Supplementary Material for: Predicting Individual Behavior with Social Networks

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Appendix A: Additional Figures and Model Parameters



Figure A1 Distribution of the number of social contacts per individual (degree) in the social network. To avoid counting mailing lists, a connection was only established between users if messages were exchanged in both directions during the two-month period.



Figure A2 Distribution of age and sex in the advertising domain and how both variables relate to probability of clicking on the ad. The area of each point is proportional to the number of individuals in the corresponding category.



Figure A3 Distribution of age and sex in the fantasy football domain and how both variables relate to probability of joining the league. The area of each point is proportional to the number of individuals in the corresponding category. The dashed line indicates the overall average participation rate.



Figure A4 Distribution of age and sex in the retail domain and how both variables relate to probability of purchase in the latter, six-month prediction period. The area of each point is proportional to the number of individuals in the corresponding category. The dashed line indicates the overall average purchasing probability.

	Demo	Demo+Social
Intercept	$-4.38(.01)^{***}$	$-4.38(.01)^{***}$
Male	$.07(.02)^{***}$	$.07(.02)^{***}$
Age30	.01(.01)	.01(.01)
Age40	$.04(.01)^{***}$	$.03(.01)^{***}$
Age50	$12(.01)^{***}$	$12(.02)^{***}$
Age60	$17(.01)^{***}$	$17(.01)^{***}$
Male:Age30	.17 (.02)***	.17 (.02)***
Male:Age40	.49 (.02)***	.49 (.02)***
Male:Age50	.94 (.02)***	.95 (.02)***
Male:Age60	$1.26(.02)^{***}$	$1.26(.02)^{***}$
AdoptingContacts		.54 (.08)***
Male:AdoptingContacts		.11 (.12)
Age30:AdoptingContacts		$26(.10)^{**}$
Age40:AdoptingContacts		$22(.09)^{**}$
Age50:AdoptingContacts		$23(.10)^{**}$
Age60:AdoptingContacts		$36(.10)^{***}$
Male:Age30:AdoptingContacts		.00(.15)
Male:Age40:AdoptingContacts		05(.14)
Male:Age50:AdoptingContacts		17(.14)
Male:Age60:AdoptingContacts		07(.14)
AIC	549.09	228.11
BIC	559.05	248.02
Log Likelihood	-264.54	-94.05
Deviance	340.98	.00
Num. obs.	20	20
*** n < 0.01 ** n < 0.05 * n < 0.1		
p < 0.01, p < 0.00, p < 0.1		

Table A1Models predicting advertising response in the apparel domain. For computational efficiency, we fit the
models via binomial regression with a logistic link function, with observations having identical values
for the predictors represented as a single observation. Thus, the number of nominal observations listed
in table are considerably smaller than the actual number of people in our dataset.

	Demo	Demo+Social	Demo+Past Participation	Demo+Past Participation+Social
Intercept	$-4.56(.01)^{***}$	$-4.60(.01)^{***}$	$-5.02(.02)^{***}$	$-5.05(.02)^{***}$
Male	2.38 (.01)***	2.30 (.01)***	1.98 (.02)***	$1.95(.02)^{***}$
Age30	1.09 (.01)***	1.02 (.02)***	.76 (.02)***	.74 (.02)***
Age40	.64 (.02)***	.56 (.02)***	.27 (.02)***	.25 (.02)***
Age50	$10(.02)^{***}$	$15(.02)^{***}$	$47(.03)^{***}$	$48(.03)^{***}$
Age60	$92(.02)^{***}$	$-1.10(.02)^{***}$	$-1.38(.03)^{***}$	$-1.41(.03)^{***}$
Male:Age30	$38(.02)^{***}$	43 (.02)***	50 (.02)***	51 (.02)***
Male:Age40	$51(.02)^{***}$	$51(.02)^{***}$	54 (.02)***	$54(.02)^{***}$
Male:Age50	$72(.02)^{***}$	$70(.02)^{***}$	$66(.03)^{***}$	$66(.03)^{***}$
Male:Age60	88 (.02)***	$78(.03)^{***}$	$72(.04)^{***}$	$72(.04)^{***}$
AdoptingContacts		.28 (.01)***		.21 (.01)***
Male:AdoptingContacts		.83 (.01)***		.48 (.02)***
Age30:AdoptingContacts		.20 (.01)***		.02 (.02)
Age40:AdoptingContacts		.28 (.01)***		.06 (.02)***
Age50:AdoptingContacts		.37 (.02)***		.09 (.02)***
Age60:AdoptingContacts		$1.10(.03)^{***}$.43 (.04)***
Male:Age30:AdoptingContacts		$38(.02)^{***}$		$21(.02)^{***}$
Male:Age40:AdoptingContacts		$52(.02)^{***}$		$28(.02)^{***}$
Male:Age50:AdoptingContacts		$52(.02)^{***}$		$22(.03)^{***}$
Male:Age60:AdoptingContacts		$89(.03)^{***}$		$31(.05)^{***}$
PastAdoption			$5.37(.04)^{***}$	5.34 (.04)***
Male:PastAdoption			$-1.28(.04)^{***}$	$-1.29(.04)^{***}$
Age30:PastAdoption			35 (.04)***	39 (.04)***
Age40:PastAdoption			.18 (.04)***	.15 (.04)***
Age50:PastAdoption			.90 (.05)***	.89 (.05)***
Age60:PastAdoption			2.05 (.06)***	1.97 (.07)***
Male:Age30:PastAdoption			.57 (.04)***	.59 (.04)***
Male:Age40:PastAdoption			.50 (.04)***	.54 (.05)***
Male:Age50:PastAdoption			.55 (.06)***	.59 (.06)***
Male:Age60:PastAdoption			.40 (.07)***	.50 (.07)***
AdoptingContacts:PastAdoption				$06(.03)^{*}$
Male:AdoptingContacts:PastAdoption				$20(.02)^{***}$
Age30:AdoptingContacts:PastAdoption				.06 (.02)*
Age40:AdoptingContacts:PastAdoption				.03 (.02)
Age50:AdoptingContacts:PastAdoption				03(.03)
Age60:AdoptingContacts:PastAdoption				$26(.04)^{***}$
AIC	1714171.39	1557774.28	30056.14	10730.02
BIC	1714211.40	1557854.31	30136.17	10874.07
Log Likelihood	-857075.69	-778867.14	-15008.07	-5329.01
Deviance	1713131.02	1556713.91	28995.77	9637.65
Num. obs.	404	404	404	404
$^{***}p < 0.001, ^{**}p < 0.01, ^{*}p < 0.05$				

Table A2Models predicting joining the recreational league. As in the advertising domain, for computational
efficiency we fit models via binomial regression with a logistic link function, so the number of nominal
observations in the table is considerably smaller than the number of individuals modeled.

	Demo	Demo+Social	Demo+Past Sales	Demo+Past Sales+Social
Intercept	002(.019)	030(.019)	$313(.022)^{***}$	$328(.023)^{***}$
Male	$180(.041)^{***}$	$176(.042)^{***}$.001(.045)	006(.046)
Age30	.030(.020)	008(.021)	.006(.023)	017(.024)
Age40	$.227(.019)^{***}$	$.153(.020)^{***}$	$.130(.023)^{***}$	$.093(.024)^{***}$
Age50	$.255(.020)^{***}$.198 (.021)***	.171 (.023)***	$.144(.024)^{***}$
Age60	.220 (.020)***	.173 (.021)***	$.169(.023)^{***}$.142 (.024)***
Male:Age30	$132(.043)^{***}$	$092(.045)^{**}$	$107(.048)^{**}$	077(.049)
Male:Age40	$209(.042)^{***}$	$147(.043)^{***}$	$161(.047)^{***}$	$126(.048)^{***}$
Male:Age50	$146(.043)^{***}$	$103(.044)^{**}$	$114(.047)^{**}$	$086(.049)^{*}$
Male:Age60	$114(.043)^{***}$	$085(.044)^{*}$	$120(.047)^{**}$	$101(.048)^{**}$
AdoptingContacts	. ,	.239 (.046)***	. ,	.148 (.056)***
Male:AdoptingContacts		.083(.128)		.158(.129)
Age30:AdoptingContacts		003(.047)		.009(.057)
Age40:AdoptingContacts		052(.047)		035(.056)
Age50:AdoptingContacts		012(.047)		016(.057)
Age60:AdoptingContacts		017(.047)		010(.057)
Male:Age30:AdoptingContacts		177(.130)		207(.132)
Male:Age40:AdoptingContacts		147(.128)		183(.129)
Male:Age50:AdoptingContacts		167(.129)		200(.130)
Male:Age60:AdoptingContacts		135(.129)		168(.130)
PastSales		· · · ·	$.003 (.000)^{***}$.003 (.000)***
Male:PastSales			$002(.000)^{***}$	$002(.000)^{***}$
Age30:PastSales			$(000.000)^*$.000 (.000)**
Age40:PastSales			$.000(.000)^{***}$	$001(.000)^{***}$
Age50:PastSales			$001(.000)^{***}$	$001(.000)^{***}$
Age60:PastSales			$001(.000)^{***}$	$001(.000)^{***}$
Male:Age30:PastSales			.000 (.000)	.000 (.000)
Male:Age40:PastSales			$.000(.000)^{**}$	$.001 (.000)^{**}$
Male:Age50:PastSales			$.001(.000)^{***}$.001 (.000)***
Male:Age60:PastSales			.001 (.000)***	.001 (.000)***
AdoptingContacts:PastSales				.000(.000)
Male:AdoptingContacts:PastSales				.000(.000)
Age30:AdoptingContacts:PastSales				.000(.000)
Age 40: Adopting Contacts: PastSales				.000(.000)
Age 50: Adopting Contacts: PastSales				.000(.000)
Age60:AdoptingContacts:PastSales				.000(.000)
AIC	789964.575	785463.448	737893.363	735918.064
BIC	790077.173	785688.646	738118.560	736323.419
Log Likelihood	-394972.287 -	-392711.724	-368926.681	-367923.032
Deviance	789944.575	785423.448	737853.363	735846.064
Num. obs.	573701	573701	573701	573701
$^{***}p < 0.01, \ ^{**}p < 0.05, \ ^*p < 0.1$				

Table A3 Models predicting retail purchases