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Experience of technological and natural disasters and their impact on the perceived risk of nuclear accidents after the Fukushima nuclear disaster in Japan 2011: A cross-country analysis

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Abstract

This paper uses cross-country data compiled immediately after the Fukushima

nuclear accident to investigate how the experience of such disasters affects the

perception of the risk of nuclear accidents. Estimation results show that the perceived

risk of a nuclear accident is positively associated with experiencing technological

disasters but not with that of natural disasters.

Keywords: Technological disaster, Natural disaster, Nuclear accidents, Risk,

Fukushima accidents, Perception, Bayesian learning

JEL classification: D83, D84, Q54

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## 1. Introduction

The devastating earthquake that occurred in Japan on March 11, 2011 was followed by a tsunami that ultimately crippled the Fukushima Daiichi nuclear power plants. These unprecedented disasters drew public attention from around the world. Such nuclear disasters are likely to change citizens' views regarding the environment (Berger, 2010.). A G8 Summit was held 2 months after the Japanese disaster, and countries agreed to join forces in an effort to promote tighter international standards for nuclear safety.

A number of studies have previously considered perceptions and responses regarding low-probability events (Camere and Kunreuther, 1989; Kunreuther and Pauly, 2004). To experience a natural disaster may influence individual risk beliefs through the updating of one's risk level. From the viewpoint of rational Bayesian learning, one would expect perceived risk to increase after experiencing a disaster. The Japanese disaster shows that nuclear disaster can be caused not only by human error with regard to technology, but also by unexpected natural disasters. Therefore, one's perception regarding the risk of nuclear accidents appears to depend not only the experience of technological disasters but also on that of natural disasters.

It has been argued that people who have experienced a disaster do not sufficiently update their perceived level of risk (Tversky and Kahneman 1974; Viscurrsi and Zeckhauser 2006). As noted by Zeckhauser (1996, p. 115), "Neither humans nor society deal effectively with information, particularly probabilistic information." Thus, the issue of the relationship between the subjective risk of a nuclear accident and experiencing natural and technological disasters appears to remain open to discussion. The perceptions of citizens' regarding risk do have an influence on policy concerning

disasters (Viscurrsi and Zeckhauser, 2006; Kahn, 2007). Thus, it is worth exploring the relationship between them.

By using cross-country data collected immediately after the disaster in Japan in 2011, this paper aims to investigate how the experience of a technological disaster effects perceptions regarding the subjective risk of a nuclear accident.

#### 2. Data and Model

Definitions and the descriptive statistics of variables used in this paper are presented in Table 1. The countries used in the estimations are listed in the appendix (Table A1). Immediately after Japan's natural disaster, WIN-Gallup International (2011) conducted a survey in 47 countries regarding nuclear energy. The survey included the following question: "How high or low is your concern about the possibility of a nuclear incident in your country?" There were 5 response choices: "very high", "high", "medium", "low", and "very low". The WIN-Gallup International (2011) survey provides the responses for each county. Based on the WIN-Gallup survey data, I calculated the rate of respondents that believed there is a high (or very high) possibility of a nuclear accident—dependent variable PACCI. In addition, an alternative measure, PACCI2, is the rate of those respondents who believe there to be a very high possibility of nuclear accident. PACCI (or PACCI2) measures the subjective risk of a nuclear accident and is, therefore, used as a dependent variable. A key independent variable is the number of technological disasters that have occurred (in the respondent country) since 1990 (TDIS), which captures the experience of technological disasters. A cursory examination of Figure 1 shows that TDIS is positively associated with PACCI. The Fukushima nuclear accident was triggered by an earthquake and tsunami. Hence, the

risk of natural disaster appears to be related to the perceived risk of a nuclear accident. Therefore, the experience of a natural disaster, a further independent variable, is captured by the number of natural disasters that have occurred (in the respondent country) since 1990 (NDIS). Figure 2 indicates a slightly positive relationship between NDIS and PACCI, although the relationship is ambiguous. For a closer examination of the relationships, regression estimations were conducted. The estimated function takes the following form:

PACCI (or PACCI2)<sub>i</sub> =  $\alpha_0 + \alpha_1 \text{Ln}(\text{TDIS})_i + \alpha_2 \text{Ln}(\text{NDIS})_i + \alpha_3 \text{Ln}(\text{GDP})_i + \alpha_4 \text{Ln}(\text{POP})_i + \alpha_5 \text{Ln}(\text{GOVSIZ})_i + \alpha_6 \text{NUCLE}_i + e_i$ ,

where PACCI (or PACCI2) represents the subjective risk of a nuclear accident in country i, α represents regression parameters, and e is an error term. With the exception of the key variables explained earlier (TDIS and NDIS), the following control variables were included as independent variables. Economic factors were captured by population (POP), GDP per capita (GDP), and the size of government (GOVSIZ).² The greater the number of nuclear energy plants in an area, the higher the possibility a nuclear accident. The number of nuclear energy plants (NUCLE) is included to control for this effect. As can be seen in Table 1, the standard deviations of each independent variable are large, thus, heteroscedasticity should be taken into account. To reduce heteroscedasticity, each independent variable, with the exception of NUCLE, is in log form.³

It is more likely than not, that nuclear plants are constructed in countries where

<sup>&</sup>lt;sup>1</sup> TDIS and NDIS were obtained from the International Disaster Database http://www.emdat.be (accessed April 30, 2011).

<sup>&</sup>lt;sup>2</sup> The data was sourced from the Penn World Table (PWT 6.3). It is available at http://pwt.econ.upenn.edu/ (accessed 28 March 2011).

 $<sup>^3</sup>$  NUCLE is 0 in some countries where no nuclear plant exists and therefore cannot be expressed as a logarithm.

people perceive there to be a low probability of nuclear accidents occurring. The OLS estimation results above possibly suffer from endogeneity bias because a reverse causality occurs between the dependent variable and independent variable (NUCLE). For the purpose of controlling for this bias, instrumental variables were used to conduct a GMM 2SLS (Generalized method of moments two-stage least square) estimation.<sup>4</sup> Sufficient land area is required to build nuclear plants. Thus, densely populated countries encounter great difficulties in searching for space in which to build nuclear plants. Therefore, land area and population density were used as instrumental variables in the GMM 2SLS estimations. The data for the instrumental variables were sourced from World Development Indicators.<sup>5</sup>

## 3. Results

The estimation results for OLS are exhibited in Table 2. The results with PACCI as the dependent variable are presented in columns (1) and (2), and those for PACCI2 are shown in columns (3) and (4). The results of the GMM 2SLS estimations are shown in Table 3. The sample size was only 37 and therefore considered small. Thus, the jackknife method was used to calculate the standard error to ensure that the results were not spurious.

In Table 2, the results for Ln(TDIS) yielded the predicted positive signs, and were statistically significant in all estimations. The absolute values of Ln(TDIS) were approximately 15.4 and 15.1 in the PACCI estimation respectively columns (1) and (2),

<sup>4</sup> The GMM estimator allows for heteroscedasticity and brings efficiency gains in the presence of heteroscedasticity (Greene 2008, Ch. 15). This is why I used the GMM 2SLS rather than the 2SLS model.

<sup>&</sup>lt;sup>5</sup> The data are available from HP of the World Bank http://databank.worldbank.org/ddp/home.do (accessed 28 March 2011).

and 10.7 and 10.4 for PACCI2 in columns (3) and (4), respectively. These results imply that a 1% increase in TDIS increases PACCI by approximately 15%, whereas a 1% increase in TDIS increases PACCI2 by about 10%. In contrast, NIDS was not statistically significant although NDIS did produce the anticipated positive sign. With the exception of POP, the other control variables were not statistically significant in all estimations.

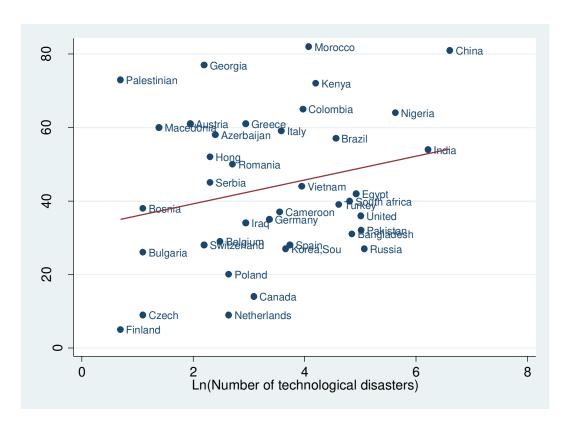
With regard to the GMM 2SLS estimation results exhibited in Table 3, an over-identification test was used to test for exogeneity in the instrumental variables. Test statistics were not significant in columns (1) and (2) and, therefore, do not reject the null hypothesis that the instrumental variables are uncorrelated with the error term. This suggests that the instrumental variables are valid. TDIS continued to yield a positive sign and be statistically significant in columns (1) and (2). Its absolute values were 14.1 for PACCI and 9.62 for PACCI2, which are similar to those shown in Table 2. On the whole, the estimation results for TDIS did not change after controlling for endogeneity bias. Thus, from the results of Tables 2 and 3, I propose that the experience of a technological disaster increases the perceived risk of a nuclear accident.

#### 4. Conclusions

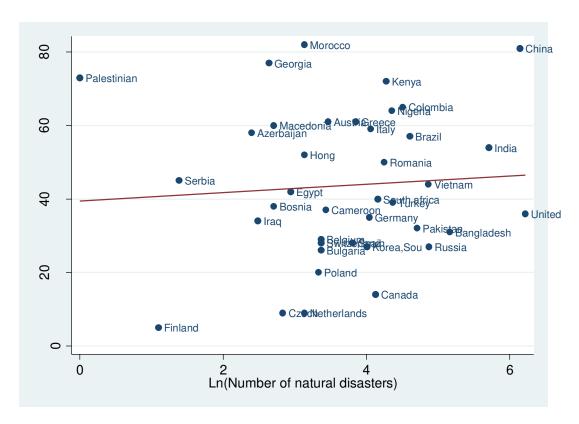
WIN-Gallup International conducted a cross-country survey on views regarding nuclear energy immediately after the Fukushima nuclear accident in Japan. Using this data, the present paper explored how the experience of a technological disaster affects the perceived risk of nuclear accidents. An analysis of the data has found that the experience of a technological disaster increases the perceived risk of a nuclear accident, whereas the experience of a natural disaster does not affect perceptions of risk.

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**Figure 1.** Association between the experience of technological disasters and the perceived risk of nuclear accidents.



**Figure 2.** Association between the experience of natural disasters and the perceived risk of nuclear accidents.

 Table 1
 Definition of variables and descriptive statistics

	Definition	Mean	Max	Min	Standard deviation
PACCI	Rate of respondents who believe that the possibility of a nuclear accident is high (or very high) (%)	42.0	82	5	21.0
PACCI2	Rate of respondents who believe that the possibility of a nuclear accident is very high (%)	20.1	48	1	13.7
TDIS	Total number of technological disasters since 1990	81.3	744	2	144.0
NDIS	Total number of natural disasters since 1990	74.7	502	1	109.4
$\operatorname{GDP}$	GDP per capita (million dollars)	1.83	4.55	0.21	1.44
POP	Population (million)	110.9	1321.8	0.3	267.8
GOVSIZ	Government size (Government expenditure/GDP) (%)	16.2	59.6	3.24	9.01
NUCLE	Number of nuclear plants.	8.41	105	0	18.9

Note: PACCI and PACCI2 were obtained from WIN-Gallup International (2011). TDIS and NDIS were sourced from the International Disaster Database (<a href="http://www.emdat.be">http://www.emdat.be</a> accessed April 30, 2011). POP, GDP, and GOVSIZ were sourced from Penn World Table 6.3. (<a href="http://pwt.econ.upenn.edu/php\_site/pwt\_index.php">http://www.euronuclear.org/info/npp-ww.htm</a> accessed April 30, 2011). NUCLE was sourced from the HP of the European Nuclear Society (<a href="http://www.euronuclear.org/info/npp-ww.htm">http://www.euronuclear.org/info/npp-ww.htm</a> accessed April 30, 2011).

 Table 2
 OLS estimation

	(1)	(2)	(4)	(5)
	Dependent	Dependent	Dependent	Dependent
	variable:	variable:	variable:	variable:
	PACCI	PACCI	PACCI2	PACCI2
Ln(TDIS)	15.4***	15.1**	10.7***	10.4**
	(3.04)	(2.61)	(2.81)	(2.57)
Ln(NDIS)	6.05	7.04	1.99	2.70
	(1.64)	(1.68)	(0.71)	(0.85)
Ln(GDP)	-3.17	-2.31	-2.17	-1.55
	(-0.76)	(-0.50)	(-0.86)	(-0.56)
Ln(POP)	-16.4***	-15.8***	-9.74**	-9.32**
	(-3.08)	(-2.75)	(-2.42)	(-2.14)
Ln(GOVSIZ)	6.97	6.94	7.28	7.26
	(0.83)	(0.79)	(1.44)	(1.35)
NUCLE		-0.10		-0.07
		(-0.48)		(-0.48)
Constant	149.0**	133.2*	79.0**	67.7
	(2.42)	(1.90)	(2.17)	(1.59)
Adjusted R <sup>2</sup>	0.25	0.23	0.30	0.28
Observations	37	37	37	37

Note: Values in parentheses are t-statistics calculated by standard errors obtained using the jackknife method. \*, \*\*, and \*\*\* denote significance at the 10%, 5% and 1 % levels, respectively.

 Table 3
 GMM 2SLS estimation

	(1)	(2)
	Dependent	Dependent
	variable:	variable:
	PACCI	PACCI2
Ln(TDIS)	14.1**	9.62*
	(2.09)	(1.88)
Ln(NDIS)	9.17	5.13
	(1.36)	(0.91)
Ln(GDP)	-0.48	0.64
	(-0.08)	(0.15)
Ln(POP)	-14.3**	-7.91
	(-2.13)	(-1.37)
Ln(GOVSIZ)	7.14	6.70
	(0.68)	(1.04)
NUCLE	-0.35	-0.34
	(-0.60)	(-0.71)
Constant	97.2	29.6
	(1.01)	(0.39)
Over-identification	0.66	0.16
test	P=0.41	P=0.68
Adjusted $R^2$	0.26	0.26
Observations	37	37

Note: Values in parentheses are t-statistics calculated by the robust standard errors obtained using the jackknife method. \* and \*\* denote significance at the 10% and 5% levels, respectively. Instrumental variables are population density and land size. The Sargan test was used for the over-identification test.

Appendix. Table A1. List of countries used in the analysis

With nuclear plants	Without nuclear plants		
Belgium	Austria		
Brazil	Azerbaijan		
Bulgaria	Bangladesh		
Canada	Bosnia and Herzegovina		
China	Cameroon		
Czech	Colombia		
Finland	Egypt		
Germany	Georgia		
India	Greece		
Korea (South)	Hong Kong		
Netherlands	Iraq		
Pakistan	Italy		
Romania	Kenya		
Russia	Macedonia		
South Africa	Morocco		
Spain	Nigeria		
Switzerland	Poland		
United States	Turkey		
	Vietnam		

Note: As in the countries listed in Table A1, surveys were also conducted in a further 10 respondent countries. The question regarding the independent variable concerning "possibility of nuclear accident" was not asked of eight of these countries. Aside from the countries without the dependent variable, data regarding the number of technological disasters were not available for two countries. Hence, these 10 countries were not included in the analysis.