

Supplementary Materials

Widespread Range Suitability and Cost Competitiveness of Electric Vehicles for Ride-hailing Drivers

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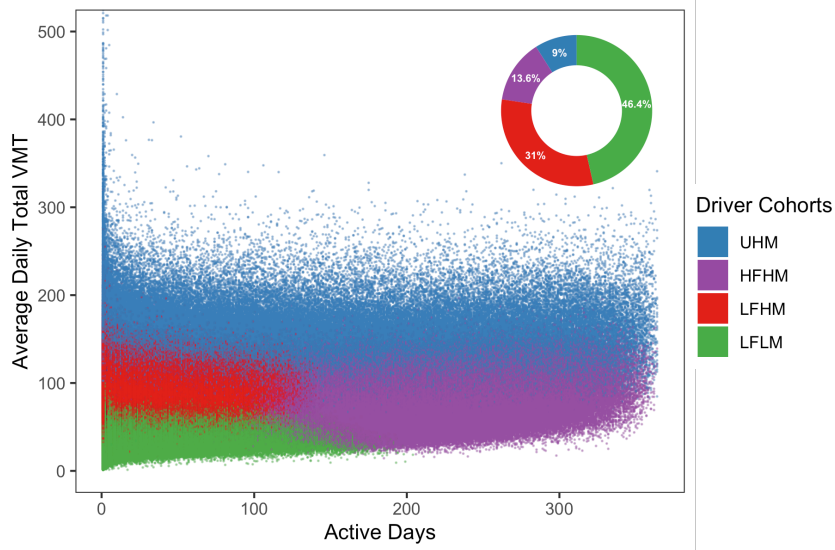


Fig. S-1. Near optimal and externally valid driver cohorts. *UHM*: Ultra High Mileage; *HFHM*: High Frequency High Mileage; *LFHM*: Low Frequency High Mileage; *LFLM*: Low Frequency Low Mileage.

Table S-1. Summary statistics of variables in the dataset for all drivers and by cohort.

	<i>All</i>	<i>UHM</i>	<i>HFHM</i>	<i>LFHM</i>	<i>LFLM</i>
Number of Active Days in 2019	59	124	190	36	24
Average Active-Day Number of Rides	5.4	12.4	5.6	6.4	3.2
Average Active-Day Observed VMT on Platform	70	145	77	86	43
Average Active-Day Occupied VMT	25	58	26	30	14
90th-percentile VMT	123	234	137	152	77
95th-percentile VMT	139	261	160	173	88
99th-percentile VMT	165	309	208	201	101
Average Active-Day Shift Duration (hr)	3.54	7.04	4.21	4.21	2.22
Observed Annual VMT on Platform	5,112	17,782	14,887	3,095	1,132
Total Annual VMT*	12,412	25,082	22,187	10,395	8,432

* Derived variable: annual observed VMT by Lyft plus 7,300 miles of personal miles.

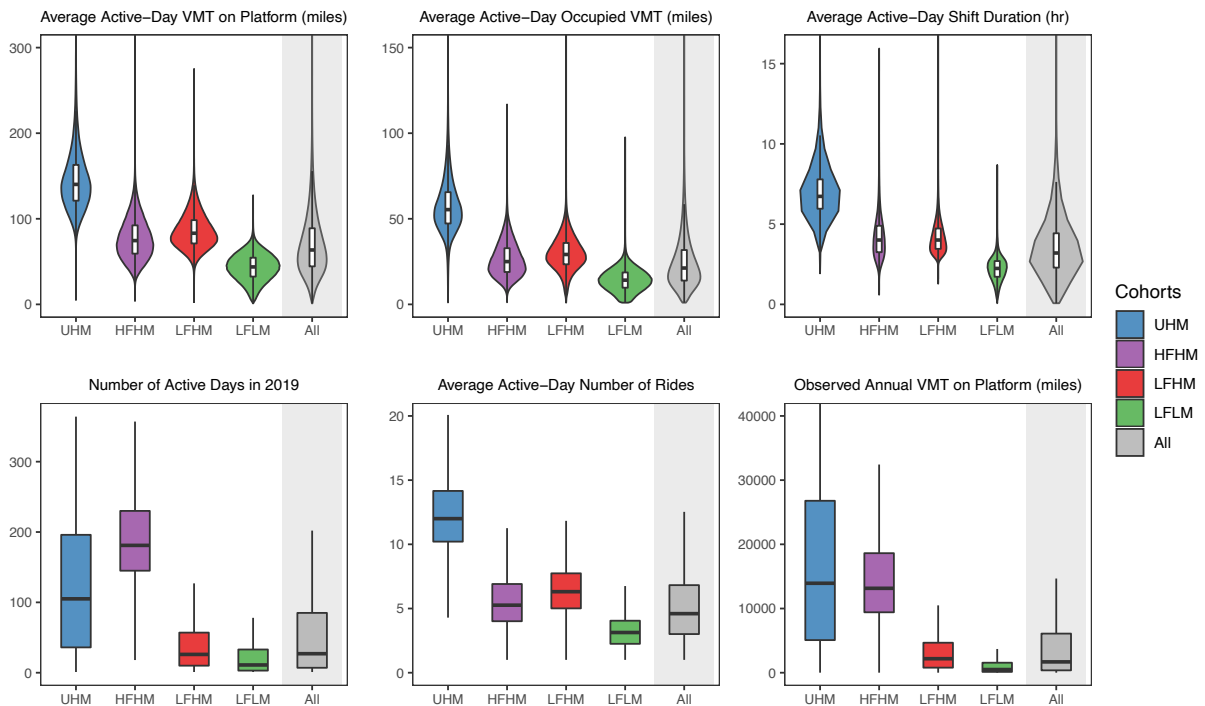


Fig. S-2. Distributions of selected variables in the dataset for all drivers and by cohort. The gray shade represents the distribution of variable for all drivers, regardless of cohort.

Supplementary Note 1: Driver clustering.

Based on computational efficiency and superior clustering power, we choose k-mean clustering, which minimizes within-cluster variances (squared Euclidean distances) of the aforementioned variables. Finally, we use several verification methods for checking the optimality of clusters (Fig. S-3). Both Elbow method and Silhouette width method suggest only two optimal clusters on selected variables and then marginal decrease in optimality with higher number of clusters (Fig. S-4). We use expert knowledge on average characteristics of resultant clusters to choose the near-optimal yet externally valid set of driver clusters. While the analysis is conducted at the individual driver level, some results are also reported on the cohort basis to provide a roadmap for identifying the ideal cohort of drivers for electrification efforts. Note that these cohorts based on the clustering method are not absolute, and drivers on the boundary of cohorts have travel patterns similar to those of either cohort.

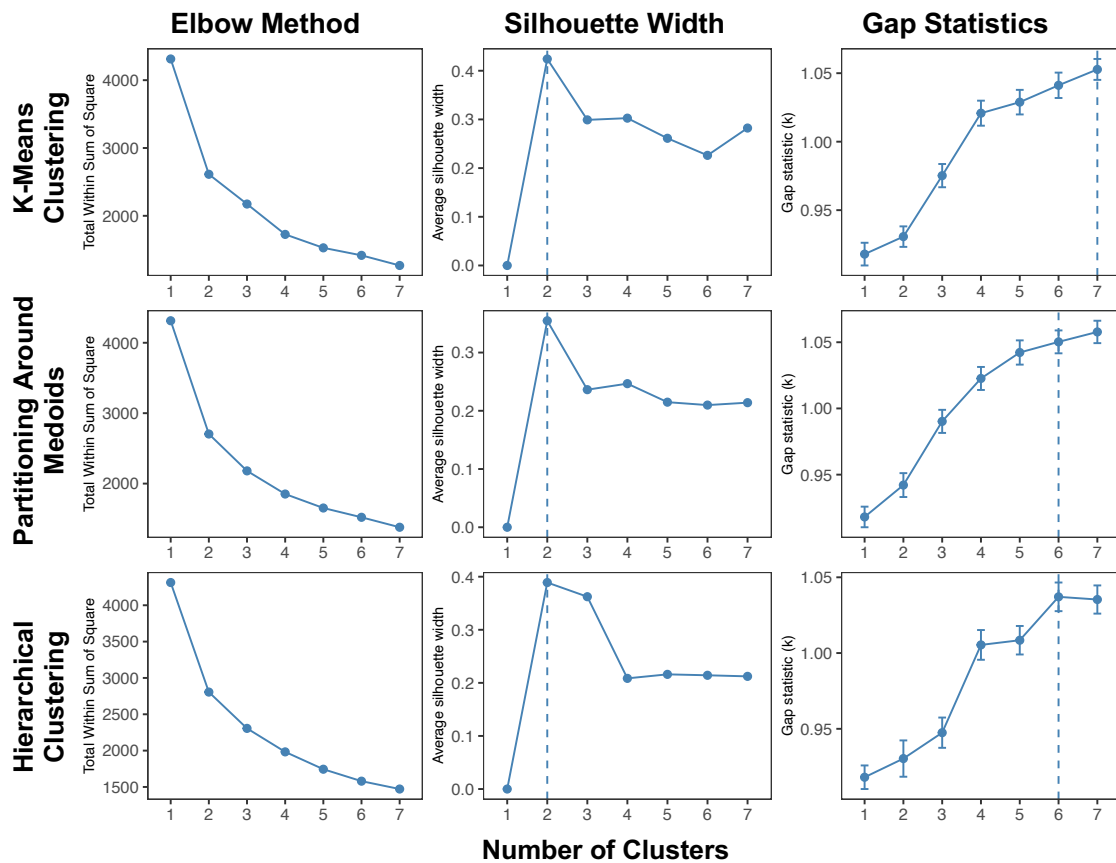


Fig. S-3. The performance of other unsupervised machine learning methods tested for defining the driver cohorts. Both Elbow method and Silhouette Width result in only two optimal clusters for all three algorithms. We use expert knowledge to choose four clusters as externally valid cohort without significantly losing the cluster optimality.

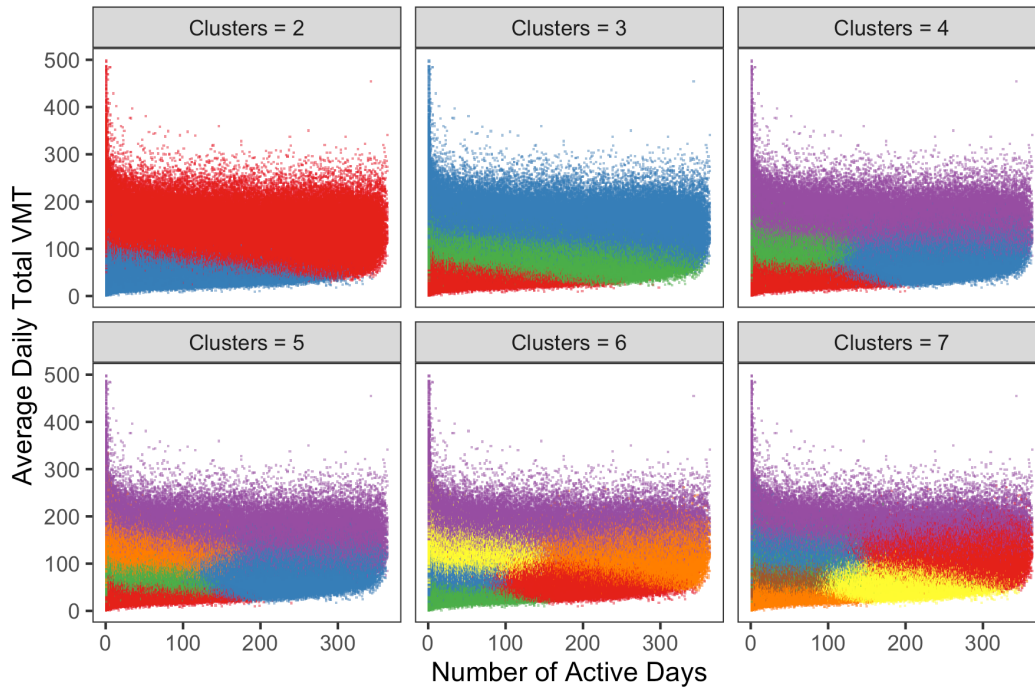


Fig. S-4. Results of different number of clusters on K-means clustering of cohorts on two variables. 4-cluster appears to have more external validity than others. The Greater number of clusters than 4 makes further cuts on low frequency low mileage drivers and does not improve the external validity.

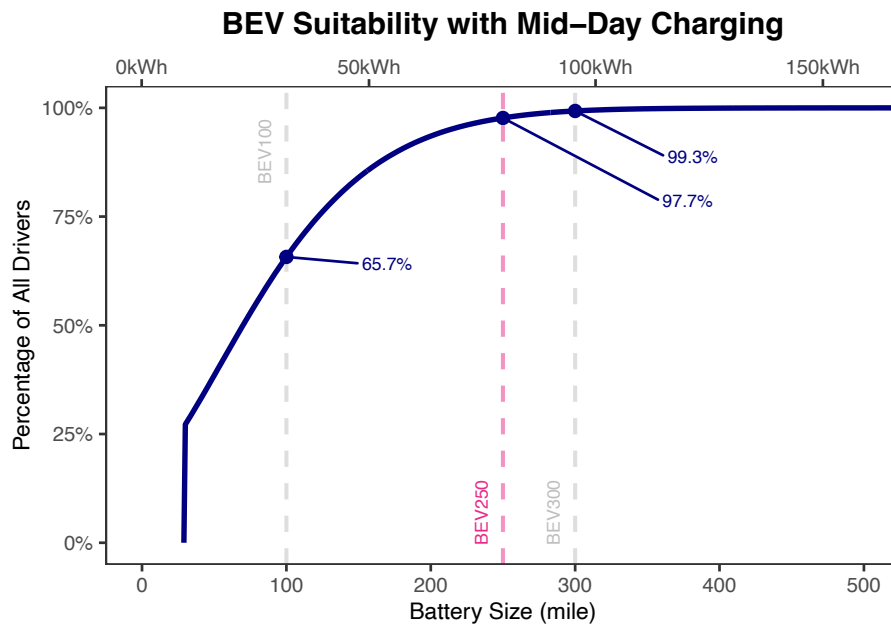


Fig. S-5. 95th-VMT BEV Suitability with midday 30-minute charging at 30 kW DCFC. We use the full sample of drivers on the Lyft platform. The procedure for producing this figure is identical to that of Fig. 1, except for the allowance of an additional 30-minute midday charge.

Table S-2. The residual value (*VRV*) of new vehicles at the end of ownership commitment period from alg.com. The residual value is expressed as the percentage of MSRP.

<i>New Models</i>	ICEV	HEV	BEV250*
MSRP	\$24,365	\$27,280	\$36,620
Mileage Per Year	3-Year Commitment		
10K miles/year	47%	56%	44%
20K miles/year	41%	51%	38%
30K miles/year	34%	45%	30%
40K miles/year	24%	39%	23%
	5-Year Commitment		
10K miles/year	32%	39%	34%
20K miles/year	23%	27%	23%
30K miles/year	10%	16%	13%
40K miles/year	1%	8%	8%

*For simplicity, we assume EV tax credits and subsidies are directly deducted from MSRP. The depreciation cost over commitment period is the difference between MSRP and residual value.

Table S-3. The residual value (*VRV*) of pre-owned vehicles at the end of ownership commitment period from alg.com

<i>Pre-owned Models*</i>	ICEV	HEV	BEV250	BEV100
Pre-owned Certified Dealer Price	\$15,632	\$18,362	\$19,144	\$11,083
Mileage Per Year	3-Year Commitment			
10K miles/year	38%	49%	51%	61%
20K miles/year	29%	42%	40%	45%
30K miles/year	19%	33%	26%	26%
40K miles/year	9%	24%	13%	8%
	5-Year Commitment			
10K miles/year	32%	42%	33%	52%
20K miles/year	17%	29%	14%	25%
30K miles/year	2%	15%	2%	3%
40K miles/year	2%	2%	2%	3%

*Kelly Blue Book estimate corresponding to "Certified Pre-Owned from Certified Dealer - Fair Purchase Price on Very Good Condition", with typical mileage of 30K at the time of purchase.

Table S-4. Estimated annual insurance costs (*I*) for new and pre-owned vehicles. We assume the insurance rate is not a function of mileage, following the methodology of AAA.

	ICEV	HEV	BEV
<i>New Models</i>	\$1,109	\$1,200	\$1,215
<i>Pre-Owned Models</i>	\$964	\$1,022	\$1,001

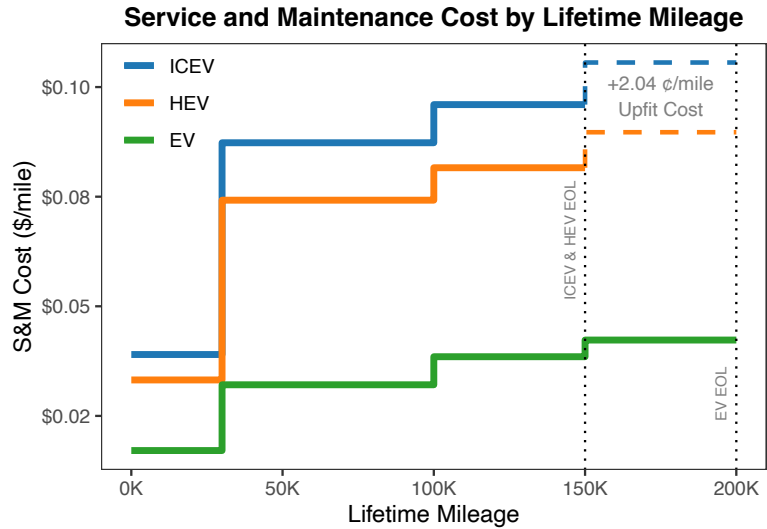


Fig. S-6. Service & Maintenance (S&M) costs per mile for different vehicle types. We assume that ICEV and HEV reach the end of their life (EOL) at 150,000 miles, while BEV reaches EOL at 200,000 miles. An upfit cost of \$0.0204/mile is assumed for any mileage after 150,000 miles for ICEVs and HEVs. No vehicle in our analysis reaches over 200,000 miles under the assumption of a 3- or 5-year commitment period.

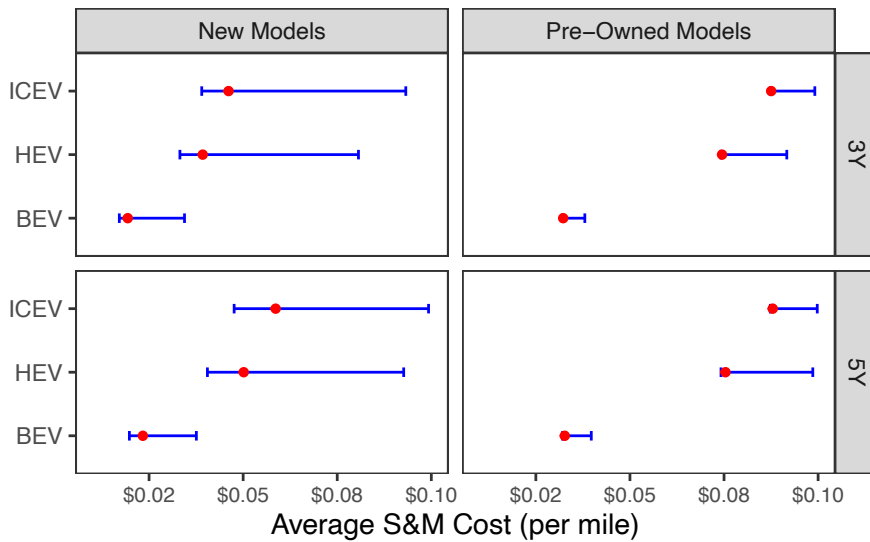


Fig. S-7. Range of mileage-weighted average S&M costs per mile for all drivers by model type, new vs. pre-owned, and commitment period length. The mileage-weighted average S&M cost additionally depends on annual mileage. Red points denote averages and whiskers show minimum and maximum.

Table S-5. 2019 average gas price and LCOC by state. Gas price includes taxes and is based on the weighted sales volume of three grades of gas, as calculated by the US Energy Information Administration.[1] The national average gas price in 2019 is \$2.763/gal, and the median is \$2.625/gal. LCOC is based on the central estimate of Borlaug et al. for each state.[2] The average LCOC is 0.15 \$/kWh nationwide, with the highest costs in Hawaii and the lowest in the Oregon, Washington DC, Delaware, and Maine. For equivalent per mile cost, 0.28 kWh/mile and 27 mile/gal are used for energy efficiency of BEV and ICEV, respectively.

<i>State</i>	<i>LCOC (\$/kWh)</i>	<i>Gas Price (G_s) (\$/gal)</i>	<i>BEV per-mile LCOC (\$/mile)</i>	<i>ICEV per-mile gas cost (\$/mile)</i>
Alabama	0.13	2.369	0.0364	0.0877
Alaska	0.25	3.516	0.0700	0.1302
Arizona	0.12	3.101	0.0336	0.1149
Arkansas	0.13	2.332	0.0364	0.0864
California	0.18	3.968	0.0504	0.1470
Colorado	0.13	2.503	0.0364	0.0927
Connecticut	0.15	3.040	0.0420	0.1126
Delaware	0.10	2.625	0.0280	0.0972
District of Columbia	0.10	3.089	0.0280	0.1144
Florida	0.15	2.698	0.0420	0.0999
Georgia	0.12	2.552	0.0336	0.0945
Hawaii	0.31	3.944	0.0868	0.1461
Idaho	0.13	2.930	0.0364	0.1085
Illinois	0.16	2.637	0.0448	0.0977
Indiana	0.15	2.491	0.0420	0.0923
Iowa	0.12	2.576	0.0336	0.0954
Kansas	0.16	2.393	0.0448	0.0886
Kentucky	0.13	2.576	0.0364	0.0954
Louisiana	0.13	2.381	0.0364	0.0882
Maine	0.10	2.723	0.0280	0.1009
Maryland	0.17	2.711	0.0476	0.1004
Massachusetts	0.23	2.955	0.0644	0.1094
Michigan	0.18	2.515	0.0504	0.0931
Minnesota	0.14	2.527	0.0392	0.0936
Mississippi	0.15	2.357	0.0420	0.0873
Missouri	0.15	2.332	0.0420	0.0864
Montana	0.15	2.784	0.0420	0.1031
Nebraska	0.15	2.613	0.0420	0.0968
Nevada	0.11	3.504	0.0308	0.1298
New Hampshire	0.12	2.808	0.0336	0.1040
New Jersey	0.15	2.845	0.0420	0.1054
New Mexico	0.14	2.479	0.0392	0.0918
New York	0.12	3.053	0.0336	0.1131
North Carolina	0.13	2.576	0.0364	0.0954
North Dakota	0.14	2.552	0.0392	0.0945
Ohio	0.15	2.393	0.0420	0.0886
Oklahoma	0.12	2.259	0.0336	0.0837
Oregon	0.10	3.480	0.0280	0.1289
Pennsylvania	0.16	3.004	0.0448	0.1113
Rhode Island	0.22	2.894	0.0616	0.1072
South Carolina	0.16	2.589	0.0448	0.0959
South Dakota	0.16	2.381	0.0448	0.0882
Tennessee	0.15	2.442	0.0420	0.0904
Texas	0.15	2.332	0.0420	0.0864
Utah	0.15	2.943	0.0420	0.1090
Vermont	0.15	2.943	0.0420	0.1090
Virginia	0.11	2.491	0.0308	0.0923
Washington	0.14	3.578	0.0392	0.1325
West Virginia	0.16	2.723	0.0448	0.1009
Wisconsin	0.12	2.503	0.0336	0.0927
Wyoming	0.15	2.906	0.0420	0.1076

Average Annual Savings (New Models)

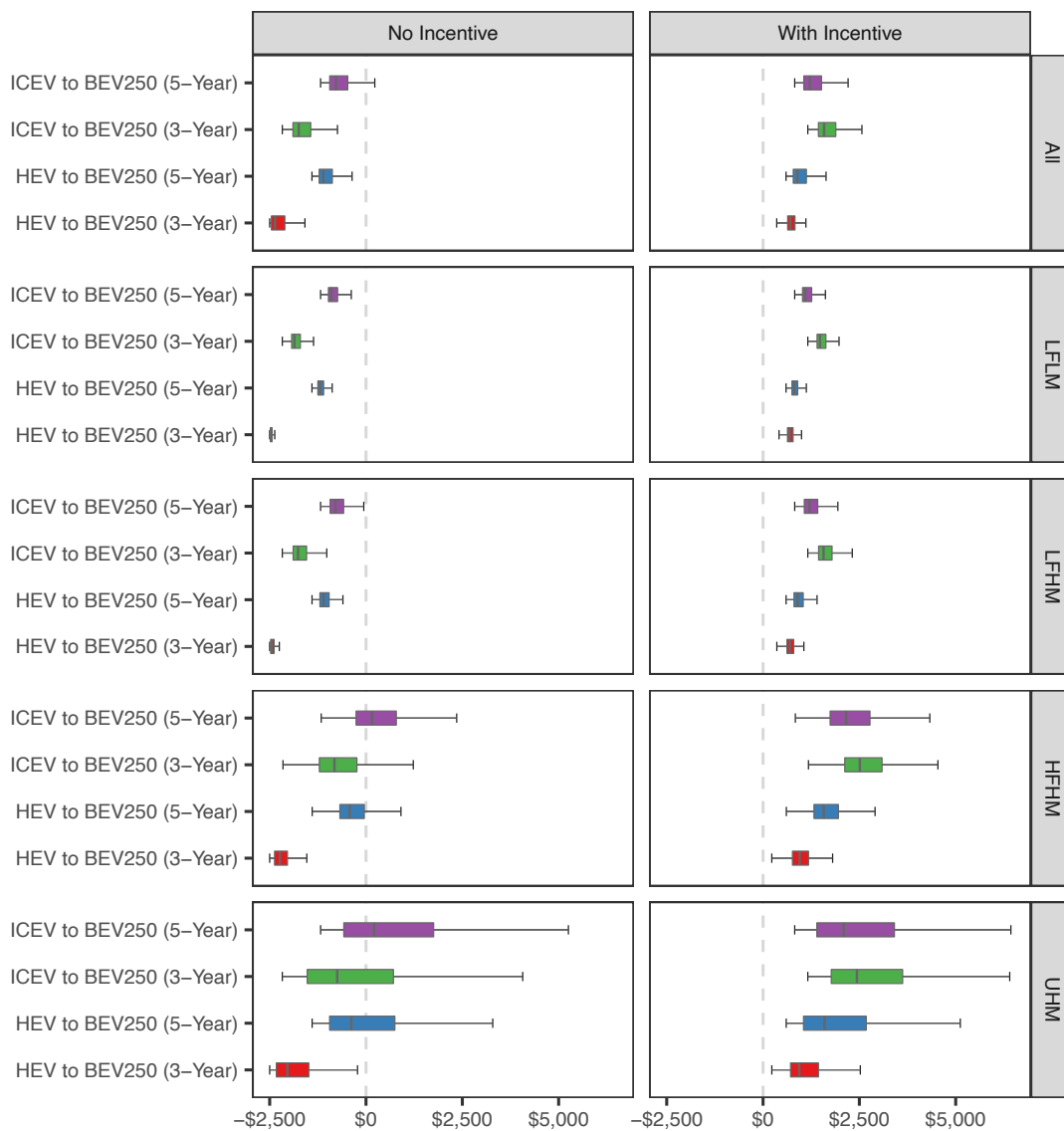


Fig. S-8. Distribution of average annual savings from switching to *new* BEVs under various scenarios. The range for all drivers is shown regardless of whether they are BEV suitable or not. Columns show with and without purchase subsidy and rows show the distribution for the cohorts. The boxes describe 25th percentiles (left hinge), medians, and 75th percentiles (right hinge) and whiskers describe 1.5 times the interquartile range.

Average Annual Savings (Pre-Owned Models)



Fig. S-9. Distribution of average annual savings from switching to *pre-preowned* BEVs under various scenarios. The range for all drivers is shown regardless of whether they are BEV suitable or not. Columns show the average savings 3- and 5-year commitment period and rows show the distribution for the cohorts. The boxes describe 25th percentiles (left hinge), medians, and 75th percentiles (right hinge) and whiskers describe 1.5 times the interquartile range.

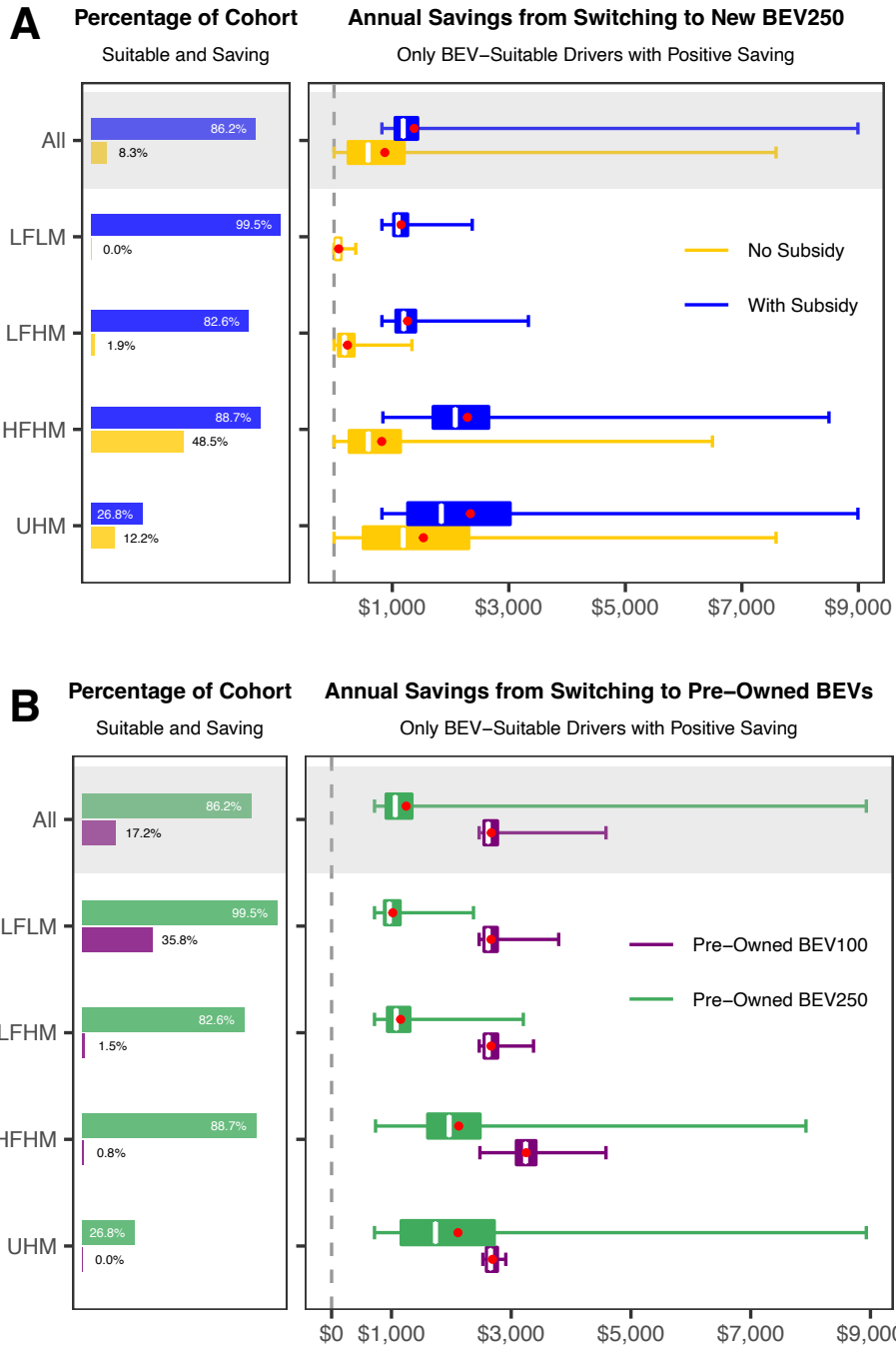


Fig. S-10. The range and distribution of annual saving from ICEV to BEV for BEV-suitable drivers with positive savings (Fig. 4 shows the full range). (A) From new ICEV to BEV250 with and without purchase subsidies under 5-year commitment period. (B) From pre-owned ICEV to pre-owned BEV250 and pre-owned BEV100 under 3-year commitment period. The red points show the average annual savings. The boxes describe 25th percentiles (left hinge), medians (white line), and 75th percentiles (right hinge) and whiskers describe absolute minimum and maximum.

Average Annual Saving from Switching to New BEV250

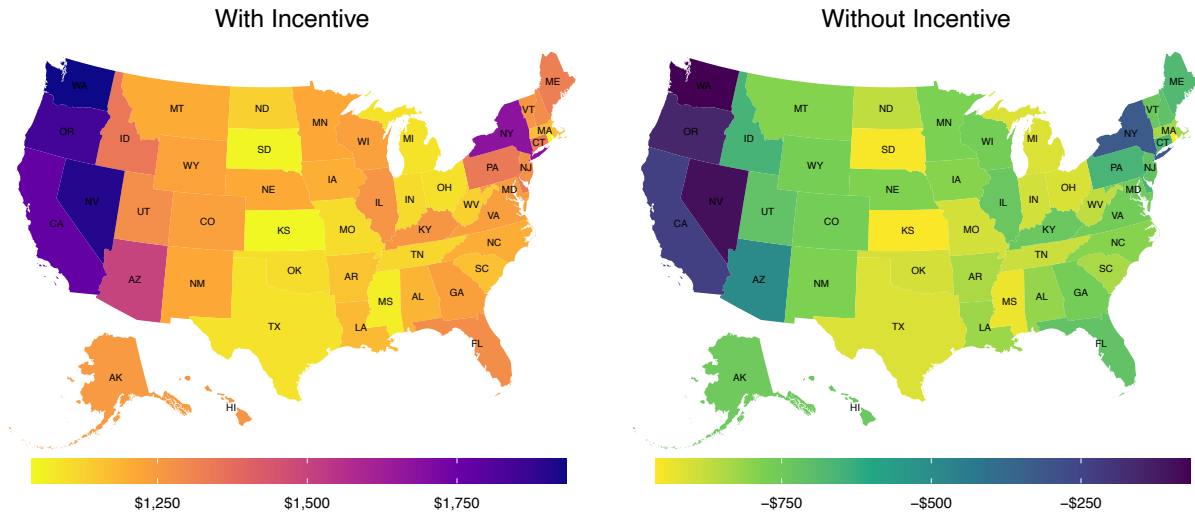


Fig. S-11. State-level average annual savings from new ICEV to new BEV250 with and without purchase subsidies under 5-year commitment period.

Supplementary Note 2: State-level average annual savings from new ICEV to new BEV250.

Fig. S-11 illustrates the state-level average annual savings from new ICEV to new BEV250 with and without purchase subsidies. With subsidies, states of WA, NE, OR, CA, and NY have the highest average annual savings. Without subsidies, Nevada's drivers return the highest savings, mostly due to the highest average mileage in the nation. States of KS, SD, MS, and RI have the lowest average annual savings in both cases. Note that, with subsidies, far more LFLM drivers in those states break even or save from switching to BEV, which changes the decomposition of the set of drivers in those states who are both BEV suitable and save from switching to BEVs.

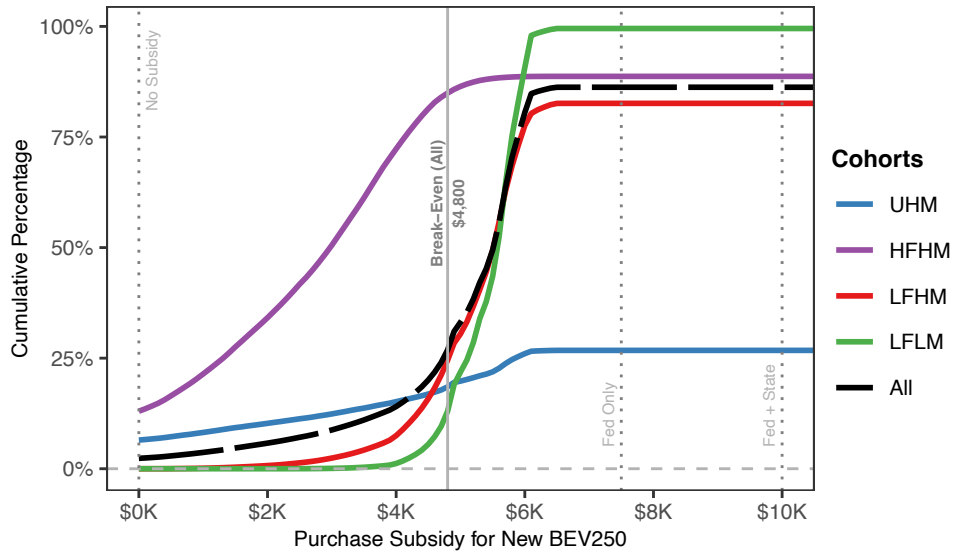


Fig. S-12. Percentage of drivers in each cohort that both find a BEV250 range-suitable and break even under a 3-year ownership commitment, as a function of subsidy level. Curves that plateau below 100% have drivers for whom a BEV250 does not have suitable range. An average driver breaks even with a minimum of \$4,800 purchase subsidy. Vertical lines indicate certain specific levels of subsidy. *Fed + State*: current level (\$10,000) for some states; *Fed Only*: \$7500 federal tax credit; *Reduced*: a scenario where tax rebate is reduced to \$5,000.

Supplementary Note 3: Sustainability implications.

The emissions conversion from gasoline to CO₂ is based on EPA measurements of 8,887 grCO_{2-eq} per gallon of gas and the fuel economy of replaced ICEV (27 miles per gallon). For the life-cycle GHG emissions we use BEV energy efficiency, data from state-level average emission factor of electricity generation from NREL’s Cambium dataset [3] and per-mile vehicle cradle-to-grave emissions (including vehicle manufacturing and battery production and end of life) for ICEV and BEV. The estimate of state-level marginal emission factor of electricity generation is for year 2020 based on short-run mid-case scenario of NREL’s Regional Energy Deployment System [3]. The US average marginal emission factor of electricity generation is 365.16 grCO_{2-eq}/kWh but varies greatly among the states. As a point of comparison, our estimate of California’s marginal emission factor for electricity generation is 192 grCO_{2-eq}/kWh which is slightly higher than the estimate of Jenn [4] (186 grCO_{2-eq}/kWh). We use a central estimate of 43 grCO_{2-eq}/mile for ICEV and a conservative estimate of 144 grCO_{2-eq}/mile for BEV including battery production for cradle-to-grave emissions excluding the use phase. Note that Cox et al., Hoekstra et al. and Elgowainy et al. estimate a range of 85-162 grCO_{2-eq}/mile for BEV as use-phase excluded cradle-to-grave emissions [5–7].

Table S-6. Implications of electrification of all drivers who are BEV250-suitable and save from switching. All figures are based on annual estimate

	All	UHM	HFHM	LFHM	LFLM
Annual Avoided Tailpipe GHG Emissions (Million Metric Tons of CO ₂ -eq)	5.72	0.85	1.52	1.55	2.34
Annual Avoided Life-Cycle GHG Emissions (Million Metric Tons of CO ₂ -eq)	4.30	0.22	1.18	1.16	1.74
Annual Electricity Consumption (TWh)	4.86	0.24	1.30	1.33	1.99

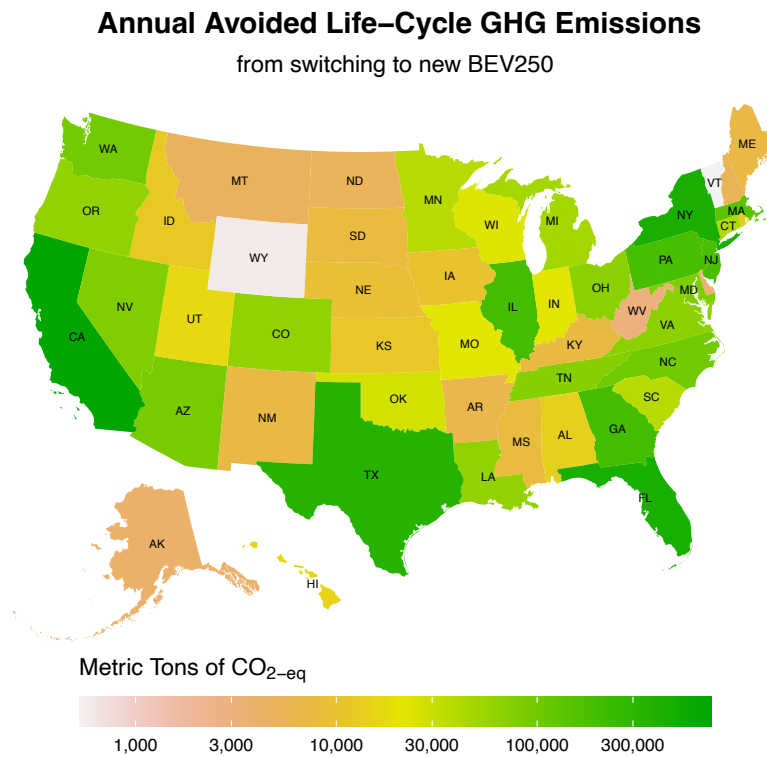


Fig. S-13. Annual avoided life-cycle GHG emissions from switching to new BEV250 across different states. We use average emission factor in each state and vehicle and battery life-cycle emissions.

Supplementary References

- [1] U.S. Energy Information Administration (EIA). The State Energy Data System (SEDS). 2019.
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