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## Research Article

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## Author Contributions Statement

Wei Yang: Methodology, Conceptualization, Formal analysis, Investigation, Writing & editing.

Jianling Qi: Data curation, Conceptualization, Funding acquisition.

Waranan, Tantiwat & Muhammad, Arif: Formal analysis, Investigation.

Yao Lu & Jin Guo: Data curation, Funding acquisition, Conceptualization.

All authors reviewed the manuscript.

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## Declaration of interests

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

## **Abstract**

Biodegradable mulch films (BDMs) technology is an environmentally-friendly substitute to traditional plastic mulch films in agricultural production. Given the high price and it is new to the market, it is not easy for farmers to accept and adopt it. This paper aims to explore the key factors affecting farmers' adoption of and willingness to pay for BDMs to understand the complex process of farmers' decision-making. This paper employs a double hurdle model to explore the multi-stage decision-making process in the adoption of BDMs using the sample of 1247 observations from Yunnan province China, where two mechanisms of decision-making (i.e., direct rejection of technology and lack of resources) were used to capture zero willingness to pay (WTP) for BDMs. The results indicate the two-stage decision-making process, where the role of technology-specific characteristics is more important than adopter-specific characteristics in the adoption of BDMs in China – training for understanding and using the technology has a positive effect on both the adoption and willingness to pay. The paper is the first attempt that empirically analyses the determinants of farmers' WTP for BDMs. It contributes to the literature on adoption analysis by 1) considering farmers' adoption choices as a two-step process by using a hurdle model and 2) addressing the importance of technology-specific characteristics on farmers' WTP for BDMs. Understanding the role of factors on different stages of farmers' decision-making could assist policymakers in designing programs, specifically tackling difficulties confronting farmers at different stages of decision-making.

**Keywords:** double hurdle model; biodegradable mulch film; farmer; China; WTP

## 1 **1. Introduction**

2 With the population growth and economic development, the global usage of plastic is rising significantly,  
3 and one of the main sources is plastic films used in agriculture for mulching (Jambeck *et al.*, 2015).  
4 Different from the developed countries, where plastic mulch is mainly used with micro-irrigation, Chinese  
5 farmers use plastic mulch on a vast scale independent of micro-irrigation, but for water conservation, weed  
6 control, and higher production (Liu *et al.*, 2014). So far, the use of plastic film mulching technology  
7 (mainly polyethylene (PE) mulches) in China has helped increase the yields of cash crops by 20% to 60%  
8 (Ingman, Santelmann, and Tilt 2015), and its land area in plastic mulch has exceeded the world's total  
9 land area in micro-irrigation (Liu *et al.*, 2014). However, the extensive use of plastic mulch films has  
10 caused severe environmental problems, because a large number of plastic mulch residues have not been  
11 appropriately recycled, directly leading to land and water pollution (Roy and Dutta, 2019). Particularly,  
12 in the high mountain areas of southwestern China, such as Sichuan and Yunnan province, the use of  
13 recycling machines is limited by the landscape, and thus recycling is associated with high labor demand  
14 and cost. Besides the geographical disadvantages, the plastic mulch film used in China is thinner than  
15 0.008 mm that is significantly lower than the international standard (e.g. 0.02 mm in Europe and Japan)  
16 (Liu *et al.*, 2014), making it more difficult to recycle the residues completely. According to a national  
17 survey in 2019, about one-third of plastic mulch films were left in the land, and 80% of plastic mulch  
18 films picked up from land were either dumped without proper treatments or burned up (Yan and Liu,  
19 2020). Consequently, although the usage of plastic mulch films has not been the largest compared to that  
20 of the northwestern area, the residues of plastic films in the southwestern provinces have been  
21 substantially high (Yan and Liu, 2020).

22 Therefore, besides encouraging farmers to adopt “best recycling practices”, the Chinese government  
23 starts introducing biodegradable mulch films (BDMs) through demonstration farms, in particular in the  
24 regions confronting the difficulty of recycling plastic film residues (e.g., southwestern China). Meanwhile,  
25 there has been a change in crop choice from planting traditional food crops (e.g., rice and wheat) to cash  
26 crops, such as fruits and vegetables, mainly due to the high return of cash crops and policy supports from  
27 the central and local government (Huang *et al.*, 2010). It is expected that the use of plastic films in  
28 southwestern China will continue to grow significantly (Yan and Liu, 2020). However, except for the  
29 demonstration farms operated by the government, few farmers have attempted to adopt BDM, mainly  
30 due to its high price that is about 3 times higher than PE films, and uncertainty about its efficiency that  
31 BDMs can be broken down easily and not having the expected functions (e.g., water conservation). It is,  
32 therefore, important to understand what drives or hinders farmers' adoption the new technology, i.e.,  
33 BDMs, to maintain crop production and productivity while reducing the environmental pollution of

34 plastic film residues. Note that many smallholder farmers rely on subsidies in technology adoption in  
35 China (Ding *et al.*, 2011; Yu and Jensen, 2010), and thus the adoption of BDMs may rely on the cost-  
36 effectiveness of the subsidy policy. Hence, prior to implementing any subsidy policy for the adoption of  
37 BDMs, policymakers need to know what determines farmers' adoption and willingness to pay for BDMs.

38 The existing studies show that farmers' adoption of sustainable practices and new technologies are  
39 contingent on a number of factors that are categorized into five broader categories, namely, farm and  
40 farmer specific characteristics, social and cultural norms, availability of support and resources, and  
41 perceived benefit. (e.g., Barham *et al.*, 2015; Dumbrell *et al.*, 2016; Mekonnen *et al.*, 2020; Pannell *et al.*,  
42 2006; Weber, 2012). Farmers' adoption is also dependent on the types of practices and technologies, but  
43 the practice or technology itself has been under-emphasized (Pannell and Zilberman, 2020). A recent  
44 study by de Oca Munguia and Llewellyn (2020) states that more emphasis should be placed on analyzing  
45 the impact of technology characteristics in adoption analysis. Additionally, when targeting developing  
46 countries, the adoption analysis needs to consider the distinctive features when analyzing farmers'  
47 adoption decision-makings (Pannell and Zilberman, 2020). For example, the majority of farms in the  
48 developing countries are small scale rather than large commercial farms (Llewellyn and Brown, 2020);  
49 the role of technologies in agricultural production is to enhance crops to further increase production and  
50 feed the poor people in developing countries (Huang *et al.*, 2002). In that way, the impact of farm and  
51 farmer characteristics on adoption behavior may be dependent on the economic-historical context (Burton  
52 2014). Besides, farmers in developing countries usually lack an adequate understanding of the new  
53 technologies due to the difficulty of accessing relevant information or training programs (Chen *et al.*,  
54 2013; Yang *et al.*, 2021). Therefore, these two gaps addressed in the adoption literature lead us to consider  
55 the technology-specific characteristics and the distinctive features of the targeted region in China in  
56 analyzing the determinants of farmers' adoption of BDMs.

57 It was not until recent years that researchers attempted to address the issue of plastic pollution from  
58 agricultural production in China. The existing studies are from the field of agriculture and environmental  
59 science, mainly focusing on measuring the amount of residual mulch films for different crops and the  
60 impact on crop yields and greenhouse gas emissions. (e.g., He *et al.*, 2018; Zhang *et al.*, 2016). To our  
61 best knowledge, no study attempts to address plastic film pollution from understanding farmers' decision-  
62 making. Till now, the literature on the determinants of farmers' decision-making of sustainable agricultural  
63 practices in China have mainly focused on good management of fertilizer, pesticide, and water quality  
64 protection (e.g. Pan *et al.*, 2017; Sun *et al.*, 2018; Wang *et al.* 2018; Wu and Hou, 2012).

65 Therefore, this study aims to investigate determinants of farmers' choices of adoption and willingness  
66 to pay for BDMs in China by using a double hurdle model, where two mechanisms of decision-making  
67 (i.e., a direct rejection of technology and lack of resources) were used to capture zero willingness to pay

68 (WTP) for BDMs. A sample of 1247 observations used in the study was sourced from a survey of farmers  
69 in Yunnan province, China, mainly because: 1) the usage of plastic mulch films is relatively high (Liu,  
70 He and Yan, 2014); 2) the central and local government starts promoting the use of BDMs. The  
71 contributions of this study are threefold. First, it addresses the importance of technology-specific  
72 characteristics and the impact on farmers' adoption and pay for the new technology, BDMs in our case.  
73 Second, the study considers farmers' adoption choices as a two-step process by using a double hurdle  
74 model, where farmers have to overcome the hurdle of whether or not to adopt and achieve the decision of  
75 how much to pay for the technology. Third, the results and findings of the study may contribute to the  
76 design of policy instruments in motivating the adoption of BDMs in China, for instance, to provide some  
77 insights into determining the value of substitutes for farmers' adoption of BDMs.

78 The rest of the paper is structured as follows. Section 2 specifies the conceptual analysis framework  
79 and econometric models used in the study. Section 3 presents the sample data and descriptive statistics of  
80 the variables used in the econometric models. The empirical results and findings are presented and  
81 discussed in Section 4, followed by the last section to conclude.

## 82 **2. Method and Data**

### 83 **2.1. Empirical Analysis Framework**

84 The decision on whether or not to adopt a specific technology and how much to pay for that technology  
85 can be made jointly or separately (Gebremedhin and Swinton, 2003). Either way can cause the censored  
86 nature of farmers' WTP, with a large proportion of zeros presented in WTP. To deal with the zero-left  
87 censored WTP, the Tobit model (Tobin 1958) and the extensions have been largely used to correct the  
88 problem of zero observations in a variety of research fields, in particular in analyzing consumer  
89 expenditure and demand (Gallet and List, 2003; Jones 1989). This paper follows the analysis framework  
90 for zero-left censored WTP from consumer expenditure (Gillingham and Tsvetanov 2019; Jones 1989),  
91 and we conceptualize the sources of farmers' zero WTP for BDMs. As shown in Figure 1, there are two  
92 main sources of zero WTP, including 1) direct rejection: the farmer directly reject adopting BDMs due to  
93 the farmer's preferences over another mulching technology, for example, PE mulch films; 2) lack of  
94 resources: the farmer would like to adopt BDMs but cannot afford to pay it.

95

96

[insert Figure 1]

97

98 Therefore, the sources of zero WTP can be modeled as:

99

$$\begin{cases} Y_1^* = \beta_1^T X_1 + \varepsilon_1 \\ Y_2^* = \beta_2^T X_2 + \varepsilon_2 \end{cases}, \quad (1)$$

100 here the two equations represent the two decision mechanisms of zero WTP.  $Y_1^*$  and  $Y_2^*$  are latent  
 101 variables representing farmers' adoption decisions and levels of WTP for BDMs;  $X_1$  and  $X_2$  represent  
 102 the independent variables impacting the two decision mechanisms, associated with unknown vector  
 103 coefficients  $\beta_1$  and  $\beta_2$ ; and  $\varepsilon_1 : N(0,1)$  and  $\varepsilon_2 : N(0,\sigma^2)$  are the random disturbances. Specially,  
 104 for the technology rejection mechanism (first equation in Equation 1), if  $Y_1^* < 0$ , biodegradable mulching  
 105 technology is not adopted as it is not considered by farmers as a relevant mulching technology. The second  
 106 equation defines the levels of WTP for BDMs, and thus, if  $Y_2^* < 0$ , BDMs are not adopted because a  
 107 negative WTP implied by resource constraints – labor, financial, and information support cannot be  
 108 realized.

109 A double hurdle model is utilized to support the decision-making context conceptualized in Figure 1.  
 110 Originally formulated by (Cragg 1971), the double hurdle model is a parametric generalization of the  
 111 Tobit model, in which two separate stochastic processes determine the decision to adopt and WTP for the  
 112 technology (Jones, 1989; Yen and Jones, 1997). In the double hurdle model,  $Y_1^*$  and  $Y_2^*$  is used to model  
 113 each decision process, and both hurdles have equations (shown in Equation 1) associated with the effects  
 114 of the included factors as the independent variables. It assumes that farmers make two sequential decisions  
 115 concerning adopting and WTP for a technology. First, the farmer decides whether or not to adopt the  
 116 technology (the first hurdle). Second, the level of WTP that shall be used in purchasing the technology  
 117 (the second hurdle). Hence, the first hurdle is a sample selection equation estimated with a Probit model  
 118 and the second hurdle involves an outcome equation, which uses a truncated model to determine the WTP  
 119 for BDMs. This second hurdle uses observations only from those respondents who indicated a positive  
 120 WTP value of the use of a technology (Martínez-Espiñeira 2006; Noltze *et al.*, 2012).

121 The double hurdle model fits our problem of measuring crowding out because it allows for the fact  
 122 that fixed costs may affect a farmer's adoption, but once the decision to adopt has been made, fixed costs  
 123 may not affect the WTP decision. Each hurdle is conditioned by factors, such as farmer and farm  
 124 characteristics (adopter-specific characteristics), technology-specific characteristics, and farmers'  
 125 awareness, perceptions, and knowledge of the technology (shown in the middle of Figure 1). Note that  
 126 these independent variables may appear in both equations or either of one, and most importantly, a  
 127 variable appearing in both equations may have opposite effects in the two equations (Jones, 1989; Yen  
 128 and Jones, 1997).

129 Combining the log-likelihood function for zero and positive WTP observations, the sample likelihood  
 130 function for this double hurdle model can then be written as (Jones 1989):

$$131 \quad \ln L = \sum_{i|Y_i=0} \ln L_i^- + \sum_{i|Y_i>0} \ln L_i^+, \quad (2)$$

132 where the first term estimates the status of  $Y_i$  (whether  $Y_i = 0$  or  $Y_i > 0$ ) capturing the contribution of a  
 133 zero observation to the sample log-likelihood function and the second term estimates WTP (the exact  
 134 value of  $Y_i$ , if  $Y_i > 0$ ). Specifically,

$$135 \quad \ln L_i^- = \ln \left( 1 - \frac{\Phi(\beta_1^T X_{1i}, \frac{\beta_2^T X_{2i}}{\sigma}; \rho)}{\Phi(\frac{\beta_2^T X_{2i}}{\sigma})} \right); \quad (3)$$

$$\ln L_i^+ = -\ln \sigma + \ln \Phi\left(\frac{e_i}{\sigma}\right) + \ln \Phi\left(\frac{\beta_1^T X_{1i} + \frac{\rho}{\sigma} e_i}{\sqrt{(1-\rho^2)}}\right) - \ln \Phi\left(\frac{\beta_2^T X_{2i}}{\sigma}\right), \quad e = Y_i - \beta_2^T X_{2i},$$

136 here,  $\Phi(\bullet)$  denotes the probability density and cumulative distribution function of an  $N(0,1)$  random  
 137 variable; and  $e$  is the “residual” of the fit representing the contribution of a positive observation to the log-  
 138 likelihood function. Note the general model shown in the above nests several other formulations. We  
 139 assume  $\rho \neq 0$ , indicating the adoption equation and WTP equation are correlated, i.e.,  $\varepsilon_1$  and  $\varepsilon_2$  are not  
 140 independent. Note when  $\rho = 0$ , the double hurdle model may collapse to the independent Cragg model  
 141 or the Tobit model nested within the independent Cragg model, with the further assumption that the  
 142 adoption probability is 1. Hence, further statistical tests will be used to test for the independence of the  
 143 adoption equation and WTP equation.

## 144 2.2. Data and variables

145 The study area Yunnan province is located in the southwest of China, which has a strong agricultural  
 146 focus. However, level land for agricultural production is scarce in Yunnan, with only 5% of the land is  
 147 under cultivation and more than 94% of the land categorized as mountainous areas (Ding *et al.*, 2011).  
 148 Given the geographical disadvantages, plastic mulch films have been extensively applied in Yunnan to  
 149 help increase production and productivity. In addition, the types of crops planted in Yunnan determine the  
 150 large use of plastic mulch films – besides having the traditional food crops, such as rice, Yunnan’s  
 151 agriculture industry is well-known for its cash crops. In particular, the tobacco industry is the main  
 152 “export” product and makes up a large part of the provincial GDP, and the flower industry takes 50% of  
 153 China’s cut flower production. Additionally, in recent years, Yunnan has developed strong competitive  
 154 potential in its fruit and vegetable industries due to its climatic and ecological advantages, and the high  
 155 demand from the market further drives the expansion of planting areas of fruit and vegetable. The



156 technology of plastic mulching helps maintain the high yield of tobacco and fruit and vegetable. Besides  
157 the difficulty of planting in upland plains and sloped hillsides, farmers find it hard to collect residual plastic  
158 mulch films. Hence, though not being ranked at the top regarding the usage of PE films, the residues of  
159 plastic films in Yunnan have been substantially high (Yan and Liu, 2020). Being one of the BDMs  
160 “pilots”, the local government has built demonstration farms to show and educate farmers regarding the  
161 utilization of BDMs. However, till now, the adoption rate of BDMs has almost been zero.

162 Data used in this study were collected through a survey from different crop growers in Yunnan  
163 province of China between July and November 2018. The objectives of the survey are to analyze pollution  
164 sources of agricultural production and understand the current status of using plastic mulch films and  
165 farmers’ adoption and pay for BDMs. The design of the questionnaires for the survey has undergone two  
166 stages. First, we conducted a pilot survey in three randomly chosen villages of Yunnan. 36 farmers were  
167 randomly chosen and interviewed by trained interviewers in person. The results of the pilot study provide  
168 a thorough understanding of the utilization of plastic mulch films in the context of Yunnan, and the initial  
169 questionnaire was tested and adjusted accordingly. Based on the results of the pilot study, a structured  
170 farm and household questionnaire was finalized and used in the second stage survey. Several trained  
171 interviewers were sent out to conduct face-to-face interviews with 1358 farmers randomly chosen from  
172 128 villages out of 26 counties in Yunnan. This provides a final sample of 1247 valid questionnaires<sup>1</sup>.  
173 Specifically, we collected the information about farmers’ adoption and willingness to pay for BDMs, their  
174 perception and knowledge of the plastic mulching technology for both PE mulch films and BDMs, and  
175 farm and farmer characteristics (e.g. farm size, age, and income). Table 1 presents a descriptive and  
176 statistical summary of the variables included in the study.

177

[Insert Table 1]

178

179

180 There are two outcome variables (dependent variables), adoption and WTP, to be included in the two  
181 equations of the double hurdle model. Figure 2 shows a clear pattern of the zero-left censored WTP, with  
182 195 farmers chose not to adopt BDMs. The average per kg WTP for BDMs was found to be 13.7 China  
183 yuan across the farmers who are willing to adopt BDMs. Although 83% of farmers stated they are willing  
184 to adopt BDMs, when all the zeros (not willing) are included in calculating the average WTP, we got a  
185 lower per kg WTP of 11.6 China yuan.

186

[Insert Figure 2]

---

<sup>1</sup> Note that 111 questionnaires were excluded as they are either incomplete or errorness.

187 The potential determinants of farmers' adoption and WTP are categorized into three groups of  
188 independent variables, including farm and farmer characteristics, technology-specific characteristics, and  
189 farmers' awareness, perception, and knowledge of plastic mulching technology. Farm and farmer  
190 characteristics include demographic factors, such as farmers' age, gender, ethnicity, and education, and  
191 income level, as well as farm-related variables of farm type, farm size, and types of crops planted. Overall,  
192 as shown in Table 1, farmers in the sample are mainly male with a relatively low education level, and the  
193 majority of the farms are small-scale, with a large proportion of them planting tobacco and fruit and  
194 vegetables. Technology-specific characteristics are closely related to farmers' awareness and knowledge  
195 of BDMs. In our case, the efficiency of BDMs is highly valued by farmers followed by the attribute of  
196 price, and only a small proportion of farmers (21%) are aware of any forms of promotions for the adoption  
197 of BDMs. Note we included the variables that are related to PE mulches, given BDMs are seen as an  
198 alternative technology to PE mulches to be adopted in the future. Although the majority of farmers notice  
199 the negative impact of PE mulches on the environment, about 61% of farmers have maintained the same  
200 level of PE usage and 28% have it increased in the past five years. This is mainly explained by their  
201 perceptions of the usefulness of plastic mulching technology in agricultural production, with  
202 approximately 96% of farmers see the technology as important. Lastly, getting a subsidy for recycling  
203 plastic residues and training opportunities may also affect their decisions on using BDMs.

### 204 **3. Empirical Results**

205 For comparison purposes, we report the results from the double hurdle model and Tobit model regarding  
206 factors affecting farmers' WTP for BDMs in Table 2: the first two columns present results from the double  
207 hurdle model for adoption and WTP and the third column presents the results from the Tobit model. For  
208 the factors affecting WTP (the second and third columns), the coefficient estimates of the double hurdle  
209 model are different from those of the Tobit model at various degrees. Notably, for example, the main crop  
210 type – tobacco is found to have no effect on farmers' WTP for BDMs in the double hurdle model but is  
211 negatively significant in the Tobit model; for the variables of "main crop", statistically significant effects  
212 are detected in the Tobit model but not in the double hurdle model. The differences in the estimation  
213 results of the two models indicate the Tobit model cannot capture the technology selection mechanism  
214 shown in Figure 1, where farmers first choose to adopt BDMs and then decide their WTP for the  
215 technology. Once they go over the first hurdle, some factors affecting the first hurdle may not affect the  
216 second hurdle of WTP. Note that the correlation coefficient  $\rho$  is positive and statistically significant,  
217 indicating the two hurdles are positively correlated, and thus the two selection mechanisms need to be  
218 considered jointly. In addition, the values of Loglik, AIC, and BIC support choosing the double hurdle  
219 model to estimate farmers' WTP for BDMs. Hence, our interpretation of the results is based on the double

220 hurdle model: in hurdle 1, farmers decide whether or not to adopt BDMs, and if one chooses to adopt,  
221 hurdle 2 considers ones' WTP associated with per kg BDMs.

222 [Insert Table 2]

223

224 For the variables of farm and farmer characteristics, the demographic factors, such as age, gender, and  
225 ethnicity, and farm characteristics, including farm size, farm type, and the number of labors, have no  
226 impact on farmers' adoption of BDMs. However, education is found to be an important determinant of  
227 both farmers' adoption and WTP. The positive effect of education indicates that farmers with higher  
228 education levels tend to be more likely to adopt BDMs and are willing to pay more for the use of BDMs.  
229 This finding is consistent with many existing studies on technology adoption for farmers across the  
230 developed and developing world (e.g., Dumbrell *et al.*, 2016; Ma *et al.*, 2012; Pannell *et al.*, 2006). For  
231 other farm characteristics, income and the type of crop, are found to only influence farmers' adoption OF  
232 BDMs. Farmers with higher household income are more likely to adopt BDMs, but they are less likely to  
233 adopt BDMs if the percentage of agricultural income increases. Intuitively, it is easier for farmers with  
234 higher household income to try out new technologies, given the high price of BDMs and the fixed cost  
235 associated with learning and training for the adoption of BDMs (Barham *et al.*, 2015; Marí *et al.*, 2019).  
236 However, farmers who mainly rely on agricultural income may concern about the high cost of BDMs and  
237 find it risky to invest in BDMs, compared to those who have more off-farm income (Gedikoglu *et al.*,  
238 2011).

239 Farmers' adoption of BDMs differs across types of crops planted. Compared to food crop growers,  
240 tobacco and other crop growers are less likely to adopt BDMs, and fruit & vegetable growers are more  
241 likely to adopt BDMs. Sastre *et al.* (2017) state that farmers often relate themselves to the crops they grow,  
242 and some are proud of growing the crops for generations – their decision-making of adopting technology  
243 and practice is dependent on the types of crops they grow. In our case, the difference across types of crops  
244 indicates a direct technology rejection of tobacco and flower growers. As the pillar industry for Yunnan's  
245 economy, the tobacco industry has its own supply chain and operation system to manage suppliers  
246 (tobacco growers). For instance, to ensure the quality and quantity of tobacco production, the industry  
247 provides a variety of support for tobacco growers, including providing low price PE mulch films, training,  
248 and subsidy (Huang *et al.*, 2010). Similarly, flower cooperatives provide flower growers with strong  
249 support, including technology training and networking. Given the support they obtain from the industry,  
250 the two groups of farmers are less likely to change from using PE mulch films to BDMs. Table 3 shows  
251 the predicted adoption (the probability) and WTP (per kg BDMs in China Yuan) across the groups of  
252 growers at the sample means of independent variables. Based on the results, the estimated probability of  
253 adopting BDMs for tobacco growers and other crop growers are relatively low compared to food crop

254 growers and fruit & vegetable growers. Regarding the predicted values of WTP, fruit & vegetable growers  
255 have the highest WTP whilst tobacco growers have the lowest. However, regardless of the types of  
256 growers, the estimated WTP is similar to the per kg price of PE films (around 10-14 China Yuan),  
257 meaning it is far less than the market price of BMDs (22-28 China Yuan).

258 [Insert Table 3]

259

260 Compared to farm and farmer characteristics, more technology-specific variables are found to affect  
261 farmers' adoption of and WTP for BDMs. Higher usage of PE mulch films increases the propensity for  
262 farmers to adopt BDMs, and farmers' WTP for BDMs is higher if they have had an increase in mulch  
263 usage in the past five years. Training is found to be an important determinant of farmers' adoption and  
264 WTP; those who obtain training for mulching technology (offered by the government or industry) are  
265 more likely to adopt and pay more for BDMs. Training in the form of workshops, field days on  
266 demonstration farms, or "farmer school" are generally regarded to an important pathway of knowledge  
267 transformation: experts can directly provide farmers with information and knowledge about new  
268 technologies; farmers can network and share knowledge with other farmers, increasing their confidence  
269 in adopting new technologies or practices (Baird et al., 2016). Receiving a subsidy for recycling residual  
270 plastic films does not affect either farmers' adoption or WTP.

271 The attributes that are directly related to BDMs may influence farmers' choices of whether or not to  
272 adopt BDMs and how much to pay for BDMs. Interestingly, farmers who see price as the most important  
273 attribute are more likely to adopt and pay more for BDMs than those who value the other attributes of  
274 BDMs, such as efficiency and quality. This finding indicates that, when price is the focus of BDMs,  
275 farmers may have already overcome the adoption barrier (e.g., technology barrier), whilst farmers who  
276 value efficiency and quality the most are still skeptical about the technology. Brand and farmers'  
277 awareness of the promotion of BDMs have no influence on their adoption and WTP.

278 All the perception variables are found to influence farmers' adoption but not WTP. Note that, if farmers  
279 are aware of the negative impact of PE mulch films on the environment, they tend to adopt an  
280 environmentally friendly technology, i.e., BDMs. Their perceptions of the usefulness of the mulching  
281 technology increase the likelihood of BDMs adoption – the more important they feel about plastic film  
282 mulching in agricultural production the more likely they adopt BDMs. This reflects the natural embedded  
283 connection between perception/ awareness and behavior change (Ajzen, 1991, Wang *et al.*, 2018).  
284 However, high perception and/ or awareness may not necessarily lead to high WTP.

#### 285 4. Discussion and Conclusion

286 This paper uses a double hurdle model to empirically analyze the determinants of farmers' adoption of  
287 BDMs in China, using the survey data from Yunnan province. It is the first attempt that explores both  
288 farmers' adoption of and WTP for BDMs. Considering using BDMs is relatively new in China and many  
289 other countries, understanding the different factors that affect farmers' decision-making is important for  
290 promoting the use of BDMs in agriculture. The double hurdle model has the advantage of modeling  
291 farmers' decision-making in two stages. We address the impact of three groups of factors, i.e., farm and  
292 farmer characteristics, technology-related and perception - and awareness-related variables, on farmers'  
293 adoption and WTP. Our results show that farmers have to overcome the first hurdle, i.e., choose to adopt  
294 or not adopt, and then decide how much they are willing to pay for BDMs. Notably, we find that  
295 education, training, and preference over the attributes of BDMs significantly affect farmers' adoption and  
296 WTP. However, once farmers decide to adopt BDMs, many other factors that affect their adoption, such  
297 as income and crop type, may not affect their WTP for BDMs.

298 Results and findings of the study may lead to several policy implications. To begin with, it is important  
299 that policymakers understand the key barriers of farmers' adoption prior to investigating their WTP for  
300 BDMs. The double hurdle modeling process provides some insights into better targeting farmers' needs  
301 at different decision-making processes. Second, the role of technology-specific characteristics is more  
302 important than adopter-specific characteristics in the adoption of BDMs in China, and possibly in other  
303 developing countries. It is not until recent years BDMs have been introduced to farmers as a substitute for  
304 PE mulching technology. Hence, to promote the adoption of BDMs, more emphasis should be placed on  
305 providing information and knowledge about technology characteristics (de Oca Munguia and Llewellyn,  
306 2020). Farmers need to have a good understanding of the technology before making decisions. Notably,  
307 training is an ideal investment for both policymakers and farmers. Besides offering information, it  
308 provides opportunities for farmers, experts, and policymakers to exchange and share knowledge,  
309 experience, and most importantly, to build trust that contributes to farmers' adoption of BDMs (Baird *et*  
310 *al.*, 2016, Yang *et al.*, 2021). Additionally, given the price of BDMs is generally higher than PE mulch  
311 films, participating in training may motivate farmers to invest more (i.e., higher WTP) in the new mulch  
312 films that are environmentally friendly and not requiring recycling efforts. Lastly, policymakers need to  
313 consider the difference in WTP amongst different crop growers, and the overall WTP (11 – 14 RMB) for  
314 BDMs is much lower than the market price (22 – 28 RMB). The government may promote the adoption  
315 of BDMs for tobacco and flower growers through tobacco firms and flower cooperatives that provide  
316 technical training and support to farmers. Similarly, when targeting other areas or countries with high  
317 usage of plastic mulch films, the local government has to relate the promotional policy to the specific  
318 crops grown by farmers.

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# Figures

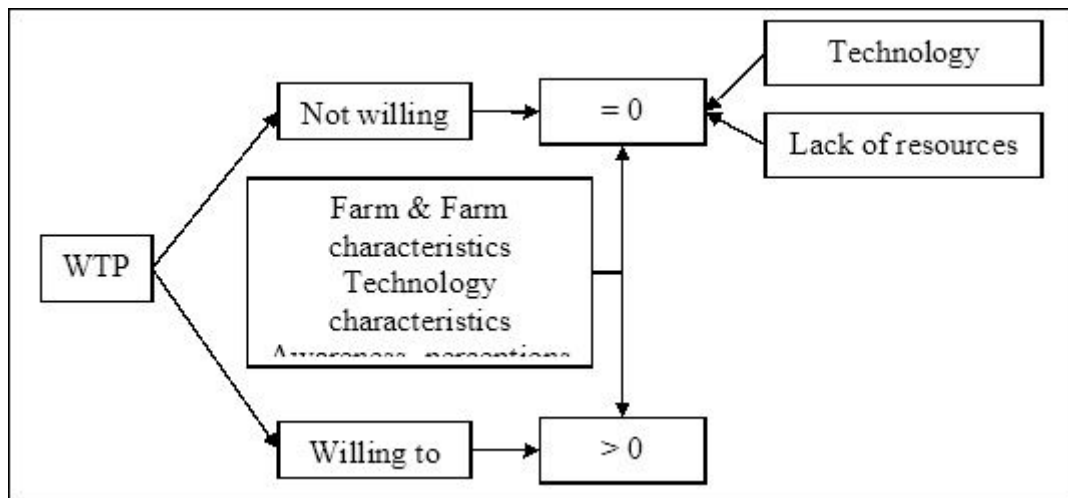


Figure 1

The conceptual analysis framework of farmers' WTP for BDMs.

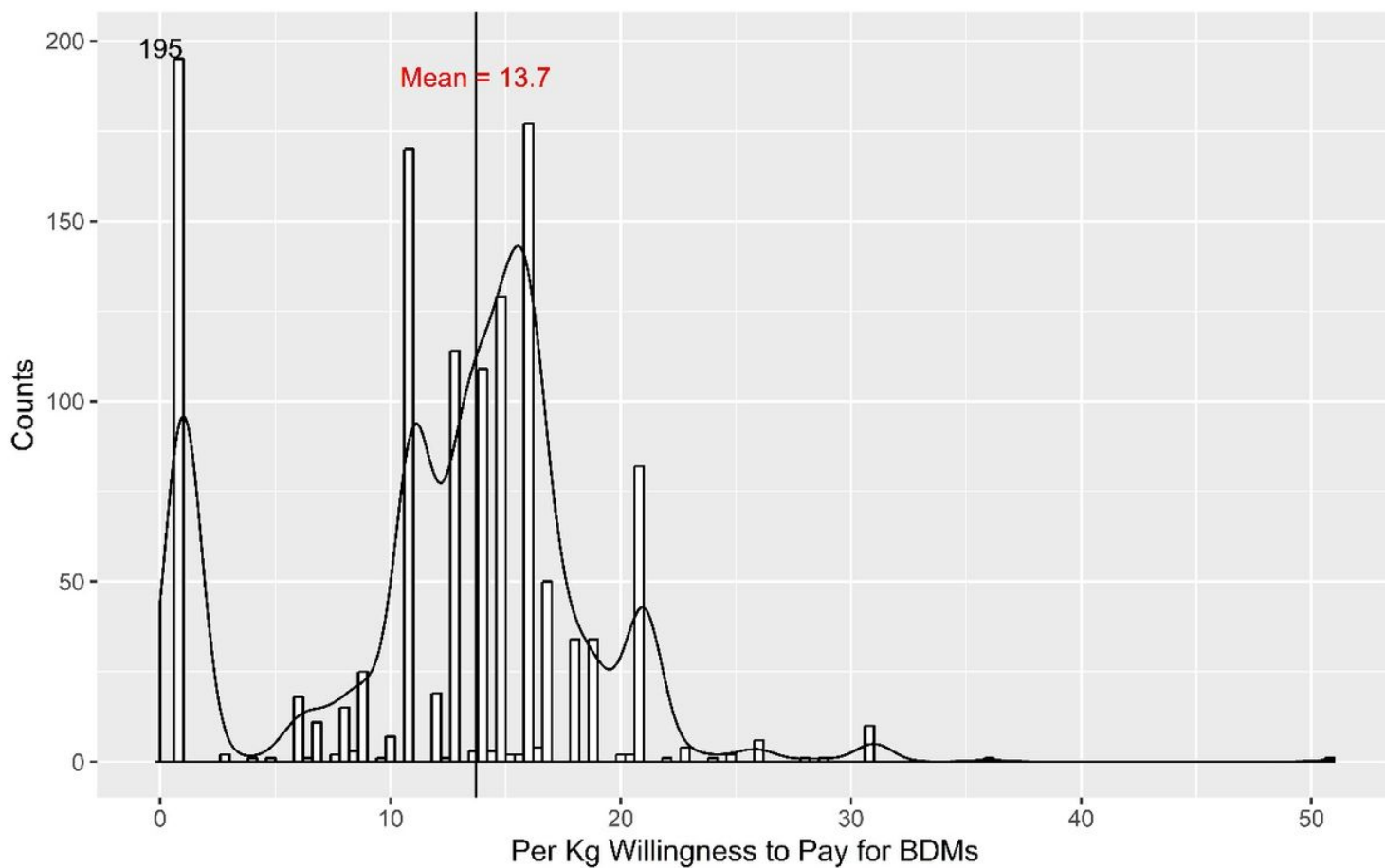


Figure 2

the distribution of farmers' WTP for BDMs.