

Comparing Proposed E-Bike Rebates in DC to the ZEV Subsidy in Maryland: Calculations Appendix

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The calculations in the January 25, 2023, article in Greater Greater Washington are based on the linear economic models created by Bigazzi and Berjisian (2021) to model e-bike demand response to subsidy. Given that these models were based on zero emission vehicle (ZEV) demand models, I used them to model both e-bike and ZEV demand from the two subsidies.

Parameters:

| Parameter | Description | Value | Notes | Source |
|-------------------|-------------------------------|-------------|-------------------------------------------------------------------------------------------------------------------------------|-----------------------------------|
| ϵ_{bike} | Elasticity of bikes | -2.0 | | Bigazzi and Berjisian, 2021 |
| ϵ_{car} | Elasticity of cars | -1.3 | Based on the Jenn et al.'s finding that a single \$1,000 subsidy generates 2.6% additional demand and that a ZEV is \$50,000. | Jenn et al., 2018 |
| r | Rebate value | Varies | See below for values | |
| n_r | Number of rebates (DC only) | 1,500 | This rebate cap is only for DC, which has 1500 \$400 rebates and 1500 \$1200 rebates. | Legislation |
| p | Base price | | Varies. See below for values. | |
| d_b | Baseline demand | | Varies. See below for values | |
| S_U | US sales | 880,000 | It is difficult to get national sales figures; this is for 2021. | Hurford, 2022 |
| P_L | Local population | | | US Census |
| P_U | US population | 331,893,760 | | US Census |
| M_L | Local bike-to-work mode share | 0.021 | | US Census |
| M_U | US bike-to-work mode share | 0.004 | | US Census |

Functions:

| | Demand-limited | Rebate-limited |
|--------------------------------------|---------------------------------------------|-----------------------------------------|
| Total demand w rebate, d_r | $d_b \left(1 - \frac{\epsilon r}{p}\right)$ | |
| Add'l demand with rebate | $-d_b \epsilon \frac{r}{p}$ | |
| Add'l sales | $-d_b \epsilon \frac{r}{p}$ | $\frac{n_r \epsilon r}{\epsilon r - p}$ |
| Portion of rebates to new purchasers | $\frac{\epsilon r}{\epsilon r - p}$ | |
| Base demand, d_b | $S_U \frac{P_L M_L}{P_U M_U}$ | |

Maryland subsidy scheme input values:

| Item | ZEVs | PIHs | Source |
|----------------------------------------|----------|----------|--------------------------------------|
| Estimated Base Price p | \$50,000 | \$45,000 | Newton, 2022 |
| Rebate Value r | \$3,000 | \$2,000 | Legislation |
| Monthly Demand Growth Rate g | 0.035 | 0.023 | MVA |
| Proportion of Rebates to New Buyers | 7.2% | 5.5% | $\frac{\epsilon r}{\epsilon r - p}$ |
| Gasoline Miles Eliminated per Purchase | 100% | 56.3% | Department of Energy |
| Per Capita VMT | 9,966 | | MVA |

Maryland subsidy scheme results:

Values after November 2022 are estimates. Fractional demand indicates a probability of sale.

| Month | Zero-Emission and Electric Vehicles (ZEVs) | | | Plug-In Hybrid Vehicles (PIHs) | | | Total Spent | Total Add'l VMT Saved |
|---------|--------------------------------------------|-----------|--------------|--------------------------------|-----------|--------------|-------------|-----------------------|
| | Total ZEVs (baseline) | ZEV d_b | Add'l Demand | Total PIHs (baseline) | PIH d_b | Add'l Demand | | |
| 10/2022 | 37,214 | 1,048 | | 20,133 | 347 | | | |
| 11/2022 | 39,517 | 2,303 | | 20,777 | 644 | | | |
| 12/2022 | 40,894 | 1,377 | | 21,254 | 477 | | | |
| 1/2023 | 42,318 | 1,425 | 111.2 | 21,742 | 488 | 28.2 | \$5,639,738 | 3,103,699 |
| 2/2023 | 43,792 | 1,474 | 56.5 | 22,242 | 499 | 14.2 | \$2,860,262 | |
| | Total | | 167.7 | | | 42.4 | \$8,500,000 | |

Note the drop-off from January to February. This is because the subsidy scheme is capped at \$8.5 million, most of which is spent in the first month of the program. To finish the money, I estimated how many vehicles might be sold daily in February by dividing the monthly values by 28. I then determined approximately how much subsidy would be spent each day and divided the remaining value of the subsidy by that amount. This estimates the subsidy will last an additional 14 days in February before running out, and these are roughly the number of vehicles that would be stimulated to be sold on those days.

To determine the VMT savings, I multiplied the number of vehicle sales induced, the gasoline miles saved, and the average per capita annual miles travelled for both the ZEV and the PIH sales:

$$3,103,699 = (167.7 \times 1.00 \times 9966) + (42.4 \times 0.563 \times 9966)$$

District of Columbia subsidy scheme results and input values:

| Item | High-income value | Low-income value | District-Wide | Source |
|-------------------------------------|-------------------|------------------|---------------|-------------------------------------|
| Estimated Number of Residents P_L | 386,329 | 283,721 | 670,050 | US Census |
| Estimated Base Demand d_b | 5,378 | 3,949 | 9,327 | $d_b = S_U \frac{P_L M_L}{P_U M_U}$ |

| | | | | |
|----------------------------------------|-------------|-------------|-------------|---------------------------------------------|
| Estimated Base Price p | \$2,000 | \$1,600 | n/a | Estimate |
| Rebate Value r | \$400 | \$1,200 | n/a | Legislation |
| Total Demand with Rebate d_r | 7,529 | 9,874 | 17,402 | $d_b \left(1 - \frac{\epsilon r}{p}\right)$ |
| Additional Demand with Rebate | 2,151 | 5,924 | 8,075 | $-d_b \epsilon \frac{r}{p}$ |
| Additional Sales s_r | 429 | 900 | 1,329 | $\frac{n_r \epsilon r}{\epsilon r - p}$ |
| Proportion of Rebates to New Buyers | 28.6% | 60.0% | 49.9% | $\frac{\epsilon r}{\epsilon r - p}$ |
| Total Cost | \$2,322,800 | \$5,818,800 | \$8,141,837 | $r(d_b + s_r)$ |
| Gasoline Miles Eliminated per Purchase | 37% | | | Soöderberg et. al, 2020 |
| Per Capita VMT | 5,341 | | | FHWA |
| VMT Saved | 4,730,600 | | | $\left(\frac{1329 \times 5341}{3}\right)$ |

While the bike-to-work value is assumed to be equal for both high- and low-income residents, I estimated the number of residents above and below 80% of the median income in DC by using Census household data. This allowed me to break out the baseline demand for the District and get more accurate results.

Base price for an e-bike varies by income for two reasons. First, given that higher-income households are more likely to be able to purchase more expensive e-bikes, \$2,000 is the median price of a relatively cheap e-bike. Second, the low-income household subsidy is capped at 75% of the total value of the e-bike, which implies a base price of \$1,600.

The Maryland alternative values, wherein the state funded e-bike rebates instead of ZEV and PIH subsidies, is the same, though I had to tweak the number of rebates offered to fit it under the budget:

| Item | High-income value | Low-income value | District-Wide | Source |
|--------------------------------------|-------------------|------------------|---------------|---------------------------------------------|
| Estimated Number of Residents P_L | 2,599,219 | 3,565,910 | 6,165,129 | US Census |
| Local Bike-to-Work Percentage, M_L | 0.2% | | | US Census |
| Estimated Base Demand d_b | 5,378 | 3,949 | 9,327 | $d_b = S_U \frac{P_L M_L}{P_U M_U}$ |
| Estimated Base Price p | \$2,000 | \$1,600 | n/a | Estimate |
| Rebate Value r | \$400 | \$1,200 | n/a | Legislation |
| Total Demand with Rebate d_r | 4,824 | 11,819 | 16,643 | $d_b \left(1 - \frac{\epsilon r}{p}\right)$ |
| Additional Demand with Rebate | 1,378 | 7,091 | 8,469 | $-d_b \epsilon \frac{r}{p}$ |
| Additional Sales s_r | 496 | 1,042 | 1,538 | $\frac{n_r \epsilon r}{\epsilon r - p}$ |
| Proportion of Rebates to New Buyers | 28.6% | 60.0% | 49.9% | $\frac{\epsilon r}{\epsilon r - p}$ |
| Total Cost | \$1,576,740 | \$6,922,822 | \$8,499,562 | $r(d_b + s_r)$ |

| | | |
|----------------------------------------|-----------|-------------------------------------------|
| Gasoline Miles Eliminated per Purchase | 37% | Soöderberg et. al, 2020 |
| Per Capita VMT | 5,341 | FHWA |
| VMT Saved | 4,730,600 | $\left(\frac{1329 \times 5341}{3}\right)$ |