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What Drives Media Reporting of Food Safety Events? Evidence from U.S. Meat Recalls

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Abstract

This paper examines how the characteristics of a recall affect the volume of media coverage about that recall. We link data on media reports to a comprehensive list of virtually all recalls of meat products over the period 2001–2012. We find considerable evidence that, up to a point, the characteristics of a food recall significantly affect reporting about that recall. Specifically a one percent increase in the volume of meat recalled results in a 0.1 percent increase in media coverage. In addition, we find that media coverage is significantly larger for incidents related to bacterial contamination as compared to other types of recalls.

Keywords: Food Safety, Media Bias, Meat Recall

1. The First Section

Between 2000 and 2012, the USDA Food Safety Inspection Service (USDA-FSIS) issued 864 recalls, or roughly one every five to six days. Recalls varied in size and scope ranging from 1 lb to 14 million lbs. While, some products are recalled for minor violations, others are recalled due to potentially life threatening contamination. However, the USDA-FSIS (and the FDA) do not communicate directly with consumers. As a result, they rely on the news media to transmit important food safety information to consumers in a timely and accurate manner. Given the number of recalls, and competition with other events for coverage, not all recalls can or should be given media attention. This paper asks how the characteristics of an FSIS recall drives

reporting about the recall itself. Put differently, are certain types of recall more likely to garner media attention?

In this paper, we investigate how the characteristics of a recall – e.g. its size and scope, the cause of the recall, the risk posed to human health – affect the extent to which an event is covered in the media. In other words, is there systematic bias in the way in which recalls are covered in the media? Is there a threshold below which recalls will not be covered? We then ask, given recall characteristics, how does media coverage translate into consumer awareness of recalls. To this end, we use an index of Google searches related food recalls in the weeks following a recall as a proxy for consumer interest.

The way in which media coverage affects consumer purchasing behavior has been extensively studied (Smith, van Ravenswaay, & Thompspon 1988; Burton & Young 1996; Verkeke & Ward 2001; Dahlgran and Fairchild 2002; ; Piggott & Marsh 2004; Marsh, Schroeder & Mintert 2004; Mazzochi 2006; Chang & Just 2007; Taylor, Phaneuf & Piggott 2009; Taylor, Klaiber & Kuchler 2013). Counts of newspaper articles are widely used by agricultural economists as a proxy for consumer awareness of food safety events. These measures, which are often formalized as media indices, are subsequently incorporated into structural models of demand. In sum, this literature finds that recalls have a significant short run impact, but with the notable exception of the U.K. BSE crisis, their long run impact may be small. Our research question differs from previous work in that it investigates the drivers behind media indices rather than the effect of media indices on purchasing behavior. To the extent that media coverage differs significantly as a function of recall characteristics, this paper will shed light on previous findings.

This paper also contributes to a broader literature at the intersection of economics and political science on media reporting bias. Previous work in this vein has looked at how newspapers cover statements by political actors (Groseclose & Mylow 2005; Baron 2006; Gentzkow & Shapiro 2010). Perhaps most closely related to this paper is work by Swinnen, McCluskey, & Francken (2005), which looks at detailed counts of media reporting for two food safety crisis in three newspapers in Belgium. Our question is fundamentally different and we add to the literature by considering *all* meat recalls across all major newspapers over a 12 year period in the United States. Moreover, our analysis extends beyond crisis and explicitly examines the differences in reporting between crisis and non-crisis events.

This paper makes two main contributions. To the best of our knowledge, this paper is the first to explicitly study how recalls are covered in the U.S.

50 media. Understanding this relationship is important to economists who use
51 media indices as explanatory variables in structural demand models, as well
52 as to policy makers who rely on the media to convey important health in-
53 formation to individuals. Second we investigate the extent to which media
54 reporting affects consumer awareness of recalls by studying data from Google
55 searches.

56 Background

57 In the United States, the Department of Agriculture’s Food Safety Inspec-
58 tion Service (USDA-FSIS) oversees the recalls of meat and poultry products
59 produced by federally inspected establishments. Once FSIS has determined
60 that the recalled product is potentially available for purchase by consumers,
61 the agency issues a recall release to ”media wire services, media outlets in ar-
62 eas that received recalled products, the FSIS e-mail subscription service, and
63 the @USDAfoodsafety Twitter feed” (USDA-FSIS 2012). The recall release
64 is also posted on the USDA-FSIS website.

65 Data Construction

66 The data used in the paper comes from three distinct sources. First, the
67 meat and poultry recall data are from the USDA Food Safety and Inspection
68 Service (FSIS) meat recall archive that is publicly available online. Data on
69 media reports of food recall news are from LexisNexis® Academic. Finally,
70 an index of the volume of Google queries related to meat recalls were obtained
71 from an online utility provided by Google, named Google Trends. We now
72 describe our data construction process in detail.

73 *FSIS Meat Recall Data*

74 We obtain information on recalls from the FSIS recall archive. Each recall
75 consists of press releases, reports and notification issued by the FSIS for
76 the recalled meat and poultry products. From the recall archive we collect
77 characteristics of the recall such as the quantity of product recalled, the
78 severity of the recall, and the reasons for the recall.

79 There were 864 meat and poultry food product recalls in the last 12 years.
80 Table 1 summarizes the characteristics of FSIS product recalls over the period
81 2001 to 2012. Note that information on the quantity recalled for 31 products
82 was missing and as a result are dropped from the analysis. We divide the

83 remaining 833 recalled products into four broad categories, namely: beef,
84 pork, poultry, processed meat and others. Note that the processed meat
85 category includes products such as hot dogs, luncheon meats, sausages and
86 the others category includes those products which are either a combination
87 of meats or not identified as containing any specific meat, such as soup,
88 burritos, pizza etc. Over the period in question, food products containing
89 beef are the most frequently recalled, followed by food products containing
90 poultry and processed meat products. Infection due to *Listeria* is the most
91 frequent cause of a recall followed by infection due to *E. Coli*.

92 FSIS classifies food product recalls according to the severity of the health
93 threat they present to the consumers. Class 1 level of severity refers to the
94 situation where use of the product will cause severe health consequences or
95 death. Class 2 refers to the situation where there is low probability of adverse
96 health effects and class 3 refers to the situation where the use of the product
97 will cause no adverse health effects (USDA FSIS). Almost 71 percent of all the
98 food products recalled were of severity class 1 and 19 percent were of severity
99 class 2, the remaining 10 percent were class 3. The FSIS recall information
100 provides the region and states where the food products were recalled. Regions
101 were constructed following the US Census Region and Divisions of the United
102 States. Most of the recalls happened in the southern states followed by the
103 midwest and the northeast.

104 *News Data and Classification*

105 Our key outcome variable is a count of newspaper articles and transcripts
106 of broadcast media related to a given recall. Counts of articles and transcripts
107 from all media sources were collected using the LexisNexis® Academic search
108 engine. Note that LexisNexis® is a well-known online archive of legal and
109 media documents. Indeed, much of the previous work in the area has also
110 used LexisNexis® data (Taylor and Phaneuf, 2009; Piggott and Marsh, 2004;
111 Burton and Young, 1996).

112 We began by querying the LexisNexis® database for media reports con-
113 taining keywords related to food product recalls, over the period 2001–2012.
114 This generated an exhaustive list of 52,168 articles plausibly related to food
115 recalls. One problem with this approach is that it generates a relatively
116 large number of false positives, articles that contain relevant keywords but
117 are unrelated to a given recall. To address this, we use the natural language

Table 1: Meat and Poultry FSIS recalls, 2001-2012

	Beef	Pork	Poultry	Processed Meat	Others	Total
Number of Recalls	295	104	179	140	115	833
<u>Reason for Recall</u>						
Listeria	32	41	59	56	21	209
E. Coli	165	1	0	5	3	174
Salmonella	8	5	11	3	5	32
Undeclared Allergen	16	22	59	33	36	166
Contamination	18	16	28	16	19	97
UnderProcessing/Undercooking	6	5	10	8	5	34
Mislabeled and others	50	14	17	22	26	129
<u>Severity of Recall</u>						
Class 1	227	70	129	104	68	598
Class 2	39	20	42	21	37	159
Class 3	29	14	8	15	10	76
<u>Region for Recall</u>						
West	79	28	54	31	33	225
NorthEast	77	27	62	41	31	238
South	107	37	53	50	37	284
MidWest	88	31	60	33	30	242
Nationwide	55	17	43	21	28	164

118 processing capabilities in Python¹ to construct a probability that an entry
 119 in the media database refers to a given recall². We exclude articles where
 120 the matching probability falls below 0.5. The result of this exercise was a
 121 list of articles related to each recall published from 2001 to 2012. For each
 122 article, we construct a number of variables from its text, including the article
 123 headline, number of times a key-word appeared in the article, date published
 124 and, source of news.

125 We then merge the information on media reports to the FSIS recall

¹See <http://nltk.org/> for further details.

²Details of the Naive Bayes matching algorithm are beyond the scope of this short paper but will be provided in the final draft. See Tumarkin & Whitelaw (2001) and Antweiler & Frank (2004) for applications to finance.

126 database. This was done by matching the date and keywords specific to
 127 a given recall to the date and keywords in a news item. The result was a
 128 potentially one to many (or zero) match of recalls to media reports. Thus
 129 our primary dataset consists of the product recall information along with the
 130 count of published articles that relate to each product recall. This type of
 131 aggregated media index has been used as a measure of consumers' awareness
 132 of product recalls in previous research (Burton and Young, 1996; Marsh et.
 133 al., 2004; Taylor and Phaneuf, 2009).

134 Table 2 summarizes the news items and the product recall information.
 135 Over the period 2001 to 2012 we linked 9551 distinct articles to USDA-FSIS
 136 recalls. Almost 44 percent were related to beef products recall followed by
 137 processed meat products recall. Recalls due to E. Coli infection had a 29
 138 percent news coverage and 77 percent of the articles were related to the
 139 recalls belonging to the class 1 severity.

140 *Google Trend Data*

As a measure of consumer awareness of meat recalls, we use information on internet searches provided by Google in the form of Google Trends data. Specifically, Google Trends gives an index of the volume of queries entered into Google in a given geographic region, here the United States. The query index is calculated as:

$$\text{Query Share} = \frac{\text{Number of queries for a search term in a region for that time period}}{\text{Total number of queries in that region for that time period}}$$

141 The maximum query share in the time period specified is normalized to be
 142 100 and the query share at the initial date being examined is normalized to
 143 be zero. As a result, the index depends on period considered³.

144 The index is scaled to run from 1 to 100 and if search levels fall below a
 145 certain point, Google reports a zero. According to Google's history archive,
 146 Google Trends was introduced in the United States in 2006. This may explain
 147 why the trends peak start from 2006 onwards. The index reaches a maximum
 148 of 100 during the 2008 Westland meat recall, which was the largest recall in
 149 United States history.

150 In our subsequent analysis we analyze the data on internet search in
 151 weekly form. For each recall, we link the Google Index in a given week and

³For futher details on the construction of the Google Index, see Choi and Varian (2009, 2012).

Table 2: Media Information variables, 2001-2012

	Mean	Standard Deviation	Percentage	Maximum
Beef	14.27	17.29	0.44	193
Pork	9.70	11.61	0.09	67
Poultry	12.12	12.58	0.21	105
Processed Meat	9.68	14.58	0.16	111
Others	10.74	13.77	0.13	105
<u>Reason for Recall</u>				
Listeria	9.47	11.01	0.17	105
E. Coli	16.28	16.11	0.29	85
Salmonella	19.62	21.94	0.06	105
Undeclared Allergen	10.83	13.44	0.16	111
Contamination	10.90	10.39	0.09	77
UnderProcessing/Undercooking	14.20	17.52	0.04	85
Mislabeled and others	9.66	18.03	0.14	193
<u>Severity of Recall</u>				
Class 1	12.67	14.65	0.77	111
Class 2	11.59	17.97	0.18	193
Class 3	7.39	6.42	0.05	27
<u>Region for Recall</u>				
West	11.64	13.08	0.29	105
NorthEast	11.48	13.87	0.25	111
South	12.31	14.86	0.32	111
MidWest	11.67	12.67	0.31	88
Nationwide	18.26	23.07	0.22	193

152 the following week to the week in which a product was recalled. In this way
 153 we capture a direct measure of consumer interest in food recalls. Note that
 154 when multiple recalls occur in a two week period (as occurs not infrequently),
 155 we confound popular interest in one recall with popular interest in another
 156 recall. This is a limitation of the Google Trends data and is an issue in the
 157 empirical work that follows.

158 2. Empirical Methods

159 We regress the natural logarithm of our measure of media attention de-
 160 scribed above on the characteristics of a given recall.

$$\begin{aligned}
 \ln(Count)_i = & \alpha \ln(Quantity)_i + \sum_{j=1}^{J-1} \beta_j Product_{ij} + \sum_{k=1}^{K-1} \gamma_k Class_{ik} + \\
 & \sum_{l=1}^{L-1} \theta_l Cause_{il} + \Phi_i + \Psi_i + u_i
 \end{aligned} \tag{1}$$

161 where, *Quantity* corresponds to the number of pounds of product recalled,
 162 *Product* refers to the type of meat recalled (beef, pork, poultry, processed
 163 meats), *Class* corresponds to the USDA-FSIS definition of class of recall, and
 164 *Cause* refers to the underlying reason for the recall (bacterial contamination –
 165 salmonella, listeria, ecoli –, undeclared allergen, mislabeling, contamination).
 166 Finally, we include year (Φ_i) fixed effects to control for national trends and
 167 region (Ψ_i) fixed effects to capture time invariant regional unobservables⁴.
 168 Note that in all cases we report heteroskedasticity consistent standard errors
 169 (White 1980).

170 To study the consumer awareness of meat recalls – to the extent that
 171 this can be measured by the volume of internet search activity – we regress
 172 the Google search index on the set of explanatory variables described above,
 173 and the aforementioned measure of media activity, $\ln(Count)$. As noted
 174 above, Google reports zero when the level of search activity falls below an
 175 undisclosed threshold. To account for the truncation, we use a tobit model
 176 and set the truncation point just below the lowest observed value of the index
 177 over the period⁵.

178 Results

179 Table 3 contains our main results. The first column contains the results
 180 for our main specification presented in Eq 1. In short, we find, perhaps not
 181 surprisingly, that the characteristics of a recall explain an important share –

⁴Note that results are robust to including state instead of region fixed effects

⁵The choice of truncation point is admittedly adhoc, though results are robust to choices slightly above and below this minimum.

roughly 30% – of the variation in media coverage of a given recall, though a substantial portion of the variation remains unexplained.

Relative to the omitted category (Others), a beef recall receives 26.9%⁶ more coverage and a poultry recall receives roughly 23.4% more coverage. Both of these are significantly different from zero at all conventional levels. The severity of a recall also has a statistically significant and economically important effect on the volume of reporting. Recalls that USDA-FSIS classifies as level one receives 52% more coverage and recalls classified as level two receive roughly 38.8% more coverage. These are relative to recalls classified as level 3, i.e. recalls that do not represent a threat to human health. Recalls caused by bacterial contamination, specifically Salmonella and E.coli, received a greater volume of coverage, 54.4% and 40.6% respectively. Finally, we see that the total quantity of product recalled has an important effect on coverage. A one percent increase in quantity recalled yields a 0.1 percent increase in coverage.

The second column of table 3 contains the estimation results for Google index as the dependent variable. Results show that a beef recall generates a significant amount of Google searches by the consumers as compared to recall of products in the other category. The total quantity of product recalled has significant effect on the Google searches. The media index (count of articles related to the recall) shows an increase in the consumer Google searches for the recall. This implies that the media coverage of the recalls is responsible for generating awareness among the consumers, atleast to some extent.

Discussion and Conclusion

While it is impossible to say what the optimal level of media coverage of a recall should be, our results suggest that the media may under report recalls that pose a higher risk to consumer health (Class 1 recalls) than those which have lower or no health risks. According to our estimates, recalls focused on products that may severely impact human health attract only 52% more articles than recalls of products that do not present any serious health threat. The volume effect also seems small. The

Suggests that there may be some nonlinear relationship. All recalls get some small coverage. Coverage scales with severity and quantity recalled.

⁶Marginal effects are computed as $\exp(0.239) - 1 = 0.269$

Table 3: Results for Measure of Media Attention

	(1) Log(Article Count)	(2) Log(Keyword Count)	(3) Google Index
Beef	0.239** (2.09)	0.759*** (3.19)	9.959** (2.18)
Pork	0.110 (0.90)	0.594** (2.57)	-0.0546 (-0.02)
Poultry	0.212** (2.04)	0.631*** (2.85)	2.751 (1.02)
Processed Meat	-0.0383 (-0.34)	0.175 (0.78)	2.193 (0.70)
Severity 1	0.420*** (3.22)	0.592** (2.36)	2.244 (0.52)
Severity 2	0.329** (2.51)	0.295 (1.11)	3.508 (0.82)
Log(Quantity)	0.106*** (8.02)	0.135*** (5.29)	1.453*** (2.65)
Number of Regions	0.00949 (1.02)	0.00284 (0.18)	-0.279 (-1.15)
Listeria	0.0387 (0.32)	0.251 (1.09)	4.835 (1.44)
Ecoli	0.341** (2.52)	0.964*** (4.13)	-4.467 (-0.87)
salmonella	0.435** (2.44)	0.503 (1.23)	-6.755 (-1.17)
Allergen	0.143 (1.27)	0.378 (1.64)	1.505 (0.53)

Table 3: Continued: Results for Measure of Media Attention

	(1)	(2)	(3)
	Log(Article Count)	Log(Keyword Count)	Google Index
Contamination	0.148 (1.34)	0.458** (2.08)	-3.902 (-0.90)
Underprocessing	0.127 (0.67)	0.462* (1.75)	-4.291 (-0.69)
Log(Article Count)			3.261** (2.14)
Observations	833	833	458

t statistics in parentheses

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$

215 Above a certain point, it looks like there is a limit on the amount of media
216 coverage.

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