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## Stated Farmers' Preferences and Willingness to Pay for Climate Resilient Potato Varieties in Kenya: A Discrete Choice Experiment.

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<b>Abstract:</b>	<p>Despite sustained efforts by various research organizations in developing and disseminating climate resilient varieties, adoption of climate resilient potato varieties (CRPVs) remains low in Sub-Saharan Africa. This has been majorly attributed to limited coordination between formal research institutions and farmers hence sidelining farmers' preferences especially smallholder farmers. Considering farmer preferences in the breeding process may yield optimal combination of varietal attributes hence increasing adoption. Therefore, this study used a discrete choice experiment to investigate farmers' preferences and mean Willingness to Pay (WTP) for various attributes of CRPV. Results indicate that farmers have a strong preference for high resistance to pests and diseases as compared to other attributes which include low water requirements, short maturation period and high yield. Despite farmers preferring low prices for CRPV attributes, we also note that they were low price responsive. This study emphasizes on the need for participatory breeding efforts that embed traits preferred by farmers hence satisfying the demands of different population segments based on age, gender and education level.</p>

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Thank you for reviewing

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4 **Stated Farmers' Preferences and Willingness to Pay for Climate Resilient Potato Varieties**  
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6 **in Kenya: A Discrete Choice Experiment.**

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25  
26 **Abstract**  
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28 Despite sustained efforts by various research organizations in developing and disseminating  
29 climate resilient varieties, adoption of climate resilient potato varieties (CRPVs) remains low in  
30 Sub-Saharan Africa. This has been majorly attributed to limited coordination between formal  
31 research institutions and farmers hence sidelining farmers' preferences especially smallholder  
32 farmers. Considering farmer preferences in the breeding process may yield optimal combination  
33 of varietal attributes hence increasing adoption. Therefore, this study used a discrete choice  
34 experiment to investigate farmers' preferences and mean Willingness to Pay (WTP) for various  
35 attributes of CRPV. Results indicate that farmers have a strong preference for high resistance to  
36 pests and diseases as compared to other attributes which include low water requirements, short  
37 maturation period and high yield. Despite farmers preferring low prices for CRPV attributes, we  
38 also note that they were low price responsive. This study emphasizes on the need for participatory  
39 breeding efforts that embed traits preferred by farmers hence satisfying the demands of different  
40 population segments based on age, gender and education level.  
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53 **Key Words:** Climate-resilient-potato-varieties; Preferences, Willingness-to-pay; Discrete-choice  
54 experiment.  
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## 1.0 Introduction

Potato production in Sub-Saharan Africa is highly threatened by climate change and variability. Reduced precipitation due to prolonged drought has adversely affected seasonal yields. On the other hand, global warming has exacerbated emergence of new pests and diseases, cases of late blight and bacterial wilt. A high reduction of potato yields by 56% was reported during the 2016-2017 drought in Kenya (International Potato Center, 2017).

Previous studies however show that, potato production levels can double up without expanding the area under production by developing, disseminating and adoption of climate resilient varieties (Paker *et al.*, 2019). This has prompted research organization such as KALRO and CIP into developing potato varieties that are climate smart. These varieties are said to be high-yielding and resistant to various pests and diseases. Despite various efforts by the government of Kenya to promote breeding programs for improved potato varieties and their dissemination, there is limited coordination between formal research institutions and farmers. This has resulted into sidelining farmers' preferences especially smallholder farmers leading to lower adoption rates (Sánchez *et al.*, 2017).

A study by Kimathi *et al.* (2021) shows that adoption of climate resilient potato varieties is still significantly low among potato farmers from Meru County, Kenya, at an actual population adoption rate of 6.3% against a potential adoption rate of 30% if all farmers were to be exposed to these climate resilient varieties. However, limited evidence exists regarding how climate resilient potato variety attributes influence adoption. Sibiya *et al.* (2013) argues that considering farmer preferences and priorities in the breeding process may yield optimal combination of varietal attributes hence increasing adoption rates. Cramer (2018) recommends that systems for accelerated delivery of climate resilient varieties need to be massively upgraded so as to embrace the full procedure from trait discovery to varietal deployment and seed system development. One of the identified gaps and needs for improved climate smart breeding include knowledge in terms of better understanding of the future trait preferences of different food system actors (Balié *et al.*, 2019).

In this research, we use a discrete choice experiment to investigate farmers' preferences and mean Willingness to Pay for various attributes of climate resilient potato varieties; information that is valuable in designing efficient participatory breeding programs that may result into development

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4 of farmer-preferred climate resilient potato varieties and hence boosting adoption rates. The study  
5 further explores preference heterogeneity of climate resilient potato attributes based on age,  
6 education level and gender of the farmer.  
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## 10 **2.0 Methodology**

### 11 **2.1 Study Area and Data**

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14 The study was carried out in Meru County, Kenya, which has an overall land area of 693,620  
15 hectares (Ha) and the altitude lies between 2230-2900m above the sea level. The estimated potato  
16 producing area is 17,534Ha with a production of 196,434T (CIP,2019). The county is made up of  
17 four main agro-ecological zones characterizing the area; the upper and lower highlands where  
18 potatoes are mainly grown, and the upper and lower midlands (Jaetzold *et al.*, 2007).  
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24 Data used for this study was cross-sectional obtained from household survey at farm level.  
25 Structured questionnaires were administered to farmers by well-trained enumerators. The sample  
26 was drawn from smallholder potato farmers in the different agro-ecological zones of Meru County  
27 using multistage sampling technique. In the first stage, Meru County was selected purposively  
28 because it is among the highest potato producing counties in Kenya. The second stage involved  
29 purposive selection of three out of the nine sub-counties based on potato production levels and  
30 climatic conditions which include; Imenti South, Imenti Central and Buuri Sub-counties. In the  
31 third stage, four wards (Abothuguchi West, Abogeta West, Kiirua/Naari, Kibirichia) were  
32 randomly selected from the three sub-counties and finally, a random sample of 384 farmers was  
33 selected from the wards using simple random sampling technique.  
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### 43 **2.2 Experimental Design**

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45 Revealed and stated preference methods are the two main approaches in eliciting individual  
46 preferences for priority setting in economic evaluation. Revealed preference approach involves  
47 analysis of individual preferences revealed by real market behavior. This approach is limited in  
48 that attributes are usually collinear in market data, making it difficult to predict the effect of  
49 independent variation in an attribute. On the other hand, stated preference approach involves  
50 asking consumers to state their preference for hypothetical scenarios that comprise different  
51 attributes with different levels. Stated preference approach is usually preferred as it addresses the  
52 multicollinearity problem of the revealed preference approach by allowing for sufficient variation  
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4 in the data. Some of the stated preference methods include Contingent Valuation Method (CVM)  
5 and Discrete Choice Modelling (DCM).  
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9 Among all the variants of DCM, Discrete Choice Experiment (DCE) is the most popular due to its  
10 validated economic theory. It is based on the Lancaster theory of consumption which states that  
11 consumers derive utility from characteristics or attributes of a product rather than the product as a  
12 whole (Lancaster, 1966). In this study, DCE method was used to determine the key attributes  
13 preferred by farmers for climate resilient potato varieties and willingness to pay for each attribute.  
14 First order interactions for attributes and socio-economic characteristics of farmers were analyzed  
15 to account for sources of preference heterogeneity. Farmers were presented with various  
16 hypothetical scenarios where they were supposed to choose different attributes described by their  
17 levels as presented in choice sets.  
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26 The experiment involved various stages:  
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### 28 *2.2.1 Selection of Attributes and levels*

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31 In the first stage, potential attributes and the attribute levels were identified based on the  
32 characteristics of climate resilient potato varieties and literature. A Focus Group Discussion  
33 comprising of 30 experienced farmers, decentralized seed multipliers, village-based potato  
34 advisors, agricultural extension officers and researchers was conducted to validate the attributes  
35 and attribute levels proposed for the experiment. Table 1 shows the potential attributes and their  
36 levels based on literature and validated by FGD.  
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42 **Table 1: Attributes and Attribute Levels of Climate Resilient Potato Varieties**

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45 <b>Attributes</b>	46 <b>Description</b>	47 <b>Levels</b>
48 Resistance to pests and diseases	49 Whether the variety is resistant to pests and diseases or not	50 Yes 51 No
52 Water Requirements	53 The amount of water the variety requires to grow	54 High 55 Low
56 Yield per Hectare	57 Yield in Tonnes per Hectare	58 High (40T/Ha) 59 Low (20T/Ha)

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Maturation Period	How long it takes for the potato to be ready for harvest	Short (<3 months) Long (>3 months )
Input Price (50kg bag)	Input price for every 50kg bag of potato seeds	2000, 2500, 3000

One of the negative effects of climate change is increased emergence of pests and diseases. Climate change factors such as higher humidity, increased temperatures and unseasonal rainfalls are said to increase the severity of potato diseases such late blight, bacterial wilt, potato leafroll virus and potato virus Y (Quiroz *et al.*, 2018). Therefore, resistance of climate resilient potato variety to pests and diseases was included as a potential attribute and was described by two levels; whether the variety preferred is resistant to pests and diseases or not. The amount of water required to grow CRPVs was also included as a potential attribute and was described by two levels; high or low water requirements. Reduced precipitation due to climate change has necessitated use of irrigation as an adaptation practice by farmer (Esayas *et al.*, 2019). However, most smallholder farmers lack enough resources to invest in irrigation equipment. More so, lack of access to adequate farming water limits smallholder farmers from adopting irrigation as an adaptation strategy.

The yield attribute was also included to capture the preferred yield per hectare by farmers. It was described using two levels; High yields (40T/Ha) and Low yields (20T/Ha). Mukherjee *et al.*, (2017) reported that the average potato productivity in India is 23T/Ha whereas International Potato Center in its 2017 annual report argued that the average productivity of quality seeds was 35-45T/Ha. To ensure food security through increased productivity without increasing the size of farming land, there is need to breed improved varieties that have high yield. Previous studies show that some farmers preferred varieties with short maturation period while others preferred varieties that mature late to ensure food security and sustainable income. Kolech *et al.* (2015) reported that farmers preferred potato varieties that are drought resistant, resistant to pests and diseases and mature early. Maturation period was included as a potential attribute described by two levels; short (<3months) and long (>3months) to determine the preferred maturation duration for CRPVs by smallholder potato farmers in Meru county.

The last attribute was the price attribute. This is the input price or buying price for every 50kg bag of CRPV seeds. The input price was set around the mean of the buying price for the most popular variety in the county named *Shangi*. The price for the certified *Shangi* variety is around 2500

Kenya shillings per 50Kg bag seed as reported from FGD. The input price was set to have three levels; 2000, 2500 and 3000 Kenya Shillings. The price attribute was an important attribute used in obtaining willingness to pay a premium for the other attributes (Sánchez *et al.*, 2017).

### 2.2.2 Designing choice cards

The second stage involved designing of choice cards whereby choice cards were constructed based on their attributes and attribute levels. Each choice card had two unlabeled scenarios of climate resilient potato varieties and one opt-out option showing a farmer who is not willing to uptake a climate resilient potato variety. D-efficient design was used to reduce the number of choice cards. The SAS software was used to generate the design as is the most widely applied package for experimental designs (Kjær, 2005). The design had a D-efficiency of 96.85% which was a relatively good measure of D-Optimality. A design is said to be D-Optimal if it yields data that enables estimation of parameters with low standard errors and the design can extract the maximum amount of required information from the respondents. The A-efficiency was 93.10% implying that variance matrix generated estimates that were sufficiently reliable. The G-efficiency was 100% which makes precise response predictions for choice experiments (Oyinbo *et al.*, 2019; Otieno *et al.*, 2019; Alemu *et al.*, 2017; Kessels *et al.*, 2006). The design generated 12 choice situations which were blocked into 4 profiles and each farmer was presented with 3 choice cards each representing a choice set. An example of a choice card is presented in Figure 1.

Climate resilient potatoes attributes	Option A	Option B	Neither A nor B
Drought and Disease Resistant	Not	Resistant	
Water Requirements	Low	High	
Yield	40T/Ha	20T/Ha	
Maturation Period	Long (> 3 months)	Short (<3 months)	
Input Price per 50kg bag	2500	2000	
Which option would you prefer?			

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4 **Figure 1: Choice card sample**  
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10 **2.2.3 Econometric Model**  
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13 The final stage involved econometric analysis of DCE which was based on the Random utility  
14 theory (Louviere *et al.*, 2010). This means that the total utility derived from adopting CRPVs can  
15 be decomposed into a deterministic and unobservable stochastic error components depending on  
16 the attributes preferred by farmers. Mixed logit model using 100 Halton draws was used in the  
17 estimation of farmers' preferences. Lecocq (2008), argued that 100 Halton draws were sufficient  
18 to yield unbiased estimates considering accuracy and time factors. The mixed logit model was  
19 used because it accounts for preference heterogeneity by allowing for variations across  
20 respondents for coefficients of variables that enter the model. The model also does not exhibit the  
21 restrictive IIA (Independence of Irrelevant Attributes) property as in the conditional logit model  
22 and accounts for correlation in unobserved utility over repeated choices (Van den Broeck *et al.*,  
23 2017). The utility ( $U_{ijk}$ ) derived by farmer  $i$  from choosing alternative  $j$  in the choice card  $k$  was  
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$$U_{ijk} = \alpha_i ASC + \beta_i X_{ijk} + \varepsilon_{ijk} \tag{1}$$
  
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$$\beta_i = \delta_i \beta + \gamma \eta_i + \delta_i \eta_i$$

39 Where;

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41  $X_{ijk}$ = Vector of attributes

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43  $\beta_i$ =Vector of individual specific parameters

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45  $\varepsilon_{ijk}$ =idiosyncratic error

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47  $\beta$ = Vector of mean attribute utility weight

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49  $\delta_i$ = Person-specific scale heterogeneity of the idiosyncratic error

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51  $\eta_i$ = Vector of individual specific deviation from the mean

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53  $\gamma$ = Scalar parameter governing variance of residual taste heterogeneity

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55 ASC= Alternative Specific Constant

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57 ASC accounts for when a farmer opts out and it captures attributes not included in the choice  
58 experiment hence catering for non-response bias. It is usually a dummy variable; 0=uptake of  
59 climate resilient variety and 1=otherwise.  
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The specification of the choice probability equation was;

$$P_{ijkl} = \frac{\exp(\beta_i X_{ijk})}{\sum_{j=1}^J \exp(\beta_i X_{ijk})} \quad (2)$$

To identify sources of heterogeneity, interactions between preferences and farmer characteristics were computed.

Values of Willingness to Pay (WTP) for different varietal attribute levels were derived as: -

$$\text{Marginal WTP} = - \frac{\beta_{\text{attribute}}}{\beta_{\text{price}}} \quad (3)$$

STATA Version 15 was used to estimate parameters for the mixed logit model. The variables used in analysis and how they enter the model are shown in Table 2. All the variables entered the model as random parameters assuming a normal distribution except for the price attribute that was specified as fixed and assumed to have a lognormal distribution in order to facilitate estimation of WTP. Hole *et al.* (2012) argued that restricting the sign of coefficients to be either positive or negative for all respondents maybe desirable in some cases and thus, lognormal distribution acts as an alternative to normal distribution.

**Table 2: Variables used in Econometric Analysis**

Variable	Description
Resistance	Variety is resistant to pests and diseases (1=Yes, 0=Otherwise)
Water Requirements	The amount of water the variety requires to grow (1=Yes, 0=Otherwise)
Yield	Yield of the variety in Tonnes per Hectare (1=Yes, 0=Otherwise)
Maturation Period	How long it takes for the potato to be ready for harvest (1=Yes, 0=Otherwise)
Price	Input price for every 50kg bag of potato seeds (2000, 2500, 3000)

### 3.0 Results and Discussion

#### 3.1 Farmers' preferences for improved CRPVs

Table 3 shows results for farmers' preferences for improved CRPVs. The log likelihood was -417.018 whereas chi-square was 108.83 and statistically significant at 1% level of significance. This shows that all the variables included in the econometric model were statistically sufficient and the model had a good fit. All the variables included in the model were positive and significant

at 1% except for input price which was negative and significant at 10% level of significance. This further reveals that the attributes and attribute levels considered in the model were essential in determining farmer preferences for CRVPs. The standard deviations for all parameter estimates were statistically significant at 1% level of significance with the exception of the standard deviation for High yield which was significant at 5% level of significance. This shows the presence of preference heterogeneity among farmers who participated in the experiment validating the suitability of mixed logit model for analysis of this objective.

**Table 3: Farmers' preferences for improved CPRVs**

Variable	Coefficient	Standard error	p-Value
Resistant to pests and diseases	7.755	1.933	0.000***
Low water requirement	2.341	0.661	0.000***
High yield (30T/Ha)	2.061	0.540	0.000***
Short maturation period(<3 months)	2.017	0.546	0.000***
Input price (per kg)	-0.024	0.014	0.085*
<i>Derived standard deviations of parameter distributions</i>			
Resistant to pests and diseases	5.275	3.90	0.000***
Low water requirement	2.455	0.700	0.000***
High Yield	-1.787	0.761	0.019**
Short Maturation Period(<3 months)	-1.488	0.537	0.006***
<b>Goodness of fit</b>			
Log Likelihood	-417.018		
LR Chi2 (4)	108.83***		
n (respondents)	384		
n (choices)	3456		

\*\*\*, \*\*, \* =level of significance at 1%, 5% and 10% respectively.

Farmers not only consider productivity of potato varieties when making decision on whether to adopt or not, but they also consider the adaptability of potato varieties to the changing factors of

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4 climate that have adverse effects on potato crop. Results indicate that farmers preferred CRPVs  
5 that were resistant to pests and diseases. The estimated coefficient for resistant was positive and  
6 significant at 1% significance level. The magnitude of the coefficient was high (7.755) and almost  
7 thrice the magnitude of all other attributes. This reveals that even though farmers preferred potato  
8 varieties with high yield, short maturation period and low water requirements; resistance to pests  
9 and diseases was their most preferred attribute. This can be explained by changes in climatic  
10 conditions for instance, increased temperature and humidity which increase the severity of pests  
11 and diseases such as late blight and bacterial wilt. Therefore, in attempt for farmers to adapt to  
12 climate change and increase resilience, they seek for potato varieties that are highly resistant to  
13 pests and diseases. This finding is consistent with previous studies Gamboa *et al.* (2018), Sánchez  
14 *et al.* (2017) and Kassie *et al.* (2017) who reported that farmers preferred varieties that were  
15 resistant to pests and diseases to reduce yield loss amid climate change.  
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27 The estimated coefficient for low water requirement was positive and significant at 1% level of  
28 significance. The magnitude of the coefficient (2.341) indicate that low water requirement was the  
29 second most preferred attribute after resistance to pests and diseases. Due to the negative effects  
30 of factors of climate change such as unseasonal rainfall and reduced precipitation, farmers prefer  
31 potato varieties that require less water to grow so as to ensure yield stability even in seasons of  
32 poor rainfall. More so, farmers from drier agro-ecological zones are usually limited in terms of  
33 access to water hence varieties with low water requirements were more favorable. Similar results  
34 were reported by Asrat *et al.* (2010) who highlighted that farmers preferred varieties that were  
35 tolerant to environmental stress factors such as poor rainfall.  
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44 The coefficient of high yield (2.061) was positive and significant at 1% level of significance  
45 revealing that potato farmers from Meru County preferred varieties that were high yielding  
46 (30T/Ha). This finding was as expected for a rational decision maker trying to maximize utility.  
47 High yielding varieties are preferred by farmers as they ensure food security, increased income  
48 from sales and hence reduced household poverty. Climate change leads to yield loss and reduced  
49 productivity. Kivuva *et al.* (2014) argued that farmers preferred high yielding varieties in order to  
50 counter production constraints. The attribute of short maturation period (<3 months) had a positive  
51 and significant coefficient at 1% level of significance and a magnitude of 2.017. This indicates  
52 that potato farmers preferred varieties that matured faster. Varieties that mature fast are usually  
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4 less affected by adverse effects of climate change such as poor rainfall, frost, potato blight and  
5 drought (Gamboa *et al.*, 2018).  
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9 The price attribute was negative and significant at 10% significance level. This indicates that  
10 farmers preferred lower prices for CRPVs holding all other factors constant. This finding was as  
11 expected since the sign of the price attribute was actually imposed by choosing log normal  
12 distribution. Similar results were reported by Van den Broeck *et al.*, (2017) and Pambo *et al.*,  
13 (2014). However, the absolute magnitude of the price coefficient was relatively small revealing  
14 that potato farmers in Meru county were low-price responsive. A small change in price did not  
15 affect their preferences for other CRPV attributes. This was contrary to (Wanyama *et al.*, 2019)  
16 who reported high-price responsiveness for low income consumers. The contrast shows that  
17 farmers in Meru county valued environmentally adaptable potato varieties despite the price mark  
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### 27 **3.2 Sources of Preference Heterogeneity**

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29 The middle part of Table 3 shows significant values for the standard deviations of the random  
30 estimates which reveals presence of preference heterogeneity. This means that preferences for  
31 CRPV attributes were allowed to vary across different farmers with similar observed  
32 characteristics. To show possible sources of preference heterogeneity, interactions between  
33 random parameters and farmer characteristics were estimated using mixed logit model (Pambo *et*  
34 *al.*, (2014). After iterative process of model estimation and comparison using simulated log  
35 likelihood procedure, interactions that were significant and produced a good fit for the model were  
36 reported in Table 4. These included; gender\_HY which represented the interaction between gender  
37 of the farmer and High Yield (HY) attribute, Age\_Lwr which was the code for interaction between  
38 age and low water requirement attribute, Educ\_HY which was the code for education and high  
39 yield, and finally Educ\_Res which represented the code for education and resistant attribute.  
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50 The top part of Table 4 shows the mean estimates for preferred CRPV attributes (discussed in  
51 Table 3), the middle part shows the interactions whereas the third part shows the standard  
52 deviations for the parameter estimates. The lower panel of Table 4 shows the goodness of fit for  
53 the model. Comparing results of the two tables (Table 3 and 4), inclusion of interactions improves  
54 the overall model fit since the log likelihood reduces to -425.380.  
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**Table 4: Potential Sources of Preference Heterogeneity**

Variable	Coefficient	Standard error	p-Value
Resistant to pests and diseases	5.240	0.935	0.000***
Low water requirement	2.858	0.728	0.000***
High Yield (30T/Ha)	0.054	0.643	0.933
Short Maturation Period(<3 months)	2.336	0.715	0.001***
Seed Price (per kg)	-0.024	0.013	0.057*
<b>Interactions</b>			
gender_HY	0.513	0.297	0.085*
Age_Lwr	-0.020	0.012	0.081*
Educ_Res	-0.221	0.118	0.061*
Educ_HY	0.222	0.092	0.015**
<b>Derived standard deviations of parameter distributions</b>			
Resistant to pests and diseases	2.788	0.327	0.000***
Low water requirement	1.333	0.275	0.000***
High Yield	0.248	0.174	0.155
Short Maturation Period(<3 months)	2.037	0.695	0.003***
<b>Goodness of fit</b>			
Log Likelihood	-425.380		
LR Chi2 (4)	87.40***		
n (respondents)	384		
n (choices)	3456		

\*\*\*, \*\*, \* =level of significance at 1%, 5% and 10% respectively.

The interaction between gender and high yield denoted by gender\_HY had a positive and significant coefficient at 10% level of significance. This shows that being male increased the preference for high yielding potato varieties by 51.3%. This can be explained by the fact that most male farmers are business-oriented and practice agribusiness unlike female farmers who in most cases farm potato for household food and nutrition security. Male farmers prefer potato varieties

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4 with higher yield so as to boost their income levels from increased sales. This finding is consistent  
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6 with Patel-Campillo *et al.* (2018) who argued that females majored in potato farming mainly for  
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8 the food security of the household.  
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10 The interaction between age of the farmer and low water requirement attribute denoted by  
11 Age\_Lwr had a negative and significant coefficient at 10% significance level. Age reduced the  
12 preference for potato varieties requiring low water to grow by 2%. Older farmers shifted their  
13 preference from low water requirements. This can be explained by the fact that older farmers are  
14 usually more endowed in terms of resources than younger farmers and therefore can afford to  
15 acquire irrigation equipment making water requirements a less preferred attribute. Simtowe *et al.*  
16 (2016) argued that older farmers are less constrained in terms of financing farming practices.  
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23 The interaction between education and resistant to pests and diseases denoted as Educ\_Res yielded  
24 a negative and significant coefficient at 10% level of significance. Farmers' preference for resistant  
25 varieties shifted downwards by 22.1% with increase in education level. More educated farmers did  
26 not place much importance to the resistance attribute besides the attribute being the most preferred  
27 generally. A possible explanation is that educated farmers are more knowledgeable in terms  
28 available pesticides suitable for use and are aware of their appropriate application. More so,  
29 educated farmers have alternative sources of income and therefore are able to afford such  
30 recommended pesticides hence do not place much importance on the resistance attribute. This  
31 finding is contrary to Chandio *et al.* (2018) who argued that education enhanced farmer ability to  
32 recognize risks associated with climate change such as severity of pests and diseases hence  
33 preferring varieties that are resistant. Further, results show that the interaction between education  
34 and high yield had a positive and significant coefficient at 5% level of significance revealing that  
35 preference for high yielding varieties increased by 22.2% with higher level of education.  
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48 This study revealed that gender, age and education level of the farmer were significant sources of  
49 preference heterogeneity for preferred CPRV attributes including resistant to pests and diseases,  
50 yield and water requirements. However, the derived standard deviation for the maturation period  
51 attribute was still significant at 1% implying that heterogeneity in preference for maturation period  
52 was caused by other factors other than the socio-economic factors included in the model.  
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### 3.3 Willingness to pay for Improved Climate Resilient Potato Varieties

This sub-section presents results for the estimation of Willing To Pay for CRPVs. The price attribute represented the purchasing price for CPRV seeds and was captured as price per 50kg bag which is the most popular recommended packaging method for potato seeds in Kenya. However, for purpose of favorable econometric modelling, the price variable entered the model as Kenya Shillings per Kilogram implying that the WTP values should be multiplied by 50 since the price variable was divided by 50 during estimation.

The price coefficient enabled estimation of Marginal Rate of Substitution (MRS) between improved CPRVs attributes and money interpreted as marginal willingness to pay for a change in each attribute. The effect of each attribute was not predetermined and therefore, the willingness to pay values could take any sign. Positive values show the amount farmers would be willing to pay to acquire preferred attributes whereas negative values indicate the discount farmers would demand for accepting less preferred attributes for CRPVs. Table 5 shows the estimated WTP values for each of the CPRV attributes.

**Table 5: WTP Values for CPRV Attributes**

Variables	Marginal WTP	Lower CI	Upper CI
Resistant to pests and diseases	327.740	-88.740	741.221
Low water requirements	98.954	-33.167	231.075
High Yield (30T/Ha)	87.083	-29.208	203.373
Short Maturation Period(<3 months)	85.256	-28.913	199.424

CI, Confidence Interval at 95% confidence level

The willingness to pay estimates highlight the extent to which potato farmers value CRPV attributes. A first observation is that WTP values for CRPV attributes average around the same value except for the resistant to pests and diseases attribute whose WTP value was over three times more than all other attributes. This means that farmers were willing to pay more for the resistant to pests and diseases attribute than for any other attribute despite the price mark up. This is justified as pests and diseases comprise of the most important challenge facing potato production in Kenya.

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4 (Maligalig *et al.*, 2018; Kivuva *et al.*, 2014) reported that the main factors constraining potato  
5 production include occurrence of pests and diseases and limited access to quality seeds.  
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9 Looking further into the details of Table 4.15, potato farmers in Meru County would be willing to  
10 pay an average of Ksh 327.740 per Kg for varieties resistant to pests and diseases, Ksh 98.954 per  
11 Kg for varieties that have low water requirements, Ksh 87.083 per Kg for high yielding varieties  
12 (30T/Ha) and above, and Ksh 85.256 per kg for varieties with short maturation period (<3 months).  
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14 However, it should be noted that this analysis was based on stated preference data which is subject  
15 to hypothetical bias. Thus, WTP values should be interpreted as high preferences rather than a  
16 strategy to develop feasible price mark-up for CRPVs (Gamboa *et al.*, 2018).  
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#### 22 **4.0 Conclusion and Recommendations**

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24 Results reveal that farmers have a strong preference for CRPVs with high resistance to pests and  
25 diseases as the most important potato crop trait as indicated by the high value of willingness to pay  
26 and a high coefficient value. Other CRPV preferred attributes include low water requirements,  
27 short maturation period and high yield. Farmers also prefer lower prices for CRPVs but were low-  
28 price responsive. Preference heterogeneity varies by socioeconomic characteristics. Male farmers  
29 prefer high yielding varieties, older farmers (>35 years) shifted their preference from low water  
30 requirement attribute and the more educated a farmer was, the less the preference for resistance to  
31 pests and diseases attribute. Breeding efforts should embed traits for CRPVs preferred by farmers.  
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33 Results indicate that CRPVs have a high potential for diffusion should ongoing breeding programs  
34 focus on development of potato varieties that are highly resistant to pests and diseases, have high  
35 yielding potential with low water requirements and short maturation period (<3months).  
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50 area.  
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## References

- Alemu, M. H., Olsen, S. B., Vedel, S. E., Kinyuru, J. N., & Pambo, K. O. (2017). Can insects increase food security in developing countries? An analysis of Kenyan consumer preferences and demand for cricket flour buns. *Food Security*, 9(3), 471-484. <https://doi.org/10.1007/s12571-017-0676-0>
- Asrat, S., Yesuf, M., Carlsson, F. & Wale, E. (2010). Farmers' preferences for crop variety traits: Lessons for on-farm conservation and technology adoption. *Ecological Economics*, 69(12), 2394-2401. <https://doi.org/10.1016/j.ecolecon.2010.07.006>
- Balié, J., Cramer, L., Friedmann, M., Gotor, E., Jones, C., Kozicka, M., & Thornton, P. (2019). Exploring opportunities around climate-smart breeding for future food and nutrition security. *CGIAR Research Program on Climate Change, Agriculture and Food Security (CCAFS)*.
- Cramer L. (2018). Access to Early Generation Seed: Obstacles for Delivery of Climate-Smart Varieties. In: Rosenstock T, Nowak A, Girvetz E (eds). *The Climate-Smart Agriculture Papers*. Springer, 87-98.
- Chandio, A. A., & Yuansheng, J. I. A. N. G. (2018). Determinants of adoption of improved rice varieties in northern Sindh, Pakistan. *Rice Science*, 25(2), 103-110. <https://doi.org/10.1016/j.rsci.2017.10.003>
- Esayas, B., Simane, B., Teferi, E., Ongoma, V., & Tefera, N. (2019). Climate variability and farmers' perception in Southern Ethiopia. *Advances in Meteorology*, 2019. <https://doi.org/10.1155/2019/7341465>
- Gamboa, C., Van den Broeck, G. & Maertens, M. (2018). Smallholders' Preferences for Improved Quinoa Varieties in the Peruvian Andes. *Sustainability*, 10(10), 3735. <https://doi.org/10.3390/su10103735>
- Hole, A. R., & Kolstad, J. R. (2012). Mixed logit estimation of willingness to pay distributions: a comparison of models in preference and WTP space using data from a health-related choice experiment. *Empirical Economics*, 42(2), 445-469. <https://doi.org/10.1007/s00181-011-0500-1>
- International Potato Center. (2017). Accelerated value chain development program. Root crops quarter 3 of year 2 report, *International Potato Center*, Lima, Peru.

- 1  
2  
3  
4 Jaetzold, R., Schmidt, H., Hornet, Z.B., Shisanya, C.A. (2007). Farm Management Handbook  
5 of Kenya. Natural Conditions and Farm Information. vol. 11/C. Eastern Province, 2nd  
6 ed. *Ministry of Agriculture/GTZ*, Nairobi, Kenya.
- 7  
8  
9  
10 Kassie, G. T., Abdulai, A., Greene, W. H., Shiferaw, B., Abate, T., Tarekegne, A., & Sutcliffe,  
11 C. (2017). Modeling Preference and Willingness to Pay for Drought Tolerance (DT) in  
12 Maize in Rural Zimbabwe. *World Development*. 94, 465–477.  
13 <https://doi.org/10.1016/j.worlddev.2017.02.008>  
14
- 15  
16  
17 Kessels, R., Goos, P., & Vandebroek, M. (2006). A comparison of criteria to design efficient  
18 choice experiments. *Journal of Marketing Research*, 43(3), 409-419.  
19 <https://doi.org/10.1509%2Fjmk.43.3.409>  
20
- 21  
22  
23 Kimathi, S. M., Ayuya, O. I., & Mutai, B. (2021). Adoption of climate-resilient potato varieties  
24 under partial population exposure and its determinants: Case of smallholder farmers in  
25 Meru County, Kenya. *Cogent Food & Agriculture*, 7(1), 1860185.  
26 <https://doi.org/10.1080/23311932.2020.1860185>  
27
- 28  
29  
30 Kivuva, B. M., Musembi, F. J., Githiri, S. M., Yengo, C. G., & Sibiyi, J. (2014). Assessment  
31 of production constraints and farmers' preferences for sweet potato genotypes. *Journal*  
32 *of Plant Breeding and Genetics*, 2(1), 15-29.
- 33  
34  
35 Kjær, T. (2005). A review of the discrete choice experiment-with emphasis on its application  
36 in health care. *Health Economics*.
- 37  
38  
39 Kolech, S. A., Halseth, D., De Jong, W., Perry, K., Wolfe, D., Tiruneh, F. M., & Schulz, S.  
40 (2015). Potato variety diversity, determinants and implications for potato breeding  
41 strategy in Ethiopia. *American Journal of Potato Research*, 92(5), 551-566.  
42 <https://doi.org/10.1007/s12230-015-9467-3>  
43
- 44  
45  
46 Lancaster, K. J. (1966). A new approach to consumer theory. *Journal of political*  
47 *economy*, 74(2), 132-157.
- 48  
49  
50 Lecocq, S. (2008). Variations in choice sets and identification of Mixed Logit models: Monte  
51 Carlo evidence. *Variations in choice sets and identification of mixed logit models:*  
52 *Monte Carlo evidence (2008)*.
- 53  
54  
55 Louviere, J.J., Flynn, T.N., Carson, R.T. (2010). Discrete Choice Experiments Are Not  
56 Conjoint Analysis. *Journal of Choice Modelling*, 3(3), 57-72.  
57 [https://doi.org/10.1016/S1755-5345\(13\)70014-9](https://doi.org/10.1016/S1755-5345(13)70014-9)  
58  
59  
60  
61  
62  
63  
64  
65

- 1  
2  
3  
4 Maligalig, R., Umbeger, W., Demont, M., & Peralta, A. (2018). Farmer preferences for rice  
5 varietal trait improvements in Nueva Ecija, Philippines: A latent class cluster approach.  
6 Invited paper presented at the 2018 *International Association of Agricultural*  
7 *Economist Conference*, British Columbia. <http://dx.doi.org/10.22004/ag.econ.277476>  
8  
9  
10 Mukherjee, D. (2017). Improved Agronomic Practices and Input Use Efficiency for Potato  
11 Production under Changing Climate: Improved Practices for Potato Production.  
12 In *Sustainable Potato Production and the Impact of Climate Change* (pp. 105-132).  
13 IGI Global. DOI: 10.4018/978-1-5225-1715-3.ch005  
14  
15  
16  
17  
18  
19 Otieno, D. J., & Oluoch-Kosura, W. (2019). Assessment of local stakeholders' preferences for  
20 foreign land lease design attributes in Kenya: A participatory choice-based survey  
21 approach. *Heliyon*, 5(10), e02730. <https://doi.org/10.1016/j.heliyon.2019.e02730>  
22  
23  
24  
25  
26  
27  
28  
29  
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53  
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55  
56  
57  
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62  
63  
64  
65
- Oyinbo, O., Chamberlin, J., Vanlauwe, B., Vranken, L., Kamara, Y. A., Craufurd, P., & Maertens, M. (2019). Farmers' preferences for high-input agriculture supported by site-specific extension services: Evidence from a choice experiment in Nigeria. *Agricultural systems*, 173, 12-26. <https://doi.org/10.1016/j.agsy.2019.02.003>
- Pambo, K. O., Otieno, D. J., & Okello, J. J. (2014). *Consumer awareness of food fortification in Kenya: The case of vitamin-A-fortified sugar* (No. 138-2016-2041).
- Parker, M. L., Low, J. W., Andrade, M., Schulte-Geldermann, E. & Andrade-Piedra, J. (2019). Climate Change and Seed Systems of Roots, Tubers and Bananas: The Cases of Potato in Kenya and Sweet potato in Mozambique. In *The Climate-Smart Agriculture Papers* (pp. 99-111). Springer, Cham.
- Patel-Campillo, A., & García, V. B. S. (2018). Un/associated: Accounting for gender difference and farmer heterogeneity among Peruvian Sierra potato small farmers. *Journal of rural studies*, 64, 91-102. <https://doi.org/10.1016/j.jrurstud.2018.10.005>
- Quiroz, R., Ramírez, D. A., Kroschel, J., Andrade-Piedra, J., Barreda, C., Condori, B., ... & Perez, W. (2018). Impact of climate change on the potato crop and biodiversity in its center of origin. *Open Agriculture*, 3(1), 273-283. <https://doi.org/10.1515/opag-2018-0029>
- Republic of Kenya. (2013) Meru County Integrated Development Plan 2013-2017. Nairobi, Kenya: *Government Printers*.

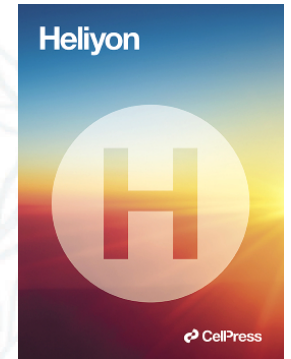
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61  
62  
63  
64  
65

Sánchez, B. I., Kallas, Z., & Gil Roig, J. M. (2017). Farmer preference for improved corn seeds in Chiapas, Mexico: A choice experiment approach. *Spanish Journal of Agricultural Research*, 15(3). <http://dx.doi.org/10.5424/sjar/2017153-11096>

Sibiya, J., Tongoona, P., Derera, J., & Makanda, I. (2013). Farmers' desired traits and selection criteria for maize varieties and their implications for maize breeding: A case study from KwaZulu-Natal Province, South Africa. *Journal of Agriculture and Rural Development in the Tropics and Subtropics (JARTS)*, 114(1), 39-49.

Simtowe, F., Asfaw, S., & Abate, T. (2016). Determinants of agricultural technology adoption under partial population awareness: the case of pigeonpea in Malawi. *Agricultural and Food Economics*, 4(1), 7. <https://doi.org/10.1186/s40100-016-0051-z>

Van den Broeck, G., Vlaeminck, P., Raymaekers, K., Velde, K. V., Vranken, L. & Maertens, M. (2017). Rice farmers' preferences for fairtrade contracting in Benin: Evidence from a discrete choice experiment. *Journal of cleaner production*, 165, 846-854. <https://doi.org/10.1016/j.jclepro.2017.07.128>



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