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Who perceive seasonality change? A case of the Meghna basin, Bangladesh

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Who perceive seasonality change? A case of the Meghna basin, Bangladesh

Moinul Islam^{*} Koji Kotani^{†,‡,§,¶,}

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Abstract

Global climate change is a scientifically demonstrated phenomenon, but there are great discrepancies in different societies about how people perceive it. It has been claimed that people's correct perceptions to climate change are necessary for mitigation and adaptation; however, most research in this regard has focused on knowledge about temporal trends of climate: no reports have examined people's perceptions of climatic regularity and patterns, i.e., seasonality change. In this study, we investigate people's perception to seasonality change in the Meghna basin, Bangladesh where catastrophic flooding occurs and the number of seasons in an annual calendar year is reported to have decreased from six to four. We conduct interviews with 7 experts and surveys with 1011 respondents. With the data, we empirically characterize participants' perceptions to seasonality change in relation to sociodemographic factors and life experiences. The results show that dependence on natural resources in profession, experiences of natural disasters and life history in the residential area shape people's accurate perception to climate seasonality.

Key Words: Perception to seasonality change; natural resource dependence; experiences of natural disasters; life history; Bangladesh

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1 **Introduction**

Climate change impacts on the seasonal distribution of the weather anomalies worldwide. Such 2 changes of weather will impact more in the areas where water availability and timing are important 3 factors to control ecosystem (Weltzin et al., 2003). Climate change altered the seasonal distribu-4 tions of rainfall in tropical areas (Easterling et al., 2000, Feng et al., 2013). Bangladesh is located 5 in the tropical region and climatically endangered (Mechler and Bouwer, 2015). Islam and Kotani 6 (2016) identify that, in Bangladesh, seasonality of temperature and precipitation patterns have 7 changed over the last 25 years. Formerly, there were six seasons in Bangladesh and now there 8 are just four. One of the main concerns of this seasonal change in Bangladesh is that many poor 9 inhabitants will fall in the poverty trap due to their dependance on nature in subsistence life. To 10 ensure the seasonal change adaptation and sustainable development, it is important to understand 11 the possible important factors that shape individuals seasonal change perception. Therefore, this 12 paper address the issue by conducting field survey in rural and urban areas of Bangladesh. 13

A group of research focus on the importance of personal experience of natural disaster on cli-14 mate change perception based on the primary information by using questionnaire survey in various 15 disaster-prone countries in the world. Frondel et al. (2017) identify that, in Germany, personal 16 experiences of adverse events are strong drivers of climate risk perception. They establish that dis-17 aster experience guides individuls to shape their climate change perceptions. Dai et al. (2015) show 18 that, in China, physical or financial damages due to extreme weather events are positively corre-19 alted with global climate change beliefs. Zaalberg et al. (2009) conduct survey on flood victims 20 and non-victims in the Netherland and identify that victims perceive themselves as more vulnera-21 ble than non-victims to future flooding. Siegrist and Gutscher (2006) select the respondents from 22 high and low flood risk areas of Switzerland and identify that respondents experiences of flooding 23 is positively related with their flood risk perceptions. Overall, these studies identify that experience 24 of disaster shaping individual's climate risk perception in different countries. 25

Several studies prioritize the importance of geographic location on correct perception of cli mate changes. Leiserowitz (2006), Gromet et al. (2013), Kahan et al. (2011), McCright and Dunlap

(2011) notice that climate change perceptions very geographically as a function of cultural and ide-28 ological factors. Generally, people live in same region for longer period of time cluster together 29 based on these factors. Akerlof et al. (2013), Egan and Mullin (2012), Goebbert et al. (2012), 30 Hamilton and Keim (2009), Howe et al. (2013), Capstick and Pidgeon (2014) conclude that ge-31 ographic location is correalted with personal experience of extreme weather events and climate 32 variability which shape individuals climate change perceptions. Overall, these studies identify that 33 geographic location of individuals can be an important factor to shape individual's climate change 34 perception. 35

Climate seasonality is likely to become more pronounced in Bangladesh, where daily life is 36 dependent on the weather anomalies. Citizens may be able to reduce the impact of seasonal vari-37 ability by adapting or adjusting their daily lives in response to weather stimuli. However, not all 38 individuals may respond similarly to seasonal changes. To estimate the impact of future climate 39 change, it is necessary to develop a predictive framework towards understanding who should adapt 40 to seasonal changes and how and why they should do so. It could be hypothesized that there are 41 several important determinants of individuals correct perception of the climate seasonality. To 42 test this hypothesis, a field survey was conducted on the local residents of Bangladesh, where the 43 dependance on natural resources in profession, natural disaster experience, and life history in the 44 residential area are identified as important determinants of the climate seasonality perception. 45

46 2 Methods

We consider people's perception to climate seasonality as a main variable of our interest. In the study region, it has been hypothesized that the temperature and rainfall trends over the last 25 years have become less variant. The changes in the seasonal patterns are documented to be more clearly visible through visual observations and experiences to climate in the region, such as shorter winter, longer summer, less rainfall and so on (Alam and Rabbani, 2007). Islam and Kotani (2016) empirically establish that the annual calendar in the study region transitions from six seasons to four seasons by applying nonparametric statistical analyses to the meteorological data over the
 last 50 years. Building upon the evidences in the study region demonstrated by past literature, we
 address who perceive such climate seasonality by statistical analyses that will follow.

56 2.1 Study area

We choose the Meghna Basin area of Bangladesh as our study area because of frequent flood-57 ing and climate seasonality. Within the Meghna Basin, we select the administrative upazilas of 58 Narsingdi Sadar and Roypura, because they are characterized by different production potentials. 59 Narsingdi Sadar includes villages 1 – 14 (figure 1). Narsingdi Sadar is located between 23°46' and 60 23°58' north and between 90°36' and 92°50' east. According to the census of 2011, the total area 61 of Narsingdi Sadar was 52742 acres where 149820 households lived, the population was 707525 62 and the population density was 3315 km^{-2} . Roypura includes villages 15 - 37 (figure 1). Roypura 63 is located between 23°55' and 24°4' north and between 90°44' and 90°59' east, and the area is 64 77 287 acres. There were 110 520 households, and the population was 535 796. The population 65 density of Roypura was 1713 km⁻². Narsingdi Sadar had a higher industrial potential and Roypura 66 a relatively high agricultural potential.¹ 67

68

[Figure 1 about here.]

The study areas had relatively uniform climate conditions with respect to temperature, high 69 humidity, and heavy rainfall. They are located at a low altitude, and the average annual temperature 70 ranges from 13 °C to 35 °C. Heavy rainfall usually occurs from June to October, which is the 71 monsoon season. Three major rivers - the Meghna, Old Brahmaputra and Arial Khan - are the 72 main water source for the area. Local agriculture is completely dependent on the natural water 73 supply. During the monsoons, flooding is a normal part of the local ecology. The 1998 flood was 74 the most serious in that region, destroying the economy of the whole area. It inundated over 70 %75 of the total land area and caused severe damage to lives and properties. In all, over 30 million 76

¹Census 2011 information is available at www.bbs.gov.bd.

people were affected as were about 1 million homesteads. Riverbanks in the Meghna Basin are not
protected by embankments. Locals have to resort to adaptation strategies to reduce the impacts of
floods.

80 **2.2 Data**

This study uses primary data collected from the field survey. We design the survey to examine 81 public perceptions to seasonality changes and the related issues: the aim is to elicit individual per-82 ceptions and obtain insight into the reasons for the correct perceptions. We use a general household 83 as the unit of analysis and regard the earning male as the decision maker. A structured question-84 naire is employed to collect the data. Care is given with the items in the survey questionnaire 85 to ensure that respondents do not require expert knowledge to answer. The data were collected 86 through interviews by research assistants using stratified random sampling. We conducted the sur-87 vey with 1020 households. However, nine households did not complete the questionnaire, and 88 we omitted them from the analysis. Thus, the final sample comprises 1011 respondents. We also 89 survey seven experts that specialize in flood control; they answer the questions related to climate 90 changes and perceptions of flood risk in the study area. We design a separate questionnaire to 91 elicit the expert responses about various issues concerning climate change and flooding there. The 92 results of the experts' interviews are not included in the statistical analysis; however, we reference 93 their responses to ensure qualitative judgement throughout the study. 94

The questionnaire for general respondents comprises the following range of items: perceptions 95 to climate change, in particular, seasonality changes; socioeconomic conditions and variables; 96 damage from and experiences of major floods; participation in disaster risk management and adap-97 tation measures; and life history in the residential area. We use the variables listed in table 1 for 98 statistical analysis. Note that we have collected the climate data concerning seasonal changes in 99 Bangladesh, demonstrating that the distributions of seasonal mean temperatures and mean rainfall 100 anomalies have become homogeneous among the seasons. More specifically, climate variation has 101 declined in the way that the annual calendar in the study area transitions from six to four seasons 102

(Islam and Kotani, 2016). Our survey is designed to check whether the respondents perceive such
 seasonality changes.

105

[Table 1 about here.]

Households in the study area are subject to the risks of seasonal flooding, and those who meet 106 the criteria and participate in the survey were compensated for their time with a gift of an equivalent 107 value. All the items in the questionnaire had to be answered. We categorize the flood experiences 108 of respondents based on their financial losses from the 1998 flood. Three questions concerning cli-109 mate change on local flood, sea level rise and climate change are posed as variables of information 110 access. Bangladeshis have a moderate level of awareness of climate change. Figure 2 presents the 111 respondents' overall climate change awareness and attitudes. Surprisingly, 78 % of respondents 112 have knowledge about climate change. However, only 50 % consider climate change as an im-113 portant issue. To obtain the respondents' perceptions of seasonality changes, we ask "Have you 114 noticed seasonal changes in your area over the last 25 years?" Most respondents (65.28 %) think 115 that such changes have not been evident, while the impact of climate change on the Meghna Basin 116 should be considerable (Mirza, 2003, Masood and Takeuchi, 2015).² 117

118

[Figure 2 about here.]

119 2.3 Statistical analysis

We apply logit regression to model the respondents' perception to climate seasonality in relation to sociodemographic variables, experiences and information access. Whether or not a respondent perceives climate seasonality is expressed as a binary-choice model of logiti regression, assuming that the perception depends on observable factors. Let Y_i be a dummy variable taking $Y_i = 1$ when respondent *i* perceives seasonality change, otherwise $Y_i = 0$. The probability for respondent *i* to perceive climate seasonality, denoted by $Prob(Y_i = 1)$, is assumed to follow

²That region will have played a major role in the occurrences of flooding in Bangladesh over the next several decades (Mirza, 2003). The peak flow of water in the basin is predicted to increase by 39.10% during the monsoons and decrease up to 26.90% in the dry season (Kamal et al., 2013).

the distribution function F evaluated at $X_i\beta$ where X_i is a $1 \times (k+1)$ vector of explanatory variables for respondent i ($X_i = (1, x_{i1}, ..., x_{ik})$) and β is a $(k+1) \times 1$ vector of parameters ($\beta = (\beta_0, \beta_1, ..., \beta_k)'$). The logit regression takes the following form of a distribution function:

$$\operatorname{Prob}(Y_i = 1) = \frac{\exp\left(X_i\boldsymbol{\beta}\right)}{1 + \exp\left(X_i\boldsymbol{\beta}\right)} \tag{1}$$

A specification of equation (1) enables us to estimate parameters β via maximum likelihood and the probability for a respondent to perceive climate seasonality in relation to explanatory variables (Wooldridge, 2016, 2010).

The explanatory variables used in this study include (1) education of household heads, (2) 123 income, (3) natural resource dependence in profession, (4) losses in the 1998 flood, (5) flood 124 information from neighbors, (6) living in an inherited house, (7) 20 years of residence in the study 125 area, (8) distance of a house from nearest river and (9) information access about local flood risk, sea 126 level rise and climate change risks. The definitions of the variables are summarized in table 1. We 127 are also interested in identifying the marginal effect of each explanatory variable on the response 128 probability of $Prob(Y_i = 1)$, holding all other variables fixed. More specifically, we estimate 129 the marginal impact on the probability for a respondent to perceive climate seasonality when one 130 explanatory (dummy) variable increases by one unit (or changes from zero to one) via equation (1). 131

132 **3 Results**

In our analysis, the perception of seasonal changes in Bangladesh is the outcome variable which is guided by different socioeconomic, geographic, demographic and regional climate knowledge variables. The estimated coefficient of the variables and the marginal effect is reported in table 2. The binary logit model results identify the relationship between the seasonal change perception and the control variables we use in the model. We notice that education of the household head is an important variable to improve the probability of correct seasonal change perception. The marginal effect of the logit model shows that educated respondent have 6 % higher probability of correct perception than those without any formal education. These findings reflect the result of Kvaløy et al.
(2012), where they identify positive correlation between education and the perception of global
warming. Education improve the concern of seasonal changes and shape individual's perception
of weather anomalies inconsistent patterns compared to previously experienced conditions.

144

[Table 2 about here.]

Income is one of the key indicators of individual's socioeconomic condition. This income level 145 impacts on seasonal change perception in Bangladesh. In our analysis education is considered 146 as a categorical variable and constructed 9 different categories after carefully observing the range 147 of national level income group classifications. The average gap between each income group is 148 approximately 5000 BDT (USD 65). Our logit regression result shows that change of income cat-149 egory from one group to the next increase the probability of correct seasonal change perception by 150 5%. This change can be related with the fact that the low income group, who are more vulnerable 151 to changing climate are less concerned about the seasonal changes in Bangladesh. 152

Livelihood and occupation in Bangladesh are not free from the dependence of nature. Tradi-153 tional agriculture, fishery and agricultural related business are highly dependent on the changes 154 of weather anomalies (e.g; rainfall, temperature). Having an occupation dependent on natural re-155 sources exerted a clear influence on perceiving climate seasonality. The regression results showed 156 that the coefficient for such occupations was positive and significant. According to the estimated 157 marginal effect, respondents with such occupations were 21 % more likely to perceive climate sea-158 sonality than other participants. Farming has a close interaction with the natural anomalies. As of 159 our knowledge, no literature empirically analyze the individual's perception on seasonality change 160 in relation to the natural resource dependence in occupation. Our result clearly shows that natural 161 resource dependence in occupations have important impact on shaping the perception of weather 162 anomalies. Climate change and natural disasters adversely impact production in many agriculture-163 dependent households. Thus, farming households are aware of climate variability and can identify 164 the most common climate factors. 165

We observe that natural disaster victims have significant positive relationship to seasonal change 166 perception. We found a clear difference between flood victims' and non-victims' knowledge and 167 perceptions of seasonal change. Recently flood losses have sharply increased in Bangladesh. For 168 instance, 1998 flood is one of the most devastating disasters Bangladesh has ever experienced and 169 affected almost the whole country. We clearly notice that the probability of making a correct re-170 sponse to changes in seasonal trends by a 1998 flood victim as significantly positive. An household 171 head who experienced losses of 1 lach BDT (1300 USD) had a 13 % higher probability of perceiv-172 ing climate seasonality than one who did not suffer any loss. In addition, residents who received 173 flood information from neighboring households had an 11% higher chance of perceiving climate 174 seasonality. These findings suggest that flood loss and information shape perceptions of climate 175 change in Bangladesh. From our empirical findings, it is evident that local flooding experience 176 provides opportunities to engage with the seasonality change perception. People may not take 177 necessary action to mitigate climate change if they lack first-hand experience of its consequences. 178 Better understanding of the mechanism that influences such perception may be relevant to commu-179 nication strategies for risk management. Individuals having direct experience of incidents directly 180 linked with seasonal change may undertake sustainable behavior. 181

We consider the residence time as a geographic factor that indicates individuals experience with 182 the local weather. Long residence in an area allows people to observe climate changes and patterns. 183 They can accurately compare previous and present conditions of climate variables. Long residence 184 also allows the sharing of experience with others. We notice that inherit householders have in-185 teraction with local residents as well as with local environment. In addition, the experience from 186 the ancestors also help to shape their personal perception in local Bangladesh. Our respondents 187 share their experience beyond the structural questionnaire and most of them noticed their ancestor 188 face less environmental variations than the present time. Our analysis suggest that individuals who 189 have an inherit house in the area have 7 % higher possibility to perceive the changing seasonality 190 in local Bangladesh compared to the individuals who don't receive any house inherently. 191

¹⁹² We also identify the distance of households from rivers as a geographic factor and it is an im-

portant variable regarding flood vulnerability. People living closer to rivers are concerned about 193 water-related disasters. Riverbanks in the Meghna Basin lack embankments. Therefore, local 194 communities are completely dependent on self-adaptation strategies. In this analysis, 53 % of re-195 spondents had been living in the study area for over 20 years. We found that longer residence could 196 increase the probability of correct perception of climate change by 6%. Long-term residence in 197 an area promotes accurate understanding of local weather trends. Longer residence helps increase 198 communication, which builds a stronger sense of community; it helps establish networks to share 199 information and inspire people to adapt. 200

We find that information about flood risk among residents guided them in developing accurate 201 perceptions of seasonal changes at the local scale. According to our estimated marginal effect, 202 knowledge of local flood risk increased the likelihood of perceiving local climate change season-203 ality by 9%. We observed no relationship between information about global climate change and 204 perceptions of local seasonal changes. Our findings suggest that efforts to enhance public infor-205 mation about global climate change are not the crucial factor in improving perceptions of local 206 seasonality change issues. Rather, information about local climate change risk is more appropriate 207 for increasing local seasonality perceptions. 208

209 4 Conclusion

The present study examine the influence of profession, disaster experience, and socio-economic 210 factors on perceiving climate seasonality. The development of mitigation and adaptation policies 211 targeting local communities is guided by perceptions of climate change. This paper offers insights 212 into adaptation to climate change by improving socioeconomic characteristics: it addresses the 213 existing lack in the literature-especially in developing regions-and it presents an approach to 214 recognizing climate seasonality. This empirical study finds that without personal experience, it is 215 difficult for individuals to accurately perceive climate change. Having an occupation that is de-216 pendent on natural resources also increases an understanding of climate seasonality. Experience of 217

flood damage raised the likelihood of correctly perceiving seasonal changes. Education also played an important role. Our results offer potentially important guidance for public risk perception. Simply considering people's risk perception may be insufficient, but it is important for resilience. To develop disaster-resilient communities towards finding better ways to manage risks for extreme and catastrophic events; perception of climate change is a key determinant in this regard.

Understanding and perceiving changed seasonality by natural-resource users is potentially im-223 portant: it has an influence on their adaptation practices. We may hypothesize that people with 224 flood experience are more likely to engage with climate change issues. Perceiving changes in local 225 weather patterns is an important precursor to proactive adaptation to climate change. Individual 226 perceptions of changes in local climate play an important role in shaping vulnerability to global 227 climate change, adaptive behavior, and support for adaptation and mitigation policies. There is a 228 wide discrepancy between climate change and accurate perception. Better education guide citi-229 zens to give greater weight to the local climate changes; that in turn shapes risk perceptions of the 230 general public. 231

Awareness of climate change is particularly important in regions where much of the population 232 is highly exposed and sensitive to the impacts of such changes. Understanding climate risks is 233 important: it can influence adaptation practices and disaster management. The World population 234 is going to be concentrated in urban cities within next 50 years and so on. Urbanization is taking 235 place in Asia and Africa, and 65-75% of the world population will be concentrated in urban cities 236 in Asia and Africa (Wigginton et al., 2016, McDonnell and MacGregor-Fors, 2016, Henderson 237 et al., 2016). However, the perception about changing seasonality and climate change may become 238 less likely to be predicted by the urban people who are less dependent on nature for their daily 239 life. Therefore, it will be very difficult for the majority of population to become concerned about 240 ongoing climate changes. This is alarming in the way that without clear climate change perception, 241 societies may become reluctant to reduce the activities which can directly and indirectly affect the 242 success of climate policies. 243

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Figure 1: The Meghna basin area of Bangladesh



Figure 2: Respondents' overall idea about climate change in the study area

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	Table 1: The	variables in statistical analysis		
Variable	Type	Description	Mean	SD
Dependent variable Seasonality change perception	Dummy	Correct perception $= 1$, otherwise 0	0.65	0.47
Independent variables Education ¹	Ordered categories	No eduction (0) to postgraduate (5)	0.11	1.19
Income ²	Ordered categories	No income (0) to over 40 000 BDT (8)	3.37	1.75
Natural resource dependence ³	Dummy	Natural resource dependence $= 1$, otherwise 0	0.95	0.22
Loss in 1998 flood	Continuous	Amount in lakh BDT ⁴	0.56	0.9
Flood info from neighbors	Dummy	Flood info from neighbors $= 1$, otherwise 0	0.13	0.34
Living in an inherited house	Dummy	Living in an inherited house $= 1$, otherwise 0	0.51	0.50
Residing over 20 years	Dummy	Residing in the area over 20 years $= 1$, otherwise 0	0.53	0.50
Living near a river	Dummy	Distance to the nearest river $\leq 1 \text{ km} = 1$, otherwise 0	0.82	0.38
Information access				
to local flood	Dummy	Information access to local flood $= 1$, otherwise 0	0.32	0.46
to sea level rise	Dummy	Information access to sea level rise $= 1$, otherwise 0	0.30	0.46
to climate change	Dummy	Information access to climate change $= 1$, otherwise 0	0.27	0.45
¹ 0: illiterate, 1: primary school 2 0: < 5000 1: $5000 - 9999$. 2:	, 2: secondary school, 10000 - 14999. 3: 1:	3: college, 4: bachelor's degree and 5: postgraduate degi 5000 - 19999 4: 20000 - 24999 5: 25000 - 29999 6: 30	ee 000 - 34	.7 .999

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⁻ 0: < 5000, 1: 5000 - 3939, 2: 10000 - 14399, 5: 15000 - 13399, 4: 20000 - 35000 - 39999 and 8: ≥ 40000 ³ It is whether a subject's profession is dependent on natural resources or not. ⁴ 1 lakh BDT = 100 000 BDT = 1300 USD

Explanatory variable	Model	Marginal effect
Socio-economic		
Education	0.308***	0.059 * * *
	(0.081)	(0.015)
Income	0.253***	0.048***
	(0.055)	(0.010)
Natural resource dependence in profession	1.102***	0.210***
	(0.309)	(0.058)
Natural disaster experience		
Loss in flood 1998	0.649 * * *	0.124 * * *
	(0.164)	(0.031)
Geographic location		
Living in an inherited house	0.360 * *	0.069 * *
	(0.157)	(0.030)
Residing over 20 years	0.322 **	0.062 * *
	(0.150)	(0.028)
Living near a river	0.359*	0.069*
	(0.187)	(0.036)
Information		
Flood info from neighbors	0.578 * *	0.110^{**}
	(0.241)	(0.046)
Information access to		
local flood	0.479*	0.091*
	(0.249)	(0.047)
sea level rise	-0.260	-0.050
	(0.251)	(0.048)
climate change	-0.188	-0.036
	(0.169)	(0.032)
Constant	-2.438^{***}	
	(0.411)	
Sample size	1011	
LR χ^2 statistic	174.680	
Log-likelihood	-565.448	
Pseudo- R^2	0.134	

Table 2: Estimation of the logit regression for respondents' perception to climate seasonality

*** significant at 1 % percent level

** significant at 5 % percent level

* significant at 10 % percent level

Standard errors are in parentheses.