
Constant	-0.058 (0.097)	-2.835*** (0.693)	-1.133** (0.557)	-1.261 (0.778)	-1.380* (0.771)
Year fixed effects	Yes	Yes	Yes	Yes	Yes
Fuel type fixed effects	Yes	Yes	Yes	Yes	Yes
State fixed effects	No	Yes	No	Yes	Yes
UZA fixed effects	Yes	No	No	No	No
S.D. of random effects			1.005*** (0.096)	1.001*** (0.095)	0.951*** (0.093)
S.D. of error term		1.489*** (0.071)	1.244*** (0.062)	1.244*** (0.062)	1.250*** (0.062)

Note: The dependent variable is the share of new buses by fuel type (diesel, CNG or hybrid) among new buses purchased in a UZA in a year. The number of observations is 3,479. Columns (3) and (5) use random effects to control for UZA-fuel type unobservables. Standard errors are in parentheses. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

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Web Appendix 1

Appendix Table 1: Impacts of Fuel Prices on Gallons Per Mile (GPM = Total Fuel Consumption/Total VMT)

	(1) Ln(GPM) All data	(2) Ln(GPM) Pop>300k	(3) Ln(GPM) Growing Cities	(4) Ln(GPM) Recession
Ln(gasoline price)	-0.002 (0.015)	0.014 (0.026)	0.003 (0.018)	-0.019 (0.022)
Ln(CNG price)	0.008 (0.011)	-0.010 (0.015)	0.007 (0.013)	0.030 (0.023)
Ln(No. of passenger/mile)	0.045** (0.019)	0.058 (0.038)	0.077*** (0.028)	0.031 (0.036)
Average bus age	0.013 (0.008)	0.015 (0.016)	0.006 (0.011)	0.009 (0.011)
Average bus age squared	-0.000 (0.000)	-0.001 (0.001)	-0.000 (0.001)	-0.000 (0.001)
Time trend	-0.007 (0.032)	0.001 (0.034)	0.000 (0.046)	-0.247 (0.242)
Time trend squared	-0.006 (0.007)	-0.009 (0.012)	-0.009 (0.008)	0.047 (0.053)
Constant	3.102*** (0.049)	3.176*** (0.106)	3.089*** (0.067)	3.331*** (0.263)
Observations	2748	1250	2087	926

Note: The dependent variable is Ln(gallons per mile). The unit of observation is UZA-year. All regressions include UZA fixed effects and census division specific time trend. Column (1) is the same as column (3) in Table 3. Column (2) focuses on UZAs with population over 300,000. Column (3) focuses on UZAs with a 10-year population growth rate over 5%, accounting for three quarters of all UZAs in the data. Column (4) only includes observations in 2001, 2002, 2008, 2009, and 2010 (recession years). The robust standard errors are clustered at the UZA level *: $p < 0.10$, **: $p < 0.05$, ***: $p < 0.01$.

Web Appendix 2

The Supply Response of Buses to Fuel Prices

We examine how bus producers respond to fuel price variations by changing the fuel economy of new buses sold. We contrast this with the supply side of the passenger car market following the strategy of Knittel (2012). While he examines the tradeoff of fuel economy and other characteristics in the new car market through the production frontier, our focus here is on how fuel prices affect fuel economy. For the passenger car market, we supplement Knittel's data with gasoline prices and examine how gasoline prices affect the tradeoff.

Let j index a vehicle model and v index vintage (year-built), our econometric analysis starts with the specification of how fuel price prices affect fuel economy of a vintage-model:

$$GPM_{jv} = \alpha_j + \beta FP_v + Z_{jv}\delta + \gamma v + \varepsilon_{jv}, \quad (1)$$

where GPM_{jv} is gallons per mile of model j and vintage v . α_j captures the mean fuel economy of model j over time and controls for average characteristics of the buses. FP_v is fuel price in year v . Z_{jv} is a vector of vehicle characteristics such as weight and horsepower. γv controls for time trend of technology progress.

For passenger cars, the results from four specifications are presented in Appendix Table 2 below. The first two use gasoline prices averaged over the past three years while the last two use the past five years. The parameter estimates on gasoline price variables are quite stable across four specifications. The first specification shows that a 10% increase in gasoline prices would lead to about a 0.0035 decrease in GPM, implying that MPG would go up from 22 to 23.8.

For buses, since we do not have information on bus characteristics, we cannot directly examine how fuel price changes affect the fuel economy of the new buses produced using the above framework. Instead, we make the inference based on the UZA-year fuel consumption data and information on the bus fleet, in particular, mileage by vintage-make. Fuel consumption of UZA c at year t is given by $Y_{ct} \equiv \sum_j \sum_v m_{ctjv} GPM_{jv}$, where m_{ctjv} is the total mileage of buses of model j and vintage v in UZA c at year t . Substituting for GPM_{jv} using equation (1), we take the following equation to data:

$$\begin{aligned}
 Y_{ct} &= \sum_j \sum_v m_{ctjv} (\alpha_j + \beta FP_v + Z_{jv}\delta + \gamma v + \varepsilon_{jv}) \\
 &= \sum_j \alpha_j \sum_v m_{ctjv} + \beta \sum_j \sum_v m_{ctjv} FP_v + \sum_j \sum_v m_{ctjv} Z_{jv} \delta + \gamma \sum_j \sum_v m_{ctjv} v + e_{ct} \\
 &= \sum_j \alpha_j M_{ct} + \beta MP_{ct} + \gamma MV_{ct} + \zeta_{ct}.
 \end{aligned} \tag{2}$$

The variables M_{ct} (total mileage), MP_{ct} (summation of mileage*gasoline price), MV_{ct} (summation of mileage*time trend) are the regressors and can be constructed based on our data. Similar to the regressions on private cars, we construct lagged 3-year average and 5-year average gasoline prices and use them to interact with vintage-make specific mileage data.

ζ_{ct} , the new composite error term, includes the interactions of mileage with other bus characteristics, which we do not observe for buses. Parameter β captures the effect of gasoline prices of bus fuel economy. If a higher gas price leads to firms to produce more fuel-efficient buses, β should be negative. In addition, to the extent that unobserved bus characteristics such as vehicle weight and horsepower are positively correlated with GPM and negatively correlated with gasoline prices, the estimate of β from the above regression would have a downward bias.

This methodology yields estimates of the supply side response of *new* buses based on the fuel economy of *all* buses (which is associated with observed total energy consumption) and fuel prices at the time when the buses were produced. Appendix Table 3 shows that the coefficient estimates on gasoline prices for bus fuel economy are neither

robust in sign nor statistically significant across specifications. This is in contrast with the finding for the passenger vehicle market.

Appendix Table 2: Fuel Economy of New Cars as a Function of Gas Prices 1980-2006

	(1) GPM	(2) GPM	(3) GPM	(4) GPM
Ln(gas price:3-year average)	-0.0035*** (0.0010)	-0.0048*** (0.0008)		
Ln(gas price:5-year average)			-0.0030** (0.0012)	-0.0039*** (0.0008)
Time trend (ln(year-1979))	- 0.0012*** (0.0003)	-0.0005 (0.0004)	-0.0007*** (0.0002)	-0.0002 (0.0003)
Time trend squared		-0.0029*** (0.0003)		-0.0015 (0.0009)
Constant	-0.0153 (0.0169)	-0.3301*** (0.0885)	-0.0027 (0.0132)	-0.1626* (0.0903)
Observations	27046	27046	27046	27046

Notes: The dependent variable is gallons per mile of each vintage-model from 1980 to 2006. All data except gasoline prices are from Knittel (2012). All regressions are from OLS and include manufacturer fixed effects. The control variables (not shown in the regressions) also include ln(curb weight), ln(horsepower), ln(torque), manual dummy, diesel dummy, turbocharge dummy and supercharged dummy. The standard errors are clustered at the manufacturer level. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

Appendix Table 3: Fuel Economy of New Buses as a Function of Gas Prices 1997-2011

	(1) GPM	(2) GPM	(3) GPM	(4) GPM
Ln(gas price:3-year average)	0.0223 (0.0418)	0.0060 (0.0438)		
Ln(gas price:5-year average)			0.0002 (0.0551)	-0.0341 (0.0640)
Time trend (ln(year-1979))	-0.0393** (0.0177)	-0.0691 (0.0751)	-0.0321 (0.0203)	-0.0858 (0.0734)
Time trend squared		0.0082 (0.0188)		0.0151 (0.0190)
Constant	1.0542*** (0.1643)	1.0611*** (0.1667)	1.0622*** (0.1647)	1.0666*** (0.1624)
Observations	3063	3063	3063	3063

Notes: The unit of observation is UZA-year. The coefficient estimates corresponds to the coefficients in equation (1) and are estimated based on equation (2). All regressions include manufacturer fixe effects, UZA fixed effects, year fixed effects, and census division specific time trend. The robust standard errors in parenthesis are clustered at the city level. *: $p < 0.10$, **: $p < 0.05$, ***: $p < 0.01$.

Appendix 3: Proof of Proposition 2

Proof: first, suppose there is no subsidy at time t . Let $g^{**} \cong 1$ and $a \cong \zeta$ sufficiently small. At time t ,

$$S_t \left(\frac{T - i(1 - g_t) - a - \mu(M_t)}{w + P_t \alpha(V^{**})}, V^{**} \right) \cong S_t \left(\frac{T - i - \zeta - \mu(M_t)}{w + P_t \alpha(V^{**})}, V^{**} \right)$$

and with a subsidy at $t+1$,

$$S_{t+1} \left(\frac{T - i(1 - g_{t+1}) - a - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) \cong S_t \left(\frac{T - \zeta - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right).$$

$$\begin{aligned} \Phi^* &= S_t \left(\frac{T - \mu(M_t)}{w + P_t \alpha(V^*)}, V^* \right) + \beta \rho \Phi_{t+1}(\cdot | V_t = V^*, g^{**}) + \beta(1 - \rho) \Phi_{t+1}(\cdot | V_t = V^*, 0) \\ &= S_t \left(\frac{T - \mu(M_t)}{w + P_t \alpha(V^*)}, V^* \right) + \beta \rho S_t \left(\frac{T - \zeta - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) \\ &\quad + \beta^2 \rho S_t \left(\frac{T - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) / (1 - \beta), \\ \Phi^{**} &= \beta S_t + S_t \left(\frac{T - i - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) / (1 - \beta) + S_t \left(\frac{T - i - \zeta - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right). \end{aligned}$$

Subtracting

$$\begin{aligned} \Phi^* - \Phi^{**} &= S_t \left(\frac{T - \mu(M_t)}{w + P_t \alpha(V^*)}, V^* \right) + \left[\beta \rho S_t \left(\frac{T - \zeta - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) - S_t \left(\frac{T - i - \zeta - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) \right] \\ &\quad + \frac{\beta}{1 - \beta} \left[\beta \rho S_t \left(\frac{T - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) - S_{t+1} \left(\frac{T - i - \mu(M_t)}{w + P_{t+1} \alpha(V^{**})}, V^{**} \right) \right]. \end{aligned}$$

The first term is positive, and for $\beta \rho$ close to 1, each of the terms in brackets is positive. Therefore $V_t = V^*$.

Appendix Table 4: Multinomial Logit Regressions of Bus Purchase Decisions

Variables	MNL		R.E. MNL		MNL		R.E. MNL	
	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.	Coef.	S.E.
Ln(pop)	-35.521	14.924	-41.801	17.692	-26.142	15.915	-31.799	18.991
Ln(area)	0.645	0.158	0.913	0.303	0.334	0.166	0.438	0.297
Ln(Gross state product per capita)	0.034	0.160	-0.003	0.312	0.297	0.169	0.342	0.293
Average bus age	-0.367	0.336	-0.062	0.650	-0.606	0.347	-0.162	0.660
Ln(mileage per bus)	0.036	0.015	0.064	0.024	0.020	0.016	0.050	0.024
Local bus dummy	-0.060	0.040	0.040	0.058	0.031	0.040	0.103	0.050
Pro-union state*Domestic bus	0.898	0.198	1.120	0.237	0.889	0.200	1.096	0.243
% of old buses from the same make	1.443	0.809	1.967	1.642	1.209	0.842	1.327	1.686
PM or Ozone nonat*diesel bus	2.306	0.112	2.307	0.143	2.341	0.113	2.341	0.147
PM or Ozone nona*cng bus	-0.036	0.099	-0.133	0.172	-0.102	0.103	-0.145	0.176
PM or Ozone nona*hybrid bus	0.415	0.200	0.520	0.297	0.383	0.202	0.489	0.299
Ln(gas price)*diesel bus	0.230	0.330	0.163	0.385	0.199	0.340	0.191	0.417
Ln(gas price)*cng bus	0.145	0.833	-0.621	1.324	-0.453	0.873	-1.112	1.371
Ln(gas price)*hybrid bus	0.792	0.999	-0.248	1.679	0.164	1.031	-0.741	1.735
Ln(cng price)*diesel bus	0.276	1.181	-0.531	1.623	-0.180	1.216	-0.875	1.667
Ln(cng price)*cng bus	-0.078	0.116	-0.237	0.141	-0.100	0.121	-0.251	0.148
Ln(cng price)*hybrid bus	-0.485	0.309	-0.558	0.459	-0.497	0.312	-0.575	0.479
% of old buses from the same fuel type	-0.554	0.418	-0.686	0.530	-0.638	0.441	-0.777	0.561
% democratic votes*hybrid bus	2.811	0.185	2.984	0.282	2.893	0.188	3.054	0.276
Ln(GSP)*hybrid bus	0.359	2.623	0.199	2.705	0.789	2.610	0.713	2.682
Capital funding per capita	1.486	1.493	1.384	1.649	0.918	1.568	0.795	1.720
Federal funding *domestic dummy	1.886	1.047	1.741	1.440	1.862	1.063	1.730	1.494
Capital funding *hybrid dummy	3.110	1.752	3.017	2.394	3.152	1.773	3.040	2.453
R.E. at UZA-make level	No		0.780	0.070	No		0.786	0.072
R.E. at UZA-fuel type level	No		0.381	0.128	No		0.337	0.145
R.E. at UZA level	No		0.104	0.312	No		0.079	0.339
Log-likelihood	-4800.7		-4724.3		-4692.4		-4620.1	

Notes: No. of observations: 2907. The choice set consists of all available make-fuel type combinations (e.g., a Gillig diesel bus). The 1st and 3rd specifications are multinomial logit regressions while the 2nd and 4th incorporate random effects at various levels. All regressions controls for year, fuel type, and make fixed effects as well as fuel-type specific time trends.