name: <unnamed> log type: smcl . use "C:\Users\Colantone\Dropbox\shared_folders\Paper_Area_B\Text_Tables\Paper\Submissions\APSR\Final_Resubmission_ > ta" (File created by user 'ajay.mishra' at Thu Feb 4 14:37:30 20) . do "C:\Users\COLANT~1\AppData\Local\Temp\STD466c 000000.tmp" . * Set up globals . global main "diesel euro4 dummy diesel dummy euro 4" . global controls "age female EDU* INC*" . global main_ass "diesel_euro4_ass dummy_diesel_ass dummy_euro_4_ass" . global main_placebo "diesel_euro5 dummy_diesel dummy_euro_5" ******* * Table 1 * * Table 1 can be produced in R, please see Replication_Code_R.R ******* * Table 2 * . * Without controls . regress vote_lega_euro \${main} if target!=3 & target!=4 & no_answer_euro==0, robust Linear regression Number of obs 602 = F(3, 598) 1.11 = Prob > F 0.3427 R-squared 0.0050 Root MSE .42996 Robust Coefficient std. err. P>|t| [95% conf. interval] vote_lega_~o t diesel_euro4 .1194786 .0750456 1.59 0.112 -.0279063 .2668635 dummy diesel -.0932496 -.2024346 .0555949 -1.68 0.094 .0159354

dummy_euro_4

_cons

-.0477564

.2769231

.0583352

.0393775

-0.82

7.03

0.413

0.000

-.1623231

.1995881

.0668103

.354258

. * Baseline

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. regress vote_lega_euro ${\text{main}}$ ${\text{controls}}$ if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression Number of obs = 602

F(23, 578) = 3.75 Prob > F = 0.0000 R-squared = 0.1301 Root MSE = .40892

vote_lega_~o	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.1830313	.0785914	2.33	0.020	.0286716	.3373909
dummy_diesel	1049461	.0570054	-1.84	0.066	216909	.0070168
dummy_euro_4	0480931	.0594245	-0.81	0.419	1648075	.0686212
age	.001885	.0016771	1.12	0.261	0014089	.005179
female	.1890018	.0392291	4.82	0.000	.1119528	.2660507
EDU1	.1826344	.0497742	3.67	0.000	.084874	.2803947
EDU2	0	(omitted)				
EDU3	.0004736	.0403185	0.01	0.991	0787151	.0796623
EDU4	.0062251	.1335002	0.05	0.963	2559796	.2684297
INC1	1382317	.1994812	-0.69	0.489	5300282	.2535647
INC2	2210545	.2106813	-1.05	0.295	6348487	.1927397
INC3	1783681	.1981399	-0.90	0.368	5675301	.2107938
INC4	.1079032	.2133229	0.51	0.613	3110794	.5268857
INC5	0650777	.1913262	-0.34	0.734	4408571	.3107017
INC6	0458264	.1911754	-0.24	0.811	4213096	.3296569
INC7	0688083	.1866818	-0.37	0.713	4354657	.2978491
INC8	0857448	.1875631	-0.46	0.648	4541331	.2826436
INC9	1435479	.1895485	-0.76	0.449	5158357	.2287398
INC10	.0241006	.1985724	0.12	0.903	3659108	.4141121
INC11	.048711	.19739	0.25	0.805	3389781	.4364001
INC12	.0576549	.1960246	0.29	0.769	3273524	.4426623
INC13	0061338	.1948657	-0.03	0.975	3888649	.3765973
INC14	.1238005	.1852829	0.67	0.504	2401092	.4877103
INC15	0	(omitted)				
INC16	.0793645	.1945586	0.41	0.683	3027636	.4614925
_cons	.0390888	.1857972	0.21	0.833	3258312	.4040087
_	I .					

^{. *} Including unkown-car

. regress vote_lega_euro \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0, robust note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Root MSE = .41685

		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4_ass	.1544376	.0779347	1.98	0.048	.001399	.3074762
dummy_diesel_ass	0819337	.0552265	-1.48	0.138	1903808	.0265134
dummy_euro_4_ass	0194904	.0591124	-0.33	0.742	1355682	.0965873
age	.0006071	.0015829	0.38	0.701	0025012	.0037155
female	.1665529	.0381233	4.37	0.000	.0916911	.2414148
EDU1	.1724035	.1294395	1.33	0.183	0817739	.4265809
EDU2	0356539	.1293472	-0.28	0.783	2896501	.2183423
EDU3	0082914	.1290414	-0.06	0.949	2616871	. 2451044
EDU4	0	(omitted)				
INC1	.0375498	.1358073	0.28	0.782	2291319	.3042315
INC2	0	(omitted)				
INC3	0217927	.1339426	-0.16	0.871	2848128	.2412274
INC4	.1647205	.154966	1.06	0.288	1395826	.4690237
INC5	.1128638	.1224952	0.92	0.357	1276773	.3534048
INC6	.1229344	.1241327	0.99	0.322	1208221	.366691
INC7	.0994985	.1220841	0.82	0.415	1402352	.3392323
INC8	.0849728	.1215182	0.70	0.485	1536498	.3235953
INC9	0034794	.1264584	-0.03	0.978	2518028	.2448441
INC10	.1817189	.1400781	1.30	0.195	0933493	.4567871
INC11	.2074839	.1333446	1.56	0.120	0543619	.4693296
INC12	.2100362	.1380874	1.52	0.129	0611231	.4811954
INC13	.2292494	.1338543	1.71	0.087	0335971	.492096
INC14	.3182555	.1214654	2.62	0.009	.0797365	.5567744
INC15	.2287107	.1935848	1.18	0.238	1514274	.6088487
INC16	.2594091	.1278177	2.03	0.043	.0084165	.5104018
dummy_car_unknown	.1702469	.0666029	2.56	0.011	.0394602	.3010336
_cons	0701454	.1874194	-0.37	0.708	4381766	. 2978858

. * Baseline, including control for past vote for Lega: Legislative 2018

. regress vote_lega_euro \${main} \${controls} vote_lega_2018 if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU4 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 583 F(24, 558) = 62.60 Prob > F = 0.0000 R-squared = 0.6010 Root MSE = .27514

_		Robust		- 1.1		
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1147724	.0467204	2.46	0.014	.023003	.2065418
dummy_diesel	023723	.03644	-0.65	0.515	0952993	.0478533
dummy_euro_4	.0066227	.0332523	0.20	0.842	0586923	.0719377
age	.0014439	.00102	1.42	0.157	0005596	.0034474
female	.1470264	.0275513	5.34	0.000	.0929094	.2011434
EDU1	088507	.1375976	-0.64	0.520	3587795	.1817655
EDU2	1051362	.1381262	-0.76	0.447	3764471	.1661746
EDU3	0849538	.1395241	-0.61	0.543	3590105	.1891029
EDU4	0	(omitted)				
INC1	0433535	.0762879	-0.57	0.570	1932001	.1064931
INC2	0668584	.0596636	-1.12	0.263	1840511	.0503344
INC3	0023031	.0861052	-0.03	0.979	1714331	.1668268
INC4	.1824029	.1120233	1.63	0.104	037636	.4024418
INC5	0713339	.0688012	-1.04	0.300	2064749	.0638071
INC6	0138449	.0716467	-0.19	0.847	1545751	.1268853

INC7	078813	.0621709	-1.27	0.205	2009307	.0433047
INC8	0694282	.0593435	-1.17	0.243	1859922	.0471357
INC9	019285	.0711701	-0.27	0.787	159079	.120509
INC10	.0435213	.071825	0.61	0.545	0975591	.1846017
INC11	.0028722	.0627428	0.05	0.964	1203687	.126113
INC12	0409834	.0636501	-0.64	0.520	1660065	.0840397
INC13	0341376	.0571087	-0.60	0.550	146312	.0780367
INC14	.1221494	.0608786	2.01	0.045	.0025701	.2417286
INC15	0	(omitted)				
INC16	.03316	.0679302	0.49	0.626	1002702	.1665903
vote_lega_2018	.8116164	.0340132	23.86	0.000	.7448068	.8784259
_cons	.0120814	.1518648	0.08	0.937	2862153	.310378

. * Baseline, including control for past vote for Lega: Regional 2018

. regress vote_lega_euro \${main} \${controls} vote_lega_regional if target!=3 & target!=4 & no_answer_euro==0 & no_ar
note: EDU2 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 551 F(24, 526) = 31.61 Prob > F = 0.0000 R-squared = 0.5767 Root MSE = .2864

vote lega euro	Coefficient	Robust std. err.	t	P> t	[95% conf.	intervall
voce_iega_euro	COETTICIENT	stu. err.		7/14	[33% COIII.	
diesel_euro4	.0941761	.0517269	1.82	0.069	0074407	.1957929
dummy_diesel	.00328	.0401764	0.08	0.935	0756459	.0822059
dummy_euro_4	.0263229	.0363235	0.72	0.469	0450339	.0976798
age	.0029109	.0009836	2.96	0.003	.0009786	.0048432
female	.1727745	.0300668	5.75	0.000	.1137088	.2318402
EDU1	.0498577	.0330373	1.51	0.132	0150436	.1147589
EDU2	0	(omitted)				
EDU3	.0532539	.0236643	2.25	0.025	.0067656	.0997421
EDU4	0271595	.2303735	-0.12	0.906	4797246	.4254056
INC1	0097669	.0944118	-0.10	0.918	1952373	.1757036
INC2	0	(omitted)				
INC3	.0382026	.0485185	0.79	0.431	0571112	.1335164
INC4	.1750409	.0922084	1.90	0.058	006101	.3561829
INC5	.0183782	.0443118	0.41	0.678	0686716	.105428
INC6	.0467729	.0614279	0.76	0.447	0739013	.1674471
INC7	.0434585	.0621593	0.70	0.485	0786524	.1655693
INC8	.0410498	.0560944	0.73	0.465	0691468	.1512463
INC9	0070477	.061342	-0.11	0.909	1275531	.1134577
INC10	.0953636	.0494111	1.93	0.054	0017037	.192431
INC11	.0064203	.0601279	0.11	0.915	1117	.1245405
INC12	.069529	.0495152	1.40	0.161	0277428	.1668008
INC13	.0536102	.0476585	1.12	0.261	0400142	.1472346
INC14	.223511	.0548733	4.07	0.000	.1157133	.3313087
INC15	.0932004	.0730371	1.28	0.202	0502799	.2366807
INC16	.0836113	.0574878	1.45	0.146	0293226	.1965452
e_lega_regional	.7791193	.0400396	19.46	0.000	.7004622	.8577764
_cons	2885823	.0652108	-4.43	0.000	4166879	1604766
	1					

. * Baseline, including control for past vote for Lega: Municipal 2018

. regress vote_lega_euro \${main} \${controls} vote_lega_municipal if target!=3 & target!=4 & no_answer_euro==0 & no_a note: EDU2 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs = 533

F(24, 508) = 27.68 Prob > F = 0.0000 R-squared = 0.4936 Root MSE = .31221

Robust vote_lega_euro Coefficient std. err. t P>|t| [95% conf. interval] .0600113 diesel_euro4 2.43 .0277116 .2635133 .1456124 0.016 -.0006682 dummy diesel .0486438 -0.01 0.989 -.0962361 .0948996 dummy_euro_4 -.0277699 .0427595 -0.65 0.516 -.1117773 .0562374 age .0022252 .0012077 1.84 0.066 -.0001475 .0045978 female .2010279 .034627 5.81 0.000 .1329982 .2690575 .0764369 .0379765 2.01 .0018265 EDU1 0.045 .1510473 EDU2 0 (omitted) EDU3 .016567 .0293663 0.56 0.573 -.0411274 .0742614 EDU4 .0935795 .1303992 0.72 0.473 -.1626085 .3497675 .002453 0.02 INC1 .1029347 0.981 -.1997771 .2046831 (omitted) INC2 0 INC3 .0567187 .0476302 1.19 0.234 -.0368578 .1502952 INC4 .2003509 .091237 2.20 0.029 .0211027 .3795991 INC5 .0725067 .0684161 1.06 0.290 -.0619066 .20692 INC6 .1163793 .0719004 1.62 0.106 -.0248794 .2576381 INC7 .0693228 .0646498 1.07 0.284 -.0576912 .1963368 INC8 .0600783 .0597806 1.00 0.315 -.0573692 .1775259 INC9 .0369859 .0795883 0.46 0.642 -.1193769 .1933487 INC10 .1266977 .0711686 1.78 0.076 -.0131234 .2665187 INC11 .077005 .0566962 1.36 0.175 -.0343829 .1883929 INC12 .0703269 .0534953 1.31 0.189 -.0347724 .1754262 INC13 .0748132 .051663 1.45 0.148 -.0266862 .1763126 0.000 INC14 .2407325 .0577384 4.17 .1272971 .3541679 -.1614097 INC15 .1038984 0.77 0.442 .1350413 .3692065 INC16 .1024436 .0669783 1.53 0.127 -.029145 .2340322 vote_lega_municipal .7299277 .0447976 16.29 0.000 .6419162 .8179391

.0808548

-3.10

0.002

-.4096635

-.0919615

-.2508125

_cons

* Table 3 *

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. * Baseline

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. regress vote_pd_euro \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust

note: EDU2 omitted because of collinearity.
note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 602 F(23, 578) = 16.07 Prob > F = 0.0000 R-squared = 0.2905 Root MSE = .38835

		Robust				
vote_pd_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0673565	.0811777	0.83	0.407	0920828	.2267958
dummy_diesel	0071384	.0557068	-0.13	0.898	1165509	.102274
dummy_euro_4	.0685883	.0626926	1.09	0.274	0545447	.1917214
age	.0078596	.0016984	4.63	0.000	.0045238	.0111954
female	062195	.0355036	-1.75	0.080	1319269	.0075369
EDU1	2904135	.0500556	-5.80	0.000	3887265	1921006
EDU2	0	(omitted)				
EDU3	0200189	.0450241	-0.44	0.657	1084497	.0684119
EDU4	2145252	.1257285	-1.71	0.088	4614657	.0324153
INC1	1584592	.1981643	-0.80	0.424	5476692	.2307507
INC2	.075999	.2192631	0.35	0.729	3546506	.5066486
INC3	2169041	.188963	-1.15	0.251	5880419	.1542338
INC4	1307243	.2120621	-0.62	0.538	5472306	.285782
INC5	1087455	.1848753	-0.59	0.557	4718549	.2543638
INC6	0564234	.1828255	-0.31	0.758	4155068	.3026599
INC7	.0960755	.1834875	0.52	0.601	264308	.456459
INC8	.0723671	.1829171	0.40	0.693	2868961	.4316304
INC9	06135	.1987421	-0.31	0.758	4516947	.3289947
INC10	2376164	.1878703	-1.26	0.206	606608	.1313753
INC11	2236768	.1890748	-1.18	0.237	5950342	.1476806
INC12	2492142	.1956537	-1.27	0.203	6334932	.1350647
INC13	3396002	.1948567	-1.74	0.082	7223137	.0431133
INC14	5304306	.1797026	-2.95	0.003	8833804	1774809
INC15	0	(omitted)				
INC16	1471699	.1880104	-0.78	0.434	5164366	.2220969
_cons	.16302	.1847704	0.88	0.378	1998832	.5259232

^{. *} Baseline, including control for past vote: Legislative 2018

. regress vote_pd_euro \${main} \${controls} vote_pd_2018 if target!=3 & target!=4 & no_answer_euro==0 & no_answer_201 note: EDU4 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression Number of obs = F(24, 558) =

F(24, 558) = 113.51 Prob > F = 0.0000 R-squared = 0.7240 Root MSE = .24322

583

		Robust				
vote_pd_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0334759	.0526587	0.64	0.525	0699576	.1369094
dummy_diesel	0315117	.0348453	-0.90	0.366	0999556	.0369322
dummy_euro_4	0021212	.043059	-0.05	0.961	0866987	.0824564
age	.0018948	.0011912	1.59	0.112	0004451	.0042347
female	.0011604	.0216043	0.05	0.957	0412753	.0435962
EDU1	0256775	.039536	-0.65	0.516	1033352	.0519801
EDU2	.0655039	.0470493	1.39	0.164	0269115	.1579193
EDU3	.0497863	.0451126	1.10	0.270	0388251	.1383976
EDU4	0	(omitted)				
INC1	136349	.147008	-0.93	0.354	4251056	.1524077
INC2	1569445	.1522266	-1.03	0.303	4559517	.1420627
INC3	1080332	.1420312	-0.76	0.447	3870144	.1709481
INC4	0861575	.1506395	-0.57	0.568	3820472	.2097322
INC5	1530372	.1394348	-1.10	0.273	4269184	.120844
INC6	1200863	.1360176	-0.88	0.378	3872553	.1470827
INC7	0953471	.1352612	-0.70	0.481	3610304	.1703363
INC8	0853615	.1348423	-0.63	0.527	3502221	.1794991
INC9	1563414	.1526971	-1.02	0.306	4562728	.1435901
INC10	1879328	.1353589	-1.39	0.166	453808	.0779424
INC11	155529	.1324198	-1.17	0.241	4156312	.1045732
INC12	1239843	.1363223	-0.91	0.363	391752	.1437834
INC13	1262123	.1565823	-0.81	0.421	433775	.1813504
INC14	2395879	.135053	-1.77	0.077	5048622	.0256864
INC15	0	(omitted)				
INC16	1044982	`.1426119	-0.73	0.464	38462	.1756235
vote pd 2018	.781718	.0379348	20.61	0.000	.7072054	.8562305
cons	.0961454	.1273019	0.76	0.450	1539041	.3461948
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. * Baseline, including control for past vote: Regional 2018

. regress vote_pd_euro \${main} \${controls} vote_pd_regional if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU2 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs = 551F(24, 526) = 108.06

F(24, 526) = 108.06 Prob > F = 0.0000 R-squared = 0.7078 Root MSE = .25298

		Robust				
vote_pd_euro	Coefficient	std. err.	t	P> t	[95% conf.	. interval]
diesel_euro4	.0562581	.057011	0.99	0.324	0557391	.1682554
dummy_diesel	0257622	.040534	-0.64	0.525	1053906	.0538662
dummy euro 4	.0021601	.0480189	0.04	0.964	0921722	.0964925
age	.0025956	.0011945	2.17	0.030	.000249	.0049423
female	0084259	.0237723	-0.35	0.723	0551262	.0382744
EDU1	0554682	.0366795	-1.51	0.131	1275245	.0165881
EDU2	0	(omitted)				
EDU3	.0102358	.0312776	0.33	0.744	0512086	.0716802
EDU4	1651738	.1411981	-1.17	0.243	4425552	.1122077
INC1	0809435	.0931254	-0.87	0.385	2638868	.1019998
INC2	0	(omitted)				
INC3	1324264	.0835505	-1.58	0.114	29656	.0317071
INC4	0332576	.1104953	-0.30	0.764	2503238	.1838087
INC5	0912534	.0569828	-1.60	0.110	2031953	.0206885
INC6	0969742	.0619333	-1.57	0.118	2186412	.0246927

INC7	0332325	.0618058	-0.54	0.591	1546489	.088184
INC8	.015495	.0537646	0.29	0.773	0901248	.1211148
INC9	1388853	.0980152	-1.42	0.157	3314346	.053664
INC10	1422172	.0726485	-1.96	0.051	284934	.0004996
INC11	0788362	.0544237	-1.45	0.148	1857507	.0280784
INC12	0787294	.0574255	-1.37	0.171	1915409	.034082
INC13	1128766	.0982971	-1.15	0.251	3059797	.0802266
INC14	1994195	.0621606	-3.21	0.001	3215329	077306
INC15	1258169	.1300361	-0.97	0.334	3812707	.1296369
INC16	0782232	.0738263	-1.06	0.290	2232538	.0668075
<pre>vote_pd_regional</pre>	.7263476	.0401024	18.11	0.000	.647567	.8051281
_cons	.0373118	.0769449	0.48	0.628	1138452	.1884688

. * Baseline, including control for past vote: Municipal 2016

. regress vote_pd_euro \${main} \${controls} vote_pd_municipal if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 533 F(24, 508) = 67.09 Prob > F = 0.0000 R-squared = 0.6996 Root MSE = .25953

	T .					
		Robust				
vote_pd_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0165936	.0643387	0.26	0.797	1098092	.1429963
dummy_diesel	0404308	.0439976	-0.92	0.359	1268704	.0460089
dummy_euro_4	.0756974	.0542025	1.40	0.163	0307914	.1821861
age	.0061487	.0014826	4.15	0.000	.0032359	.0090615
female	0251053	.0263756	-0.95	0.342	076924	.0267133
EDU1	0	(omitted)				
EDU2	.0795794	.0399738	1.99	0.047	.001045	.1581138
EDU3	.0660008	.0424167	1.56	0.120	0173329	.1493345
EDU4	25695	.1616527	-1.59	0.113	5745402	.0606402
INC1	1139343	.1453083	-0.78	0.433	3994134	.1715448
INC2	0	(omitted)				
INC3	1037755	.1427353	-0.73	0.468	3841996	.1766486
INC4	1326176	.1258907	-1.05	0.293	379948	.1147129
INC5	1507703	.1231209	-1.22	0.221	3926591	.0911185
INC6	1218543	.1266173	-0.96	0.336	3706123	.1269037
INC7	0279001	.1208853	-0.23	0.818	2653967	.2095965
INC8	0718897	.1222842	-0.59	0.557	3121347	.1683553
INC9	1386031	.1466069	-0.95	0.345	4266335	.1494274
INC10	1320984	.1297048	-1.02	0.309	3869223	.1227255
INC11	0524724	.1235772	-0.42	0.671	2952577	.1903128
INC12	0584533	.1209431	-0.48	0.629	2960634	.1791569
INC13	0877352	.1433637	-0.61	0.541	3693938	.1939235
INC14	141365	.1250377	-1.13	0.259	3870197	.1042897
INC15	.2170013	.2688475	0.81	0.420	3111885	.745191
INC16	09105	.1261936	-0.72	0.471	3389755	.1568756
<pre>vote_pd_municipal</pre>	.7391553	.0390579	18.92	0.000	.6624204	.8158902
_cons	2048511	.1244825	-1.65	0.100	449415	.0397128

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. * Forza Italia *
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. * Baseline

. regress vote_forzaitalia_euro \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression Number of obs = 602

F(23, 578) = 8.07 Prob > F = 0.0000 R-squared = 0.2617 Root MSE = .33399

	C (C; -;+	Robust		D. 1±1	[05%	
vote_forza~o	Coefficient	std. err.	t	P> t	[95% conf.	intervalj
diesel_euro4	051503	.0580359	-0.89	0.375	16549	.062484
dummy_diesel	.0304237	.0443797	0.69	0.493	0567413	.1175888
dummy_euro_4	0631221	.0370165	-1.71	0.089	1358253	.0095811
age	0018126	.0012573	-1.44	0.150	004282	.0006567
female	0970692	.0319868	-3.03	0.003	1598938	0342447
EDU1	0145138	.0334561	-0.43	0.665	0802241	.0511965
EDU2	0	(omitted)				
EDU3	0384718	.0354495	-1.09	0.278	1080973	.0311537
EDU4	1346073	.0840162	-1.60	0.110	2996217	.030407
INC1	.123227	.0596389	2.07	0.039	.0060916	.2403624
INC2	.223678	.1129139	1.98	0.048	.0019065	.4454496
INC3	.1514527	.0818114	1.85	0.065	0092311	.3121365
INC4	.0223623	.0373225	0.60	0.549	050942	.0956666
INC5	.1453818	.0546683	2.66	0.008	.038009	.2527546
INC6	.0834163	.0458571	1.82	0.069	0066506	.1734832
INC7	.0623051	.0453599	1.37	0.170	0267853	.1513955
INC8	.0828709	.0484088	1.71	0.087	0122076	.1779494
INC9	.039891	.0389753	1.02	0.307	0366595	.1164414
INC10	.0463765	.0548609	0.85	0.398	0613745	.1541276
INC11	.2625543	.0813426	3.23	0.001	.1027913	.4223174
INC12	.384412	.1065216	3.61	0.000	.1751953	.5936287
INC13	.2280887	.0866181	2.63	0.009	.0579642	.3982132
INC14	.5859414	.0609232	9.62	0.000	.4662836	.7055992
INC15	0	(omitted)				
INC16	.1521239	.0520846	2.92	0.004	.0498257	. 254422
_cons	.1323498	.069207	1.91	0.056	003578	.2682775
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^{. *} Baseline, including control for past vote: Legislative 2018

ioce. Incis omitteed because of collinearity.

[.] regress vote_forzaitalia_euro \${main} \${controls} vote_forzaitalia_2018 if target!=3 & target!=4 & no_answer_euro= note: EDU4 omitted because of collinearity. note: INC15 omitted because of collinearity.

		Robust				
vote_forzaitalia_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0032324	.0316293	0.10	0.919	0588945	.0653594
dummy_diesel	0382471	.0283595	-1.35	0.178	0939515	.0174572
dummy_euro_4	.0008944	.0250034	0.04	0.971	0482178	.0500066
age	.0004699	.0006103	0.77	0.442	0007288	.0016686
female	0197158	.0183297	-1.08	0.283	0557194	.0162878
EDU1	.0160134	.0307336	0.52	0.603	0443543	.0763811
EDU2	.0194387	.0332249	0.59	0.559	0458225	.0846999
EDU3	.0239191	.0295486	0.81	0.419	0341209	.0819591
EDU4	0	(omitted)				
INC1	0315349	.0412458	-0.76	0.445	1125509	.0494812
INC2	.0071664	.0240011	0.30	0.765	0399771	.0543099
INC3	0363264	.0530201	-0.69	0.494	1404697	.067817
INC4	009407	.0131011	-0.72	0.473	0351405	.0163265
INC5	.0305993	.028216	1.08	0.279	0248232	.0860219
INC6	.0023962	.0296694	0.08	0.936	0558812	.0606736
INC7	.0138727	.0263996	0.53	0.599	037982	.0657275
INC8	.0027384	.0143196	0.19	0.848	0253885	.0308653
INC9	0027116	.0139591	-0.19	0.846	0301304	.0247071
INC10	.0016309	.0133475	0.12	0.903	0245865	.0278483
INC11	.0514571	.0337614	1.52	0.128	0148578	.117772
INC12	.1237576	.0579046	2.14	0.033	.01002	.2374953
INC13	0104963	.0328382	-0.32	0.749	0749979	.0540052
INC14	.1091663	.0316026	3.45	0.001	.0470917	.171241
INC15	0	(omitted)				
INC16	0746893	.0398735	-1.87	0.062	1530098	.0036312
vote_forzaitalia_2018	.8474668	.0354064	23.94	0.000	.7779206	.917013
cons	0101991	.0398233	-0.26	0.798	0884209	.0680227

. * Baseline, including control for past vote: Regional 2018

. regress vote_forzaitalia_euro \${main} \${controls} vote_forzaitalia_regional if target!=3 & target!=4 & no_answer_e note: EDU2 omitted because of collinearity.
note: INC2 omitted because of collinearity.

				Robust		
interval]	[95% conf.	P> t	t	std. err.	Coefficient	<pre>vote_forzaitalia_euro</pre>
.127382	0263618	0.197	1.29	.0391309	.0505101	diesel_euro4
.000031	1403064	0.050	-1.96	.0357186	0701377	dummy_diesel
.0341963	0771479	0.449	-0.76	.0283393	0214758	dummy_euro_4
.0010432	0021678	0.492	-0.69	.0008173	0005623	age
.0310169	0583126	0.549	-0.60	.0227361	0136478	female
.0162721	0816772	0.190	-1.31	.02493	0327025	EDU1
				(omitted)	0	EDU2
.0297921	053397	0.577	-0.56	.0211733	0118025	EDU3
.0370785	1121714	0.323	-0.99	.0379871	0375465	EDU4
.1764447	1687254	0.965	0.04	.0878526	.0038596	INC1
				(omitted)	0	INC2
.2210919	086699	0.391	0.86	.0783389	.0671965	INC3
.1488641	2180248	0.711	-0.37	.0933805	0345803	INC4
.2349633	069429	0.286	1.07	.0774738	.0827671	INC5
.1571194	1458538	0.942	0.07	.0771127	.0056328	INC6

INC7	.0296288	.0793946	0.37	0.709	1263407	.1855983
INC8	.0238564	.0764711	0.31	0.755	1263698	.1740826
INC9	.0288311	.0731537	0.39	0.694	1148781	.1725403
INC10	.0449754	.075333	0.60	0.551	1030151	.1929658
INC11	.104858	.0904079	1.16	0.247	0727469	.2824628
INC12	.1640165	.1044232	1.57	0.117	0411212	.3691541
INC13	.0661278	.0778658	0.85	0.396	0868383	.219094
INC14	.1757183	.0810588	2.17	0.031	.0164796	.334957
INC15	0759552	.1246697	-0.61	0.543	3208669	.1689565
INC16	.0010245	.0856615	0.01	0.990	1672561	.1693051
vote_forzaitalia_regio~l	.7526285	.0427364	17.61	0.000	.6686736	.8365834
_cons	.0502839	.0873054	0.58	0.565	1212261	.221794

. * Baseline, including control for past vote: Municipal 2016

regress vote_forzaitalia_euro \${main} \${controls} vote_forzaitalia_municipal if target!=3 & target!=4 & no_answer_note: EDU2 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 533 F(24, 508) = 89.23 Prob > F = 0.0000 R-squared = 0.7279 Root MSE = .21176

				Robust		
interval]	[95% conf.	P> t	t	std. err.	Coefficient	vote_forzaitalia_euro
.0753602	0849974	0.906	-0.12	.0408108	0048186	diesel_euro4
.0629634	0764421	0.849	-0.19	.0354785	0067394	dummy_diesel
.0388417	0797835	0.498	-0.68	.03019	0204709	dummy_euro_4
.0012155	0021016	0.600	-0.52	.0008442	000443	age
0009104	090951	0.046	-2.00	.0229152	0459307	female
.0077026	0911852	0.098	-1.66	.0251668	0417413	EDU1
				(omitted)	0	EDU2
.0431541	0339162	0.814	0.24	.0196143	.0046189	EDU3
.0789994	0895207	0.902	-0.12	.0428882	0052606	EDU4
.0219758	2489182	0.100	-1.65	.0689422	1134712	INC1
				(omitted)	0	INC2
.0715368	1152367	0.646	-0.46	.0475337	0218499	INC3
.0049223	2571359	0.059	-1.89	.0666935	1261068	INC4
.0868369	1467391	0.615	-0.50	.0594448	0299511	INC5
.0003072	1894473	0.051	-1.96	.0482923	0945701	INC6
.0159971	1912245	0.097	-1.66	.0527377	0876137	INC7
.0054429	1864991	0.064	-1.85	.048849	0905281	INC8
0004728	2057896	0.049	-1.97	.0522529	1031312	INC9
.0148685	1548212	0.106	-1.62	.0431858	0699764	INC10
.096104	1304714	0.766	-0.30	.0576632	0171837	INC11
.1907572	0656073	0.338	0.96	.0652445	.0625749	INC12
.058286	1214692	0.490	-0.69	.0457475	0315916	INC13
.1813464	0153244	0.098	1.66	.0500525	.083011	INC14
.0322159	1627154	0.189	-1.32	.0496098	0652497	INC15
.0500709	2018401	0.237	-1.18	.0641111	0758846	INC16
.8240579	.6514069	0.000	16.79	.0439395	.7377324	ote_forzaitalia_munic~l
.2619208	.0113898	0.033	2.14	.0637599	.1366553	cons

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. * Baseline

. regress vote_m5s_euro \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression Number of obs = 602

F(23, 578) = 3.22 Prob > F = 0.0000 R-squared = 0.1346 Root MSE = .3505

		Robust				
vote_m5s_e~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	0895018	.0686121	-1.30	0.193	2242612	.0452577
dummy_diesel	.0536489	.0516961	1.04	0.300	0478862	.155184
dummy_euro_4	.0212632	.0509207	0.42	0.676	0787489	.1212754
age	005268	.0013253	-3.97	0.000	0078709	002665
female	.006274	.0326666	0.19	0.848	0578856	.0704337
EDU1	.1217444	.0436795	2.79	0.005	.0359546	.2075342
EDU2	0	(omitted)				
EDU3	0198816	.0352259	-0.56	0.573	0890681	.0493048
EDU4	.3318995	.2119532	1.57	0.118	0843929	.7481919
INC1	0435591	.1511511	-0.29	0.773	3404315	.2533133
INC2	.1045069	.1862159	0.56	0.575	2612354	.4702493
INC3	.2859461	.183118	1.56	0.119	0737118	.6456039
INC4	.0248964	.1647494	0.15	0.880	2986841	.3484768
INC5	.143163	.1553964	0.92	0.357	1620475	.4483734
INC6	.246488	.1570902	1.57	0.117	0620493	.5550252
INC7	.0784037	.1485653	0.53	0.598	2133899	.3701973
INC8	.0757106	.1493609	0.51	0.612	2176456	.3690668
INC9	.175628	.1670714	1.05	0.294	152513	.5037689
INC10	.3359523	.1698507	1.98	0.048	.0023524	.6695522
INC11	.0883139	.1539982	0.57	0.567	2141503	.3907781
INC12	.0116075	.1488893	0.08	0.938	2808225	.3040375
INC13	.2734599	.1678554	1.63	0.104	056221	.6031407
INC14	.0284433	.1451375	0.20	0.845	2566178	.3135044
INC15	0	(omitted)				
INC16	0292299	.1460221	-0.20	0.841	3160284	.2575687
_cons	.2708056	.1488962	1.82	0.069	0216379	.5632492
	1					

^{. *} Baseline, including control for past vote: Legislative 2018

Linear regression
Number of obs = 583 F(24, 558) = 43.03 Prob > F = 0.0000 R-squared = 0.6918

Root MSE = .21134

[.] regress vote_m5s_euro \${main} \${controls} vote_m5s_2018 if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2
note: EDU4 omitted because of collinearity.
note: INC15 omitted because of collinearity.

		Robust				
vote_m5s_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	004534	.0381897	-0.12	0.906	0795472	.0704791
dummy_diesel	.0197971	.0293282	0.68	0.500	03781	.0774043
dummy_euro_4	0034235	.0270838	-0.13	0.899	0566222	.0497751
age	0006305	.0009189	-0.69	0.493	0024355	.0011744
female	.0030019	.0182325	0.16	0.869	0328107	.0388146
EDU1	0777255	.0628011	-1.24	0.216	2010809	.0456299
EDU2	112606	.063843	-1.76	0.078	238008	.012796
EDU3	1169586	.0631815	-1.85	0.065	2410612	.007144
EDU4	0	(omitted)				
INC1	0850168	.0655032	-1.30	0.195	2136797	.0436462
INC2	.0266922	.0442443	0.60	0.547	0602134	.1135979
INC3	.0650597	.0431992	1.51	0.133	0197932	.1499125
INC4	1445319	.0860679	-1.68	0.094	3135885	.0245247
INC5	.0364216	.0606784	0.60	0.549	0827644	.1556076
INC6	.0350549	.0533388	0.66	0.511	0697145	.1398243
INC7	.013607	.044741	0.30	0.761	0742744	.1014885
INC8	.0103558	.0380316	0.27	0.785	0643468	.0850585
INC9	.0422468	.0732156	0.58	0.564	1015651	.1860588
INC10	.0423697	.0574011	0.74	0.461	0703788	.1551183
INC11	003951	.044431	-0.09	0.929	0912235	.0833215
INC12	0327719	.0474874	-0.69	0.490	1260478	.060504
INC13	.0867154	.071231	1.22	0.224	0531983	.2266291
INC14	0061858	.0396379	-0.16	0.876	0840436	.071672
INC15	0	(omitted)				
INC16	0248777	.0394115	-0.63	0.528	1022906	.0525353
vote_m5s_2018	.7704655	.041631	18.51	0.000	.6886928	.8522382
cons	.1423696	.0795151	1.79	0.074	0138158	.298555

. * Baseline, including control for past vote: Regional 2018

. regress vote_m5s_euro \${main} \${controls} vote_m5s_regional if target!=3 & target!=4 & no_answer_euro==0 & no_answer_euro==0 target!=4 target!=4

Linear regression Number of obs = 551F(24, 526) = 91.61

Prob > F = 0.0000 R-squared = 0.6801 Root MSE = .20909

		Robust				
vote_m5s_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0123189	.0390419	0.32	0.752	0643784	.0890161
dummy_diesel	.0247581	.0336865	0.73	0.463	0414185	.0909347
dummy_euro_4	0576437	.027329	-2.11	0.035	1113311	0039563
age	0013279	.0009097	-1.46	0.145	0031149	.0004592
female	.014286	.0194717	0.73	0.463	0239658	.0525379
EDU1	.0032673	.0300696	0.11	0.914	055804	.0623387
EDU2	0	(omitted)				
EDU3	0625311	.0218591	-2.86	0.004	1054729	0195893
EDU4	.1768029	.1852484	0.95	0.340	1871147	.5407205
INC1	0523078	.0704295	-0.74	0.458	1906655	.08605
INC2	0	(omitted)				
INC3	.0807172	.0818279	0.99	0.324	0800324	. 2414669
INC4	0874358	.0734764	-1.19	0.235	2317789	.0569074
INC5	.0437155	.0474206	0.92	0.357	0494416	.1368725
INC6	.1332921	.0617024	2.16	0.031	.0120787	. 2545056

TN07	0400=0				0.00.00	400=040
INC7	.019959	.0420583	0.47	0.635	0626639	.1025819
INC8	0081337	.0328768	-0.25	0.805	0727196	.0564522
INC9	.1147886	.0776137	1.48	0.140	0376824	.2672595
INC10	.0592778	.0476696	1.24	0.214	0343683	.1529239
INC11	.0242835	.0411589	0.59	0.555	0565724	.1051395
INC12	.0539522	.0652812	0.83	0.409	0742918	.1821962
INC13	.0255142	.0563491	0.45	0.651	0851827	.1362111
INC14	0125194	.0389894	-0.32	0.748	0891136	.0640747
INC15	0570238	.0379667	-1.50	0.134	1316088	.0175612
INC16	0303372	.0298736	-1.02	0.310	0890234	.028349
vote_m5s_regional	.8448969	.038395	22.01	0.000	.7694705	.9203232
_cons	.123243	.0559553	2.20	0.028	.0133197	.2331663

. * Baseline, including control for past vote: Municipal 2016

. regress vote_m5s_euro \${main} \${controls} vote_m5s_municipal if target!=3 & target!=4 & no_answer_euro==0 & no_answer_euro==

Linear regression

Number of obs = 533 F(24, 508) = 72.13 Prob > F = 0.0000 R-squared = 0.6852 Root MSE = .20736

		Robust				
vote_m5s_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	0108078	.0406046	-0.27	0.790	0905814	.0689658
dummy_diesel	.0219996	.031681	0.69	0.488	0402424	.0842415
dummy_euro_4	0244571	.0295863	-0.83	0.409	0825836	.0336695
age	0017321	.0009619	-1.80	0.072	0036219	.0001576
female	.0166315	.0210243	0.79	0.429	0246738	.0579369
EDU1	.0166349	.0305533	0.54	0.586	0433914	.0766613
EDU2	0	(omitted)				
EDU3	0273727	.0198061	-1.38	0.168	0662847	.0115393
EDU4	.1786925	.1557979	1.15	0.252	1273951	.48478
INC1	.0155634	.0641012	0.24	0.808	1103726	.1414994
INC2	0	(omitted)				
INC3	.1323113	.0949635	1.39	0.164	0542582	.3188809
INC4	0591825	.0705902	-0.84	0.402	1978671	.0795021
INC5	.0451808	.0576506	0.78	0.434	0680822	.1584438
INC6	.1570214	.063826	2.46	0.014	.031626	.2824168
INC7	.0371373	.0427522	0.87	0.385	0468556	.1211301
INC8	.0289558	.0414938	0.70	0.486	0525648	.1104763
INC9	.0870995	.071754	1.21	0.225	0538716	.2280707
INC10	.0189979	.0568551	0.33	0.738	0927022	.1306981
INC11	.0350073	.0422698	0.83	0.408	0480378	.1180524
INC12	0070848	.0344963	-0.21	0.837	0748577	.0606882
INC13	.0183237	.0483899	0.38	0.705	0767453	.1133926
INC14	0311636	.038905	-0.80	0.423	107598	.0452708
INC15	0443929	.0375847	-1.18	0.238	1182335	.0294477
INC16	0366492	.0352924	-1.04	0.300	1059863	.0326879
vote_m5s_municipal	.8076289	.0434895	18.57	0.000	.7221874	.8930704
_cons	.1060504	.0516584	2.05	0.041	.0045601	. 2075407

. . . ******** . * Table 4 *

. * Without controls

. regress vote_lega_euro \${main_placebo} if target!=1 & ((target==2 & class==6) | target==3) & target!=4 & no_answer

vote_lega_~o	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro5	0045225	.0757672	-0.06	0.952	1533904	.1443454
dummy_diesel	0887271	.0514334	-1.73	0.085	1897839	.0123297
dummy_euro_5	.0017553	.0538733	0.03	0.974	1040953	.1076059
_cons	.2751678	.0367355	7.49	0.000	.2029896	.347346

. * With controls

. regress vote_lega_euro \${main_placebo} \${controls} if target!=1 & ((target==2 & class==6) | target==3) & target!=4 note: EDU4 omitted because of collinearity. note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 495 F(23, 471) = 5.24 Prob > F = 0.0000 R-squared = 0.0932 Root MSE = .41475

vote lega ~o	Coefficient	Robust std. err.	t	P> t	[95% conf	interval]
Voce_lega_00	COETTICIENT	stu. err.		7/14	[95% COIII.	
diesel_euro5	0245217	.0774894	-0.32	0.752	1767895	.127746
dummy_diesel	0661468	.0558884	-1.18	0.237	1759683	.0436747
dummy_euro_5	.0177333	.0538013	0.33	0.742	0879871	.1234536
age	2.40e-06	.0016642	0.00	0.999	0032679	.0032727
female	.058915	.0390532	1.51	0.132	0178251	.1356551
EDU1	.3387144	.0694503	4.88	0.000	.2022436	.4751851
EDU2	.1576482	.0685792	2.30	0.022	.022889	.2924073
EDU3	.1770128	.0715148	2.48	0.014	.0364853	.3175403
EDU4	0	(omitted)				
INC1	.1571774	.2025042	0.78	0.438	2407462	.5551009
INC2	0	(omitted)				
INC3	.0630257	.1912607	0.33	0.742	3128043	.4388556
INC4	.2764242	.1857936	1.49	0.137	0886626	.6415111
INC5	.1206026	.1655303	0.73	0.467	2046667	.4458719
INC6	.1950878	.1698004	1.15	0.251	1385722	.5287478
INC7	.1553539	.1699139	0.91	0.361	1785293	.4892371
INC8	.2096881	.1767722	1.19	0.236	1376717	.5570479
INC9	.3048482	.1933594	1.58	0.116	0751057	.6848021
INC10	.3197418	.1815309	1.76	0.079	0369688	.6764524
INC11	.2039822	.1756657	1.16	0.246	1412033	.5491677
INC12	.1263779	.1733681	0.73	0.466	2142928	.4670487
INC13	.2703056	.1847356	1.46	0.144	0927024	.6333136

INC14	.0852682	.1642101	0.52	0.604	237407	.4079433
INC15	.2022837	.2278324	0.89	0.375	2454099	.6499773
INC16	.3614571	.1759701	2.05	0.041	.0156736	.7072406
_cons	1930923	.1977042	-0.98	0.329	5815837	.195399

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. ********** . * Table 5 *

. * Baseline

. regress vote_lega_euro compensated \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 602 F(24, 577) = 3.94 Prob > F = 0.0000 R-squared = 0.1454 Root MSE = .40565

		Robust				
vote_lega_~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
compensated	2536831	.0551045	-4.60	0.000	3619129	1454533
diesel_euro4	.2066188	.0789035	2.62	0.009	.0516458	.3615919
dummy_diesel	1097926	.0571696	-1.92	0.055	2220785	.0024932
dummy_euro_4	0476261	.0594185	-0.80	0.423	164329	.0690767
age	.001211	.0016872	0.72	0.473	0021028	.0045248
female	.1710155	.0392657	4.36	0.000	.0938944	.2481366
EDU1	.1917347	.0501053	3.83	0.000	.0933237	.2901457
EDU2	0	(omitted)				
EDU3	000288	.0405127	-0.01	0.994	0798584	.0792823
EDU4	.0009109	.1291488	0.01	0.994	2527481	.25457
INC1	1321492	.1999626	-0.66	0.509	5248926	.2605941
INC2	2190667	.2092116	-1.05	0.295	6299759	.1918424
INC3	185652	.1961351	-0.95	0.344	5708779	.1995738
INC4	.1054745	.2116482	0.50	0.618	3102204	.5211694
INC5	0644246	.1891128	-0.34	0.733	435858	.3070089
INC6	0451187	.1889371	-0.24	0.811	4162071	.3259697
INC7	0786047	.1845481	-0.43	0.670	4410727	.2838632
INC8	0809672	.1853576	-0.44	0.662	445025	.2830906
INC9	1465825	.1872386	-0.78	0.434	5143347	.2211697
INC10	.0401658	.1966381	0.20	0.838	3460479	.4263795
INC11	.0452547	.1953733	0.23	0.817	3384748	.4289842
INC12	.056468	.1936941	0.29	0.771	3239634	.4368994
INC13	.0035391	.1924831	0.02	0.985	3745139	.381592
INC14	.1496445	.1837503	0.81	0.416	2112566	.5105456
INC15	0	(omitted)				
INC16	.0786513	.1925028	0.41	0.683	2994404	.4567429
_cons	.0746624	.183906	0.41	0.685	2865444	.4358693

. * Switching from legislative

. regress sw_to_lega_18_19 compensated \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs = 483

F(24, 458) = **3.29** Prob > F = **0.0000** R-squared = **0.2350**

Root MSE = .2687

sw to 1~8 19	Coefficient	Robust std. err.	t	P> t	[95% conf.	intonvall
SW_CO_1~6_19	Coefficient	stu. em.		ראונו	[95% COIII.	Interval
compensated	1438501	.0519319	-2.77	0.006	2459045	0417957
diesel_euro4	.1651254	.0532187	3.10	0.002	.0605423	.2697085
dummy_diesel	0451719	.0404128	-1.12	0.264	1245895	.0342457
dummy_euro_4	0006281	.0371539	-0.02	0.987	0736413	.0723852
age	.001351	.0010654	1.27	0.205	0007427	.0034447
female	.1500537	.0282225	5.32	0.000	.094592	.2055153
EDU1	0	(omitted)				
EDU2	0162332	.0345141	-0.47	0.638	0840589	.0515925
EDU3	.0164158	.034147	0.48	0.631	0506884	.0835199
EDU4	.0788519	.1334576	0.59	0.555	1834133	.341117
INC1	.0339762	.0693932	0.49	0.625	1023923	.1703448
INC2	0	(omitted)				
INC3	.0783508	.0778484	1.01	0.315	0746335	.231335
INC4	.3016302	.1248254	2.42	0.016	.0563288	.5469317
INC5	.0381483	.0442776	0.86	0.389	0488642	.1251608
INC6	.0966239	.0567267	1.70	0.089	0148529	.2081007
INC7	0023187	.0355626	-0.07	0.948	0722048	.0675673
INC8	.0093434	.0349301	0.27	0.789	0592998	.0779865
INC9	.0464577	.0549637	0.85	0.398	0615546	.15447
INC10	.1122787	.0715768	1.57	0.117	028381	.2529383
INC11	.0362697	.0501559	0.72	0.470	0622945	.134834
INC12	.0438538	.0440429	1.00	0.320	0426974	.130405
INC13	.0241034	.0404141	0.60	0.551	0553166	.1035234
INC14	.2219016	.051016	4.35	0.000	.1216472	.322156
INC15	.0342942	.0592941	0.58	0.563	082228	.1508165
INC16	.0977057	.0620317	1.58	0.116	0241964	.2196079
_cons	1595185	.0705891	-2.26	0.024	2982371	0207999

^{. *} Switching from regional

note: INC4 omitted because of collinearity.

Linear regression Number of obs = 450F(24, 425) = 4.29

F(24, 425) = 4.29 Prob > F = 0.0000 R-squared = 0.3259 Root MSE = .26078

[.] regress sw_to_lega_reg_19 compensated ${\text{main}}$ \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU1 omitted because of collinearity.

		Robust				
sw_to_l~g_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
compensated	1813179	.0345886	-5.24	0.000	2493039	1133319
diesel_euro4	.1673126	.0560033	2.99	0.003	.0572346	.2773905
dummy_diesel	0115285	.0404951	-0.28	0.776	0911242	.0680671
dummy_euro_4	.004706	.0370587	0.13	0.899	0681351	.0775471
age	.0030795	.0008944	3.44	0.001	.0013215	.0048376
female	.1979188	.0306056	6.47	0.000	.1377616	.258076
EDU1	0	(omitted)				
EDU2	087796	.0348776	-2.52	0.012	1563501	019242
EDU3	0420225	.0353403	-1.19	0.235	1114861	.0274411
EDU4	.0892846	.1275966	0.70	0.484	1615144	.3400836
INC1	120409	.1331293	-0.90	0.366	3820828	.1412649
INC2	2110148	.1135747	-1.86	0.064	4342529	.0122232
INC3	160608	.1127608	-1.42	0.155	3822464	.0610303
INC4	0	(omitted)				
INC5	1787975	.1074015	-1.66	0.097	3899016	.0323067
INC6	1301146	.1186854	-1.10	0.274	363398	.1031689
INC7	1203122	.1108344	-1.09	0.278	3381641	.0975396
INC8	1281879	.1075783	-1.19	0.234	3396396	.0832638
INC9	1625968	.105862	-1.54	0.125	370675	.0454815
INC10	1116591	.1035836	-1.08	0.282	3152589	.0919408
INC11	1737437	.1087698	-1.60	0.111	3875374	.04005
INC12	1250707	.1054144	-1.19	0.236	3322693	.0821278
INC13	1378644	.1031177	-1.34	0.182	3405486	.0648198
INC14	.0678997	.104846	0.65	0.518	1381816	.273981
INC15	1469449	.1299612	-1.13	0.259	4023916	.1085019
INC16	127006	.1210399	-1.05	0.295	3649174	.1109054
_cons	0523971	.125783	-0.42	0.677	2996313	.1948372
_	I					

. * Switching from municipality

. regress sw_to_lega_16_19 compensated \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs =

Number of obs = 452 F(24, 427) = 3.83 Prob > F = 0.0000 R-squared = 0.2317 Root MSE = .30382

		Robust				
sw_to_l~6_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
compensated	1665132	.050092	-3.32	0.001	2649709	0680555
diesel_euro4	.2035352	.0667816	3.05	0.002	.0722736	.3347967
dummy_diesel	0346041	.0524736	-0.66	0.510	1377429	.0685347
dummy_euro_4	0380601	.0478593	-0.80	0.427	1321292	.056009
age	.0019015	.0012465	1.53	0.128	0005486	.0043516
female	.1817506	.0352137	5.16	0.000	.1125368	.2509645
EDU1	0	(omitted)				
EDU2	0994346	.0444044	-2.24	0.026	186713	0121563
EDU3	0572446	.0452703	-1.26	0.207	146225	.0317358
EDU4	0293098	.1405871	-0.21	0.835	3056387	.247019
INC1	.1054495	.1127477	0.94	0.350	1161601	.327059
INC2	0	(omitted)				
INC3	.0657336	.0448085	1.47	0.143	0223391	.1538063
INC4	.2211074	.1132798	1.95	0.052	0015481	.4437628
INC5	.1443635	.0663631	2.18	0.030	.0139245	.2748026

INC6	.1537129	.0745181	2.06	0.040	.007245	.3001808
INC7	.0866838	.062205	1.39	0.164	0355824	.20895
INC8	.0996759	.0584338	1.71	0.089	0151778	.2145297
INC9	.0981083	.0622349	1.58	0.116	0242165	.2204332
INC10	.1367598	.0735332	1.86	0.064	0077722	.2812919
INC11	.0335115	.046481	0.72	0.471	0578485	.1248715
INC12	.0766264	.0541323	1.42	0.158	0297726	.1830253
INC13	.0655735	.0489387	1.34	0.181	0306171	.1617642
INC14	.2781158	.0581676	4.78	0.000	.1637854	.3924462
INC15	.0465033	.152332	0.31	0.760	2529106	.3459172
INC16	.1120654	.0814948	1.38	0.170	0481155	.2722464
_cons	1635338	.0733815	-2.23	0.026	3077676	0193
	I					

********** . * Table S1-1 *

. regress vote_lega_euro \${main} \${controls} vote_lega_2018 vote_lega_regional vote_lega_municipal if target!=3 & ta > nswer_regional==0 & no_answer_municipal==0, robust note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression

Number of obs 515 F(26, 488) 85.18 Prob > F 0.0000 R-squared 0.6534 .25999 Root MSE

	Т					
		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.100259	.0467101	2.15	0.032	.0084811	.1920368
dummy_diesel	.004806	.0393905	0.12	0.903	0725901	.082202
dummy_euro_4	.0321173	.0320264	1.00	0.316	0308094	.0950439
age	.0029921	.0009075	3.30	0.001	.001209	.0047751
female	.174353	.0297458	5.86	0.000	.1159074	.2327986
EDU1	0150552	.1893595	-0.08	0.937	3871158	.3570054
EDU2	0270385	.1891341	-0.14	0.886	3986562	.3445793
EDU3	.0236278	.1907156	0.12	0.901	3510972	.3983528
EDU4	0	(omitted)				
INC1	.0129163	.0718124	0.18	0.857	1281834	.154016
INC2	0	(omitted)				
INC3	.0605248	.0428418	1.41	0.158	0236524	.1447019
INC4	.1896276	.108979	1.74	0.082	0244983	.4037536
INC5	034791	.0494214	-0.70	0.482	1318959	.062314
INC6	.0477563	.0583353	0.82	0.413	0668631	.1623757
INC7	0207436	.0534766	-0.39	0.698	1258164	.0843292
INC8	0080566	.0414493	-0.19	0.846	0894976	.0733845
INC9	0012328	.0573119	-0.02	0.983	1138414	.1113758
INC10	.0716461	.0528184	1.36	0.176	0321335	.1754258
INC11	0061397	.0501659	-0.12	0.903	1047076	.0924281
INC12	.0160326	.0443505	0.36	0.718	0711089	.103174
INC13	.0021757	.0413725	0.05	0.958	0791145	.0834658
INC14	.1648819	.0499073	3.30	0.001	.0668223	.2629416
INC15	.070224	.0970642	0.72	0.470	1204914	.2609395
INC16	.0475347	.0488887	0.97	0.331	0485237	.1435931
vote_lega_2018	.5009243	.0950567	5.27	0.000	.3141533	.6876952
vote_lega_regional	.4283746	.1087499	3.94	0.000	.2146988	.6420504
vote_lega_municipal	0232905	.1019766	-0.23	0.819	2236579	.1770768

. * PD

. regress vote_pd_euro \${main} \${controls} vote_pd_2018 vote_pd_regional vote_pd_municipal if target!=3 & target!=4
> gional==0 & no_answer_municipal==0, robust

note: **EDU4** omitted because of collinearity. note: **INC15** omitted because of collinearity.

Linear regression

Number of obs = 515 F(26, 488) = 164.18 Prob > F = 0.0000 R-squared = 0.7939 Root MSE = .21546

		Robust				
vote_pd_euro	Coefficient	std. err.	t	P> t	[95% conf.	. interval]
diesel_euro4	.017415	.0531682	0.33	0.743	0870519	.1218819
dummy_diesel	0349239	.0368441	-0.95	0.344	1073166	.0374689
dummy_euro_4	.0245181	.0474191	0.52	0.605	0686527	.1176888
age	.0025694	.0011662	2.20	0.028	.000278	.0048607
female	0073984	.0207312	-0.36	0.721	0481319	.0333351
EDU1	.1342895	.1050076	1.28	0.202	0720334	.3406124
EDU2	.1990937	.1087619	1.83	0.068	0146057	.412793
EDU3	.1931735	.107569	1.80	0.073	018182	.404529
EDU4	0	(omitted)				
INC1	2070871	.2133939	-0.97	0.332	6263713	.2121971
INC2	2444345	.2106711	-1.16	0.247	6583688	.1694999
INC3	2378999	.2092292	-1.14	0.256	6490012	.1732013
INC4	283183	.2034262	-1.39	0.165	6828824	.1165164
INC5	2781023	.2030971	-1.37	0.172	6771551	.1209504
INC6	2759308	.2052875	-1.34	0.180	6792872	.1274256
INC7	2441271	.2011332	-1.21	0.225	639321	.1510669
INC8	2273791	.1998658	-1.14	0.256	6200828	.1653246
INC9	2949516	.2200024	-1.34	0.181	7272205	.1373173
INC10	3045176	.2013439	-1.51	0.131	7001257	.0910905
INC11	243403	.1995184	-1.22	0.223	6354241	.1486182
INC12	2235915	.1995813	-1.12	0.263	6157362	.1685531
INC13	2336708	.2157248	-1.08	0.279	6575348	.1901933
INC14	3089168	.2005125	-1.54	0.124	7028912	.0850577
INC15	0	(omitted)				
INC16	2261214	.2046454	-1.10	0.270	6282163	.1759735
vote_pd_2018	.3774448	.0895692	4.21	0.000	.2014558	.5534338
<pre>vote_pd_regional</pre>	.2218785	.0903331	2.46	0.014	.0443887	. 3993684
<pre>vote_pd_municipal</pre>	.273133	.0884967	3.09	0.002	.0992514	.4470147
_cons	008657	.2274162	-0.04	0.970	4554929	.4381789

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. * Forza Italia

. regress vote_forzaitalia_euro \${main} \${controls} vote_forzaitalia_2018 vote_forzaitalia_regional vote_forzaitalia
> 0 & no_answer_2018==0 & no_answer_regional==0 & no_answer_municipal==0, robust

note: **EDU4** omitted because of collinearity. note: **INC2** omitted because of collinearity.

Linear regression

Number of obs = 515 F(26, 488) = 353.54 Prob > F = 0.0000 R-squared = 0.8458 Root MSE = .16196

-		Robust				
vote_forzaitalia_euro	Coefficient		t	P> t	[95% conf.	interval]
diesel_euro4	.0184553	.0285732	0.65	0.519	0376864	.074597
dummy_diesel	0411226	.0285182	-1.44	0.150	0971561	.0149109
dummy_euro_4	0127931	.0245233	-0.52	0.602	0609775	.0353912
age	.0005391	.0005489	0.98	0.326	0005393	.0016176
female	0233381	.0192207	-1.21	0.225	0611036	.0144274
EDU1	0106986	.0321699	-0.33	0.740	0739072	.0525101
EDU2	.0001079	.0345523	0.00	0.998	0677818	.0679976
EDU3	.0099266	.0322894	0.31	0.759	0535168	.07337
EDU4	0	(omitted)				
INC1	0543352	.0465041	-1.17	0.243	1457081	.0370377
INC2	0	(omitted)				
INC3	.0085156	.0248152	0.34	0.732	0402422	.0572734
INC4	0351057	.0322137	-1.09	0.276	0984003	.0281889
INC5	.001569	.0276349	0.06	0.955	0527291	.0558671
INC6	0226331	.0411579	-0.55	0.583	1035017	.0582355
INC7	.0034627	.0320614	0.11	0.914	0595328	.0664581
INC8	0171281	.0250123	-0.68	0.494	0662731	.0320169
INC9	0187541	.0258851	-0.72	0.469	0696141	.0321059
INC10	0050582	.0253776	-0.20	0.842	054921	.0448046
INC11	.0286653	.0414161	0.69	0.489	0527106	.1100413
INC12	.1011637	.0575617	1.76	0.079	0119357	.2142631
INC13	0202619	.0314971	-0.64	0.520	0821486	.0416248
INC14	.0707215	.0302338	2.34	0.020	.011317	.130126
INC15	0136241	.0304686	-0.45	0.655	0734899	.0462418
INC16	0724743	.0440965	-1.64	0.101	1591167	.0141681
vote_forzaitalia_2018	.6106362	.099379	6.14	0.000	.4153726	.8058998
vote_forzaitalia_regio~l	.0336882	.0883174	0.38	0.703	139841	.2072175
vote_forzaitalia_munic~l	.2663754	.0876224	3.04	0.002	.0942117	.4385391
_cons	.0157519	.044011	0.36	0.721	0707226	.1022264

^{* 5} Stan

. regress vote_m5s_euro \${main} \${controls} vote_m5s_2018 vote_m5s_regional vote_m5s_municipal if target!=3 & target
> r_regional==0 & no_answer_municipal==0, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 515 F(26, 488) = 158.78 Prob > F = 0.0000 R-squared = 0.7716 Root MSE = .17693

		Robust				
vote_m5s_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0229576	.0337405	0.68	0.497	0433369	.0892522
dummy_diesel	.0147053	.0282645	0.52	0.603	0408299	.0702404
dummy_euro_4	042862	.0231661	-1.85	0.065	0883797	.0026556
age	0003947	.0007976	-0.49	0.621	0019618	.0011725
female	.0073079	.0162656	0.45	0.653	0246515	.0392672
EDU1	1030411	.0779879	-1.32	0.187	2562746	.0501923
EDU2	1154583	.0772024	-1.50	0.135	2671484	.0362319
EDU3	1503021	.0761727	-1.97	0.049	2999691	0006352
EDU4	0	(omitted)				
INC1	000899	.0225442	-0.04	0.968	0451946	.0433966
INC2	0	(omitted)				
INC3	.0639333	.0604667	1.06	0.291	0548739	.1827405
INC4	1031508	.0807733	-1.28	0.202	2618572	.0555556
INC5	.0306328	.0518749	0.59	0.555	0712928	.1325585
INC6	.0881644	.0489789	1.80	0.072	0080712	.1844
INC7	.0167916	.0313389	0.54	0.592	0447842	.0783673
INC8	.0055452	.0240621	0.23	0.818	0417329	.0528233
INC9	.0659078	.0696896	0.95	0.345	0710209	.2028365
INC10	.0018416	.0407888	0.05	0.964	0783017	.0819848
INC11	.0155519	.0273392	0.57	0.570	0381651	.069269
INC12	.0095364	.0344825	0.28	0.782	0582161	.077289
INC13	.0214254	.0437405	0.49	0.624	0645176	.1073684
INC14	0103842	.0281978	-0.37	0.713	0657883	.0450199
INC15	0393177	.0227929	-1.72	0.085	084102	.0054667
INC16	0275718	.0221532	-1.24	0.214	0710992	.0159556
vote_m5s_2018	.3281813	.0940694	3.49	0.001	.1433502	.5130123
vote_m5s_regional	.3293957	.1183128	2.78	0.006	.0969303	.5618612
vote_m5s_municipal	.2723042	.1203303	2.26	0.024	.0358749	.5087336
_cons	.1554381	.0861082	1.81	0.072	0137506	.3246267

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. regress sw_to_lega_18_19 \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018==0, robust

Linear regression

Number of obs = 483 F(3, 479) = 5.06 Prob > F = 0.0019 R-squared = 0.0300 Root MSE = .29587

sw_to_l~8_19	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1162375	.0474483	2.45	0.015	.023005	.20947
dummy_diesel	0169704	.0322751	-0.53	0.599	0803886	.0464479
dummy_euro_4	0023135	.0351701	-0.07	0.948	0714201	.0667932
cons	.0549451	.0239871	2.29	0.022	.0078121	.102078

^{. *======*}

^{. *} From legislative 2018 *

^{. *=======*}

^{. *} Without controls

. * Including individual controls

. regress sw_to_lega_18_19 \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018==0, rob note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs 483

F(23, 459) 3.33 = Prob > F 0.0000

R-squared 0.2235 Root MSE .27042

-						
		Robust				
sw_to_1~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1514822	.0531084	2.85	0.005	.0471166	.2558479
dummy_diesel	0439853	.0404769	-1.09	0.278	1235283	.0355577
dummy_euro_4	0025755	.0373437	-0.07	0.945	0759613	.0708103
age	.0018651	.0010831	1.72	0.086	0002633	.0039936
female	.1614047	.0291034	5.55	0.000	.1042124	.218597
EDU1	0	(omitted)				
EDU2	0094208	.0332041	-0.28	0.777	0746717	.0558301
EDU3	.0245927	.0335838	0.73	0.464	0414043	.0905897
EDU4	.0874068	.1363619	0.64	0.522	1805641	.3553777
INC1	.0322451	.0655185	0.49	0.623	0965082	.1609985
INC2	0	(omitted)				
INC3	.0847396	.0796877	1.06	0.288	0718584	.2413375
INC4	.3036762	.125325	2.42	0.016	.0573944	.5499581
INC5	.0377168	.0454241	0.83	0.407	0515482	.1269819
INC6	.0961767	.0576723	1.67	0.096	0171578	.2095111
INC7	.0051461	.0370998	0.14	0.890	0677603	.0780526
INC8	.0075784	.0356499	0.21	0.832	0624789	.0776356
INC9	.0564252	.0564928	1.00	0.318	0545913	.1674417
INC10	.1014188	.0718958	1.41	0.159	039867	.2427046
INC11	.0376706	.0506835	0.74	0.458	0619299	.137271
INC12	.0426816	.0440471	0.97	0.333	0438772	.1292405
INC13	.0155368	.0399599	0.39	0.698	0629902	.0940638
INC14	.2063874	.0494206	4.18	0.000	.1092688	.3035059
INC15	.0331732	.0634538	0.52	0.601	0915227	.1578691
INC16	.0959931	.0624317	1.54	0.125	0266944	.2186805
_cons	1915342	.0724406	-2.64	0.008	3338905	049178

^{. *} Including unkown-car

. regress sw_to_lega_18_19 \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_answer_20 note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs 511 F(24, 486) 3.08 Prob > F 0.0000 R-squared 0.2125

Root MSE .26546

		Robust				
sw_to_lega_18_19	Coefficient	std. err.	t	P> t	[95% conf.	. interval]
diesel_euro4_ass	.1437451	.0517529	2.78	0.006	.0420581	. 2454321
dummy_diesel_ass	0466815	.0372051	-1.25	0.210	1197842	.0264211
dummy_euro_4_ass	0094446	.036968	-0.26	0.798	0820815	.0631923
age	.0019163	.0010199	1.88	0.061	0000876	.003920
female	.1482249	.0279148	5.31	0.000	.0933764	.2030734
EDU1	0	(omitted)				
EDU2	0096656	.0316129	-0.31	0.760	0717804	.0524493
EDU3	.0230464	.0319988	0.72	0.472	0398268	.085919
EDU4	.0875008	.1385611	0.63	0.528	1847519	.359753
INC1	.0367606	.064186	0.57	0.567	0893557	.1628769
INC2	0	(omitted)				
INC3	.0925857	.0739618	1.25	0.211	0527387	.2379
INC4	.2546332	.103177	2.47	0.014	.0519052	.4573612
INC5	.0399855	.0428526	0.93	0.351	0442136	.124184
INC6	.0962126	.0551386	1.74	0.082	0121269	.204552
INC7	.0139215	.0349068	0.40	0.690	0546654	.0825083
INC8	.013921	.0333211	0.42	0.676	0515501	.0793922
INC9	.0606685	.0510998	1.19	0.236	0397353	.161072
INC10	.0992109	.0693512	1.43	0.153	0370543	.235476
INC11	.0538349	.0466318	1.15	0.249	03779	.1454598
INC12	.0613748	.041508	1.48	0.140	0201824	.1429321
INC13	.0167214	.0370135	0.45	0.652	0560048	.089447
INC14	.207256	.0469879	4.41	0.000	.1149315	.2995804
INC15	.0305686	.0602863	0.51	0.612	0878854	.149022
INC16	.1114205	.0577838	1.93	0.054	0021164	.2249574
dummy_car_unknown	1844311	.0397818	-4.64	0.000	2625965	1062656
_cons	1816748	.0685881	-2.65	0.008	3164406	046909
	1					

^{. *=======*}

. regress sw_to_lega_reg_19 \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_regional==0, robust

Linear regression	Number of obs	=	450
	F(3, 446)	=	8.56
	Prob > F	=	0.0000
	R-squared	=	0.0489
	Root MSE	=	.30239

sw_to_l~g_19	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1151898	.0455485	2.53	0.012	.0256735	.204706
dummy_diesel	.025974	.0312546	0.83	0.406	0354505	.0873986
dummy_euro_4	.005276	.028439	0.19	0.853	050615	.061167
cons	.025974	.0182074	1.43	0.154	009809	.061757

^{. *} From regional 2018 *

^{. *===========}

^{. *} Without controls

. * Including individual controls

. regress sw_to_lega_reg_19 \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_regional==0 note: EDU1 omitted because of collinearity.

note: INC4 omitted because of collinearity.

Linear regression Number of obs = 450

F(23, 426) = 4.20Prob > F = 0.0000

R-squared = **0.3078** Root MSE = **.26397**

sw to l~g 19	Coefficient	Robust std. err.	t	P> t	[OF% conf	intervall
Sw_to_1~g_19	Coefficient	stu. em.		PYILI	[95% COIII.	Interval
diesel_euro4	.1506085	.055842	2.70	0.007	.0408485	.2603686
dummy_diesel	0098996	.040623	-0.24	0.808	089746	.0699469
dummy_euro_4	.0018974	.0375602	0.05	0.960	0719291	.0757239
age	.0037959	.0009058	4.19	0.000	.0020155	.0055763
female	.2164849	.0312395	6.93	0.000	.1550822	.2778877
EDU1	0	(omitted)				
EDU2	0734387	.0343466	-2.14	0.033	1409486	0059288
EDU3	0274401	.0346484	-0.79	0.429	0955433	.040663
EDU4	.1009306	.1342799	0.75	0.453	1630031	.3648644
INC1	1223376	.1356356	-0.90	0.368	3889359	.1442607
INC2	2158072	.1171754	-1.84	0.066	4461211	.0145066
INC3	1538549	.1153662	-1.33	0.183	3806127	.0729029
INC4	0	(omitted)				
INC5	1845408	.1109288	-1.66	0.097	4025767	.033495
INC6	1361027	.1216477	-1.12	0.264	3752072	.1030018
INC7	1175546	.1134476	-1.04	0.301	3405413	.105432
INC8	1376269	.1107416	-1.24	0.215	3552949	.080041
INC9	1661013	.1089573	-1.52	0.128	3802622	.0480596
INC10	1325327	.1067276	-1.24	0.215	3423109	.0772455
INC11	1759298	.1116698	-1.58	0.116	3954222	.0435626
INC12	1326356	.1081234	-1.23	0.221	3451574	.0798861
INC13	155812	.1062519	-1.47	0.143	3646552	.0530313
INC14	.0407618	.1075018	0.38	0.705	1705381	.2520617
INC15	1550935	.1359664	-1.14	0.255	422342	.112155
INC16	1354979	.1241962	-1.09	0.276	3796115	.1086157
_cons	0946414	.1273444	-0.74	0.458	344943	.1556602
	I .					

^{. *} Including unkown-car

. regress sw_to_lega_reg_19 \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_answer_r
note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression
Number of obs = 478 F(24, 453) = 3.83 Prob > F = 0.0000 R-squared = 0.2933

Root MSE = .25955

		B. b I				
sw_to_lega_reg_19	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
				. 1 - 1		
diesel_euro4_ass	.1525461	.0542934	2.81	0.005	.0458479	. 2592442
<pre>dummy_diesel_ass</pre>	0188188	.0373804	-0.50	0.615	0922792	.0546416
dummy_euro_4_ass	0142653	.0368852	-0.39	0.699	0867526	.058222
age	.0036225	.0008346	4.34	0.000	.0019823	.0052627
female	.2042839	.0302171	6.76	0.000	.1449007	.263667
EDU1	1001374	.1396051	-0.72	0.474	3744913	.1742165
EDU2	1702677	.1387323	-1.23	0.220	4429065	.1023711
EDU3	1273845	.1399459	-0.91	0.363	4024082	.1476391
EDU4	0	(omitted)				
INC1	.0947368	.0857435	1.10	0.270	0737675	.2632411
INC2	0	(omitted)				
INC3	.070391	.0481701	1.46	0.145	0242737	.1650557
INC4	.1924576	.0931364	2.07	0.039	.0094246	.3754906
INC5	.0346764	.037722	0.92	0.358	0394555	.1088083
INC6	.0832444	.060007	1.39	0.166	0346822	.201171
INC7	.0999105	.0561752	1.78	0.076	0104857	.2103068
INC8	.0828713	.0496381	1.67	0.096	0146782	.1804208
INC9	.0504569	.0380285	1.33	0.185	0242773	.1251912
INC10	.0801605	.0413534	1.94	0.053	0011079	.1614289
INC11	.066061	.0460268	1.44	0.152	0243914	.1565135
INC12	.1021269	.0476838	2.14	0.033	.0084179	.1958359
INC13	.0565478	.0424129	1.33	0.183	0268025	.1398982
INC14	.2523637	.0489189	5.16	0.000	.1562276	. 3484999
INC15	.0529661	.0910425	0.58	0.561	125952	.2318842
INC16	.103977	.0634249	1.64	0.102	0206665	.2286206
dummy car unknown	2077316	.0420838	-4.94	0.000	2904354	1250279
_cons	1857897	.1484914	-1.25	0.212	4776073	.1060278

^{. *=======*}

. regress sw_to_lega_16_19 \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_municipal==0, robust

Linear regression	Number of obs	=	452
_	F(3, 448)	=	5.47
	Prob > F	=	0.0011

R-squared = **0.0315** Root MSE = **.33302**

sw_to_1~6_19	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1416393	.0571263	2.48	0.014	.0293705	.2539081
dummy_diesel	0020243	.0434025	-0.05	0.963	087322	.0832734
dummy_euro_4	0313283	.0411291	-0.76	0.447	1121583	.0495017
cons	.0789474	.0310695	2.54	0.011	.0178873	.1400075

^{. *} From municipal 2016 *

^{. *=======*}

^{. *} Without controls

. * Including individual controls

. regress sw_to_lega_16_19 \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_municipal==0 note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs 452

F(23, 428) 3.86 = Prob > F 0.0000 R-squared 0.2185 Root MSE .30606

		Robust		- 1.1		
sw_to_l~6_19	Coefficient	std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.1859254	.0669761	2.78	0.006	.0542824	.3175684
dummy_diesel	0334801	.0526371	-0.64	0.525	1369395	.0699793
dummy_euro_4	0392449	.0482422	-0.81	0.416	1340659	.0555762
age	.0026523	.0012771	2.08	0.038	.0001422	.0051625
female	.1984297	.0360946	5.50	0.000	.1274851	.2693744
EDU1	0	(omitted)				
EDU2	0872332	.0433968	-2.01	0.045	1725306	0019358
EDU3	0441607	.0448433	-0.98	0.325	1323011	.0439798
EDU4	0159414	.1431951	-0.11	0.911	2973945	.2655116
INC1	.096417	.1084506	0.89	0.374	1167452	.3095791
INC2	0	(omitted)				
INC3	.0779006	.0457676	1.70	0.089	0120567	.1678579
INC4	.2249319	.1162964	1.93	0.054	0036513	.453515
INC5	.1415828	.0673763	2.10	0.036	.0091532	.2740124
INC6	.1540308	.0757078	2.03	0.043	.0052254	.3028361
INC7	.0949844	.0632868	1.50	0.134	0294072	.2193761
INC8	.0961769	.0593116	1.62	0.106	0204013	.2127551
INC9	.1008583	.0640888	1.57	0.116	0251096	.2268263
INC10	.1236821	.0739225	1.67	0.095	0216142	.2689784
INC11	.0362145	.0474426	0.76	0.446	0570349	.129464
INC12	.0750477	.053892	1.39	0.164	0308783	.1809736
INC13	.0545669	.0487678	1.12	0.264	0412873	.1504211
INC14	.2590435	.0568945	4.55	0.000	.147216	.370871
INC15	.0418679	.1663443	0.25	0.801	2850855	.3688213
INC16	.1113728	.0828623	1.34	0.180	051495	.274240
_cons	2106905	.0761256	-2.77	0.006	360317	061064

^{. *} Including unkown-car

. regress sw_to_lega_16_19 \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_answer_mu note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs 479 F(24, 454) 3.65 Prob > F 0.0000 R-squared 0.2137

Root MSE .29933

		Robust				
sw_to_lega_16_19	Coefficient		t	P> t	[95% conf.	interval]
diesel_euro4_ass	.1838481	.0640494	2.87	0.004	.057978	.3097181
dummy_diesel_ass	0386408	.0481353	-0.80	0.423	1332364	.0559549
dummy_euro_4_ass	0501638	.0469723	-1.07	0.286	1424739	.0421463
age	.0027081	.0011412	2.37	0.018	.0004655	.0049507
female	.186826	.0345525	5.41	0.000	.1189233	. 2547287
EDU1	.0131322	.1451687	0.09	0.928	2721537	. 2984182
EDU2	0705493	.1452673	-0.49	0.627	356029	. 2149304
EDU3	0284203	.1462206	-0.19	0.846	3157734	.2589329
EDU4	0	(omitted)				
INC1	.1013743	.1070203	0.95	0.344	1089424	.311691
INC2	0	(omitted)				
INC3	.0842832	.0415268	2.03	0.043	.0026747	.1658917
INC4	.1963249	.092019	2.13	0.033	.015489	.3771609
INC5	.1406377	.0648944	2.17	0.031	.013107	.2681685
INC6	.153335	.0727675	2.11	0.036	.0103322	. 2963378
INC7	.0958249	.0604941	1.58	0.114	0230583	.214708
INC8	.0993607	.0570904	1.74	0.082	0128334	. 2115549
INC9	.0963586	.0580205	1.66	0.097	0176636	.2103807
INC10	.1233451	.0702183	1.76	0.080	014648	.2613382
INC11	.0596866	.0451682	1.32	0.187	0290781	.1484513
INC12	.0919579	.0504549	1.82	0.069	0071962	.1911121
INC13	.0523002	.0461545	1.13	0.258	0384027	.1430031
INC14	.2573684	.054313	4.74	0.000	.1506323	. 3641045
INC15	.0365515	.1631755	0.22	0.823	2841215	. 3572246
INC16	.1306791	.0750747	1.74	0.082	0168579	.2782161
dummy_car_unknown	2004401	.0395662	-5.07	0.000	2781957	1226845
_cons	2146302	.1550613	-1.38	0.167	5193571	.0900966

^{. *} Figure 5 can be generated using data available in the file Data_Figure_5.xlsx

^{. *} To generate the figure one needs to use https://flourish.studio

^{. *} Specifically, the following steps have to be followed:

^{. * (1)} click on "New visualization"

^{. * (2)} select graph type "Sankey diagram"

^{. * (3)} select version "Alluvial"

^{. * (4)} a sample graph will pop up

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. * (5) click on "Data", and then "Upload data" using "Data_Figure_5.xlsx"
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- . * (6) select as "Source" column "A", and as "Target" column "B" $\,$
- . * (7) select as "Value of Link" column "C"
- . * (8) click on "Preview", and then on "Match data sheet"
- . * The same applies for Figures SI-2 and SI-3, starting from Data_Figure_SI_2.xlsx and Data_Figure_SI_3.xlsx, respe

. * Table SI-3 - Figure 6 *

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. * From municipal 2016 to legislative 2018 *

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. * Without controls

. regress sw_to_lega_16_18 \${main} if target!=3 & target!=4 & no_answer_2018==0 & no_answer_municipal==0, robust

sw_to_leg~18	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	0287834	.0449658	-0.64	0.522	1171524	.0595856
dummy_diesel	.0364757	.0374111	0.97	0.330	0370464	.1099978
dummy_euro_4	0087045	.0310839	-0.28	0.780	0697921	.0523832
cons	.0394737	.0224349	1.76	0.079	0046164	.0835638

. * Including individual controls

. regress sw_to_lega_16_18 ${\text{main}}$ \${controls} if target!=3 & target!=4 & no_answer_2018==0 & no_answer_municipal==0 note: **EDU1** omitted because of collinearity.

note: INC2 omitted because of collinearity.

sw_to_leg~18	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.017165	.045179	0.38	0.704	0716342	.1059642
dummy_diesel	.0246787	.0370628	0.67	0.506	0481681	.0975256
dummy_euro_4	0347584	.0337135	-1.03	0.303	1010222	.0315053
age	4.57e-06	.0008306	0.01	0.996	0016279	.0016371
female	.0424433	.0254603	1.67	0.096	0075987	.0924853
EDU1	0	(omitted)				
EDU2	0782763	.0337089	-2.32	0.021	144531	0120216
EDU3	0912267	.0315769	-2.89	0.004	153291	0291623
EDU4	117194	.0442434	-2.65	0.008	2041542	0302339

INC1	.0729273	.0688878	1.06	0.290	0624714	.2083261
INC2	0	(omitted)				
INC3	.0008096	.0240241	0.03	0.973	0464097	.048029
INC4	.0375475	.0242859	1.55	0.123	0101863	.0852813
INC5	.137209	.0578461	2.37	0.018	.0235127	.2509053
INC6	.0925822	.0532521	1.74	0.083	0120845	.197249
INC7	.100012	.046696	2.14	0.033	.0082311	.1917928
INC8	.1039318	.0480131	2.16	0.031	.0095622	.1983013
INC9	.0402416	.0247314	1.63	0.104	0083679	.088851
INC10	.0337366	.0240535	1.40	0.161	0135405	.0810136
INC11	.0458599	.0259251	1.77	0.078	0050959	.0968156
INC12	.044698	.0284123	1.57	0.116	0111463	.1005422
INC13	.0608557	.0281364	2.16	0.031	.0055538	.1161576
INC14	.0964136	.0319922	3.01	0.003	.033533	.1592942
INC15	003296	.0402867	-0.08	0.935	0824793	.0758873
INC16	.1040124	.0683295	1.52	0.129	030289	.2383139
_cons	.0086491	.0458118	0.19	0.850	0813939	.0986921

. * Including unkown-car

. regress sw_to_lega_16_18 \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_2018==0 & no_answer_mu note: EDU1 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs 480 F(24, 455) 0.88 Prob > F 0.6268 R-squared 0.0665 Root MSE .1983

		Robust				
sw_to_lega_16_18	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4_ass	.0217277	.0433132	0.50	0.616	063391	.1068463
dummy_diesel_ass	.0183479	.0340966	0.54	0.591	0486585	.0853542
dummy_euro_4_ass	0374091	.0326717	-1.14	0.253	1016153	.0267971
age	.0001762	.0007272	0.24	0.809	0012529	.0016052
female	.0418908	.02406	1.74	0.082	0053918	.0891734
EDU1	0	(omitted)				
EDU2	0762314	.0321979	-2.37	0.018	1395064	0129565
EDU3	0880762	.0306623	-2.87	0.004	1483334	0278189
EDU4	1170759	.0435098	-2.69	0.007	202581	0315708
INC1	.0747629	.0679638	1.10	0.272	0587989	.2083248
INC2	0	(omitted)				
INC3	.0057092	.0222318	0.26	0.797	0379806	.049399
INC4	.049116	.0218233	2.25	0.025	.0062291	.092003
INC5	.1358559	.0564589	2.41	0.017	.0249033	. 2468084
INC6	.0928985	.0518513	1.79	0.074	0089992	.1947961
INC7	.1008944	.0459446	2.20	0.029	.0106044	.1911844
INC8	.1038662	.0469174	2.21	0.027	.0116645	.1960679
INC9	.040908	.0227762	1.80	0.073	0038516	.0856676
INC10	.0348886	.0231958	1.50	0.133	0106955	.0804727
INC11	.0538273	.0255174	2.11	0.035	.0036807	.1039739
INC12	.0486889	.0271086	1.80	0.073	0045846	.1019625
INC13	.0614309	.0273542	2.25	0.025	.0076746	.1151871
INC14	.0966107	.0310432	3.11	0.002	.0356049	.1576165
INC15	0023744	.0438373	-0.05	0.957	088523	.0837742
INC16	.0978247	.061251	1.60	0.111	0225452	. 2181946
dummy_car_unknown	0473594	.0183348	-2.58	0.010	0833908	011328
_cons	.0017385	.0414154	0.04	0.967	0796507	.0831278

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. * From municipal 2016 to regional 2018 *

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. * Without controls

. regress sw_to_lega_16_reg \${main} if target!=3 & target!=4 & no_answer_regional==0 & no_answer_municipal==0, robus

Linear regression

Number of obs = 448 F(3, 444) = 0.41 Prob > F = 0.7452 R-squared = 0.0031 Root MSE = .20712

sw_to_lega~g	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	0022832	.0462091	-0.05	0.961	0930989	.0885326
dummy_diesel	.0101405	.0388784	0.26	0.794	066268	.086549
dummy_euro_4	0240253	.0343338	-0.70	0.484	0915023	.0434518
cons	.0547945	.0267558	2.05	0.041	.0022108	.1073782

. * Including individual controls

. regress sw_to_lega_16_reg \${main} \${controls} if target!=3 & target!=4 & no_answer_regional==0 & no_answer_municipunote: EDU4 omitted because of collinearity.

note: INC4 omitted because of collinearity.

Linear regression

Number of obs = 448 F(23, 424) = 0.87 Prob > F = 0.6392 R-squared = 0.0735 Root MSE = .20434

sw to lega~g	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.0360649	.0563403	0.64	0.522	0746762	.146806
dummy_diesel	.028705	.0416433	0.69	0.491	053148	.110558
dummy_euro_4	057254	.0384179	-1.49	0.137	1327674	.0182593
age	0022034	.0010884	-2.02	0.044	0043427	0000641
female	.0201276	.0251842	0.80	0.425	0293739	.0696291
EDU1	1267181	.1623063	-0.78	0.435	4457433	.192307
EDU2	1241521	.1612919	-0.77	0.442	4411834	.1928791
EDU3	1490061	.1589194	-0.94	0.349	461374	.1633618
EDU4	0	(omitted)				
INC1	.1296743	.0903878	1.43	0.152	0479896	.3073382
INC2	.0106339	.0298949	0.36	0.722	0481267	.069394
INC3	018629	.0245237	-0.76	0.448	0668323	.0295742
INC4	0	(omitted)				
INC5	.1338396	.0611618	2.19	0.029	.0136216	.2540576
INC6	.0620607	.0562952	1.10	0.271	0485918	.1727131
INC7	.0107307	.0383858	0.28	0.780	0647194	.0861809
INC8	.0354829	.0382737	0.93	0.354	0397468	.1107126
INC9	.0412279	.0535708	0.77	0.442	0640695	.1465252
INC10	.0348764	.0545083	0.64	0.523	0722637	.1420165
INC11	.0850501	.0666107	1.28	0.202	0458782	.2159784
INC12	0170758	.0227481	-0.75	0.453	0617889	.0276373
INC13	.0003888	.0243719	0.02	0.987	0475159	.0482935

INC14	0001993	.0326195	-0.01	0.995	0643153	.0639166
INC15	0289609	.0496204	-0.58	0.560	1264936	.0685717
INC16	.1519181	.0778473	1.95	0.052	0010966	.3049329
_cons	.2379456	.1617428	1.47	0.142	079972	.5558631

. * Including unkown-car

. regress sw_to_lega_16_reg \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_regional==0 & no_answer_regiona

Linear regression

Number of obs 474 F(24, 449) 0.84 Prob > F 0.6886 0.0691 R-squared Root MSE .19929

sw to lega 16 reg	Coefficient	Robust std. err.	t	P> t	[95% conf	. interval]
	COETTICIENT	stu. err.		7/14	[93% COIII.	. Incervar
diesel_euro4_ass	.0409513	.0545196	0.75	0.453	066194	.1480966
<pre>dummy_diesel_ass</pre>	.0207708	.0388197	0.54	0.593	0555201	.0970617
dummy_euro_4_ass	0563745	.0371419	-1.52	0.130	129368	.0166189
age	0018069	.0009949	-1.82	0.070	0037621	.0001482
female	.0194068	.0240931	0.81	0.421	0279424	.0667559
EDU1	1234517	.1619478	-0.76	0.446	4417214	.1948181
EDU2	12463	.1608457	-0.77	0.439	4407338	.1914738
EDU3	1459175	.1586855	-0.92	0.358	457776	.1659411
EDU4	0	(omitted)				
INC1	.1218535	.0847952	1.44	0.151	0447911	.2884982
INC2	0	(omitted)				
INC3	0144257	.0235488	-0.61	0.540	0607052	.0318538
INC4	.0038917	.0251537	0.15	0.877	0455419	.0533254
INC5	.1199438	.0591207	2.03	0.043	.0037561	.2361314
INC6	.051496	.0481089	1.07	0.285	0430505	.1460425
INC7	.0043647	.0418842	0.10	0.917	0779486	.0866781
INC8	.0276359	.0412072	0.67	0.503	0533471	.1086188
INC9	.0388694	.0486266	0.80	0.425	0566946	.1344335
INC10	.0291871	.0629955	0.46	0.643	0946155	.1529896
INC11	.0752667	.0587215	1.28	0.201	0401364	.1906698
INC12	0151	.0308431	-0.49	0.625	0757148	.0455148
INC13	0038067	.0291771	-0.13	0.896	0611473	.053534
INC14	0049323	.0339961	-0.15	0.885	0717435	.0618788
INC15	0346127	.0394271	-0.88	0.380	1120973	.042872
INC16	.1249634	.0699885	1.79	0.075	0125823	.2625092
dummy_car_unknown	0585482	.0262649	-2.23	0.026	1101655	0069308
_cons	.2271047	.1627461	1.40	0.164	0927339	.5469433

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. * Recycled materials

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. regress recycled_materials \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 705 F(23, 681) = 6.41 Prob > F = 0.0000 R-squared = 0.1460 Root MSE = .74415

		Robust				
recycled_m~s	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0108437	.1292791	0.08	0.933	2429898	.2646772
dummy_diesel	.0891133	.107548	0.83	0.408	1220523	.3002789
dummy_euro_4	.1212364	.0941056	1.29	0.198	0635355	.3060083
age	0075863	.0029582	-2.56	0.011	0133946	0017779
female	0127913	.0629644	-0.20	0.839	136419	.1108363
EDU1	1792736	.2615011	-0.69	0.493	6927188	.3341716
EDU2	109631	.2599327	-0.42	0.673	6199968	.4007347
EDU3	0052341	.2575552	-0.02	0.984	5109319	.5004637
EDU4	0	(omitted)				
INC1	.0214809	.2440217	0.09	0.930	4576444	.5006063
INC2	0	(omitted)				
INC3	.0592984	.2719178	0.22	0.827	4745995	.5931963
INC4	.4259932	.2601995	1.64	0.102	0848964	.9368828
INC5	.2008553	.2329962	0.86	0.389	2566219	.6583325
INC6	.0965158	.2208851	0.44	0.662	3371819	.5302135
INC7	.0071316	.2353705	0.03	0.976	4550075	.4692707
INC8	.0430919	.2382404	0.18	0.857	4246821	.5108658
INC9	.0438252	.2315971	0.19	0.850	4109049	.4985553
INC10	.1369898	.2846541	0.48	0.630	4219153	.6958949
INC11	.2621512	.2508723	1.04	0.296	2304249	.7547274
INC12	.29631	.2736956	1.08	0.279	2410786	.8336986
INC13	.3484173	.2647021	1.32	0.189	171313	.8681475
INC14	.5435689	.2291837	2.37	0.018	.0935774	.9935604
INC15	4011436	.3398642	-1.18	0.238	-1.068451	.2661639
INC16	.1276622	.2295001	0.56	0.578	3229505	.5782749
_cons	3.940152	.3781369	10.42	0.000	3.197697	4.682606

^{. *} Short showers

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. regress short_shower \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 705 F(23, 681) = 3.77 Prob > F = 0.0000 R-squared = 0.0968 Root MSE = .82474

		Robust				
short_shower	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.160258	.1502396	1.07	0.286	1347305	.4552466
dummy_diesel	0716561	.1086449	-0.66	0.510	2849752	.1416631
dummy_euro_4	.0024467	.1129155	0.02	0.983	2192576	.2241511
age	.0042714	.0031196	1.37	0.171	0018538	.0103966
female	.0048683	.0701518	0.07	0.945	1328715	.1426082
EDU1	.4650469	.3250613	1.43	0.153	1731959	1.10329
EDU2	.2303885	.3270569	0.70	0.481	4117726	.8725496
EDU3	.3416668	.3236125	1.06	0.291	2937313	.977065
EDU4	0	(omitted)				
INC1	0729574	.3219923	-0.23	0.821	7051743	.5592595
INC2	0	(omitted)				
INC3	0489824	.3302768	-0.15	0.882	6974655	.5995007
INC4	1923748	.329553	-0.58	0.560	8394368	.4546871
INC5	1201747	.2634889	-0.46	0.648	637523	.3971736
INC6	1844381	.2713788	-0.68	0.497	7172778	.3484016
INC7	2554354	.2709271	-0.94	0.346	7873881	.2765174
INC8	3679959	.2724002	-1.35	0.177	9028409	.1668492
INC9	3096889	.300906	-1.03	0.304	9005039	.2811261
INC10	0495765	.3063715	-0.16	0.871	6511227	.5519697
INC11	2352805	.3149982	-0.75	0.455	8537649	.3832039
INC12	.4588141	.2805036	1.64	0.102	0919417	1.00957
INC13	.2109967	.3136852	0.67	0.501	4049096	.8269029
INC14	.1797439	.2789329	0.64	0.520	3679278	.7274157
INC15	6656518	.4276226	-1.56	0.120	-1.505269	.1739653
INC16	358373	.2703329	-1.33	0.185	8891591	.1724131
_cons	3.545144	.4536929	7.81	0.000	2.654339	4.435949

. * Eco-mode

. regress eco_mode \${main} \${controls} if target!=3 & target!=4, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705F(23, 681) = 2.79

Prob > F = 0.0000 R-squared = 0.0812 Root MSE = .81049

		Robust				
eco_mode	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	2416154	.1466776	-1.65	0.100	52961	.0463792
dummy_diesel	.1416843	.1059268	1.34	0.181	0662982	.3496667
dummy_euro_4	.0801488	.1080954	0.74	0.459	1320914	.2923891
age	0015495	.0033528	-0.46	0.644	0081325	.0050335
female	0695701	.0693179	-1.00	0.316	2056725	.0665324
EDU1	0109089	.3293818	-0.03	0.974	6576348	.6358169
EDU2	.010856	.3264469	0.03	0.973	6301073	.6518193
EDU3	.1292534	.3237773	0.40	0.690	5064683	.7649751
EDU4	0	(omitted)				
INC1	.2673387	.2847366	0.94	0.348	2917284	.8264058
INC2	0	(omitted)				
INC3	.1618251	.2667032	0.61	0.544	3618342	.6854844
INC4	.1508939	.2348849	0.64	0.521	3102917	.6120795
INC5	.2852681	.2065957	1.38	0.168	120373	.6909092
INC6	.1580757	.2130864	0.74	0.458	2603095	.5764609
INC7	.1035687	.2123514	0.49	0.626	3133735	.5205109

INC8	.0553907	.2057751	0.27	0.788	3486392	.4594206
INC9	.1757533	. 2491454	0.71	0.481	3134321	.6649387
INC10	.0649494	.2791956	0.23	0.816	4832382	.613137
INC11	.1926103	.2354723	0.82	0.414	2697285	.6549492
INC12	.5464981	.2485686	2.20	0.028	.0584453	1.034551
INC13	.4761923	.2284356	2.08	0.037	.0276696	.924715
INC14	.5029904	.2125022	2.37	0.018	.0857523	.9202286
INC15	6343526	.3921699	-1.62	0.106	-1.40436	.1356548
INC16	.0684231	.2119493	0.32	0.747	3477295	.4845757
_cons	3.830906	.4181559	9.16	0.000	3.009876	4.651936

. * Bottles

. regress water_bottle \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705

F(23, 681) = 3.82 Prob > F = 0.0000 R-squared = 0.1568 Root MSE = 1.0281

		Robust				
water_bottle	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	0273303	.1848144	-0.15	0.882	3902047	.3355442
dummy_diesel	0412512	.1468916	-0.28	0.779	3296661	.2471637
dummy_euro_4	.2024465	.142483	1.42	0.156	0773123	.4822052
age	030991	.0043989	-7.05	0.000	0396279	022354
female	0066495	.0869264	-0.08	0.939	1773256	.1640265
EDU1	3835509	.3378835	-1.14	0.257	-1.04697	.2798677
EDU2	3415115	.3348523	-1.02	0.308	9989785	.3159555
EDU3	2487204	.3307795	-0.75	0.452	8981906	.4007498
EDU4	0	(omitted)				
INC1	0164456	.418242	-0.04	0.969	8376444	.8047531
INC2	0	(omitted)				
INC3	0875058	.44066	-0.20	0.843	9527212	.7777096
INC4	.1051554	.4287731	0.25	0.806	7367207	.947031
INC5	.0854831	.3825599	0.22	0.823	6656556	.8366217
INC6	.0649926	.3796885	0.17	0.864	6805082	.8104934
INC7	.0558593	.3729457	0.15	0.881	6764023	.7881209
INC8	2717362	.3881784	-0.70	0.484	-1.033906	.4904343
INC9	.0853493	.3984204	0.21	0.830	6969307	.8676292
INC10	0049807	.4204492	-0.01	0.991	8305132	.8205518
INC11	.1930586	.4011219	0.48	0.630	5945257	.9806428
INC12	.3788791	.395926	0.96	0.339	3985032	1.156262
INC13	.220979	.4237379	0.52	0.602	6110107	1.052969
INC14	.2169753	.3756772	0.58	0.564	5206495	.9546002
INC15	2038906	.5570095	-0.37	0.714	-1.297553	.8897717
INC16	0362852	.3809197	-0.10	0.924	7842033	.7116329
_cons	5.518566	.5499487	10.03	0.000	4.438767	6.598365

. * Climate neutrality

. regress climate_neutrality \${main} \${controls} if target!=3 & target!=4, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705

F(23, 681) = 2.30 Prob > F = 0.0005 R-squared = 0.0402 Root MSE = .28231

climate_ne~y	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
	0000574	0497090	0.10		104505	000000
diesel_euro4	0089574	.0487089	-0.18	0.854	104595	.0866802
dummy_diesel	041508	.0357047	-1.16	0.245	1116125	.0285966
dummy_euro_4	.0069288	.0352085	0.20	0.844	0622015	.0760591
age	0011111	.0009937	-1.12	0.264	0030621	.0008399
female	0209388	.0231397	-0.90	0.366	0663724	.0244949
EDU1	.0478372	.1560876	0.31	0.759	2586336	.354308
EDU2	.0947328	.155996	0.61	0.544	2115582	.4010238
EDU3	.0973601	.1558998	0.62	0.533	2087419	.4034621
EDU4	0	(omitted)				
INC1	.2161947	.1257282	1.72	0.086	0306667	.4630561
INC2	0	(omitted)				
INC3	.2029338	.1285725	1.58	0.115	0495123	.45538
INC4	.1943941	.130823	1.49	0.138	0624708	.451259
INC5	.1858703	.1202069	1.55	0.123	0501503	.421891
INC6	.1770798	.1221236	1.45	0.148	0627043	.4168639
INC7	.1843424	.1215991	1.52	0.130	0544119	.4230966
INC8	.230884	.1180414	1.96	0.051	0008847	.4626528
INC9	.2916984	.1142611	2.55	0.011	.067352	.5160448
INC10	.2462789	.1224008	2.01	0.045	.0059506	.4866071
INC11	.1903184	.1224275	1.55	0.121	0500622	.430699
INC12	.2199451	.1230542	1.79	0.074	0216662	.4615564
INC13	.150093	.1296659	1.16	0.247	1045	.4046861
INC14	.2361448	.119902	1.97	0.049	.0007228	.4715668
INC14	.187516	.1430302	1.31	0.190	0933172	.4683491
INC15	.1617903	.1201411	1.35	0.179	0741012	.3976818
	.7145879	.2025649	3.53	0.000	.3168611	1.112315
_cons	./1450/9	. 2023049	3.33	9.000	.2100011	1.112313

. * Climate transition positive for citizens

. regress green_policies_positive \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.

note: **INC2** omitted because of collinearity.

Linear regression Number of obs = 705F(23, 681) = 1.78

Prob > F = 0.0144 R-squared = 0.0819 Root MSE = .30409

green_poli~e Coefficient std. err. t P> t [95% conf. interval diesel_euro4 .0353323 .0542309 0.65 0.515 0711476 .14181 dummy_diesel .0866788 .0411503 -2.11 0.036 1674756 09588 dummy_euro_4 0024638 .03383978 -0.06 0.949 0778561 .07292 age 0016338 .001086 -1.50 0.133 0037662 .000499 female 0488674 .0256259 -1.91 0.057 0991826 .00144 EDU1 .1889977 .1908279 0.99 0.322 185684 .563679 EDU3 .2290361 .1897838 1.21 0.228 1435956 .60166 EDU4 0 (omitted) 1NC1 .3204389 .1411934 2.27 0.024 .0432122 .597669 INC2 0 (omitted) 1NC3 .3813249 .1377987 2.77 0.006 .1107636 .651880							
diesel_euro4 diesel_euro4 diesel_euro4 dummy_diesel 0866788 .0411503 -2.11			Robust				
dummy_diesel 0866788 .0411503 -2.11 0.036 1674756 005888 dummy_euro_4 0024638 .0383978 -0.06 0.949 0778561 .072928 age 0016338 .001086 -1.50 0.133 0037662 .000498 female 0488674 .0256259 -1.91 0.057 0991826 .00144 EDU1 .1889977 .1908279 0.99 0.322 185684 .563678 EDU2 .2526085 .1900559 1.33 0.184 1205575 .625776 EDU3 .2290361 .1897838 1.21 0.228 1435956 .60166 EDU4 0 (omitted) 1 1NC1 .3204389 .1411934 2.27 0.024 .0432122 .597669 INC2 0 (omitted) 1 1NC3 .3813249 .1377987 2.77 0.006 .1107636 .651880 INC3 .3442036 .1298356 2.65 0.008	green_poli~e	Coefficient	std. err.	t	P> t	[95% conf.	interval]
dummy_euro_4 0024638 .0383978 -0.06 0.949 0778561 .072928 age 0016338 .001086 -1.50 0.133 0037662 .000498 female 0488674 .0256259 -1.91 0.057 0991826 .00144 EDU1 .1889977 .1908279 0.99 0.322 185684 .563678 EDU2 .2526085 .1900559 1.33 0.184 1205575 .625774 EDU3 .2290361 .1897838 1.21 0.228 1435956 .60166 EDU4 0 (omitted)	diesel_euro4	.0353323	.0542309	0.65	0.515	0711476	.1418121
age	dummy_diesel	0866788	.0411503	-2.11	0.036	1674756	0058821
female 0488674 .0256259 -1.91 0.057 0991826 .00144 EDU1 .1889977 .1908279 0.99 0.322 185684 .563679 EDU2 .2526085 .1900559 1.33 0.184 1205575 .625774 EDU3 .2290361 .1897838 1.21 0.228 1435956 .60166 EDU4 0 (omitted) 0 .0024 .0432122 .59766 INC1 .3204389 .1411934 2.27 0.024 .0432122 .59766 INC2 0 (omitted)	dummy_euro_4	0024638	.0383978	-0.06	0.949	0778561	.0729286
EDU1 .1889977 .1908279 0.99 0.322185684 .563679 EDU2 .2526085 .1900559 1.33 0.1841205575 .625774 EDU3 .2290361 .1897838 1.21 0.2281435956 .601669 EDU4 0 (omitted) INC1 .3204389 .1411934 2.27 0.024 .0432122 .597669 INC2 0 (omitted) INC3 .3813249 .1377987 2.77 0.006 .1107636 .651880 INC4 .3227907 .1436797 2.25 0.025 .0406824 .604899 INC5 .3442036 .1298356 2.65 0.008 .0892774 .599129 INC6 .3236281 .1329906 2.43 0.015 .0625072 .584740 INC7 .3370817 .1309777 2.57 0.010 .0799131 .594250 INC8 .3830584 .1289955 2.97 0.003 .1297816 .636331 INC9 .3857197 .1341377 2.88 0.004 .1223466 .649099 INC10 .4232245 .1321952 3.20 0.001 .1636654 .682780 INC11 .4118606 .1277857 3.22 0.001 .1636654 .682780 INC12 .3907189 .1325213 2.95 0.003 .1305195 .650910 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591390 INC14 .422897 .1300364 3.25 0.001 .1675765 .678210 INC15 .2766821 .1717271 1.61 0.1080604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .547790	age	0016338	.001086	-1.50	0.133	0037662	.0004986
EDU2 .2526085 .1900559 1.33 0.1841205575 .625774 EDU3 .2290361 .1897838 1.21 0.2281435956 .60166 EDU4 0 (omitted) INC1 .3204389 .1411934 2.27 0.024 .0432122 .597669 INC2 0 (omitted) INC3 .3813249 .1377987 2.77 0.006 .1107636 .651880 INC4 .3227907 .1436797 2.25 0.025 .0406824 .604899 INC5 .3442036 .1298356 2.65 0.008 .0892774 .599129 INC6 .3236281 .1329906 2.43 0.015 .0625072 .584740 INC7 .3370817 .1309777 2.57 0.010 .0799131 .594250 INC8 .3830584 .1289955 2.97 0.003 .1297816 .636331 INC9 .3857197 .1341377 2.88 0.004 .1223466 .649099 INC10 .4232245 .1321952 3.20 0.001 .1636654 .682780 INC11 .4118606 .1277857 3.22 0.001 .1636654 .682780 INC12 .3907189 .1325213 2.95 0.003 .1305195 .650910 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591390 INC14 .422897 .1300364 3.25 0.001 .1675765 .678210 INC15 .2766821 .1717271 1.61 0.1080604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .5477900	female	0488674	.0256259	-1.91	0.057	0991826	.0014478
EDU3	EDU1	.1889977	.1908279	0.99	0.322	185684	.5636795
EDU4	EDU2	.2526085	.1900559	1.33	0.184	1205575	.6257745
INC1	EDU3	.2290361	.1897838	1.21	0.228	1435956	.6016677
INC2	EDU4	0	(omitted)				
INC3	INC1	.3204389	.1411934	2.27	0.024	.0432122	.5976657
INC4 .3227907 .1436797 2.25 0.025 .0406824 .604899 INC5 .3442036 .1298356 2.65 0.008 .0892774 .599129 INC6 .3236281 .1329906 2.43 0.015 .0625072 .584744 INC7 .3370817 .1309777 2.57 0.010 .0799131 .594250 INC8 .3830584 .1289955 2.97 0.003 .1297816 .636333 INC9 .3857197 .1341377 2.88 0.004 .1223466 .649093 INC10 .4232245 .1321952 3.20 0.001 .1636654 .682783 INC11 .4118606 .1277857 3.22 0.001 .1609593 .662763 INC12 .3907189 .1325213 2.95 0.003 .1305195 .650913 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591393 INC14 .422897 .1300364 3.25 0.001 .1675765 .67821 INC15 .2766821 .1717271 1.61 <t< td=""><td>INC2</td><td>0</td><td>(omitted)</td><td></td><td></td><td></td><td></td></t<>	INC2	0	(omitted)				
INC5	INC3	.3813249	.1377987	2.77	0.006	.1107636	.6518863
INC6 .3236281 .1329906 2.43 0.015 .0625072 .584744 INC7 .3370817 .1309777 2.57 0.010 .0799131 .594250 INC8 .3830584 .1289955 2.97 0.003 .1297816 .636333 INC9 .3857197 .1341377 2.88 0.004 .1223466 .649093 INC10 .4232245 .1321952 3.20 0.001 .1636654 .682783 INC11 .4118606 .1277857 3.22 0.001 .1609593 .662763 INC12 .3907189 .1325213 2.95 0.003 .1305195 .650913 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591393 INC14 .422897 .1300364 3.25 0.001 .1675765 .67821 INC15 .2766821 .1717271 1.61 0.108 0604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .547793	INC4	.3227907	.1436797	2.25	0.025	.0406824	.6048993
INC7 .3370817 .1309777 2.57 0.010 .0799131 .594250 INC8 .3830584 .1289955 2.97 0.003 .1297816 .636333 INC9 .3857197 .1341377 2.88 0.004 .1223466 .649093 INC10 .4232245 .1321952 3.20 0.001 .1636654 .682783 INC11 .4118606 .1277857 3.22 0.001 .1609593 .662763 INC12 .3907189 .1325213 2.95 0.003 .1305195 .650913 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591393 INC14 .422897 .1300364 3.25 0.001 .1675765 .67821 INC15 .2766821 .1717271 1.61 0.108 0604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .547793	INC5	.3442036	.1298356	2.65	0.008	.0892774	.5991298
INC8	INC6	.3236281	.1329906	2.43	0.015	.0625072	.5847489
INC9 .3857197 .1341377 2.88 0.004 .1223466 .649099 INC10 .4232245 .1321952 3.20 0.001 .1636654 .68278 INC11 .4118606 .1277857 3.22 0.001 .1609593 .66276 INC12 .3907189 .1325213 2.95 0.003 .1305195 .65091 INC13 .3168554 .1398265 2.27 0.024 .0423126 .59139 INC14 .422897 .1300364 3.25 0.001 .1675765 .67821 INC15 .2766821 .1717271 1.61 0.108 0604962 .613866 INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779	INC7	.3370817	.1309777	2.57	0.010	.0799131	.5942503
INC10 .4232245 .1321952 3.20 0.001 .1636654 .68278: INC11 .4118606 .1277857 3.22 0.001 .1609593 .66276: INC12 .3907189 .1325213 2.95 0.003 .1305195 .65091: INC13 .3168554 .1398265 2.27 0.024 .0423126 .59139: INC14 .422897 .1300364 3.25 0.001 .1675765 .67821: INC15 .2766821 .1717271 1.61 0.1080604962 .61386: INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779:	INC8	.3830584	.1289955	2.97	0.003	.1297816	.6363353
INC11 .4118606 .1277857 3.22 0.001 .1609593 .66276: INC12 .3907189 .1325213 2.95 0.003 .1305195 .65091: INC13 .3168554 .1398265 2.27 0.024 .0423126 .59139: INC14 .422897 .1300364 3.25 0.001 .1675765 .67821: INC15 .2766821 .1717271 1.61 0.1080604962 .61386: INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779:	INC9	.3857197	.1341377	2.88	0.004	.1223466	.6490928
INC12 .3907189 .1325213 2.95 0.003 .1305195 .650918 INC13 .3168554 .1398265 2.27 0.024 .0423126 .591398 INC14 .422897 .1300364 3.25 0.001 .1675765 .678218 INC15 .2766821 .1717271 1.61 0.1080604962 .613866 INC16 .2888349 .1318882 2.19 0.029 .0298786 .5477988	INC10	.4232245	.1321952	3.20	0.001	.1636654	.6827836
INC13 .3168554 .1398265 2.27 0.024 .0423126 .591393 INC14 .422897 .1300364 3.25 0.001 .1675765 .678213 INC15 .2766821 .1717271 1.61 0.1080604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .547793	INC11	.4118606	.1277857	3.22	0.001	.1609593	.6627618
INC14 .422897 .1300364 3.25 0.001 .1675765 .67821 INC15 .2766821 .1717271 1.61 0.1080604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779	INC12	.3907189	.1325213	2.95	0.003	.1305195	.6509183
INC15 .2766821 .1717271 1.61 0.1080604962 .613860 INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779	INC13	.3168554	.1398265	2.27	0.024	.0423126	.5913983
INC16 .2888349 .1318882 2.19 0.029 .0298786 .54779	INC14	.422897	.1300364	3.25	0.001	.1675765	.6782174
	INC15	.2766821	.1717271	1.61	0.108	0604962	.6138604
_cons .4470781 .2347877 1.90 0.0570139166 .90807	INC16	.2888349	.1318882	2.19	0.029	.0298786	.5477912
	_cons	.4470781	.2347877	1.90	0.057	0139166	.9080729

. * Clicked on website

. regress dummy_zeroco2_click \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705F(23, 681) = 2.73

Number of obs = 705 F(23, 681) = 2.73 Prob > F = 0.0000 R-squared = 0.0474 Root MSE = .28125

		Robust				
dummy_zero~k	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel euro4	0402373	.046319	-0.87	0.385	1311824	.0507079
dummy diesel	.055747	.0364635	1.53	0.127	0158474	.1273415
dummy_euro_4	.0129951	.0312414	0.42	0.678	048346	.0743362
age	0014965	.00092	-1.63	0.104	0033029	.0003099
female	0088005	.0255325	-0.34	0.730	0589323	.0413313
EDU1	.114066	.0344688	3.31	0.001	.0463881	.181744
EDU2	.0487782	.0338052	1.44	0.150	0175967	.115153
EDU3	.0334636	.0318737	1.05	0.294	029119	.0960462
EDU4	0	(omitted)				
INC1	1105298	.080886	-1.37	0.172	2693458	.0482861
INC2	0	(omitted)				

INC3	.0590321	.1141473	0.52	0.605	1650909	.2831551
INC4	0184565	.1031332	-0.18	0.858	2209538	.1840407
INC5	0409731	.0887867	-0.46	0.645	2153016	.1333554
INC6	.0033572	.0941981	0.04	0.972	1815964	.1883108
INC7	0176518	.0907513	-0.19	0.846	1958379	.1605342
INC8	0615055	.0877265	-0.70	0.483	2337525	.1107414
INC9	082651	.0895799	-0.92	0.357	2585371	.0932351
INC10	1090537	.0849639	-1.28	0.200	2758764	.057769
INC11	0038539	.0945759	-0.04	0.968	1895493	.1818414
INC12	1063879	.0851351	-1.25	0.212	2735468	.060771
INC13	.1062452	.1119665	0.95	0.343	1135959	.3260862
INC14	.0365885	.0888732	0.41	0.681	13791	.2110869
INC15	1169999	.0842656	-1.39	0.165	2824516	.0484518
INC16	.0042384	.0920231	0.05	0.963	1764447	.1849215
_cons	.0897905	.1056565	0.85	0.396	1176611	.2972421

. * Watched video

. regress dummy_watch_video \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs 705 F(23, 681) 7.85 Prob > F = 0.0000 R-squared Root MSE 0.1571 Root MSE .43941

		Robust				
dummy_watc~o	Coefficient	std. err.	t	P> t	[95% conf.	. interval]
diesel_euro4	0454322	.0758283	-0.60	0.549	1943175	.1034531
dummy_diesel	0360598	.0556509	-0.65	0.517	1453278	.0732082
dummy_euro_4	0717763	.0561216	-1.28	0.201	1819686	.0384159
age	.0010214	.0015637	0.65	0.514	0020488	.0040916
female	.1036117	.0385568	2.69	0.007	.0279071	.1793162
EDU1	.0783756	.1871328	0.42	0.675	2890511	.4458022
EDU2	.0909902	.1873049	0.49	0.627	2767743	.4587547
EDU3	.0566248	.1869019	0.30	0.762	3103485	.4235981
EDU4	0	(omitted)				
INC1	.2743691	.1494869	1.84	0.067	0191415	.5678797
INC2	0	(omitted)				
INC3	.1144751	.1429598	0.80	0.424	1662199	.3951701
INC4	.0608335	.1436045	0.42	0.672	2211273	.3427943
INC5	.1393792	.1248571	1.12	0.265	105772	.3845303
INC6	.1338991	.1244256	1.08	0.282	1104048	.378203
INC7	0464083	.1161801	-0.40	0.690	2745226	.181706
INC8	.0211766	.1189788	0.18	0.859	2124328	. 254786
INC9	.0782754	.1345825	0.58	0.561	1859712	.3425219
INC10	.1302819	.1411089	0.92	0.356	1467788	.4073426
INC11	.1868835	.1313481	1.42	0.155	0710123	.4447794
INC12	.0341498	.1323599	0.26	0.796	2257328	.2940323
INC13	.234422	.1397078	1.68	0.094	0398878	.5087317
INC14	.4914303	.1214146	4.05	0.000	.2530385	.7298222
INC15	0712389	.1593312	-0.45	0.655	3840783	.2416005
INC16	.1155307	.1227597	0.94	0.347	1255023	.3565637
_cons	.0819615	.2377152	0.34	0.730	3847812	.5487042

. * Social media

•

. regress dummy_social \${main} \${controls} if target!=3 & target!=4, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705

F(23, 681) = 8.65 Prob > F = 0.0000 R-squared = 0.1839 Root MSE = .44893

		Robust				
dummy_social	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1407076	.0799246	1.76	0.079	0162206	.2976359
dummy_diesel	.0192343	.0592304	0.32	0.745	0970617	.1355304
dummy_euro_4	.0393676	.0590395	0.67	0.505	0765537	.1552888
age	0124496	.0015425	-8.07	0.000	0154782	0094209
female	1072664	.0381299	-2.81	0.005	1821327	0324001
EDU1	.3546741	.227638	1.56	0.120	0922826	.8016307
EDU2	.3936459	.2263638	1.74	0.082	050809	.8381007
EDU3	.3952869	.2242932	1.76	0.078	0451024	.8356762
EDU4	0	(omitted)				
INC1	1981903	.1657516	-1.20	0.232	5236359	.1272554
INC2	0	(omitted)				
INC3	2686688	.1680465	-1.60	0.110	5986203	.0612828
INC4	2989241	.1587125	-1.88	0.060	6105487	.0127005
INC5	0365418	.1374152	-0.27	0.790	3063502	.2332667
INC6	.0064161	.1373116	0.05	0.963	2631889	.2760212
INC7	1726491	.1390667	-1.24	0.215	4457002	.100402
INC8	1239229	.1382415	-0.90	0.370	3953536	.1475078
INC9	2372341	.1545263	-1.54	0.125	5406392	.0661711
INC10	2184776	.1556278	-1.40	0.161	5240456	.0870904
INC11	1625923	.142184	-1.14	0.253	4417641	.1165795
INC12	1114092	.1483492	-0.75	0.453	402686	.1798676
INC13	0094351	.1482383	-0.06	0.949	3004942	.2816239
INC14	1563203	.1378498	-1.13	0.257	4269819	.1143413
INC15	1944253	.1888749	-1.03	0.304	5652725	.1764219
INC16	2482991	.1411997	-1.76	0.079	5255382	.0289399
_cons	.8948214	.2646385	3.38	0.001	.375216	1.414427

. * Podcast

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note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 705

F(23, 681) = 18.88 Prob > F = 0.0000 R-squared = 0.1868 Root MSE = .45766

[.] regress dummy_podcast \${main} \${controls} if target!=3 & target!=4, robust

		Robust				
dummy_podc~t	Coefficient	std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.2195355	.0824521	2.66	0.008	.0576446	.3814263
dummy_diesel	0214624	.0596762	-0.36	0.719	1386338	.0957091
dummy_euro_4	0166512	.061065	-0.27	0.785	1365495	.1032472
age	007395	.0016259	-4.55	0.000	0105873	0042027
female	1204622	.0397404	-3.03	0.003	1984907	0424338
EDU1	.4089942	.0645348	6.34	0.000	.282283	.5357053
EDU2	.5729875	.0650329	8.81	0.000	.4452983	.7006766
EDU3	.4995579	.063891	7.82	0.000	.3741108	.6250049
EDU4	0	(omitted)				
INC1	1316338	.1637608	-0.80	0.422	4531705	.1899029
INC2	0	(omitted)				
INC3	1193319	.1619589	-0.74	0.461	4373306	.1986668
INC4	1487367	.1569895	-0.95	0.344	4569784	.159505
INC5	0607615	.1346294	-0.45	0.652	3251	.203577
INC6	0353662	.1379417	-0.26	0.798	3062082	.2354759
INC7	0550038	.1335758	-0.41	0.681	3172737	.2072661
INC8	.0106524	.1348851	0.08	0.937	2541883	. 2754931
INC9	1475578	.1465043	-1.01	0.314	4352122	.1400965
INC10	1754899	.1554036	-1.13	0.259	4806176	.1296378
INC11	009746	.1412644	-0.07	0.945	2871121	.26762
INC12	.035698	.1448328	0.25	0.805	2486744	.3200704
INC13	.0645152	.1436572	0.45	0.654	217549	.3465794
INC14	0219954	.1350107	-0.16	0.871	2870827	.2430918
INC15	3659301	.1787848	-2.05	0.041	7169657	0148945
INC16	1956745	.1352171	-1.45	0.148	461167	.069818
_cons	.4289969	.1635358	2.62	0.009	.107902	.7500917

. * Buy tree

. regress dummy_buy \${main} \${controls} if target!=3 & target!=4, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs

Number of obs = 705 F(23, 681) = 21.85 Prob > F = 0.0000 R-squared = 0.2974 Root MSE = .4211

		Robust				
dummy_buy	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel euro4	.3732277	.0777729	4.80	0.000	.2205242	.5259312
dummy diesel	0647061	.0556414	-1.16	0.245	1739555	.0445433
dummy euro 4	0703876	.0584901	-1.20	0.229	1852301	.044455
age	0058768	.0016087	-3.65	0.000	0090355	0027182
female	0257527	.0354258	-0.73	0.468	0953097	.0438043
EDU1	0205038	.199902	-0.10	0.918	413002	.3719945
EDU2	.031067	.1999527	0.16	0.877	3615309	.4236649
EDU3	.0381247	.1985413	0.19	0.848	3517018	.4279513
EDU4	0	(omitted)				
INC1	.0556699	.1400506	0.40	0.691	2193129	.3306527
INC2	0	(omitted)				
INC3	1198604	.1336662	-0.90	0.370	3823078	.142587
INC4	.3029336	.1428599	2.12	0.034	.0224348	.5834323
INC5	.2093331	.1156545	1.81	0.071	0177492	.4364154
INC6	.175857	.1141283	1.54	0.124	0482286	.3999426
INC7	.2042112	.1113987	1.83	0.067	0145148	.4229373

INC8	.2405894	.1119537	2.15	0.032	.0207735	.4604053
INC9	.1062873	.129258	0.82	0.411	1475048	.3600794
INC10	.1792188	.1354765	1.32	0.186	0867831	.4452207
INC11	.2342258	.1198701	1.95	0.051	0011337	.4695852
INC12	.3391731	.1224882	2.77	0.006	.0986731	.579673
INC13	.243663	.1294005	1.88	0.060	0104088	.4977349
INC14	.471467	.108317	4.35	0.000	.2587915	.6841425
INC15	0137293	.1609901	-0.09	0.932	3298258	.3023672
INC16	.0125659	.114447	0.11	0.913	2121454	.2372773
_cons	.5415242	.2349179	2.31	0.021	.0802739	1.002775

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. * Clicked on website

. regress dummy_genitori_click $\{\{\{\}\}\}\}$ if target!=3 & target!=4, robust note: EDU4 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression

dummy_geni~k Coefficient std. err. t P> t [95% conf. interded] diesel_euro4 dummy_diesel .0155633 .0274279 0.57 0.571 0382901 .069 def dummy_diesel .0289839 .0233145 -1.24 0.214 0747609 .016 def dummy_euro_4 .0342839 .0200849 -1.71 0.088 .0737198 .005 def female 0043654 .0004052 -1.40 0.163 001361 .000 def female 0114266 .0129675 -0.88 0.379 0368876 .014 def EDU1 .0155298 .0145169 1.07 0.285 0129735 .04 def EDU3 001243 .0134939 0.48 0.631 0199322 .032 def EDU4 0 (omitted) 0 (omitted) 0 .0287 0070682 .023 def INC2 0 (omitted) .0040552 .0436408 1.06 0.292 0396315 .131 def .131 def <th></th> <th></th> <th>Robust</th> <th></th> <th></th> <th></th> <th></th>			Robust				
dummy_diesel 0289839 .0233145 -1.24 0.214 0747609 .016 dummy_euro_4 0342839 .0200849 -1.71 0.088 0737198 .005 age 0005654 .0004052 -1.40 0.163 001361 .000 female 0114266 .0129675 -0.88 0.379 0368876 .014 EDU1 .0155298 .0145169 1.07 0.285 0129735 .04 EDU2 .0064643 .0134439 0.48 0.631 0199322 .032 EDU3 001243 .0138935 -0.09 0.929 0285222 .026 EDU4 0 (omitted) 0 .0083983 .0078772 1.07 0.287 0070682 .023 INC2 0 (omitted) .0108534 .0398074 2.53 0.012 .0226934 .179 INC5 .1008534 .0398074 2.53 0.012 .0226934 .179 INC6 <t< th=""><th>dummy_geni~k</th><th>Coefficient</th><th></th><th>t</th><th>P> t </th><th>[95% conf.</th><th>interval]</th></t<>	dummy_geni~k	Coefficient		t	P> t	[95% conf.	interval]
dummy_euro_4 0342839 .0200849 -1.71 0.088 0737198 .005 age 0005654 .0004052 -1.40 0.163 001361 .000 female 0114266 .0129675 -0.88 0.379 0368876 .014 EDU1 .0155298 .0145169 1.07 0.285 0129735 .04 EDU2 .0064643 .0134439 0.48 0.631 0199322 .032 EDU3 001243 .0138935 -0.09 0.929 0285222 .026 EDU4 0 (omitted) .0083983 .0078772 1.07 0.287 0070682 .023 INC2 0 (omitted) .008172 .0070682 .023 .023 INC3 .0460552 .0436408 1.06 0.292 0396315 .131 INC4 .0001283 .0076813 0.02 0.987 0149537 .015 INC5 .1008534 .0398074 2.53 0.	diesel_euro4	.0155633	.0274279	0.57	0.571	0382901	.0694167
age 0005654 .0004052 -1.40 0.163 001361 .000 female 0114266 .0129675 -0.88 0.379 0368876 .014 EDU1 .0155298 .0145169 1.07 0.285 0129735 .04 EDU2 .0064643 .0134439 0.48 0.631 0199322 .032 EDU3 001243 .0138935 -0.09 0.929 0285222 .026 EDU4 0 (omitted) .0287 0070682 .023 INC1 .0083983 .0078772 1.07 0.287 0070682 .023 INC2 0 (omitted) .007070682 .023 .023 INC2 0 (omitted) .017 .0287 0070682 .023 INC3 .0460552 .0436408 1.06 0.292 0396315 .131 INC4 .0001283 .0076813 0.02 0.987 0149537 .015 INC5 .1	dummy_diesel	0289839	.0233145	-1.24	0.214	0747609	.0167931
female 0114266 .0129675 -0.88 0.379 0368876 .014 EDU1 .0155298 .0145169 1.07 0.285 0129735 .04 EDU2 .0064643 .0134439 0.48 0.631 0199322 .032 EDU3 001243 .0138935 -0.09 0.929 0285222 .026 EDU4 0 (omitted) .018772 1.07 0.287 0070682 .023 INC2 0 (omitted) .01877 1.07 0.287 0070682 .023 INC3 .0460552 .0436408 1.06 0.292 0396315 .131 INC4 .0001283 .0076813 0.02 0.987 0149537 .015 INC5 .1008534 .0398074 2.53 0.012 .0226934 .179 INC6 .0225308 .0206356 1.09 0.275 0179862 .063 INC7 .0102951 .0087551 1.18 0.240	dummy_euro_4	0342839	.0200849	-1.71	0.088	0737198	.0051519
EDU1 .0155298 .0145169 1.07 0.2850129735 .04 EDU2 .0064643 .0134439 0.48 0.6310199322 .032 EDU3001243 .0138935 -0.09 0.9290285222 .026 EDU4 0 (omitted) INC1 .0083983 .0078772 1.07 0.2870070682 .023 INC2 0 (omitted) INC3 .0460552 .0436408 1.06 0.2920396315 .131 INC4 .0001283 .0076813 0.02 0.9870149537 .015 INC5 .1008534 .0398074 2.53 0.012 .0226934 .179 INC6 .0225308 .0206356 1.09 0.2750179862 .063 INC7 .0102951 .0087551 1.18 0.2400068952 .027 INC8 .0399625 .0238919 1.67 0.0950069481 .086 INC9 .0452575 .0361874 1.25 0.2110257947 .116 INC10 .0786493 .0520677 1.51 0.1310235833 .180 INC11 .0118808 .0093124 1.28 0.2020064037 .030 INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	age	0005654	.0004052	-1.40	0.163	001361	.0002302
EDU2	female	0114266	.0129675	-0.88	0.379	0368876	.0140344
EDU3001243 .0138935 -0.09 0.9290285222 .026 EDU4 0 (omitted) INC1 .0083983 .0078772 1.07 0.2870070682 .023 INC2 0 (omitted) INC3 .0460552 .0436408 1.06 0.2920396315 .131 INC4 .0001283 .0076813 0.02 0.9870149537 .015 INC5 .1008534 .0398074 2.53 0.012 .0226934 .179 INC6 .0225308 .0206356 1.09 0.2750179862 .063 INC7 .0102951 .0087551 1.18 0.2400068952 .027 INC8 .0399625 .0238919 1.67 0.0950069481 .086 INC9 .0452575 .0361874 1.25 0.2110257947 .116 INC10 .0786493 .0520677 1.51 0.1310235833 .180 INC11 .0118808 .0093124 1.28 0.2020064037 .030 INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	EDU1	.0155298	.0145169	1.07	0.285	0129735	.044033
EDU4 0 (omitted) INC1 .0083983 .0078772 1.07 0.2870070682 .023 INC2 0 (omitted) INC3 .0460552 .0436408 1.06 0.2920396315 .131 INC4 .0001283 .0076813 0.02 0.9870149537 .015 INC5 .1008534 .0398074 2.53 0.012 .0226934 .179 INC6 .0225308 .0206356 1.09 0.2750179862 .063 INC7 .0102951 .0087551 1.18 0.2400068952 .027 INC8 .0399625 .0238919 1.67 0.0950069481 .086 INC9 .0452575 .0361874 1.25 0.2110257947 .116 INC10 .0786493 .0520677 1.51 0.1310235833 .180 INC11 .0118808 .0093124 1.28 0.2020064037 .030 INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	EDU2	.0064643	.0134439	0.48	0.631	0199322	.0328607
INC1	EDU3	001243	.0138935	-0.09	0.929	0285222	.0260363
INC2	EDU4	0	(omitted)				
INC3	INC1	.0083983	.0078772	1.07	0.287	0070682	.0238647
INC4	INC2	0	(omitted)				
INC5	INC3	.0460552	.0436408	1.06	0.292	0396315	.1317418
INC6	INC4	.0001283	.0076813	0.02	0.987	0149537	.0152102
INC7	INC5	.1008534	.0398074	2.53	0.012	.0226934	.1790134
INC8	INC6	.0225308	.0206356	1.09	0.275	0179862	.0630478
INC9	INC7	.0102951	.0087551	1.18	0.240	0068952	.0274853
INC10 .0786493 .0520677 1.51 0.1310235833 .180 INC11 .0118808 .0093124 1.28 0.2020064037 .030 INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC8	.0399625	.0238919	1.67	0.095	0069481	.0868731
INC11 .0118808 .0093124 1.28 0.2020064037 .030 INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC9	.0452575	.0361874	1.25	0.211	0257947	.1163097
INC12 .0120486 .009455 1.27 0.203006516 .030 INC13 .0430546 .034364 1.25 0.2110244174 .110 INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC10	.0786493	.0520677	1.51	0.131	0235833	.1808819
INC13	INC11	.0118808	.0093124	1.28	0.202	0064037	.0301653
INC14 .0276332 .0135888 2.03 0.042 .0009523 .054 INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC12	.0120486	.009455	1.27	0.203	006516	.0306131
INC15 .0817201 .0854244 0.96 0.3390860068 .249 INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC13	.0430546	.034364	1.25	0.211	0244174	.1105266
INC16 .0068317 .0067132 1.02 0.3090063493 .020	INC14	.0276332	.0135888	2.03	0.042	.0009523	.0543142
	INC15	.0817201	.0854244	0.96	0.339	0860068	. 2494469
_cons .0447238 .0278532 1.61 0.1090099646 .099	INC16	.0068317	.0067132	1.02	0.309	0063493	.0200127
	_cons	.0447238	.0278532	1.61	0.109	0099646	.0994122

. * Newsletter

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. regress dummy_newsletter \${main} \${controls} if target!=3 & target!=4, robust

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 705 F(23, 681) = 17.04 Prob > F = 0.0000 R-squared = 0.2459 Root MSE = .43039

d	C ((; -;+	Robust	_	D. 1±1	[OF% (
dummy_news~r	Coefficient	std. err.	t	P> t	[95% conf.	intervalj
diesel_euro4	.2814751	.0791158	3.56	0.000	.126135	.4368152
dummy_diesel	0656289	.0564937	-1.16	0.246	1765518	.0452939
dummy_euro_4	.0006155	.0609966	0.01	0.992	1191486	.1203795
age	0044142	.0017201	-2.57	0.010	0077915	0010369
female	0220316	.0358917	-0.61	0.540	0925032	.0484401
EDU1	140469	.197706	-0.71	0.478	5286555	.2477175
EDU2	1193762	.1976341	-0.60	0.546	5074217	.2686692
EDU3	0700055	.196349	-0.36	0.722	4555276	.3155167
EDU4	0	(omitted)				
INC1	0559301	.1583659	-0.35	0.724	3668741	. 255014
INC2	0	(omitted)				
INC3	167915	.158573	-1.06	0.290	4792657	.1434358
INC4	.1108758	.1592405	0.70	0.486	2017854	.423537
INC5	.0798849	.1313658	0.61	0.543	1780458	.3378155
INC6	.1858817	.1302068	1.43	0.154	0697733	.4415368
INC7	.081559	.1303937	0.63	0.532	174463	. 337581
INC8	.1177354	.131067	0.90	0.369	1396086	.3750794
INC9	.0560145	.1483932	0.38	0.706	2353487	.3473776
INC10	.081709	.1533478	0.53	0.594	2193823	.3828003
INC11	.1064764	.1366434	0.78	0.436	1618167	.3747695
INC12	.1580539	.139654	1.13	0.258	1161502	.432258
INC13	.1870093	.1397468	1.34	0.181	087377	.4613957
INC14	.3703853	.1257791	2.94	0.003	.1234238	.6173468
INC15	2240805	.1641892	-1.36	0.173	5464584	.0982975
INC16	0680927	.1315748	-0.52	0.605	3264338	.1902484
_cons	.7250443	.2454533	2.95	0.003	.2431081	1.206981

. * Donation

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note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 705 F(23, 681) = 53.57 Prob > F = 0.0000 R-squared = 0.4533 Root MSE = .37564

[.] regress dummy_donation \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.

		Robust		- 1.1	50-0/	
dummy_dona~n	Coefficient	std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.4344912	.0667541	6.51	0.000	.3034226	.5655599
dummy_diesel	0270609	.048614	-0.56	0.578	1225122	.0683905
dummy_euro_4	0545249	.0483033	-1.13	0.259	1493663	.0403164
age	003573	.0013567	-2.63	0.009	0062368	0009091
female	049834	.0310693	-1.60	0.109	1108372	.0111692
EDU1	.0374138	.1603124	0.23	0.816	2773522	.3521798
EDU2	.0503041	.1604519	0.31	0.754	2647357	.3653439
EDU3	.1136996	.1592826	0.71	0.476	1990444	.4264436
EDU4	0	(omitted)				
INC1	1714105	.1216494	-1.41	0.159	4102636	.0674425
INC2	0	(omitted)				
INC3	2715381	.1130235	-2.40	0.017	4934544	0496218
INC4	0216342	.133179	-0.16	0.871	283125	.2398567
INC5	0920038	.1096608	-0.84	0.402	3073177	.1233101
INC6	.0306704	.1087124	0.28	0.778	1827813	.2441221
INC7	.0710824	.1104966	0.64	0.520	1458724	.2880373
INC8	.056419	.1119614	0.50	0.614	163412	.2762501
INC9	1193468	.121604	-0.98	0.327	3581106	.119417
INC10	.107594	.1351474	0.80	0.426	1577616	.3729497
INC11	.1372801	.120761	1.14	0.256	0998285	.3743888
INC12	.1869313	.1217061	1.54	0.125	0520331	.4258956
INC13	.1010787	.1233501	0.82	0.413	1411134	.3432709
INC14	.4026315	.1098571	3.67	0.000	.1869322	.6183308
INC15	1645029	.1292933	-1.27	0.204	4183643	.0893584
INC16	1101251	.1095353	-1.01	0.315	3251926	.1049425
_cons	.3776939	.2025501	1.86	0.063	0200039	.7753917

. regress gov_firms_responsibility \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 705 F(23, 681) = 3.97 Prob > F = 0.0000 R-squared = 0.1100 Root MSE = .4359

		Robust				
gov_firms_~y	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.362217	.0795907	4.55	0.000	.2059443	.5184897
dummy_diesel	158334	.0588978	-2.69	0.007	2739772	0426909
dummy_euro_4	0557829	.0590674	-0.94	0.345	171759	.0601933
age	.0028614	.0017723	1.61	0.107	0006184	.0063413
female	0130425	.0369655	-0.35	0.724	0856226	.0595377
EDU1	1588936	.1619008	-0.98	0.327	4767784	.1589911
EDU2	1415295	.1606236	-0.88	0.379	4569066	.1738475
EDU3	2018552	.1593129	-1.27	0.206	5146587	.1109483
EDU4	0	(omitted)				
INC1	0971638	.1641721	-0.59	0.554	419508	.2251804
INC2	0	(omitted)				
INC3	.0887294	.1540564	0.58	0.565	2137531	.3912119
INC4	.1892049	.1504063	1.26	0.209	1061109	.4845207

INC5	0104077	.1385645	-0.08	0.940	2824726	.2616572
INC6	04118	.138554	-0.30	0.766	3132242	.2308643
INC7	.1249656	.1356231	0.92	0.357	141324	.3912553
INC8	.0990127	.1359591	0.73	0.467	1679367	.3659622
INC9	0234527	.1558932	-0.15	0.880	3295418	.2826364
INC10	.0916889	.1546709	0.59	0.554	2120003	.3953781
INC11	.1691833	.1425094	1.19	0.236	1106274	.4489939
INC12	.1370024	.1442566	0.95	0.343	1462387	.4202436
INC13	.0838609	.1495877	0.56	0.575	2098477	.3775695
INC14	.1452458	.1343707	1.08	0.280	1185849	.4090764
INC15	.0210363	.1954467	0.11	0.914	3627143	.4047869
INC16	.0155958	.1359729	0.11	0.909	2513807	.2825723
_cons	.6534129	.2288432	2.86	0.004	. 20409	1.102736

. regress taxes_eco_friendly \${main} \${controls} if target!=3 & target!=4, robust
note: EDU4 omitted because of collinearity.

note: ${\bf INC2}$ omitted because of collinearity.

Linear regression Number of obs

F(23, 681) = 13.01 Prob > F = 0.0000 R-squared = 0.2490 Root MSE = .43836

705

Robust Coefficient std. err. t P>|t| [95% conf. interval] taxes eco ~y diesel_euro4 .1123808 .0788473 1.43 0.155 -.0424322 .2671938 dummy_diesel .0353833 .0570478 0.62 0.535 -.0766274 .1473941 dummy_euro_4 .0793609 .0578581 1.37 0.171 -.0342408 .1929627 age -.0012165 .0016846 -0.72 0.470 -.0045241 .0020911 female -2.82 -.1057647 .0374435 0.005 -.1792834 -.0322461 EDU1 -.0929375 .196701 -0.47 0.637 -.4791508 .2932759 EDU2 .0050816 .1971337 0.03 0.979 -.3819812 .3921444 EDU3 .090249 .1955615 0.46 0.645 -.293727 .474225 EDU4 0 (omitted) INC1 -.0060253 .1410283 -0.04 0.966 -.2829277 .2708772 INC2 0 (omitted) INC3 .0511898 .1496229 0.34 0.732 -.2425878 .3449674 INC4 .0134778 .1453519 0.09 0.926 -.271914 .2988696 INC5 .0052217 0.04 0.967 .1256015 -.241391 .2518344 INC6 .1199149 .1256428 0.95 0.340 -.1267788 .3666086 INC7 .1424337 .1242305 1.15 0.252 -.101487 .3863545 INC8 .2344277 .1235354 1.90 0.058 -.0081284 .4769837 INC9 .0819296 .1388893 0.59 0.555 -.1907731 .3546323 .1279491 INC10 .1477742 0.87 0.387 -.1621987 .418097 INC11 .2002191 .1350557 1.48 0.139 -.0649564 .4653946 INC12 .150144 .1426486 1.05 0.293 -.1299399 .4302278 INC13 .1414157 .143843 0.98 0.326 -.1410134 .4238448 0.003 INC14 .3661359 .123079 2.97 .1244759 .6077959 INC15 .0016328 .1735916 0.01 0.992 -.3392063 .3424718 -.019114 INC16 .1235077 -0.15 0.877 -.2616156 .2233876 .380941 .2430255 1.57 0.117 -.0962282 .8581103 _cons

. regress pay_eco_friendly ${\mathcal F}_{\alpha} \$ if target!=3 & target!=4, robust

note: **EDU4** omitted because of collinearity. note: **INC2** omitted because of collinearity.

Linear regression Number of obs 705

F(23, 681) = 7.15 Prob > F = 0.0000 R-squared = 0.1660 Root MSE = .43159

		Robust				
pay_eco_fr~y	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1317076	.0783292	1.68	0.093	0220881	.2855033
dummy_diesel	0517565	.0567732	-0.91	0.362	1632281	.0597151
dummy_euro_4	.0416532	.059408	0.70	0.483	0749916	.158298
age	0003804	.001651	-0.23	0.818	0036221	.0028613
female	.0028068	.0363942	0.08	0.939	0686516	.0742652
EDU1	1396947	.1805834	-0.77	0.439	4942618	.2148724
EDU2	0412318	.180116	-0.23	0.819	3948811	.3124175
EDU3	.011356	.1789873	0.06	0.949	3400773	.3627894
EDU4	0	(omitted)				
INC1	0044528	.1558752	-0.03	0.977	3105065	.3016009
INC2	0	(omitted)				
INC3	.2917364	.1590951	1.83	0.067	0206394	.6041122
INC4	.0731125	.1658506	0.44	0.659	2525275	.3987525
INC5	.0152016	.136169	0.11	0.911	2521599	.2825632
INC6	.2656692	.1360528	1.95	0.051	0014642	.5328026
INC7	.2680767	.1345124	1.99	0.047	.0039679	.5321854
INC8	.3306842	.1332585	2.48	0.013	.0690373	.592331
INC9	.3445898	.1454982	2.37	0.018	.0589108	.6302687
INC10	.2173444	.1499719	1.45	0.148	0771185	.5118074
INC11	.1809423	.1445174	1.25	0.211	102811	.4646956
INC12	.2481801	.1512807	1.64	0.101	0488525	.5452126
INC13	.2772341	.1465312	1.89	0.059	0104732	.5649414
INC14	.3901237	.1310798	2.98	0.003	.1327547	.6474928
INC15	.1732128	.1895662	0.91	0.361	1989917	.5454173
INC16	.170419	.1356653	1.26	0.209	0959535	.4367916
_cons	.4622117	.2334378	1.98	0.048	.0038674	.9205561

. * Table SI-8 * *********

. * Without controls

. regress vote_lega_euro diesel_euro4 if target!=3 & target!=4 & no_answer_euro==0 & fuel==1, robust

Linear regression Number of obs 376

F(1, 374) = 2.31 Prob > F = 0.1293 R-squared = 0.0055 Root MSE = .42502

vote_lega_~o	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.0717222	.0471792	1.52	0.129	0210475	.1644919
_cons	.1836735	.0392193	4.68	0.000	.1065554	.2607915

. * Including individual controls

. regress vote_lega_euro diesel_euro4 \${controls} if target!=3 & target!=4 & no_answer_euro==0 & fuel==1, robust note: EDU1 omitted because of collinearity.

note: **EDU1** omitted because of collinearity. note: **INC2** omitted because of collinearity. note: **INC15** omitted because of collinearity.

Linear regression

Number of obs = 376 F(20, 355) = 4.57 Prob > F = 0.0000 R-squared = 0.1956 Root MSE = .39235

		Dala at				
vote lega ~o	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval
diesel_euro4	.1580012	.0555281	2.85	0.005	.0487957	.2672066
age	.002703	.0026204	1.03	0.303	0024505	.0078564
female	.2330868	.0582212	4.00	0.000	.118585	.3475885
EDU1	0	(omitted)				
EDU2	2401182	.0719722	-3.34	0.001	3816637	0985727
EDU3	2503943	.0711208	-3.52	0.000	3902654	1105232
EDU4	1033747	.275356	-0.38	0.708	6449089	.4381595
INC1	3946509	.293048	-1.35	0.179	9709792	.1816775
INC2	0	(omitted)				
INC3	514162	.2367375	-2.17	0.031	9797462	0485777
INC4	1065227	.2629588	-0.41	0.686	6236757	.4106303
INC5	2813137	.2599937	-1.08	0.280	7926351	.2300078
INC6	2125078	.2504061	-0.85	0.397	7049737	.2799581
INC7	3068207	.2350093	-1.31	0.193	7690061	.1553647
INC8	313264	.2368341	-1.32	0.187	7790383	.1525102
INC9	365199	.2432977	-1.50	0.134	8436851	.1132871
INC10	2534847	.2429315	-1.04	0.297	7312506	.2242812
INC11	1386916	.2558368	-0.54	0.588	6418379	.3644548
INC12	1978445	.2398552	-0.82	0.410	6695604	.2738713
INC13	20175	. 2451755	-0.82	0.411	6839291	.2804291
INC14	0821287	.2356797	-0.35	0.728	5456327	.3813753
INC15	0	(omitted)				
INC16	1567358	.2599636	-0.60	0.547	6679981	.3545264
_cons	.3155282	.2488481	1.27	0.206	1738736	.80493

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. * On the baseline, including control for past vote for Lega: Legislative 2018

. regress vote_lega_euro diesel_euro4 \${controls} vote_lega_2018 if target!=3 & target!=4 & no_answer_euro==0 & no_a

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.
note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 370 F(21, 348) = 38.09 Prob > F = 0.0000 R-squared = 0.5781 Root MSE = .28514

-						
		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.130436	.0413648	3.15	0.002	.0490796	.2117924
age	.0033575	.0018042	1.86	0.064	000191	.006906
female	.253228	.0415275	6.10	0.000	.1715516	.3349045
EDU1	1396663	.3111484	-0.45	0.654	7516343	.4723016
EDU2	179585	.3117024	-0.58	0.565	7926427	.4334727
EDU3	1598683	.3098115	-0.52	0.606	7692069	.4494704
EDU4	0	(omitted)				
INC1	1604048	.1711048	-0.94	0.349	4969345	.1761249
INC2	0	(omitted)				
INC3	1887917	.109157	-1.73	0.085	4034821	.0258988
INC4	0732101	.1011376	-0.72	0.470	2721279	.1257077
INC5	2158599	.1212291	-1.78	0.076	4542938	.0225739
INC6	0895937	.124377	-0.72	0.472	3342188	.1550314
INC7	1305691	.0987736	-1.32	0.187	3248374	.0636991
INC8	1133965	.0956773	-1.19	0.237	3015751	.074782
INC9	0773801	.1237075	-0.63	0.532	3206885	.1659284
INC10	0585456	.0921245	-0.64	0.526	2397364	.1226452
INC11	0693604	.0972967	-0.71	0.476	2607241	.1220032
INC12	1376808	.1008294	-1.37	0.173	3359924	.0606308
INC13	0867279	.096112	-0.90	0.367	2757614	.1023057
INC14	.0275325	.0979245	0.28	0.779	1650658	.2201308
INC15	0	(omitted)				
INC16	0419199	.1304472	-0.32	0.748	2984839	.2146442
vote_lega_2018	.7950874	.0477151	16.66	0.000	.7012411	.8889337
_ cons	.0148977	.3314904	0.04	0.964	637079	.6668744
_						

. * On the baseline, including control for past vote for Lega: Regional 2018

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.
note: INC4 omitted because of collinearity.

Linear regression

Number of obs = 357 F(21, 335) = 19.70 Prob > F = 0.0000 R-squared = 0.5247 Root MSE = .30312

[.] regress vote_lega_euro diesel_euro4 \${controls} vote_lega_regional if target!=3 & target!=4 & no_answer_euro==0 &

	1					
		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1464441	.0434267	3.37	0.001	.0610208	.2318674
age	.0050925	.0017857	2.85	0.005	.00158	.0086051
female	.2733136	.0465151	5.88	0.000	.181815	.3648121
EDU1	.2495893	.5294252	0.47	0.638	7918276	1.291006
EDU2	.1733778	.5299506	0.33	0.744	8690724	1.215828
EDU3	.2056501	.5303861	0.39	0.698	8376569	1.248957
EDU4	0	(omitted)				
INC1	2106074	.1547131	-1.36	0.174	514939	.0937243
INC2	0	(omitted)				
INC3	1276011	.1041883	-1.22	0.222	3325469	.0773446
INC4	0	(omitted)				
INC5	0935746	.0986694	-0.95	0.344	2876643	.1005151
INC6	017538	.1123803	-0.16	0.876	2385978	.2035219
INC7	0577561	.0842888	-0.69	0.494	2235581	.1080458
INC8	005647	.0867917	-0.07	0.948	1763724	.1650784
INC9	1062992	.1100333	-0.97	0.335	3227425	.1101441
INC10	.0102381	.0844802	0.12	0.904	1559405	.1764166
INC11	0574691	.101676	-0.57	0.572	2574729	.1425348
INC12	0362339	.0827716	-0.44	0.662	1990514	.1265837
INC13	006915	.085213	-0.08	0.935	1745349	.1607049
INC14	.1340089	.0841624	1.59	0.112	0315446	.2995623
INC15	.0997678	.1154655	0.86	0.388	1273611	.3268966
INC16	.0086786	.1258651	0.07	0.945	2389069	.2562641
vote_lega_regional	.7572394	.0656932	11.53	0.000	.6280162	.8864626
_cons	5147052	.5495338	-0.94	0.350	-1.595677	.5662667
	I .					

. \ast On the baseline, including control for past vote for Lega: Municipal 2016

. regress vote_lega_euro diesel_euro4 \${controls} vote_lega_municipal if target!=3 & target!=4 & no_answer_euro==0 & note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.
note: INC3 omitted because of collinearity.

Linear regression

Number of obs = 349 F(21, 327) = 32.83 Prob > F = 0.0000 R-squared = 0.4747 Root MSE = .31744

		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel euro4	.142733	.0465799	3.06	0.002	.051099	.2343671
age	.0052173	.0022357	2.33	0.020	.0008191	.0096155
female	.2879489	.0517773	5.56	0.000	.1860903	.3898075
EDU1	0	(omitted)				
EDU2	1024421	.054998 8	-1.86	0.063	2106383	.005754
EDU3	061926	.0571725	-1.08	0.280	1743983	.0505463
EDU4	.0484336	.3078232	0.16	0.875	55713	.6539973
INC1	023365	.2041628	-0.11	0.909	4250033	.3782734
INC2	0	(omitted)				
INC3	0	(omitted)				
INC4	.0953286	.1052786	0.91	0.366	1117801	.3024373
INC5	.0821045	.1460663	0.56	0.574	2052437	.3694528
INC6	.1399268	.1333937	1.05	0.295	1224912	.4023449
INC7	.0392646	.0789894	0.50	0.619	116127	.1946562
INC8	.0657237	.0861662	0.76	0.446	1037864	.2352337
INC9	.0150058	.1304427	0.12	0.908	2416071	.2716186

INC10 INC11 INC12 INC13 INC14 INC15 INC16 vote_lega_municipal	.0806072 .0473775 .0359116 .0722753 .203437 .2715401 .0483029 .7602726	.0752907 .0935894 .0812426 .083427 .0848984 .0848392 .1238875 .067572	1.07 0.51 0.44 0.87 2.40 3.20 0.39	0.285 0.613 0.659 0.387 0.017 0.002 0.697 0.000	067508 1367358 1239125 091846 .0364209 .1046407 1954142 .6273419	.2287224 .2314907 .1957357 .2363967 .370453 .4384395 .2920199 .8932032
_cons	3300161	.1319893	-2.50	0.013	5896714	0703608

. * Baseline on the switching regressions - Legislative 2018

. regress sw_to_lega_18_19 diesel_euro4 \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018==0 note: EDU1 omitted because of collinearity.

note: **INC2** omitted because of collinearity. note: **INC4** omitted because of collinearity.

Linear regression Number of obs =

Number of obs = 316 F(20, 295) = 3.97 Prob > F = 0.0000 R-squared = 0.3181 Root MSE = .28112

Robust sw_to_l~8_19 Coefficient std. err. t P>|t| [95% conf. interval] diesel_euro4 .1581692 .0456003 3.47 0.001 .0684261 .2479123 age .004801 .0019217 2.50 0.013 .001019 .008583 .2684593 female .0465071 5.77 0.000 .1769316 .359987 EDU1 0 (omitted) EDU2 .0084138 .0500912 0.17 0.867 -.0901676 .1069952 .1284054 .0363571 0.78 -.0556913 EDU3 .0467716 0.438 .780835 EDU4 .1506616 .3202039 0.47 0.638 -.4795118 INC1 -.0926848 .1830377 -0.51 0.613 -.4529099 .2675404 INC2 0 (omitted) INC3 -.108709 .1084614 -1.00 0.317 -.3221652 .1047471 INC4 0 (omitted) .0745435 INC5 -.1422824 -1.29 0.198 -.3591084 .1101736 INC6 -.0104934 .1270186 -0.08 0.934 -.2604708 .239484 INC7 -.082974 .0848447 -0.98 0.329 -.2499515 .0840036 INC8 -.0738784 .0859401 -0.86 -.2430118 0.391 .095255 INC9 -.0174936 .1162981 -0.15 0.881 -.2463727 .2113854 INC10 -.0117927 .082132 -0.14 0.886 -.1734316 .1498462 INC11 -.0767407 .0877111 -0.87 0.382 -.2493596 .0958781 INC12 -.0551152 .0848702 -0.65 0.517 -.222143 .1119127 -0.86 INC13 -.0740652 .086246 0.391 -.2438006 .0956702 INC14 .0806663 .0884535 0.91 0.363 -.0934137 .2547462 INC15 .0298896 .1396295 0.21 0.831 -.2449066 .3046857 INC16 -.0078817 .1500844 -0.05 0.958 -.3032534 .28749 -.3113304 .1280434 -.0593362 -2.43 0.016 -.5633247 _cons

. * Baseline on the switching regressions - Regional 2018

. regress sw_to_lega_reg_19 diesel_euro4 \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_region note: EDU1 omitted because of collinearity. note: INC2 omitted because of collinearity. note: INC15 omitted because of collinearity.

Linear regression

Number of obs 309 F(19, 288) Prob > F R-squared 0.4068 Root MSE .27879

		Robust				
sw_to_l~g_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1927858	.0492967	3.91	0.000	.0957583	.2898133
age	.0076923	.0018354	4.19	0.000	.0040799	.0113047
female	.3375997	.0487596	6.92	0.000	.2416292	.4335701
EDU1	0	(omitted)				
EDU2	0890645	.0630383	-1.41	0.159	2131387	.0350098
EDU3	0421724	.0617776	-0.68	0.495	1637653	.0794205
EDU4	.4236179	.0791892	5.35	0.000	.2677549	.5794809
INC1	3782054	.19143	-1.98	0.049	7549847	0014262
INC2	0	(omitted)				
INC3	2521688	.1762425	-1.43	0.154	5990554	.0947179
INC4	105394	.201303	-0.52	0.601	5016057	.2908177
INC5	2876493	.183506	-1.57	0.118	6488323	.0735337
INC6	1165359	.1875002	-0.62	0.535	4855804	.2525086
INC7	1782995	.1584731	-1.13	0.261	4902118	.1336128
INC8	1372086	.1608634	-0.85	0.394	4538256	.1794083
INC9	1793297	.1610231	-1.11	0.266	4962611	.1376017
INC10	1216435	.1570297	-0.77	0.439	4307148	.1874279
INC11	1969191	.1641652	-1.20	0.231	5200348	.1261966
INC12	1521965	.1594081	-0.95	0.340	4659491	.1615561
INC13	1658971	.1609208	-1.03	0.303	4826271	.1508328
INC14	0200518	.1601812	-0.13	0.900	3353262	.2952225
INC15	0	(omitted)				
INC16	1400997	.2060177	-0.68	0.497	5455911	.2653917
_cons	3031803	.1737744	-1.74	0.082	6452091	.0388485

^{. *} Baseline on the switching regressions - Municipal 2016

. regress sw_to_lega_16_19 diesel_euro4 \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_municip

note: EDU1 omitted because of collinearity. note: INC2 omitted because of collinearity. note: INC4 omitted because of collinearity.

Linear regression

Number of obs 313 F(19, 292) Prob > F R-squared 0.3276 Root MSE .31103

		Robust				
sw_to_1~6_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1820431	.0507139	3.59	0.000	.0822319	.2818542
age	.0059851	.0024241	2.47	0.014	.0012141	.010756
female	.3181335	.0528604	6.02	0.000	.2140979	.4221691
EDU1	0	(omitted)				
EDU2	0860521	.0596061	-1.44	0.150	2033641	.0312598
EDU3	0363945	.0614303	-0.59	0.554	1572968	.0845079
EDU4	.0442226	.3141687	0.14	0.888	5740996	.6625447
INC1	1573775	.2555652	-0.62	0.539	6603609	.3456059
INC2	0	(omitted)				
INC3	0912023	.1300863	-0.70	0.484	347228	.1648234
INC4	0	(omitted)				
INC5	.0353595	.1762837	0.20	0.841	3115881	.3823071
INC6	.0346858	.1557248	0.22	0.824	2717995	.3411711
INC7	0931472	.1025716	-0.91	0.365	2950206	.1087262
INC8	0537455	.1073674	-0.50	0.617	2650575	.1575666
INC9	0223986	.1303381	-0.17	0.864	2789198	.2341226
INC10	0423167	.1011777	-0.42	0.676	2414467	.1568133
INC11	1159236	.1108348	-1.05	0.296	3340598	.1022127
INC12	0775594	.1034869	-0.75	0.454	2812342	.1261155
INC13	0865828	.1052412	-0.82	0.411	2937102	.1205447
INC14	.0671851	.1064769	0.63	0.529	1423745	.2767447
INC15	.2349816	.1121121	2.10	0.037	.0143314	.4556318
INC16	1183658	.1645727	-0.72	0.473	4422649	.2055333
_cons	2985561	.1566928	-1.91	0.058	6069465	.0098343

. *********** . * Table SI-9 *

. regress vote_lega_euro km_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 602 F(29, 572) = 4.18 Prob > F = 0.0000 R-squared = 0.1634 Root MSE = .40312

		Robust				
vote_lega_e~o	Coefficient	std. err.	t	P> t	[95% conf.	. interval]
km_less_1k	0202976	.1414461	-0.14	0.886	2981147	.2575195
km_1k_to_5k	028093	.1085512	-0.26	0.796	2413005	.1851146
km 5k to 10k	0917916	.1028986	-0.89	0.373	2938968	.1103137
km_10k_to_20k	1347337	.1028493	-1.31	0.191	336742	.0672746
km_20k_to_30k	.0426395	.1090709	0.39	0.696	1715889	.2568679
km more 30k	2541756	.1228166	-2.07	0.039	4954022	012949
diesel euro4	.1570227	.0801143	1.96	0.050	0003315	.3143768
dummy diesel	0962313	.0562318	-1.71	0.088	2066772	.0142147
dummy euro 4	0656862	.0611008	-1.08	0.283	1856955	.0543231
age	.0023603	.0017237	1.37	0.171	0010251	.0057458
female	.1621021	.0395396	4.10	0.000	.0844415	.2397627
EDU1	.1704414	.05132	3.32	0.001	.0696427	. 27124
EDU2	0	(omitted)				
EDU3	0002075	.0420635	-0.00	0.996	0828252	.0824102
EDU4	0275702	.1229177	-0.22	0.823	2689954	.2138549

INC1	1141109	.209157	-0.55	0.586	5249202	.2966984
INC2	2290901	.2213217	-1.04	0.301	6637925	.2056123
INC3	1815046	.2077935	-0.87	0.383	589636	.2266269
INC4	.0886887	.217556	0.41	0.684	3386174	.5159949
INC5	0580511	.1984692	-0.29	0.770	4478684	.3317663
INC6	0359096	.1978615	-0.18	0.856	4245334	.3527141
INC7	0338196	.1934058	-0.17	0.861	4136919	.3460527
INC8	0485872	.1936506	-0.25	0.802	4289403	.3317659
INC9	1080102	.1987113	-0.54	0.587	4983031	.2822827
INC10	.0253456	.2038703	0.12	0.901	3750802	.4257714
INC11	.039851	.2042543	0.20	0.845	3613289	.441031
INC12	.0668731	.2075596	0.32	0.747	3407988	.4745451
INC13	.0161794	.2048514	0.08	0.937	3861733	.418532
INC14	.1482788	.1991844	0.74	0.457	2429433	.5395009
INC15	0	(omitted)				
INC16	.0947789	.1996177	0.47	0.635	2972941	.4868519
_cons	.1148178	.1967771	0.58	0.560	271676	.5013117

. regress vote_lega_euro ten_km \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: ${\bf INC15}$ omitted because of collinearity.

Linear regression

Number of obs = 602 F(24, 577) = 3.71 Prob > F = 0.0000 R-squared = 0.1328 Root MSE = .40865

		Robust				
vote_lega_~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
ten_km	0522204	.040753	-1.28	0.201	1322628	.0278219
diesel_euro4	.193353	.0793038	2.44	0.015	.0375937	.3491123
dummy_diesel	0987909	.056819	-1.74	0.083	2103882	.0128063
dummy_euro_4	0542926	.0601704	-0.90	0.367	1724723	.0638872
age	.0018669	.0016976	1.10	0.272	0014674	.0052012
female	.1836975	.0394201	4.66	0.000	.1062731	.2611219
EDU1	.1785728	.0500288	3.57	0.000	.080312	.2768336
EDU2	0	(omitted)				
EDU3	002976	.040463	-0.07	0.941	0824488	.0764968
EDU4	0004415	.1369648	-0.00	0.997	2694519	.2685688
INC1	1533383	.2009824	-0.76	0.446	5480846	.2414079
INC2	2355605	.2106144	-1.12	0.264	6492248	.1781038
INC3	1866144	.1993445	-0.94	0.350	5781438	.204915
INC4	.0968874	.2133392	0.45	0.650	3221286	.5159034
INC5	0761807	.1921843	-0.40	0.692	4536469	.3012854
INC6	0547207	.1918708	-0.29	0.776	4315711	.3221297
INC7	0659847	.1875369	-0.35	0.725	4343228	.3023535
INC8	0894521	.1881859	-0.48	0.635	459065	.2801608
INC9	1543709	.1914838	-0.81	0.420	530461	.2217193
INC10	.0214239	.1979769	0.11	0.914	3674193	.4102671
INC11	.0364907	.197224	0.19	0.853	3508738	.4238552
INC12	.0552262	.1970442	0.28	0.779	3317851	.4422375
INC13	008347	.1959381	-0.04	0.966	393186	.3764919
INC14	.1348137	.1862706	0.72	0.470	2310374	.5006648
INC15	0	(omitted)				
INC16	.0675673	.195215	0.35	0.729	3158513	.4509859
_cons	.0728957	.1872234	0.39	0.697	2948268	.4406182

regress sw_to_lega_18_19 km_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018==0

note: EDU1 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 483

F(29, 453) = **3.92** Prob > F = **0.0000** R-squared = **0.3263**

Root MSE = .25356

		Robust				
sw_to_le~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
km_less_1k	215579	.0980473	-2.20	0.028	4082629	0228951
km_1k_to_5k	1574114	.0988809	-1.59	0.112	3517336	.0369108
km_5k_to_10k	1348035	.0953535	-1.41	0.158	3221936	.052586
km_10k_to_20k	153839	.0967372	-1.59	0.112	3439484	.036270
km_20k_to_30k	.067884	.10735	0.63	0.527	1430818	.2788497
km_more_30k	2935509	.1131093	-2.60	0.010	5158349	0712669
diesel_euro4	.1074224	.0535022	2.01	0.045	.0022791	.2125656
dummy_diesel	0495244	.0379071	-1.31	0.192	12402	.0249713
dummy_euro_4	0205673	.0400113	-0.51	0.607	0991981	.0580634
age	.0028768	.0010984	2.62	0.009	.0007183	.0050353
female	.1481451	.0279806	5.29	0.000	.0931572	.203133
EDU1	0	(omitted)				
EDU2	.0081849	.0336026	0.24	0.808	0578515	.0742213
EDU3	.0453422	.0335421	1.35	0.177	0205753	.1112596
EDU4	.0315522	.1252145	0.25	0.801	2145212	.2776257
INC1	.0450605	.0716699	0.63	0.530	0957862	.1859073
INC2	0	(omitted)				
INC3	.0647867	.0840119	0.77	0.441	1003148	.2298882
INC4	.2424417	.1261861	1.92	0.055	0055412	.490424
INC5	.0126332	.0510939	0.25	0.805	0877773	.1130436
INC6	.073754	.0632824	1.17	0.244	0506094	.198117
INC7	.0057534	.045283	0.13	0.899	0832375	.0947443
INC8	.0196678	.0431285	0.46	0.649	065089	.104424
INC9	.0702252	.0610728	1.15	0.251	0497959	.1902463
INC10	.0402326	.080666	0.50	0.618	1182934	.1987586
INC11	0155004	.0546656	-0.28	0.777	1229301	.0919292
INC12	.0059553	.0532459	0.11	0.911	0986843	.1105949
INC13	022567	.0489334	-0.46	0.645	1187317	.073597
INC14	.1707785	.0595165	2.87	0.004	.0538159	.287741
INC15	0401763	.101624	-0.40	0.693	2398893	.1595368
INC16	.0776275	.0711914	1.09	0.276	0622788	.2175337
_cons	0691901	.1221951	-0.57	0.572	3093298	.170949

. regress sw_to_lega_18_19 ten_km \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018= note: EDU1 omitted because of collinearity. note: INC2 omitted because of collinearity.

Linear regression
Number of obs = 483 F(24, 458) = 3.18 Prob > F = 0.0000 R-squared = 0.2238

Root MSE = .27068

		Robust				
sw_to_l~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
ten_km	.0104782	.0295225	0.35	0.723	0475382	.0684946
diesel euro4	.1483766	.0535052	2.77	0.006	.0432304	.2535228
dummy_diesel	0450045	.0406193	-1.11	0.268	1248278	.0348189
dummy_euro_4	000011	.0381945	-0.00	1.000	0750693	.0750472
age	.0018732	.0010741	1.74	0.082	0002377	.003984
female	.1621628	.0291728	5.56	0.000	.1048336	.2194921
EDU1	0	(omitted)				
EDU2	0101197	.0334456	-0.30	0.762	0758455	.0556061
EDU3	.0248127	.0336133	0.74	0.461	0412427	.0908681
EDU4	.0883409	.13607	0.65	0.517	1790581	.3557398
INC1	.0325624	.0655562	0.50	0.620	0962658	.1613906
INC2	0	(omitted)				
INC3	.083802	.0789355	1.06	0.289	0713186	.2389225
INC4	.3025349	.1251281	2.42	0.016	.0566385	.5484312
INC5	.0367086	.0446959	0.82	0.412	0511257	.124543
INC6	.0949961	.0580265	1.64	0.102	0190351	.2090274
INC7	.0018109	.0387647	0.05	0.963	0743678	.0779896
INC8	.0050953	.0365098	0.14	0.889	0666521	.0768428
INC9	.0564024	.0562183	1.00	0.316	0540754	.1668802
INC10	.0983582	.0743925	1.32	0.187	0478347	.2445511
INC11	.036553	.0511273	0.71	0.475	0639201	.1370262
INC12	.0404064	.0445762	0.91	0.365	0471929	.1280057
INC13	.0138285	.0401773	0.34	0.731	0651262	.0927831
INC14	.2015829	.0518213	3.89	0.000	.0997459	.3034199
INC15	.0308644	.0630449	0.49	0.625	0930288	.1547576
INC16	.0954	.0626439	1.52	0.128	0277051	.2185051
_cons	1956603	.0705145	-2.77	0.006	3342323	0570882

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. **********
. * Table SI-10 *

. regress vote_lega_euro use_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: ${\bf INC15}$ omitted because of collinearity.

		Robust				
vote_lega_~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
use_day	.1707104	.0767983	2.22	0.027	.0198705	.3215504
use_week	.3940392	.075163	5.24	0.000	.246411	.5416673
use month	.3516128	.0905128	3.88	0.000	.173836	.5293895
use_year	.4382875	.2268186	1.93	0.054	0072082	.8837832
diesel euro4	.2283582	.0787751	2.90	0.004	.0736357	.3830807
dummy_diesel	0912386	.055052	-1.66	0.098	1993665	.0168893
dummy euro 4	0847309	.0608697	-1.39	0.164	2042853	.0348236
age	.0002352	.0017263	0.14	0.892	0031555	.0036259
female	.1449545	.0391683	3.70	0.000	.0680239	.2218852
EDU1	.1830121	.0497226	3.68	0.000	.0853517	.2806725
EDU2	0	(omitted)				
EDU3	0167973	.0407534	-0.41	0.680	0968412	.0632467

EDU4	.0574325	.1029569	0.56	0.577	1447856	.2596507
INC1	1554385	.1835065	-0.85	0.397	5158645	.2049876
INC2	0935623	.1933922	-0.48	0.629	473405	.2862803
INC3	1565654	.1781714	-0.88	0.380	5065128	.1933819
INC4	.162811	.1987445	0.82	0.413	2275442	.5531662
INC5	0120042	.1727362	-0.07	0.945	3512763	.3272679
INC6	.002527	.1739051	0.01	0.988	3390409	.3440949
INC7	.0162319	.168907	0.10	0.923	3155194	.3479831
INC8	0205608	.1693378	-0.12	0.903	3531582	.3120365
INC9	0988124	.1726414	-0.57	0.567	4378984	.2402735
INC10	.0562142	.1809858	0.31	0.756	299261	.4116894
INC11	.1005274	.1808874	0.56	0.579	2547546	.4558093
INC12	.1140592	.1787116	0.64	0.524	2369492	.4650677
INC13	.0551641	.1798952	0.31	0.759	2981691	.4084973
INC14	.2058462	.1668031	1.23	0.218	1217727	.5334652
INC15	0	(omitted)				
INC16	.1258596	.1773092	0.71	0.478	2223943	.4741136
_cons	1942427	.1901127	-1.02	0.307	567644	.1791587

. regress vote_lega_euro everyweek \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression

Number of obs = 602 F(24, 577) = 3.63 Prob > F = 0.0000 R-squared = 0.1318 Root MSE = .40886

_		Robust				
vote_lega_~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
everyweek	0616356	.0625048	-0.99	0.324	1844003	.061129
diesel_euro4	.1896048	.0784935	2.42	0.016	.035437	.3437727
dummy_diesel	1018365	.0567925	-1.79	0.073	2133818	.0097088
dummy_euro_4	0500125	.0591473	-0.85	0.398	1661827	.0661578
age	.0016173	.0017117	0.94	0.345	0017447	.0049793
female	.1891028	.0391684	4.83	0.000	.1121727	.2660329
EDU1	.1842727	.0496594	3.71	0.000	.0867375	.2818079
EDU2	0	(omitted)				
EDU3	0024821	.0403104	-0.06	0.951	0816551	.0766909
EDU4	.0020993	.1338774	0.02	0.987	2608472	.2650458
INC1	1398783	.2029589	-0.69	0.491	5385066	.25875
INC2	2135698	.2139336	-1.00	0.319	6337533	.2066138
INC3	1843406	.2013076	-0.92	0.360	5797255	.2110443
INC4	.1061194	.214997	0.49	0.622	3161527	.5283915
INC5	0668901	.1947221	-0.34	0.731	4493407	.3155606
INC6	0470626	.1945356	-0.24	0.809	4291467	.3350216
INC7	0630379	.1900391	-0.33	0.740	4362905	.3102148
INC8	0821475	.191095	-0.43	0.667	4574742	.2931791
INC9	1464703	.1934109	-0.76	0.449	5263455	.2334048
INC10	.0288247	.2016013	0.14	0.886	3671372	.4247866
INC11	.042243	.2013389	0.21	0.834	3532036	.4376895
INC12	.0580367	.1991106	0.29	0.771	3330333	.4491067
INC13	003616	.1982149	-0.02	0.985	3929267	.3856947
INC14	.1305589	.1886948	0.69	0.489	2400535	.5011714
INC15	0	(omitted)				
INC16	.0829563	.19803	0.42	0.675	3059912	.4719038
_cons	.10126	.2009949	0.50	0.615	2935108	.4960308
_						

. regress vote_lega_euro workcar_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.

note: INC15 omitted because of collinearity.

R-squared = **0.1698** Root MSE = **.40086**

		Robust				
vote_lega_e~o	Coefficient	std. err.	t	P> t	[95% conf.	interval]
workcar_day	119542	.0578062	-2.07	0.039	2330795	0060044
workcar_week	.0925826	.0621807	1.49	0.137	0295469	.214712
workcar_month	.0930742	.0795129	1.17	0.242	0630976	. 249246
workcar_year	.0295522	.1147854	0.26	0.797	1958983	.2550028
diesel_euro4	.2042852	.0818524	2.50	0.013	.0435185	.3650519
dummy_diesel	0902122	.0553005	-1.63	0.103	1988282	.0184037
dummy_euro_4	0646571	.0607815	-1.06	0.288	1840384	.0547242
age	.0010953	.0018008	0.61	0.543	0024416	.0046322
female	.1521209	.0390792	3.89	0.000	.0753653	.2288765
EDU1	.173117	.0503879	3.44	0.001	.0741498	.2720843
EDU2	0	(omitted)				
EDU3	0117551	.0405563	-0.29	0.772	0914119	.0679016
EDU4	0247034	.1133345	-0.22	0.828	2473043	.1978974
INC1	1360684	.197432	-0.69	0.491	5238457	.2517089
INC2	1105962	.2042825	-0.54	0.588	5118285	. 2906361
INC3	1703656	.1892765	-0.90	0.368	5421246	.2013934
INC4	.1199787	.2112338	0.57	0.570	2949067	.534864
INC5	0336984	.1870243	-0.18	0.857	4010339	.333637
INC6	0059254	.1865874	-0.03	0.975	3724028	.360552
INC7	0174548	.1817669	-0.10	0.924	3744642	.3395545
INC8	0515112	.1831584	-0.28	0.779	4112535	.3082312
INC9	1405221	.1882215	-0.75	0.456	5102091	.2291649
INC10	.0280202	.1935777	0.14	0.885	3521868	.4082271
INC11	.0769551	.1933614	0.40	0.691	3028271	.4567373
INC12	.0979069	.1918124	0.51	0.610	2788328	.4746467
INC13	.0436501	.1922624	0.23	0.820	3339735	.4212737
INC14	.1698911	.1804075	0.94	0.347	1844483	.5242305
INC15	0	(omitted)				
INC16	.1147832	.1897697	0.60	0.546	2579445	.487511
cons	.0848277	.1910171	0.44	0.657	29035	.4600053

. regress vote_lega_euro work_everyweek \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0, robust note: EDU2 omitted because of collinearity.
note: INC15 omitted because of collinearity.

Linear regression
Number of obs = 602 F(24, 577) = 3.76 Prob > F = 0.0000 R-squared = 0.1345

Root MSE = .40823

		Robust				
vote_lega_euro	Coefficient	std. err.	t	P> t	[95% conf.	interval]
work_everyweek	0749307	.0455502	-1.65	0.101	1643951	.0145337
diesel_euro4	.2124175	.0811824	2.62	0.009	.0529685	.3718666
dummy_diesel	1065743	.0565114	-1.89	0.060	2175674	.0044187
dummy_euro_4	0618169	.060342	-1.02	0.306	1803337	.0566998
age	.0011674	.00173	0.67	0.500	0022304	.0045652
female	.1861594	.039364	4.73	0.000	.1088452	.2634737
EDU1	.1822699	.0496084	3.67	0.000	.0848348	. 279705
EDU2	0	(omitted)				
EDU3	0009294	.0402858	-0.02	0.982	080054	.0781953
EDU4	0041493	.1349662	-0.03	0.975	2692343	.2609357
INC1	1493887	.1924744	-0.78	0.438	5274246	.2286473
INC2	1842629	.2055215	-0.90	0.370	5879244	.2193986
INC3	1717894	.1898253	-0.90	0.366	5446221	.2010434
INC4	.1159658	.2066071	0.56	0.575	2898278	.5217595
INC5	0504664	.1838454	-0.27	0.784	4115541	.3106213
INC6	0314969	.1842756	-0.17	0.864	3934296	.3304359
INC7	0460969	.1795735	-0.26	0.798	3987943	.3066006
INC8	066454	.1807478	-0.37	0.713	4214578	.2885498
INC9	1397291	.1825537	-0.77	0.444	4982798	.2188216
INC10	.0339929	.1914588	0.18	0.859	3420483	.410034
INC11	.0543728	.1904131	0.29	0.775	3196145	.42836
INC12	.0779236	.189144	0.41	0.681	293571	.4494182
INC13	.0231483	.1881654	0.12	0.902	3464243	.392721
INC14	.1558875	.1783258	0.87	0.382	1943593	.5061343
INC15	0	(omitted)				
INC16	.0864903	.1874043	0.46	0.645	2815875	.454568
_cons	.1040741	.183604	0.57	0.571	2565395	.4646876
_	1					

. regress sw_to_lega_18_19 use_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018== note: EDU1 omitted because of collinearity. note: INC2 omitted because of collinearity.

Linear regression	Number of obs	=	483
	F(27, 455)	=	4.20
	Prob > F	=	0.0000
	R-squared	=	0.3155
	Root MSE	=	.25502

sw_to_1~8_19	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
use_day	0806769	.0777024	-1.04	0.300	2333769	.0720231
use_week	.1292145	.0750143	1.72	0.086	0182029	.2766319
use_month	.1282395	.0856337	1.50	0.135	0400471	.2965261
use_year	.0192108	.0895645	0.21	0.830	1568007	.1952222
diesel_euro4	.2007218	.0544188	3.69	0.000	.0937784	.3076653
dummy_diesel	0225506	.0384669	-0.59	0.558	0981455	.0530443
dummy_euro_4	0474107	.0380717	-1.25	0.214	122229	.0274075
age	.0002851	.0011287	0.25	0.801	001933	.0025032
female	.1206058	.0256967	4.69	0.000	.0701068	.1711047
EDU1	0	(omitted)				
EDU2	0115758	.0350768	-0.33	0.742	0805085	.0573569
EDU3	0025844	.0342041	-0.08	0.940	0698019	.0646331
EDU4	.1215102	.103673	1.17	0.242	082227	.3252474
INC1	1010692	.0804059	-1.26	0.209	2590823	.0569438
INC2	0	(omitted)				
INC3	0298188	.0780934	-0.38	0.703	1832873	.1236497
INC4	.2225939	.1308106	1.70	0.090	0344741	.4796619

INC5	0240505	.0529646	-0.45	0.650	1281362	.0800351
INC6	.0034923	.0680654	0.05	0.959	1302692	.1372538
INC7	0352997	.0488108	-0.72	0.470	1312223	.0606229
INC8	0479775	.0482191	-0.99	0.320	1427372	.0467821
INC9	0320821	.0694052	-0.46	0.644	1684766	.1043124
INC10	.0082073	.0757127	0.11	0.914	1405826	.1569972
INC11	0366304	.0576996	-0.63	0.526	1500212	.0767603
INC12	0239311	.0578739	-0.41	0.679	1376644	.0898023
INC13	0574742	.0523802	-1.10	0.273	1604113	.0454628
INC14	.1634951	.0548341	2.98	0.003	.0557356	.2712546
INC15	075516	.0769194	-0.98	0.327	2266773	.0756453
INC16	0055892	.0727449	-0.08	0.939	1485469	.1373684
_cons	0396602	.1126779	-0.35	0.725	2610938	.1817735

. regress sw_to_lega_18_19 everyweek \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_20 note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression

Number of obs = 483 F(24, 458) = 3.23 Prob > F = 0.0000 R-squared = 0.2305 Root MSE = .26951

	I					
		Robust				
sw_to_l~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
everyweek	0883036	.0469147	-1.88	0.060	1804984	.0038912
diesel_euro4	.1602018	.0529307	3.03	0.003	.0561847	.2642189
dummy_diesel	0398615	.0393285	-1.01	0.311	1171481	.0374251
dummy_euro_4	0040796	.036322	-0.11	0.911	0754581	.0672989
age	.0014693	.0010995	1.34	0.182	0006914	.00363
female	.1609392	.0289709	5.56	0.000	.1040069	.2178715
EDU1	0	(omitted)				
EDU2	0119942	.0329056	-0.36	0.716	0766589	.0526705
EDU3	.0163576	.0328599	0.50	0.619	0482172	.0809324
EDU4	.0792006	.1363524	0.58	0.562	1887534	.3471546
INC1	.0172687	.0676416	0.26	0.799	1156577	.1501952
INC2	0	(omitted)				
INC3	.0707843	.0820996	0.86	0.389	0905543	.232123
INC4	.290155	.1202649	2.41	0.016	.0538155	.5264945
INC5	.0267994	.0449127	0.60	0.551	0614611	.11506
INC6	.0877966	.05839	1.50	0.133	026949	.2025421
INC7	.0054602	.0400743	0.14	0.892	073292	.0842124
INC8	.0038692	.0384264	0.10	0.920	0716447	.0793832
INC9	.04449	.0582383	0.76	0.445	0699574	.1589374
INC10	.0974939	.0691425	1.41	0.159	038382	.2333698
INC11	.0169	.0514671	0.33	0.743	0842409	.1180408
INC12	.0361382	.0465261	0.78	0.438	055293	.1275694
INC13	.0094561	.0424152	0.22	0.824	0738965	.0928087
INC14	.2096961	.0518528	4.04	0.000	.1077972	.311595
INC15	.0204369	.0717533	0.28	0.776	1205697	.1614435
INC16	.0902836	.0650866	1.39	0.166	0376218	.218189
_cons	0903043	.0885614	-1.02	0.308	2643413	.0837327

. regress sw_to_lega_18_19 workcar_* \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_20

note: **EDU1** omitted because of collinearity. note: **INC2** omitted because of collinearity.

Linear regression Number of obs = 483

F(27, 455) = **4.27** Prob > F = **0.0000** R-squared = **0.3235**

Root MSE = .25352

		Dalamat				
		Robust				
sw_to_le~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
workcar_day	0995564	.0364517	-2.73	0.007	1711909	0279218
workcar_week	.144694	.0407432	3.55	0.000	.0646259	.2247621
workcar_month	.1084388	.0794184	1.37	0.173	0476336	.2645112
workcar_year	.0591692	.0952984	0.62	0.535	1281104	. 2464489
diesel_euro4	.186206	.0580788	3.21	0.001	.07207	.3003421
dummy_diesel	0342316	.039519	-0.87	0.387	111894	.0434307
dummy_euro_4	0376069	.0378879	-0.99	0.321	112064	.0368501
age	.0013752	.0011524	1.19	0.233	0008895	.00364
female	.1225782	.0260434	4.71	0.000	.071398	.1737584
EDU1	0	(omitted)				
EDU2	0034549	.0369128	-0.09	0.925	0759955	.0690857
EDU3	.0149142	.0344586	0.43	0.665	0528037	.082632
EDU4	.0616239	.1081985	0.57	0.569	1510069	.2742546
INC1	0911848	.0787209	-1.16	0.247	2458864	.0635168
INC2	0	(omitted)				
INC3	031425	.0731374	-0.43	0.668	1751539	.1123039
INC4	.1995199	.1373468	1.45	0.147	0703927	.4694326
INC5	0568321	.0466018	-1.22	0.223	1484136	.0347495
INC6	.0193342	.0613532	0.32	0.753	1012366	.1399049
INC7	0593902	.0409902	-1.45	0.148	1399438	.0211635
INC8	0759296	.0401754	-1.89	0.059	1548818	.0030227
INC9	0592829	.0673918	-0.88	0.379	1917208	.0731549
INC10	0202888	.0719265	-0.28	0.778	1616381	.1210605
INC11	0593007	.0506066	-1.17	0.242	1587522	.0401509
INC12	0446146	.0495785	-0.90	0.369	1420458	.0528167
INC13	0732027	.0467029	-1.57	0.118	1649829	.0185774
INC14	.1231335	.0471463	2.61	0.009	.0304819	.215785
INC15	1052665	.061216	-1.72	0.086	2255677	.0150347
INC16	.0044638	.0678666	0.07	0.948	1289071	.1378347
_cons	0516398	.0816294	-0.63	0.527	2120572	.1087776

. regress sw_to_lega_18_19 work_everyweek \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_euro==0 to note: EDU1 omitted because of collinearity.

note: INC2 omitted because of collinearity.

Linear regression
Number of obs = 483 F(24, 458) = 3.20 Prob > F = 0.0000 R-squared = 0.2274

Root MSE = .27005

		Robust				
sw_to_leg~8_19	Coefficient	std. err.	t	P> t	[95% conf.	interval]
work_everyweek	0507285	.0339835	-1.49	0.136	1175115	.0160544
diesel_euro4	.1760923	.059332	2.97	0.003	.0594956	. 292689
dummy_diesel	0454877	.0404076	-1.13	0.261	124895	.0339196
dummy_euro_4	0156356	.0399414	-0.39	0.696	0941267	.0628555
age	.0014162	.0011571	1.22	0.222	0008576	.00369
female	.1585477	.0290539	5.46	0.000	.1014522	.2156432
EDU1	0	(omitted)				
EDU2	0122846	.0331491	-0.37	0.711	0774278	.0528587
EDU3	.021283	.0333547	0.64	0.524	0442643	.0868302
EDU4	.0775663	.1366892	0.57	0.571	1910494	.346182
INC1	.0047994	.0709445	0.07	0.946	1346178	.1442166
INC2	0	(omitted)				
INC3	.0669736	.07488	0.89	0.372	0801773	.2141246
INC4	.2881912	.1269112	2.27	0.024	.0387907	.5375917
INC5	.029605	.0452017	0.65	0.513	0592234	.1184334
INC6	.0837023	.0596007	1.40	0.161	0334225	.2008271
INC7	0034889	.0392868	-0.09	0.929	0806936	.0737158
INC8	0031446	.037921	-0.08	0.934	0776653	.0713761
INC9	.0385006	.0598037	0.64	0.520	079023	.1560242
INC10	.0870622	.0697421	1.25	0.213	049992	.2241164
INC11	.0197072	.0515174	0.38	0.702	0815326	.1209469
INC12	.0343623	.0459195	0.75	0.455	0558768	.1246014
INC13	.0113469	.0418021	0.27	0.786	0708008	.0934946
INC14	.2056621	.0509259	4.04	0.000	.1055846	.3057395
INC15	.0187163	.0632168	0.30	0.767	1055147	.1429473
INC16	.0787455	.0623403	1.26	0.207	0437629	.201254
cons	1237529	.0906983	-1.36	0.173	3019893	.0544835
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. regress switch_descriptive \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018_rob==0, robust

switch_des~e	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1251704	.048229	2.60	0.010	.0304053	.2199354
dummy_diesel	0265414	.0334758	-0.79	0.428	0923179	.039235
dummy_euro_4	0118846	.0362749	-0.33	0.743	083161	.0593919
cons	.0645161	.0255803	2.52	0.012	.0142535	.1147788

. regress switch_descriptive \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_2018_rob==

note: EDU4 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 486

F(23, 462) = **3.34** Prob > F = **0.0000**

R-squared = **0.2140** Root MSE = **.27379**

		Robust				
switch_des~e	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.16575	.0543168	3.05	0.002	.0590115	.2724886
dummy_diesel	0547889	.0412773	-1.33	0.185	1359035	.0263257
dummy_euro_4	0154954	.0387477	-0.40	0.689	0916389	.0606481
age	.0018942	.0010815	1.75	0.081	000231	.0040194
female	.1556982	.0296144	5.26	0.000	.0975026	.2138938
EDU1	0806443	.1364948	-0.59	0.555	3488719	.1875834
EDU2	0977102	.137907	-0.71	0.479	3687129	.1732924
EDU3	0628611	.1390068	-0.45	0.651	336025	.2103028
EDU4	0	(omitted)				
INC1	.03649	.0647756	0.56	0.573	0908012	.1637812
INC2	0	(omitted)				
INC3	.0849295	.0779943	1.09	0.277	068338	.238197
INC4	.3073508	.1255514	2.45	0.015	.0606283	.5540733
INC5	.0748452	.0553478	1.35	0.177	0339195	.1836098
INC6	.098931	.0567925	1.74	0.082	0126726	.2105346
INC7	.0094808	.0358605	0.26	0.792	0609891	.0799508
INC8	.011065	.0343152	0.32	0.747	0563682	.0784983
INC9	.0563486	.0535139	1.05	0.293	0488123	.1615094
INC10	.1039323	.0703757	1.48	0.140	0343639	.2422285
INC11	.0406544	.0496686	0.82	0.413	05695	.1382589
INC12	.0467995	.042905	1.09	0.276	0375136	.1311126
INC13	.0187816	.0386573	0.49	0.627	0571844	.0947475
INC14	.2134103	.0488898	4.37	0.000	.1173364	.3094843
INC15	.0335153	.0640818	0.52	0.601	0924126	.1594432
INC16	.1003589	.0618087	1.62	0.105	0211022	.2218199
_cons	1002812	.1504579	-0.67	0.505	3959479	.1953854

. regress switch_descriptive \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_answer_note: EDU4 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression Number of obs =

F(24, 489) = 3.09 Prob > F = 0.0000 R-squared = 0.2040 Root MSE = .26863

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		Robust				
switch_descript~e	Coefficient		t	P> t	[95% conf.	interval]
diesel_euro4_ass	.1571823	.0529318	2.97	0.003	.0531804	.2611842
<pre>dummy_diesel_ass</pre>	0565426	.0379912	-1.49	0.137	1311888	.0181035
dummy_euro_4_ass	0215844	.0382952	-0.56	0.573	0968278	.053659
age	.0019349	.001018	1.90	0.058	0000653	.0039351
female	.1431563	.028364	5.05	0.000	.087426	.1988866
EDU1	0809144	.1387746	-0.58	0.560	3535826	.1917537
EDU2	0982438	.1402582	-0.70	0.484	3738269	.1773392
EDU3	0645162	.1414958	-0.46	0.649	342531	.2134986
EDU4	0	(omitted)				
INC1	.0097801	.0860669	0.11	0.910	1593265	.1788867
INC2	0309095	.0611023	-0.51	0.613	150965	.089146
INC3	.0611688	.088288	0.69	0.489	1123018	.2346395
INC4	.2275625	.1069851	2.13	0.034	.0173552	.4377697
INC5	.0444475	.0745266	0.60	0.551	1019842	.1908793
INC6	.0681986	.0748659	0.91	0.363	0789001	.2152972
INC7	0129091	.0580403	-0.22	0.824	1269483	.1011302
INC8	013412	.05802	-0.23	0.817	1274112	.1005872
INC9	.029425	.0703616	0.42	0.676	1088234	.1676735
INC10	.0707565	.0787467	0.90	0.369	0839672	.2254801
INC11	.0256506	.0666759	0.38	0.701	1053561	.1566572
INC12	.0338188	.0599231	0.56	0.573	0839198	.1515575
INC13	0102948	.0590658	-0.17	0.862	1263489	.1057593
INC14	.183051	.0634129	2.89	0.004	.0584556	.3076464
INC15	0	(omitted)				
INC16	.0847776	.0803289	1.06	0.292	0730549	.24261
dummy car unknown	1855606	.0393925	-4.71	0.000	2629601	1081612
_cons	0595879	.1567131	-0.38	0.704	3675021	.2483262

. regress switch_descriptive_reg \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_regional_rob==0, r

Linear regression	Number of obs	=	475
_	F(3, 471)	=	7.98
	Prob > F	=	0.0000
	R-squared	=	0.0438
	Root MSE	=	.304

switch_des~g	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1177514	.046019	2.56	0.011	.0273234	.2081795
dummy_diesel	.0240964	.0333342	0.72	0.470	0414057	.0895985
dummy_euro_4	0071591	.0288906	-0.25	0.804	0639296	.0496114
_cons	.0361446	.0205743	1.76	0.080	0042842	.0765733

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. regress switch_descriptive_reg \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_region note: EDU4 omitted because of collinearity.

note: INC15 omitted because of collinearity.

R-squared = **0.2701** Root MSE = **.27143**

		Robust				
switch_des~g	Coefficient	std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1585404	.0547945	2.89	0.004	.0508562	.2662245
dummy_diesel	0136357	.0411147	-0.33	0.740	0944358	.0671644
dummy_euro_4	0144936	.0374001	-0.39	0.699	0879937	.0590066
age	.0032814	.0009248	3.55	0.000	.001464	.0050989
female	.1924495	.0303631	6.34	0.000	.1327789	.2521202
EDU1	0899322	.1332275	-0.68	0.500	3517559	.1718915
EDU2	1732652	.1316148	-1.32	0.189	4319197	.0853893
EDU3	1361793	.1327782	-1.03	0.306	39712	.1247615
EDU4	0	(omitted)				
INC1	.0212121	.1106251	0.19	0.848	1961925	.2386168
INC2	0545011	.0875727	-0.62	0.534	2266023	.1176002
INC3	0146372	.0891724	-0.16	0.870	1898822	.1606077
INC4	.1510666	.1234495	1.22	0.222	091541	.3936741
INC5	.0072449	.0937889	0.08	0.938	1770725	.1915624
INC6	.0523576	.0986302	0.53	0.596	1414743	.2461894
INC7	.0359001	.0918494	0.39	0.696	1446059	.2164061
INC8	.0215012	.0892244	0.24	0.810	153846	.1968484
INC9	0083127	.0847709	-0.10	0.922	1749077	.1582824
INC10	.0243853	.0837842	0.29	0.771	1402706	.1890413
INC11	020368	.0871161	-0.23	0.815	1915718	.1508358
INC12	.0793255	.0986304	0.80	0.422	1145068	.2731578
INC13	.0059995	.0857885	0.07	0.944	1625953	.1745943
INC14	.2150553	.0887131	2.42	0.016	.040713	.3893976
INC15	0	(omitted)				
INC16	.0195341	.1010918	0.19	0.847	1791354	.2182035
cons	1146643	.1601777	-0.72	0.474	4294516	.2001231

. regress switch_descriptive_reg \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_answer_euro==

note: INC15 omitted because of collinearity.

R-squared = 0.2394 Root MSE = .27212

switch_descript~g	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4_ass dummy_diesel_ass dummy_euro_4_ass age female EDU1 EDU2	.1322324 0002021 0215215 .0031999 .1721551 1000495 1779017	.0576287 .0417074 .0368683 .0008486 .0302006 .1413704	2.29 -0.00 -0.58 3.77 5.70 -0.71	0.022 0.996 0.560 0.000 0.000 0.479 0.204	.018996208215410939651 .0015325 .112813137783234528833	.2454687 .08175 .050922 .0048674 .2314972 .1777333

	1					
EDU4	0	(omitted)				
INC1	.0312493	.1045813	0.30	0.765	1742455	.2367441
INC2	0411822	.0799231	-0.52	0.607	1982255	.1158611
INC3	0003872	.0807057	-0.00	0.996	1589682	.1581937
INC4	.1264483	.1040275	1.22	0.225	0779582	.3308549
INC5	.0163274	.086339	0.19	0.850	1533226	.1859774
INC6	.0574828	.0912681	0.63	0.529	1218526	.2368182
INC7	.042704	.0850056	0.50	0.616	124326	.209734
INC8	.0309491	.0820376	0.38	0.706	130249	.1921472
INC9	0032363	.0770029	-0.04	0.966	1545415	.1480689
INC10	.0218825	.0765077	0.29	0.775	1284496	.1722146
INC11	.0454712	.0860981	0.53	0.598	1237055	.2146479
INC12	.0854578	.0907189	0.94	0.347	0927985	.2637141
INC13	.0046228	.0784823	0.06	0.953	1495893	.158835
INC14	.2126295	.0819177	2.60	0.010	.051667	.3735919
INC15	0	(omitted)				
INC16	.0436125	.0912712	0.48	0.633	1357289	.2229539
dummy_car_unknown	1539084	.0586528	-2.62	0.009	269157	0386598
_cons	0987383	.1630233	-0.61	0.545	4190674	.2215909

. regress switch_descriptive_mun \${main} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_municipal_rob==0,

Linear regression	Number of obs	=	492
	F(3, 488)	=	5.16
	Prob > F	=	0.0016
	R-squared	=	0.0242
	Root MSE	=	.33365

switch_des~n	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1431272	.0561249	2.55	0.011	.032851	.2534035
dummy_diesel	006917	.0437494	-0.16	0.874	0928774	.0790435
dummy_euro_4	0537084	.0399051	-1.35	0.179	1321155	.0246986
cons	.0978261	.0310993	3.15	0.002	.0367209	.1589312

.
. regress switch_descriptive_mun \${main} \${controls} if target!=3 & target!=4 & no_answer_euro==0 & no_answer_munici
note: EDU1 omitted because of collinearity.

note: INC15 omitted because of collinearity.

Linear regression	Number of obs	=	492
_	F(23, 468)	=	3.85
	Prob > F	=	0.0000
	R-squared	=	0.1899
	Root MSE	=	.31044

	I					
switch_des~n	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
diesel_euro4	.1844073	.062747	2.94	0.003	.0611066	.307708
dummy_diesel	0359289	.0488211	-0.74	0.462	1318645	.0600068
dummy_euro_4	0650738	.0450037	-1.45	0.149	1535081	.0233605
age	.0012119	.001249	0.97	0.332	0012424	.0036662
female	.1626308	.0342819	4.74	0.000	.0952654	.2299963
EDU1	0	(omitted)				
EDU2	1369632	.0425322	-3.22	0.001	220541	0533855
EDU3	0963403	.0437612	-2.20	0.028	182333	0103475
EDU4	0325144	.1341872	-0.24	0.809	2961984	.2311696
INC1	159593	.1873602	-0.85	0.395	5277644	.2085785

INC2	2514945	.1714468	-1.47	0.143	5883953	.0854062
INC3	1460877	.1842245	-0.79	0.428	5080972	.2159218
INC4	0508537	.1874115	-0.27	0.786	419126	.3174186
INC5	0728526	.1789158	-0.41	0.684	4244303	.2787251
INC6	0739448	.1760707	-0.42	0.675	4199318	.2720423
INC7	1261598	.1681618	-0.75	0.453	4566054	.2042859
INC8	1071435	.1708961	-0.63	0.531	4429621	.2286752
INC9	1346105	.1709755	-0.79	0.431	4705852	.2013642
INC10	1118787	.1710795	-0.65	0.513	4480579	.2243004
INC11	1882631	.1658841	-1.13	0.257	5142329	.1377066
INC12	1494151	.1638167	-0.91	0.362	4713224	.1724923
INC13	1570503	.1645092	-0.95	0.340	4803183	.1662178
INC14	.0604393	.1681192	0.36	0.719	2699226	.3908013
INC15	0	(omitted)				
INC16	1185678	.1774772	-0.67	0.504	4673187	.2301831
_cons	.1469747	.1831223	0.80	0.423	212869	.5068184

. regress switch_descriptive_mun \${main_ass} \${controls} dummy_car_unknown if target!=3 & no_answer_euro==0 & no_ans note: EDU1 omitted because of collinearity.
note: INC2 omitted because of collinearity.

Linear regression Number of obs = 522 F(24, 497) = 3.52 Prob > F = 0.0000 R-squared = 0.1745Root MSE = .30946

Robust Coefficient switch_descript~n std. err. P>|t| [95% conf. interval] t diesel euro4 ass .1644503 .06333 2.60 0.010 .0400228 .2888777 .0644729 dummy_diesel_ass -.0290264 .0475884 -0.61 0.542 -.1225256 .0179466 dummy_euro_4_ass -.0688166 .04416 -1.56 0.120 -.1555799 .001132 .0011786 0.96 0.337 -.0011837 .0034477 age female .1447436 .0334476 4.33 0.000 .0790274 .2104598 EDU1 0 (omitted) EDU2 -.1395202 .0407244 -3.43 0.001 -.2195335 -.059507 EDU3 -.0916992 .0429679 0.033 -2.13 -.1761204 -.007278 EDU4 -.0264444 .1379563 -0.19 0.848 -.2974938 .2446051 INC1 .0952529 .0821677 1.16 0.247 -.0661859 .2566918 INC2 (omitted) -.0462293 INC3 .1011995 .075037 1.35 0.178 .2486283 INC4 .1650693 .0847757 1.95 0.052 -.0014936 .3316323 INC5 .1741652 .064452 2.70 0.007 .0475332 .3007973 INC6 .1714194 .0626937 2.73 0.006 .048242 .2945967 INC7 .1266645 .0534299 2.37 0.018 .0216882 .2316409 INC8 .1439119 0.012 .0569418 2.53 .0320357 .2557882 INC9 .1067125 .0529289 2.02 0.044 .0027205 .2107044 INC10 .1323404 .0648833 2.04 0.042 .0048611 .2598197 .1099418 .0463869 0.018 .0188033 .2010803 INC11 2.37 INC12 .1006458 .0467992 2.15 0.032 .0086972 .1925944 INC13 .0860594 .0425214 2.02 0.044 .0025156 .1696032 INC14 .3050518 .0525627 5.80 0.000 .2017793 .4083243 INC15 .3734892 .1863822 2.00 0.046 .0072951 .7396833 0.037 INC16 2.09 .265952 .1369731 .0656465 .0079943 dummy_car_unknown -.0960499 .0590524 -1.63 0.104 -.2120729 .0199732 -.0874763 .0727611 -1.20 0.230 -.2304336 .055481 _cons

. ***************
. * Table SI-14 *
. ***********

. * Predicting the probability to vote for Lega just based on controls, only on unaffected car owners

. regress vote_lega_euro age female EDU2-EDU4 INC2-INC16 if target==2 & no_answer_euro==0,robust

vote_lega_~o	Coefficient	Robust std. err.	t	P> t	[95% conf.	interval]
age	.0008513	.0020195	0.42	0.674	0031228	.0048254
female	.0677995	.0483678	1.40	0.162	0273797	.1629787
EDU2	1379676	.0640771	-2.15	0.032	2640601	011875
EDU3	1659439	.0714132	-2.32	0.021	3064724	0254153
EDU4	2940399	.0670894	-4.38	0.000	42606	1620198
INC2	035188	.1380863	-0.25	0.799	3069175	.2365415
INC3	.0084852	.1387773	0.06	0.951	264604	.2815744
INC4	.1727354	.14708	1.17	0.241	1166922	.4621629
INC5	.0894967	.1116577	0.80	0.423	1302261	.3092194
INC6	.1380499	.1133428	1.22	0.224	0849887	.3610885
INC7	.1469624	.1216323	1.21	0.228	0923886	.3863133
INC8	.1422324	.1279225	1.11	0.267	1094966	.3939615
INC9	0243039	.1192914	-0.20	0.839	2590484	.2104407
INC10	.1669218	.1323465	1.26	0.208	0935128	.4273564
INC11	.2427207	.1405237	1.73	0.085	0338052	.5192465
INC12	.0804941	.1312807	0.61	0.540	1778432	.3388314
INC13	.08338	.1382304	0.60	0.547	1886332	.3553932
INC14	.0650164	.1093265	0.59	0.552	1501189	.2801518
INC15	.1390528	.1870233	0.74	0.458	2289761	.5070818
INC16	.2264764	.1191212	1.90	0.058	0079331	.4608859
_cons	.1352901	.1344144	1.01	0.315	1292138	.3997941

. predict prob, xb

. * Obtaining figures reported in footnote 24 in the main text

.
. sum prob if target==1

Variable	0bs	Mean	Std. dev.	Min	Max
prob	293	.1684818	.0977771	046683	.4900079

. sum prob if target==2

Variable	0bs	Mean	Std. dev.	Min	Max
prob	412	.2377102	.1102705	0654115	.5096574

end of do-file

. log close
 name: <unnamed>

log type: smcl