Online Program Engagement and Audience Size during Television Ads = Web Appendices =

Web Appendix 1: Interviews with Experts from the Advertising Industry

We conducted interviews with five professionals in the television media buying and advertising industries to gain insights on how advertisers obtain information on audience size. We began our interviews via email with the question: "how long does it typically take you to know how many viewers saw a specific television ad?" Our conversations validate claims from industry citations (e.g., Crupi 2019; Friedman 2012; Lafayette 2018b; Schwarz 2019; Story 2007) that (1) data on audience size during individual ads is very rare for firms to obtain as it is almost always aggregated to either the ad break or program level, and (2) aggregate data on audience size is not available in real-time but rather takes several days for firms to obtain in the best-case scenarios. We showcase representative quotes from our interviews, alongside relevant expertise of the five professionals below (identifying information is masked to preserve anonymity).

Professional 1, television media planner and broadcast traffic manager (5 years): It will often take 90 days (one quarter) to get the full Post (post-buy analysis) based on Nielsen ratings. This likely won't be specific to certain ad slots themselves, but rather by program. So, then the buyer can review planned versus delivered GRPs for an entire schedule across multiple networks and say, we delivered 110% against our planned levels. Although, I do know in certain cases buyers can obtain an overnight rating for more premium placements (they'll know the next day or two after the Oscars what the rating was) upon request.

Professional 2, national television ad buyer (5 years) and media planner at MediaCom (4 years): With TV, you typically get Nielsen ratings 3-4 weeks after air. You can get overnight ratings, but they are not final and will not reflect C3 (Commercial Live + 3 days) or C7 (Commercial Live + 7 days) - which is how majority of the ratings are purchased/measured. It's typically not specific ad slot though. Depends on the network/program, but most program ratings from what I've been told historically are within certain time periods. I think for broadcast, it is quarter hour/half hour- 15-30 min increments.

Professional 3, executive director of research at ABC Television (8 years):

Advertisers get viewership data from third party companies, mostly Nielsen. Nielsen has a quick one-day turnaround for program ratings (not commercial ratings though). However, given that viewers watch programs delayed with DVRs and given that multiple stakeholders in the industry have an incentive to count all of the viewing, advertisers can't get final viewing numbers for C3 and C7 until about 8 days after airing. Even then, you can't get a rating for your specific commercial; you get a rating for where your commercial lived. The most granular data is typically viewership numbers for a 60 second block, so it does not give you viewership for a specific ad.

Professional 4, media manager of television ad buys (1 year) and brand manager (1 year): I believe actual live viewership is confirmed at the spot level within 7 days (although usually less) since many agencies individually confirm that each spot has run with the networks. Also depends on which viewership metrics you want - I've heard of Nielsen C3 (live + 3 days digital recorder playback) & C7 (live + 7 days).

Professional 5, media associate at ad agency Starcom (3 year): It takes about 2-3 days for advertisers to know how many viewers saw their ad. We use a program called Lake5 that congregates the information for us to pull from. Networks aggregate program ratings for viewership for ads within a show.

We conducted a follow-up interview with Professional 1 who has experience with programmatic television ad buys to improve our understanding of what information advertisers have available in such ad buys. This interview provided further support of claims from industry reports that audience size data is not typically leveraged in programmatic ad buys (e.g., Chordia 2018; Peterson 2019). Specifically, Professional 1 noted that audience size data is not used as ad buys typically do not occur at the program-level. Rather, programmatic ad buys are "based on audience segments and television genres but not specific shows." This interview provided additional support to our claim that, even in a programmatic ad buying world, advertisers could benefit from real-time insights on audience size as gained through readily available online program engagement (OPE) data.

Web Appendix 2: Additional Descriptive Statistics

Web Appendix Table 2.1 shows the correlations among the variables in our model. Web Appendix Table 2.2 illustrates the variation of our two measures of OPE both within and across programs.

	Web Appendix Table 2.1: Correlation table																				
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21
1 AudienceSizeBeg i	1.0000																				
2 AudienceSizeEnd	.9995	1.0000																			
3 LogProgram																					
WOMVolume _i	3006	3007	1.0000																		
4 LogProgram	0429	0274	0011	1 0000																	
WOMDeviation $_i$	0428	03/4	0011	1.0000																	
5 Ad position	0033	.0172	.0137	.1119	1.0000																
6 Break position	0121	0126	0840	.1378	0095	1.0000															
7 Ad length	.0128	.0047	.0170	0723	1930	.0128	1.0000														
8 Ads on other	- 0288	0291	.0399	0003	0283	- 0129	0574	1.0000													
networks	.0200	.02)1	.0377	.0003	.0203	.012)	.0374	1.0000													
9 Ad stock (same	- 0893	0903	0627	- 0804	0138	.0151	0870	0407	1 0000												
channel)	.0075	.0702	.0027	.000.	.0150	.0101	.00,0	.0.07	1.0000												
10 Ad stock																					
(different	.0569	.0540	.0459	0425	0722	.1140	.1498	0323	.3034	1.0000											
broadcast																					
channel)	0644	.0640	- 0847	0028	0044	0073	.0129	.0319	0175	0274	1.0000										
11 Monday 12 Tuesday	.0644	.0040	0847	.0348	0322	.0075	.0129	.0060	01/3		2379	1 0000									
12 Tuesday 13 Wednesday	.0200	.0198	0440	0570	0322			0180	0251		2379		1 0000								
14 Thursday	0520	0518	.3007	0030	0252	0004	.0009	.0201	.0221				2215	1 0000							
15 Friday		1495		.0191		0032								2004	1 0000						
16 Saturday	.0109		0178	.0068										0151		1 0000					
17 Sunday	.0997		1060	.0099		.0237								1710			1.0000				
18 Viewer episode																					
ratings	2140	2136	.1553	0200	0154	0259	.0286	0460	.1454	.0793	.0532	.0461	1349	0164	0693	0490	.1407	1.0000			
19 Half-hour break	.0939	.0942	0394	.0686	.0078	.1715	0007	.0110	0192	.0276	0129	.0092	0110	.0027	.0097	0120	.0048	0291	1.0000		
20 Season premiere	.0288	.0271	.1523	1235	0320	.0091	.0587	.0209	.1034	.0407	.0063	0447	.0211	.0288	0134	0100	.0026	.0877	0085	1.0000	
21 Fall finale	0486	0476	.0099	.0543	.0095	0110	0020	0133	0101	0448	0729	.0043	0420	.0258	.0317	0106	.0703	.0843	0043	1110	1.0000

Notes: See Table 2 in the main manuscript for variable definitions. The model also includes fixed effects for the ad creative and the program in which the ad airs.

Web Appendix Table 2.2: Descriptive statistics for OPE measures across programs

Program		ProgramW	VOMVolume	i	Pr	ogramW	OMDeviation	on i
	Min	Mean	(SD)	Max	Min	Mean	(SD)	Max
2 BROKE GIRLS	21.61	50.36	(47.25)	231.67	-75.63	-2.99	(17.73)	60.33
48 HOURS	7.00	9.22	(1.49)	10.73	-4.65	1.53	(5.81)	13.52
ALMOST HUMAN	77.89	137.09	(47.26)	231.00	-90.33	-25.28	(24.18)	70.45
AMAZING RACE	31.50	53.35	(25.51)	138.34	-39.70	10.07	(24.77)	104.06
AMERICAN DAD	18.50	51.71	(46.97)	187.29	-109.56	-4.83	(44.15)	249.50
AMERICA'S NEXT TOP MODEL	32.37	71.72	(30.88)	201.57	-44.27	8.78	(26.42)	132.09
ARROW	53.10	191.27	(50.80)	345.23	-159.23	7.47	(78.50)	253.80
BACK IN THE GAME	8.11	25.62	(13.82)	65.38	-22.57	-2.54	(7.92)	13.89
BEAUTY AND THE BEAST	99.88	130.85	(18.43)	198.00	-55.10	5.70	(26.42)	96.17
BETRAYAL	27.17	61.10	(21.76)	120.60	-44.00	-8.42	(16.88)	41.04
BIG BANG THEORY	47.50	158.14	(156.51)	793.50	-174.46	-4.53	(45.09)	194.33
BIGGEST LOSER	29.20	47.00	(26.84)	173.25	-77.25	42	(17.90)	86.08
BLACKLIST	102.74	184.49	(41.51)	284.28	-145.33	8.72	(114.37)	504.40
BLUE BLOODS	22.68	61.16	(39.55)	285.50	-113.60	-12.85	(23.57)	46.07
BOB'S BURGERS	33.14	59.31	(23.86)	161.86	-75.86	-14.89	(15.51)	3.96
BONES	51.25	136.24	(58.85)	343.43	-96.86	-4.57	(44.79)	187.52
BROOKLYN NINE-NINE	29.88	69.26	(44.20)	231.10	-75.44	-10.96	(20.44)	44.85
CARRIE DIARIES	25.48	42.25	(18.98)	103.17	-47.52	1.45	(19.76)	85.12
CASTLE	160.58	273.49	(121.33)	770.08	-321.29	-41.19	(81.57)	266.14
CHICAGO FIRE	169.89	299.96	(90.69)	602.21	-361.94	12.17	(142.36)	630.02
CRAZY ONES	15.93	30.71	(23.17)	121.75	-40.75	-4.09	(9.25)	15.27
CRIMINAL MINDS	70.82	201.87	(175.83)	1169.14	-449.14	-8.67	(83.06)	280.77
CSI	14.21	39.21	(30.42)	187.75	-59.83	-6.41	(20.84)	153.09
DADS	39.69	62.25	(22.18)	144.78	-19.72	-1.97	(9.01)	23.13
DANCING WITH THE STARS	127.98	235.88	(182.68)	1075.77	-307.49	49.17	(185.87)	1273.82
DATELINE	15.69	21.40	(3.67)	32.50	-10.81	66	(4.97)	11.81
DRACULA	69.14	121.84	(57.75)	273.59	-82.59	-16.06	(28.71)	79.35
ELEMENTARY	32.02	51.26	(21.76)	145.29	-52.64	-6.41	(14.22)	39.00
FAMILY GUY	56.00	189.36	(152.67)	777.89	-153.00	62.08	(501.21)	3507.11
GLEE	379.02	1409.34	(1396.89)	5885.69	-2271.52	-215.73	(572.38)	2846.53
GOLDBERGS	25.48	55.69	(34.56)	176.91	-43.06	-6.02	(15.47)	24.41
GOOD WIFE	.00	36.31	(20.65)	114.00	-58.00	9.74	(25.59)	144.45
GREY'S ANATOMY	127.17	368.95	(314.46)	1758.67	-803.83	-34.85	(175.69)	739.19
GRIMM	57.44	101.65	(48.44)	271.55	-77.50	-13.62	(21.39)	34.75
HART OF DIXIE	12.26	23.37	(10.76)	64.44	-27.48	-2.66	(9.72)	47.00
HAWAII FIVE-0	13.34	29.20	(23.71)	135.67	-28.88	-4.32	(9.58)	28.40
HOSTAGES	23.62	53.65	(35.82)	164.30	-100.07	-6.05	(26.66)	88.88
HOW I MET YOUR MOTHER	73.04	208.36	(262.77)	1255.25	-400.58	29.68	(163.75)	659.50
IRONSIDE	16.68	54.88	(33.91)	126.11	-60.11	-15.86	(12.89)	5.70
LAST MAN STANDING	4.71	13.34	(7.31)	31.57	-13.18	-1.86	(4.38)	15.14
LAW & ORDER: SVU	120.23	373.16	(288.81)	1616.00	-603.94	-8.88	(165.98)	587.95
LUCKY 7	16.15	21.89	(2.42)	25.18	-12.18	-1.68	(6.13)	8.53

Web Appendix Table 2.2: Descriptive statistics for OPE measures across programs (continued)

Program		_	VOMV olume			-	OMDeviatio	•
	Min	Mean	(SD)	Max	Min	Mean	(SD)	Max
MARVEL'S AGENTS OF	112.89	294.25	(266.21)	1557.18	-415.54	-8.42	(103.66)	259.65
S.H.I.E.L.D.								
MASTERCHEF JUNIOR	20.03	36.25	(18.53)	89.00	-25.26	3.35	(11.03)	32.38
MENTALIST	17.21	55.53	(33.35)	135.75	-30.09	4.69	(27.05)	136.34
MICHAEL J. FOX SHOW	10.00	36.10	(36.23)	153.60	-57.60	-7.79	(12.78)	19.50
MIDDLE	53.50	74.21	(15.89)	129.29	-39.55	-1.53	(13.17)	18.82
MILLERS	6.78	17.08	(13.50)	61.00	-15.14	-1.61	(6.01)	11.82
MINDY PROJECT	39.63	96.22	(48.00)	280.50	-73.50	-2.34	(51.37)	222.22
MODERN FAMILY	57.83	143.72	(76.31)	391.87	-136.93	-33.24	(40.16)	68.28
MOM	.00	2.52	(2.61)	10.50	-5.54	.04	(2.07)	6.60
NASHVILLE	93.68	159.44	(83.90)	518.09	-207.05	-4.48	(60.88)	175.30
NCIS: LOS ANGELES	18.36	47.77	(48.20)	237.93	-94.00	3.62	(42.36)	261.07
NCIS	43.19	116.03	(134.74)	824.00	-65.29	3.91	(59.40)	350.52
NEIGHBORS	17.79	24.18	(4.94)	38.74	-12.38	-2.31	(5.80)	11.59
NEW GIRL	74.53	230.86	(166.72)	904.00	-397.50	-39.04	(77.05)	65.00
ONCE UPON A TIME IN	30.75	77.33	(56.73)	218.93	-58.31	-6.77	(16.68)	28.66
WONDERLAND			· ·				,	
ONCE UPON A TIME	283.19	419.71	(102.32)	766.00	-196.00	38.18	(175.51)	972.68
ORIGINALS	206.03	359.15	(225.33)	1290.58	-645.58	13.61	(167.77)	631.73
PARENTHOOD	26.65	52.29	(28.96)	152.00	-86.00	-3.11	(22.69)	61.88
PARKS AND RECREATION	55.03	103.39	(44.74)	273.00	-105.00	-11.86	(32.15)	36.05
PERSON OF INTEREST	27.87	59.66	(30.78)	212.00	-86.00	2.70	(46.26)	307.83
RAISING HOPE	7.21	13.27	(4.71)	27.20	-11.07	60	(4.23)	10.80
REIGN	68.83	160.38	(59.15)	297.43	-87.59	10.55	(57.56)	305.44
REVENGE	165.18	291.03	(159.78)	1098.44	-433.95	-14.84	(98.43)	303.06
REVOLUTION	32.50	100.85	(38.06)	237.50	-78.75	-1.57	(33.97)	129.51
SCANDAL	1366.00	3104.47	(1510.19)	8778.50	-2212.50	185.85	(1191.36)	6502.9
SEAN SAVES THE WORLD	15.05	27.94	(11.88)	67.83	-21.08	-2.44	(6.79)	8.48
SHARK TANK	66.69	87.87	(14.80)	151.80	-54.80	.41	(22.98)	91.05
SIMPSONS	35.54	122.23	(82.40)	358.75	-119.22	-17.72	(44.45)	152.67
SLEEPY HOLLOW	214.42	382.51	(118.32)	754.50	-245.38	-49.21	(97.99)	261.89
SUPER FUN NIGHT	18.04	55.71	(48.99)	199.64	-73.65	-4.91	(22.63)	59.95
SUPERNATURAL	348.38	568.25	(194.75)	1116.44	-508.85	73.93	(293.36)	1359.6
SURVIVOR	21.43	167.92	(100.67)	575.60	-212.11	17.98	(92.65)	412.27
TOMORROW PEOPLE	45.89	92.68	(39.08)	226.17	-76.00	-8.67	(42.28)	297.97
TROPHY WIFE	24.00	34.74	(7.13)	54.82	-17.82	-1.89	(9.43)	32.54
TWO AND A HALF MEN	5.39	10.30	(4.81)	31.80	-7.40	56	(3.21)	6.60
UNDERCOVER BOSS	5.00	12.95	(5.20)	32.25	-15.25	9.37	(18.31)	82.47
VAMPIRE DIARIES	250.74	502.48	(369.48)	2014.00	-821.22	56.92	(251.96)	1070.2
VOICE	200.28	589.38	(329.25)	2722.63	-558.87		(2563.71)	
WE ARE MEN	15.26	42.37	(19.79)	65.60	-21.00	-2.37	(15.16)	17.96
WELCOME TO THE FAMILY	16.04	23.41	(4.18)	29.83	-8.57	-2.22	(3.97)	6.17
X FACTOR	66.90	669.40	(345.01)	2369.00		15.19	(233.08)	1342.4
Total	.00	266.56	(510.46)			23.91	(667.86)	

Web Appendix 3: Supplementary Analyses

Web Appendix 3.1: Alternative audience size measures

As discussed in the main manuscript, the 30-second intervals of the audience size data do not always line up with an ad's beginning and end. As a remedy, we determine audience size at the beginning and end of ad instance i using three approaches. For our primary approach in the main analysis, we treat the audience size data as constant such that the audience size for a given program recorded at 8:00:00 PM would be used to represent the number of households tuned into that program from 8:00:00 PM-8:00:29 PM (Constant Audience Size Measure). We also estimate two alternative approaches. First, we consider an operationalization that uses the audience size estimate recorded closest to a given ad's start or end time (Closest Audience Size Measure). For example, if an ad began airing at 8:00:16 PM, then the audience size recorded at 8:00:30 PM, instead of 8:00:00 PM, would be used to represent the number of households tuned in at the start of that ad. Second, we consider an approach that measures audience size using the last audience size estimate recorded before the start of ad i and the first estimate recorded after its end (First/Last Audience Size Measure). In Web Appendix Table 3.1.1, we provide examples of how audience size at the start and end of a given ad varies across these three approaches for three different ad start times and two different ad lengths. We estimate our model with these two alternative approaches to measuring audience size and the key results, shown in Web Appendix Table 3.1.2, are consistent with those of our proposed approach.

Web Appendix Table 3.1.1: Examples of different approaches to measuring audience size at beginning and end of ads

		Constant Audien	ce Size Measure	Closest Audiend	ce Size Measure	First/Last Audier	nce Size Measure
Ad begins	Ad length	Measurement window for	Measurement window for <i>AudienceSize</i>	Measurement window for AudienceSize	Measurement window for	Measurement window for	Measurement window for
		AudienceSize Beg _i	End_i	Beg_i	$Audience Size \ End_i$	AudienceSize Beg _i	AudienceSize End _i
8:00:14	15 secs	8:00:00 -8:00:29	8:00:00 -8:00:29	8:00:00 -8:00:29	8:00:30 -8:00:59	8:00:00 -8:00:29	8:00:30 -8:00:59
8:00:14	30 secs	8:00:00 -8:00:29	8:00:30 -8:00:59	8:00:00 -8:00:29	8:00:30 -8:00:59	8:00:00 -8:00:29	8:01:00 -8:01:29
8:00:16	15 secs	8:00:00 -8:00:29	8:00:30 -8:00:59	8:00:30 -8:00:59	8:00:30 -8:00:59	8:00:00 -8:00:29	8:01:00 -8:01:29
8:00:16	30 secs	8:00:00 -8:00:29	8:00:30 -8:00:59	8:00:30 -8:00:59	8:01:00 -8:01:29	8:00:00 -8:00:29	8:01:00 -8:01:29
8:00:30	15 secs	8:00:30 -8:00:59	8:00:30 -8:00:59	8:00:30 -8:00:59	8:01:00 -8:01:29	8:00:30 -8:00:59	8:01:00 -8:01:29
8:00:30	30 secs	8:00:30 -8:00:59	8:01:00 -8:01:29	8:00:30 -8:00:59	8:01:00 -8:01:29	8:00:30 -8:00:59	8:01:00 -8:01:29

Web Appendix Table 3.1.2: Key results from alternative audience size measure analyses

Variable	0.00.	est Audien e measure		First/La Size	ce		
	Est	imate (SE))	Estimate (SE)			
Ad position	.005	(.000)	**	.010	(.000)	**	
OPE							
$LogProgramWOMVolume_i$	9.323	(2.461)	**	14.660	(3.863)	**	
$LogProgramWOMDeviation_i$.000	(000.)	**	.001	(000.)	**	
Interaction with ad position							
$LogProgramWOMVolumei \times Ad$ position	192	(.625)		-1.253	(.981)		
$LogProgramWOMDeviationi \times Ad$ position	000	(.000)	**	.000	(000.)	**	
Adjusted R-squared		.9997		.9992			

^{*} *p* < .10, ** *p* < .05

Web Appendix 3.2: Alternative operationalizations of the dependent variable

For robustness, we test five alternative dependent variables: (1) our primary outcome without the log transformation: $AudienceSizeEnd_i$ (with $AudienceSizeBeg_i$ as a control variable); (2) percentage change in audience size $(AudienceSizePC_i = (AudienceSizeEnd_i - AudienceSizeBeg_i)/AudienceSizeBeg_i)$; (3) log of the percentage change in audience size $(LogAudienceSizePC_i = log(AudienceSizePC_i + 1))^1$; (4) ratio of audience size from the end to the beginning of the ad $(AudienceSizeRatio_i = AudienceSizeEnd_i/AudienceSizeBeg_i)$; and (5) log of the ratio of audience size from the end to the beginning of the ad $(LogAudienceSizeRatio_i = log(AudienceSizeRatio_i + 1))$. In models (2)-(5), audience size at the beginning of ad i is not included as a control since it is incorporated into the outcome measure. The key results, shown in Web Appendix Table 3.2, are consistent with our proposed approach.

¹ If the percentage change is negative, we take the log transformation of the absolute value of $AudienceSizePC_i$ plus 1 and then multiply this by -1.

Web Appendix Table 3.2: Key results from alternative operationalizations of the dependent variable

Variable	AudienceSizeEnd _i Estimate (SE)			Audie	enceSizePC	i	LogAudienceSizePC _i			AudienceSize Ratio _i			_	LogAudience SizeRatio _i	
				Estimate (SE)			Estimate (SE)			Est	imate (SE))	Est	Estimate (SE)	
Ad position OPE	3.4E+04	(6.4E+02)	**	.710	(.010)	**	.304	(.003)	**	.007	(.000.)	**	.004	(.000)	**
$LogProgramWOMVolume_i$	3.9E+07	(1.8E+07)	**	575.100	(285.700)	**	173.600	(98.270)	*	5.754	(2.857)	**	2.928	(1.452)	**
LogProgramWOMDeviation _i Interaction with ad position	1.9E+03	(3.4E+02)	**	.055	(.005)	**	.021	(.002)	**	.001	(.000)	**	.000	(.000)	**
$LogProgramWOMVolume_i \times Ad\ position$	-4.2E+07	(4.7E+06)	**	-82.350	(72.790)		-13.030	(25.040)		825	(.728)		422	(.370)	
LogProgramWOMDeviation _i × Ad position	-8.5E+02	(1.7E+02)	**	018	(.003)	**	005	(.001)	**	000	(.000)	**	000	(.000)	**
Adjusted R-squared	.9986		.5143			.6162			.5143			.5117			

Notes: Measures for OPE and ad position are mean-centered for ease of interpretation. * p < .10, ** p < .05

Web Appendix 3.3: Proposed mechanism

Web Appendix Table 3.3 shows the results from the proposed mechanism tests, as detailed in the main manuscript.

Web Appendix Table 3.3: Key results from tests of the proposed mechanism

Variable	-	-involvem ondition	ent	Lower-involvement condition			
	Esti	mate (SE)	ı	Est	imate (SE)		
Peak primetime versus	other time	s in prim	etim	e			
Ad position	.008	(.000.)	**	.007	(.00.)	**	
OPE							
$LogProgramWOMVolume_i$	17.230	(5.245)	**	4.575	(4.36)		
$LogProgramWOMDeviation_i$.001	(000.)	**	.000	(.00)	**	
Interaction with ad position							
$LogProgramWOMVolume_i \times Ad$ position	705	(1.316)		-1.658	(1.07)		
$LogProgramWOMDeviation_i \times Ad$ position	000	(000.)	**	000	(.00)	**	
Adjusted R-squared		.9995			.9995		
Second-half of episode ve	ersus first	-half of e	pisod	le			
Ad position	.008	(000.)	**	.006	(000.)	**	
OPE							
$LogProgramWOMVolume_i$	13.140	(5.100)	**	10.490	(4.086)	**	
$LogProgramWOMDeviation_i$.001	(000.)	**	.001	(000.)	**	
Interaction with ad position							
$LogProgramWOMVolume_i \times Ad$ position	-1.140	(1.299)		.266	(1.035)		
$LogProgramWOMDeviation_i \times Ad$ position	000	(000.)	**	.000	(000.)	**	
Adjusted R-squared		.9994			.9996		
Older programs ver	sus newer	· progran	18				
Ad position	.007	(000.)	**	.007	(000.)	**	
OPE							
$LogProgramWOMVolume_i$	8.005	(3.496)	**	6.618	(13.010)		
$LogProgramWOMDeviation_i$.001	(000.)	**	.000	(000.)	**	
Interaction with ad position							
$LogProgramWOMVolume_i \times Ad\ position$	867	(.907)		.178	(2.095)		
$LogProgramWOMDeviation_i \times Ad$ position	000	(000.)	**	000	(.000)	**	
Adjusted R-squared		.9994		.9996			

^{*} *p* < .10, ** *p* < .05

Web Appendix 3.4: Ad WOM as a measure of attention paid to ads

To test whether ads aired in more social episodes or after more social moments see less attention, we use a measure of ad WOM: the change in the relative volume of brand-related WOM by audience size after its ad was aired compared to before it was aired. We obtain this data for the 248 brands in our data at the second-level from Crimson Hexagon. Akin to our approach for program-related WOM, we tally the brand-related Twitter mentions, capturing Tweets mentioning the brand, a hashtag featuring the brand name, a hashtag included in the brand's ad, or the brand's Twitter handle. We model ad WOM as per Equation (1), that is, as a function of OPE (volume and deviation), ad position, and the ad and program control variables from our main model. For ease of interpretation, we mean-center the measures for OPE and ad position.

In contrast to concerns that viewers of social episodes might pay less attention to ads even though they do not change the channel, the results in Web Appendix Table 3.4 reveal a marginally significant positive relationship between OPE volume and ad WOM (β = .0006, p = .054). This positive relationship is stronger for earlier ads in an ad break (β = -.0003, p < .001). We do not find a significant relationship between OPE deviation and ad WOM (β = -.0000, p = .430), and this relationship does also not depend on ad position (β = .0000, p = .882).

Web Appendix Table 3.4: Key results from analysis of ad WOM

Variable	Estimate (SE)					
Ad position	0000	(.0000)				
OPE						
$LogProgramWOMVolume_i$.0006	(.0003) *				
$LogProgramWOMDeviation_i$	0000	(0000)				
Interaction with ad position						
$LogProgramWOMVolume_i \times Ad\ position$	0003	(.0001) **				
$LogProgramWOMDeviation_i \times Ad position$	0000	(0000)				
Adjusted R-squared		.14				

^{*} p < .10, ** p < .05

Web Appendix 3.5: Alternative operationalizations of OPE volume

For robustness, we test three alternative operationalizations of OPE volume: (1) main specification of OPE volume without the log transformation; (2) absolute measure of OPE volume, that is, our main specification but without dividing by the number of viewers at the beginning of the focal ad and without taking the log of this value; and (3) log transformation of absolute measure of OPE volume, that is, specification as per (2) but with a log transformation. The key results, shown in Web Appendix Table 3.5, are consistent with our main results and illustrate the robustness of our findings to alternative operationalizations of OPE volume.

Web Appendix Table 3.5: Key results from alternative operationalizations of OPE volume

Variable	measu	in volume re without sformatio	log		lute volu 1easure	те		Log of absolute volume measure			
	Estimate (SE) Estimate (SE)						Esti	mate (SI	Ξ)		
Ad position	.007	(.000.)	**	.007	(.000.)	**	.007	(.000.)	**		
OPE											
$LogProgramWOMVolume_i$	9.476	(2.988)	**	.000	(000.)	**	.002	(.001)	**		
$LogProgramWOMDeviation_i$.001	(000.)	**	.001	(.000)	**	.001	(.000)	**		
Interaction with ad position											
$LogProgramWOMVolume_i \times Ad\ position$	358	(.759)		.000	(.000.)		.000	(.000.)	**		
LogProgramWOMDeviation _i × Ad position	000	(.000)	**	000	(.000.)	**	000	(.000.)	**		
Adjusted R-squared	.9995				.9995		.9995				

^{*} *p* < .10, ** *p* < .05

Web Appendix 3.6: Alternative operationalizations of OPE deviation

In our conceptual framework, we conjecture that OPE deviations just prior to an ad will have the most meaningful relationship with audience size during ads, and, in our main analysis, we operationalize this variable as the difference between the volume of program-related Tweets in the minute before a focal ad begins airing and the average number of per-minute program-related Tweets between the start of the episode and when the focal ad begins airing. We test nine alternative operationalizations of this measure where we vary the time window used pre-ad from 2 to 10 minutes. These analyses not only allow us to probe our conjecture that OPE deviations closer to the ad airing are more meaningful for ad audience size, but they also allow us to explore the dynamics of OPE and examine the robustness of our results. As shown in Web Appendix Table 3.6.1, we find that the positive relationship between positive OPE deviations and ad audience size remains significant using the 2-minute to 8-minute measurement windows but then becomes insignificant. Overall, these results support our argument that positive OPE deviations just prior to an ad will have the most meaningful relationship with ad audience size while also showing that our results regarding OPE deviation are robust to alternative operationalizations.

Web Appendix Table 3.6.1: Key results from alternative pre-ad time windows for OPE deviation

Operationalization of LogProgramWOMDeviation _i	Estimate (SE) of LogProgram WOMDeviation _i
Using 2-minute window pre-ad	.00033 (.00006) **
Using 3-minute window pre-ad	.00029 (.00006) **
Using 4-minute window pre-ad	.00027 (.00007) **
Using 5-minute window pre-ad	.00027 (.00007) **
Using 6-minute window pre-ad	.00024 (.00007) **
Using 7-minute window pre-ad	.00019 (.00008) **
Using 8-minute window pre-ad	.00014 (.00008) *
Using 9-minute window pre-ad	.00012 (.00008)
Using 10-minute window pre-ad	.00006 (.00008)

Notes: * p < .10, ** p < .05

As further robustness tests, we explore 11 alternative operationalizations of OPE deviation: (1) main specification with a more narrow baseline (i.e., log of the difference between the volume of program-related Tweets in the minute before ad i airs (window p) and the average number of per-minute program-related Tweets in the 5-minute window preceding window p); (2) dichotomous operationalization to capture a *spike* in OPE (variable equals 1 if the volume of program-related Tweets in the minute before ad i is greater than or equal to a 25% increase relative to the average number of per-minute program-related Tweets between the start of the episode and ad i; 0 otherwise); (3) dichotomous operationalization to capture a *spike* in OPE (variable equals 1 if the volume of program-related Tweets in the minute before ad i is greater than or equal to a 50% increase relative to the average number of per-minute program-related Tweets between the start of the episode and ad i; 0 otherwise); (4) ProgramWOMDeviationRatio_i, the volume of program-related Tweets in the minute before ad i airs divided by the average number of per-minute program-related Tweets between the start of the episode and ad i; (5) log of ProgramWOMDeviationRatio; (6) ProgramWOMDeviationPC_i, the percentage of program-related Tweets in the minute before ad i (operationalized as the volume of program-related Tweets in the minute before ad i divided by the volume of programrelated Tweets between the start of the episode and ad i; (7) log of $Program WOMD eviation PC_i$; (8) $Program WOMD eviation PC_i$ using the 2-minute window before ad i (rather than the 1-minute window); (9) log of *ProgramWOMDeviationPC*_i using the 2-minute window; (10) *ProgramWOMDeviationPC_i* using the 5-minute window before ad i; and (11) log of *ProgramWOMDeviationPC*^{*i*} using the 5-minute window.

The key results, shown in Web Appendix Table 3.6.2, are consistent with our main model and illustrate the robustness our findings to alternative operationalizations of OPE deviation.

Web Appendix Table 3.6.2: Key results from alternative operationalizations of OPE deviation

Variable	LogProgramWOM Deviation; with narrow baseline Estimate (SE)		Deviat with 259	Estimate (SE)		tionSpike i % threshold hate (SE)	Deviati	$amWOM$ $fonRatio_i$ $ate (SE)$	Deviat	$gramWOM$ $ionRatio_i$ $ate (SE)$
Ad position OPE	.007	(.000) **	.007	(.000) **	.007	(.000) **	.007	(.000) **		(.000) **
LogProgramWOMVolume i	8.383	(2.998) **	8.368	(3.003) **	8.094	(3.009) **	7.365	(3.002) **	8.367	(2.991) **
$LogProgramWOMDeviation_i$.001	(.000) **	.004	(.000) **	.004	(.001) **	.003	(.000) **	.009	(.001) **
Interaction with ad position LogProgramWOMVolume i × Ad position	253	(.766)	.410	(.760)	.472	(.762)	.542	(.761)	.534	(.759)
LogProgramWOMDeviation _i × Ad position	000	(.000) **	001	(.000) **	001	(.000) **	001	(.000) **	003	(.001) **
Adjusted R-squared	.9995		.9995		.9995		.9995		.9995	

			LogProgramWOM DeviationPC $_i$ with		_	camWOM	$\begin{array}{c c} LogProgramWOM \\ \hline DeviationPC_i \text{ with} \end{array}$		_	Ü		ramWOM nPC; with
Variable		ite window		te window		ite window		e window		te window		e window
<u> </u>	Estimate (SE)		Estimate (SE)		Estin	nate (SE)	Estim	ate (SE)	Estim	ate (SE)	Estim	ate (SE)
Ad position	.007	** (000.)	.007	** (000.)	.007	(.000) **	.007	(.000) **	.007	(.000) **	.007	** (000.)
OPE												
$LogProgramWOMVolume_{i}$	8.292	(3.005) **	8.356	(2.977) **	8.354	(3.002) **	8.415	(2.988) **	8.228	(3.000) **	8.134	(2.995) **
$LogProgramWOMDeviation_i$.015	(.004) **	.004	(.000) **	.005	(.003) *	.002	(.000) **	.004	(.002) **	.002	(.001) **
Interaction with ad position												
$LogProgramWOMVolume_i$ ×	100	(762)	012	(754)	111	(761)	027	(757)	000	(761)	124	(750)
Ad position	.199	(.762)	.013	(.754)	.111	(.761)	.037	(.757)	.088	(.761)	.124	(.759)
$LogProgramWOMDeviation_i \times$	008	(.002) **	001	(.000) **	007	(.001) **	001	(.000) **	004	(.000) **	001	(.000) **
Ad position	008	(.002)	001	(.000)	00 /	(.001) **	001	(.000) **	004	(.000) **	001	(.000)
Adjusted R-squared	.9995		.9995 .9995		.9995		.9995		.9995		.9995	

^{*} *p* < .10, ** *p* < .05

Web Appendix 3.7: Program characteristics that associate with social moments

Our analyses show that positive OPE deviations – that is, social moments in an episode – associate with increased ad audience size. We conduct a follow up analysis in which we explore how episode characteristics associate with social moments. We present the results from this analysis, which we discuss in the main manuscript, below in Web Appendix Table 3.7.

Web Appendix Table 3.7: Relationship between program characteristics and OPE deviation

Variable	Estimate (SE)					
Intercept	-3.13	(.45)	**			
Fixed episode characteristics						
Day of the week (Baseline: Friday)						
Monday	03	(.13)				
Tuesday	.27	(.13)	**			
Wednesday	32	(.14)	**			
Thursday	.20	(.14)				
Saturday	.53	(1.28)				
Sunday	.05	(.15)				
Program genre (Baseline: Slice-of-						
life)						
Comedy	46	(.27)	*			
Drama/adventure	-1.80	(.21)	**			
News	.73	(1.01)				
Suspense/mystery	-2.01	(.45)	**			
Special episode						
Fall finale	.43	(.12)	**			
Season premiere	-1.33	(.13)	**			
Viewer episode rating	.35	(.06)	**			
Characteristics that vary within episodes						
$LogProgramWOMVolume_i$	285.36	(321.04)				
Ads on other networks	02	(.08)				
Break position	1.60	(.25)	**			
Break position × comedy	-1.57	(.39)	**			
Break position × drama/adventure	.68	(.30)	**			
Break position × news	-1.77	(1.46)				
Break position × suspense/mystery	1.42	(.71)	**			
Half-hour break	.51	(.11)	**			
Adjusted R-squared	.0660					

Notes: * p < .10, ** p < .05

Web Appendix 3.8: Exclusion of final ad break

Television viewers may show different patterns of behavior for ad breaks that air after a program has ended. We therefore consider a robustness analysis in which we exclude ads that aired in the final ad break of a program, i.e., the only ad break that may occur after the program has ended. The key results from this alternative analysis, shown in Web Appendix Table 3.8, are consistent with those of our main analysis and provide evidence that ads that air in the final ad break of a program do not alter the results we observe.

Web Appendix Table 3.8: Key results from main model estimation excluding ads aired in a program's final ad break

Variable	Estimate (SE)			
Ad position	.007	(.000.)	**	
OPE				
$LogProgramWOMVolume_i$	10.100	(3.211)	**	
$LogProgramWOMDeviation_i$.001	(.000)	**	
Interaction with ad position				
$LogProgramWOMVolume_i \times $	153	(.822)		
Ad position				
$LogProgramWOMDeviation_i \times$	000	(.000)	**	
Ad position	000	(.000)	-	
Adjusted R-squared	.9996			

^{*} p < .10, ** p < .05