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5  
6 **Unraveling heterogeneity in Farmers' adoption of Mobile Phone Technologies:  
7 A systematic review.**

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14  
15 **Abstract**

16 Adoption and usage of mobile phones enable information and knowledge flows among value chain players,  
17 including farmers, and therefore contribute to improved efficiency. To ensure the successful  
18 implementation of MPTs in agriculture, farmers should embrace them. We present a systematic review of  
19 determinants, constraints, methods, indicators, and measures of MPT adoption among farmers and a  
20 methodological quality assessment of the included studies. Findings from 53 studies showed significant  
21 heterogeneity in research. Most studies targeted developing regions in Africa and Asia and only a few in  
22 developed countries. Although studies mainly focused on MPT adoption, they varied significantly in the  
23 outcome indicators assessed and their measurement. This heterogeneity in the conceptualization and  
24 measurement of adoption indicators underscores the need for standardized approaches in future studies.  
25 There is also sub-optimal use of established psychological-behavioral theories to underpin MPT adoption.  
26 Cross-sectional designs and quantitative approaches dominate the research landscape. Concerning  
27 farmers' MPT adoption, key drivers were education, age, gender, perceived ease of use, perceived  
28 usefulness, perceived cost, performance and effort expectancy, attitude, skills, and knowledge. Factors  
29 such as mobile phone cost, inadequate infrastructure, and language barrier constitute major hurdles  
30 constraining farmers' adoption of MPT. The generally low quality of the reviewed studies suggests that  
31 future studies should invest in transparently providing the study objectives, methods, and interpretation  
32 of their findings. This systematic review contributes to a better understanding of farmers' MPT adoption  
33 drivers and suggests areas for future research. It provides relevant information to policy-makers, public  
34 and private sector agencies, mobile phone companies and app developers, researchers, agricultural  
35 extension workers, academicians, and other stakeholders when designing and implementing policies for  
36 MPT adoption in agriculture.

37 **Keywords:** Adoption, Farmers, Mobile phone Technology (MPTs), Quality appraisal; Mixed-Method  
38 Appraisal tool (MMAT), Systematic review

39  
40 **1. Introduction**

41 The last two decades have seen a rapid advancement in the global use of information and communication  
42 technologies (ICT) (Aker, Ghosh, & Burrell, 2016). The Mobile Phone Technology (MPT) component of ICTs  
43 has spread incredibly quickly in both developed and developing countries, supposedly because of its cost-  
44 effectiveness, lower infrastructure demand, and relative ease of use (Kabbiri, Dora, Kumar, Elepu, &  
45 Gellynck, 2018). Existing research recognizes the critical roles played by MPTs, such as mobile (smart)  
46 phones and related applications (apps), in enhancing agricultural productivity among farmers (Ezeoha,  
47 Obi, Igwe, & Ezeruigbo, 2020; Schulz, Prior, Kahn, & Hinch, 2021).

48 Among these roles, MPTs facilitate fast communication via data, voice, and SMS services among farmers  
49 – connecting them to markets. Likewise, there is evidence that MPTs are crucial for addressing the issue  
50 of information asymmetry among farmers and other actors within the agricultural value chain and  
51 contribute to lowering farmers’ transaction costs (Aker et al., 2016; Nakasone, Torero, & Minten, 2014).  
52 By increasing farmers’ timely access to information on agricultural best practices, weather and disease  
53 forecasts, or market prices, mobile phones might boost agricultural yields and farmers’ income (Okello,  
54 Kirui, Njiraini, & Gitonga, 2012). Existing research indicates that MPT use in agriculture could enhance  
55 social relationships among farmers (Baardewijk, 2017; Butt, 2015) and address barriers to traditional  
56 extension methods (Aldosari et al., 2017b). However, the adoption of such technologies among farmers  
57 remains low (Bonke, Fecke, Michels, & Musshoff, 2018).

58 [Farmers’ adoption of technology is complex and often involves a cognitive process that leads to a](#)  
59 [motivated decision on whether to adopt it or not](#) (Shikuku, Okello, Sindi, Low, & McEwan, 2017). Extant  
60 literature highlights the intricate interaction of factors influencing farmers’ decision-making (Shikuku et  
61 al., 2017). [For example, Misaki et al., 2018 found that lack of trust and transparency, limited farmers’](#)  
62 [involvement in the early stages of invention, poor mobile infrastructure, and designing MPTs in foreign](#)  
63 [languages hinder mobile phone use by farmers in developing countries.](#) Other factors influencing mobile  
64 technology adoption are age, sex, education, experience, and household size (Benjamin Yao Folitse,  
65 Manteaw, Dzandu, Obeng-Koranteng, & Bekoe, 2018). Despite the existence of a few theoretical and  
66 empirical research on farmers’ adoption and use of MPTs, scientific understanding of the factors  
67 influencing the decision to own a mobile phone and utilize it for agricultural purposes remains limited  
68 (Asravor, Boakye, & Essuman, 2021). What is even less clear is the nature of the effects of these  
69 determinants on MPT adoption in farming (H. G. Hoang & Drysdale, 2021). Much uncertainty still exists  
70 about the relationship between various determinants and farmers’ adoption of MPTs. For example, (S. P.  
71 Thar, Ramilan, Farquharson, Pang, & Chen, 2021) reported a negative effect of age on the use of MPTs by  
72 farmers. However, Benjamin Yao Folitse et al. (2018) found a positive effect, and other studies found no  
73 effect of age on MPT adoption (Krell et al., 2020; Sikundla, Mushunje, & Akinyemi, 2018). Similarly, the  
74 effect of education on MPT adoption is mixed. Education positively influenced MPT adoption in some  
75 studies (e.g., Mwalukasa, Mlozi, & Sanga, 2018; Nikam, Kumar, & Kingsly, 2021) and had no significant  
76 influence in other studies (Filippini, Marescotti, Demartini, & Gaviglio, 2020; Narine, Harder, & Roberts,  
77 2019).

78 Besides the inconclusiveness of studies on the effect of various factors on technology adoption among  
79 farmers, they are also inconsistent with the indicators and measurement of adoption – thus presenting a  
80 methodological challenge. [Owing to this complexity, the factors that drive farmers’ adoption of MPTs must](#)  
81 [be understood. Such relevant insights would facilitate decision-making and implementation by policy](#)  
82 [makers, boost the uptake of MPTs, and subsequently generate better agricultural outcomes](#) (Mwalupaso,  
83 Wang, Xu, & Tian, 2019).

84 Published reviews on ICTs, especially MPTs in agriculture, present several limitations. First, except for a  
85 few of these targeted reviews (e.g., Mendes et al., 2020; Misaki, Apiola, Gaiani, & Tedre, 2018;  
86 Pongnumkul, Chaovalit, & Surasvadi, 2015), most studies on this topic are non-systematic. Second, their  
87 focus is narrow and often restricted to one specific sub-sector of agriculture or mobile phone application,  
88 e.g., precision agriculture (Mendes et al., 2020); smartphone applications for the sustainability of  
89 agricultural landscapes (Inwood & Dale, 2019); specific mobile phone-enabled services among smallholder  
90 farmers (Baumüller, 2017) and pastoralists (Parlasca, 2021), smartphone-based sensors (Pongnumkul et  
91 al., 2015); or consider the general state of ICTs (Aker, 2011; Nakasone et al., 2014). [Third, previous](#)  
92 [literature reviews on farmers’ adoption of mobile phone technology services and applications have](#)  
93 [primarily targeted developing countries](#) (e.g., Baumüller, 2017; Duncombe, 2015; Keerthi & Gautam, 2021;  
94 Misaki et al., 2018).

95 Thus, without disregarding the contributions of past studies on MPT, this review adds to the current body  
96 of knowledge by (1) focusing on MPT adoption among farmers and (2) evaluating the methodological  
97 quality of studies on farmer MPT adoption. The specific objectives of the study were to (a) analyze and  
98 synthesize the key determinants and constraints of MPT adoption among farmers; (b) evaluate the various  
99 measures, methods, and theories employed by studies to understand MPT adoption; (c) assess farmers'  
100 perception regarding different MPTs and their associated usage, (d) appraise study quality and (e) identify  
101 and suggest areas for future research and recommendations for scaling up MPT adoption among farmers.  
102 Our systematic review differs from past reviews on mobile phone technologies in four ways. Firstly, it  
103 provides the first study that uses a methodologically sound review procedure to systematically and  
104 comprehensively examine the adoption of MPT by farmers. Secondly, this is the first review to assess the  
105 methodological quality of farmers' MPT adoption studies based on the Mixed-Method appraisal tool.  
106 Thirdly, our review provides a deeper understanding of the various measures, methods, statistical  
107 techniques, and behavioral theories used to evaluate MPT adoption and its' motivations. Fourthly, the  
108 review considers the full range of smartphone applications, MPTs, and all types of farmers and includes  
109 studies from developing and developed country contexts, which provides a broader perspective of the  
110 determinants of MPT adoption. The result of this study will advance understanding of the critical factors  
111 influencing farmers' adoption and use of MPTs, which is vital for the conceptualization and measurement  
112 of prospective studies. Moreover, the findings are relevant for effectuating novel MPTs among farmers  
113 and the agricultural food chain.

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## 116 2. Methodology

### 117 2.1 Literature search strategy

118 A systematic literature review of published evidence on farmers' adoption of MPT was undertaken using  
119 Moher, Liberati, Tetzlaff, and Altman (2009)'s PRISMA approach. Before launching the systematic review  
120 process, a review protocol was developed, registered, and published on Open Science Framework (DOI:  
121 [10.17605/OSF.IO/AGR6Y](https://doi.org/10.17605/OSF.IO/AGR6Y), available at <https://osf.io/agr6>)

122 In line with the general search strategy for systematic reviews, we applied a search query with keywords  
123 developed from literature (Aker, 2011; El Ayadi et al., 2020) and expert consultations in the Web of Science  
124 electronic database on 10th September 2021. To complete our dataset and identify studies that our  
125 electronic search strategy in the Web of Science database could have missed, we conducted a citation  
126 search of high-quality articles based on the MMAT criteria in google scholar. As an additional step, we  
127 examined the reference list of the included studies to identify more studies.

128 The search syntax used a combination of terms referring to "Farmers" or similar, "Adoption" or equivalent,  
129 and "Mobile phone Technology" or similar. Thus, the primary studies were identified using the following  
130 specific syntax: Farm\* OR Smallholder\* OR Small-holder\* OR "Small holder\*" OR "Primary producer\*" OR  
131 "Small producer\*" OR Agriculture OR Pastoral\* OR "Farming household\*" OR "Farm household\*" OR  
132 "landholder\*" AND "mobile Phone\*" OR "Smartphone\*" OR "Smart phone\*" OR "Smart-phone\*" OR  
133 "cellular" OR "cellphone\*" OR "cell phone\*" OR "cell-phone\*" OR "mobile device\*" OR "smartphone-  
134 based\*" OR "Smartphone-enabled\*" OR "mobile phone-based\*" OR "Mobile Phone-enabled\*" OR  
135 "Mobile cellular\*" OR "Mobile technol\*" OR android\* OR "ios" OR "Short Messaging Service\*" OR SMS  
136 OR ICT\* OR "mobile app\*" OR "mobile application\*" OR "Smart phone app\*" OR "Smartphone app\*" OR  
137 "smart-phone app\*" OR "Smart phone application\*" OR "Smartphone application\*" OR "smart-phone  
138 application\*" AND Adopt\* OR use OR accepta\* OR behavior\* OR behaviour\* OR choice OR choos\* OR  
139 attitude\* OR decision\* OR "decision mak\*" OR WTP OR willingness\* Or willingness-to-pay OR willingness-  
140 to-accept OR WTA OR willingness-to-adopt OR willingness-to-try OR prefer\*

141

### 142 2.2 Eligibility criteria and screening

143 The Joanna Briggs Institute’s population, concept, and context (PCC) framework (see Table 1) was utilized  
144 to define the primary concepts of this study (Madlabana, Mashamba-Thompson, & Petersen, 2020). To be  
145 included in the review, studies had to meet all the inclusion criteria (see Supplementary Table 3), that is,  
146 studies published in peer-reviewed journals with full-text versions available, based on primary data,  
147 written in English, and focused on farmers as the target population, assessed adoption of MPT, and  
148 published from the year 2000 onwards when mobile phones started increasing rapidly across the globe,  
149 including in developing regions (Baumüller, 2017). We did not restrict studies based on the study design.  
150 Figure 1 shows the detailed PRISMA flow chart for the systematic review.

151 [Figure 1]

152 Our search in the web of science database yielded 4191 articles that we imported to Endnote Desktop  
153 (version X9). Once in Endnote, we removed nine duplicates and subjected the remaining articles to a  
154 rigorous selection process. This selection process involved the screening of (1) 4180 article titles for the  
155 presence of relevant keywords, (2) 90 abstracts to scrutinize further the relevance of studies, and (3) in-  
156 depth reading of the 65 full-text articles. At this stage, 30 articles met the inclusion criteria. In total, we  
157 identified 52 studies through the citation search in google scholar and reference list search, of which we  
158 eventually included 21 studies in the review after the screening process. We updated our electronic  
159 database search on 5<sup>th</sup> June 2022 and included two more articles. Thus, we extracted data from 53 studies  
160 (see supplementary Table 1).

161  
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163

### 164 **2.3 Data extraction and analysis**

165 Before data extraction, a standardized sheet was created in Microsoft Excel and utilized during the full-  
166 text review to facilitate coding details. This data extraction sheet captured different characteristics of the  
167 included studies (e.g., detailed reference, study settings), the mobile phone technology studied, sample  
168 characteristics, study methodology, adoption indicators and measurement, the determinants and  
169 constraints of adoption, and MMAT quality score of the included studies (see supplementary Table 4 for  
170 detailed information on extracted data).

171 The variation in methods, models, and adoption indicators employed to assess MPT adoption across the  
172 included studies made a statistical meta-analysis impossible. Thus, the systematic review adopted a  
173 summative approach (Grant & Booth, 2009) to compile and analyze the extracted data. Following Kamrath,  
174 Wesana, Broring, and De Steur (2019), a qualitative content analysis, often used in analyzing text data, was  
175 applied to categorize the variables used by the included studies. The percentages relating to the variables  
176 in Figure 2 are calculated based on the total number of included studies.

177 [Figure 2]

178  
179

### 180 **2.4 Appraising the quality of included studies**

181 The methodological quality of the included studies was assessed using the well-established Mixed Method  
182 appraisal tool (MMAT), version 2018 (Hong et al., 2018). This approach facilitates the simultaneous  
183 appraisal of all research study designs, including mixed, quantitative, and qualitative methods (Crowe &  
184 Sheppard, 2011; Parikh, Aparo, Nordhagen, & De Steur, 2022). Additionally, the MMAT criteria enable the  
185 evaluation of study objectives, research questions, methodology, reporting, and discussion of results for  
186 appropriateness and clarity (Simera, Moher, Hoey, Schulz, & Altman, 2010).

187 Before assessing the methodological quality, the included studies were subjected to two screening  
188 questions: (1) Are there clear research questions? and (2) Do the collected data allow to address the  
189 research questions? Afterward, the studies were rated for the appropriate study design category as  
190 denoted – sections 1 for qualitative studies, 4 for quantitative descriptive studies, and 5 for mixed methods  
191 studies. A score of 1 was allotted to each question when it met the quality criteria, a score of 2 when it did

191 not meet the requirements, or a score of 3 when the reviewer could not tell (see footnote of  
192 [supplementary Table 2](#) for the detailed MMAT checklist). Similar to Parikh et al. (2022), MMAT scores  
193 representing the number of criteria met (rated ‘yes’) were divided by five and translated into percentages.  
194 Studies scoring at least 4 (80%) are appraised as “high quality,” scores between 3-4 (60-80%) are appraised  
195 as “moderate quality,” and studies that scored below 3(60%) are considered to be of “low quality.” Based  
196 on the recommendation of (Hong et al., 2018), this study neither calculated an overall quality score from  
197 the ratings of each criterion nor excluded studies with low methodology quality. Thus, unlike Pace et al.  
198 (2012), our quality appraisal aimed to illuminate the caliber of research on MPT adoption among farmers  
199 and inform the citation search stage. [Supplementary Table 2](#) shows the detailed quality appraisal for the  
200 included studies.

201

### 202 **3 Results**

#### 203 **3.1 Study characteristics and methods**

204 A total of 4243 studies were identified, of which 53 articles were included in the systematic review (Figure  
205 1; [Supplementary Table 1](#)). The key characteristics are discussed below (see [Table 5](#)).

206 **Publication year.** The included articles were published from 2011 to 2022, with a clear trend of an  
207 increasing number of publications each year. 2021 had the highest number of published articles (n=11  
208 equivalent to 22.6%)

209 **Geographic distribution.** More than half (53%) of the studies focused on Africa, followed by 28% in Asia,  
210 13% in Europe, 4% in the Americas, and just 1.8% in Oceania (see Figure 2). Nigeria (n=7), Ghana, Ethiopia,  
211 and Tanzania, each with four studies, dominated farmer MPT studies on the African continent, while India  
212 (n=3), Bangladesh, Pakistan, and Vietnam, each with two studies, dominated studies on the Asian  
213 continent. Studies in Europe were dominated by Germany (n=4).

214 **Study settings.** While 43% of the studies did not specify their setting, 47% targeted farmers in rural areas.  
215 Only 6% and 4% of the investigations targeted farmers in urban and peri-urban-rural regions, respectively.

216 **Farming activity and production scale of farmers.** Most studies (51%) included in this review neither  
217 specified the farming activity undertaken by the farmers nor the scale of production (64%). Nonetheless,  
218 the remaining 49% of studies targeted farmers engaged in crop production (21%), livestock production  
219 (17%), and mixed farming (11%). Regarding production scale, 30% of the targeted farmers were small-  
220 scale farmers (SSF), 2% were medium-scale farmers, and 4% of the studies targeted a mixture of small,  
221 medium, and large-scale farmers.

222 **Research designs and methods.** Three-quarters (75.4%) of the studies in this review utilized quantitative  
223 descriptive approaches to describe or analyze the factors associated with farmers’ MPT adoption, while  
224 19% applied mixed-method techniques, and 5.6 % were qualitative studies. [All \(100%\) of the studies](#)  
225 [employed cross-sectional research designs. Although both developing and developed countries focused](#)  
226 [on mixed-method and quantitative descriptive approaches, qualitative methods were only applied by](#)  
227 [studies targeting developing countries.](#)

228 The quantitative studies collected data on farmers’ socio-demographic characteristics, perceptions, and  
229 attitudes about using mobile phone applications in farming, drivers, and constraints to MPT adoption.  
230 Most quantitative studies used face-to-face interviews aided by semi-structured and structured open-  
231 ended questionnaires to collect data. What was also fascinating was the increased use of mobile  
232 applications or software to collect research data among some quantitative descriptive studies. Examples  
233 of such software include CommCare (S. P. Thar et al., 2021), GeoODK mobile phone application (Molina-  
234 Maturano et al., 2021), and Qualtrics (Victor, Nic, & Xiaomeng, 2021). However, none of the studies  
235 reported on the effectiveness of such methods.

236 Sampling methods, including multi-stage procedures, random selection, convenience sampling, purposive  
237 sampling, systematic sampling, stratified systematic sampling, non-probability quota sampling, or a  
238 combination of one or more of the stated sampling methods, were used to identify study respondents.

239 Most (77%) of the quantitative descriptive studies analyzed their data using descriptive statistics such as  
240 means, percentages, and frequencies. Others (28%) used inferential statistics and different types of  
241 regression analyses (42%), including logistic regression, Tobit regression, linear regression, probit model,  
242 and other methods such as cluster analysis and structural equation modeling. It is worth noting that some  
243 studies did not specify their sampling procedures and the statistical techniques employed to analyze data.  
244 The qualitative studies employed similar sampling methods and obtained identical information as the  
245 quantitative ones through Focused group discussions (FGD), structured interview schedules, and Key  
246 Informant Interviews. These qualitative data were analyzed using ethnography, phenomenology, and  
247 thematic analysis applying inductive approaches to obtain qualitative data.

248 The mixed-method studies used a combination of quantitative and qualitative designs.

249 **Types of MPTs.** Most studies (36.9%) assessed the adoption of smartphones and mobile phones, while  
250 26.4% of studies considered the adoption and use of smartphone applications supporting farmers'  
251 decision-making (DSAs) (see Table 2). Such DSAs included crop protection apps, mobile-based agricultural  
252 extension, Agri-info App, market information systems, herd/ livestock management apps, camera function  
253 apps, disease and weather forecast apps, and Agri (mobile-based advisory) apps. 24.5% of studies  
254 examined text-messaging (SMS) applications, 13.2% investigated social media applications such as  
255 Facebook, Twitter, IMO, and WhatsApp, and another 13.2% assessed applications that provide news and  
256 information. Other app categories (15%), such as call/video, geotag, finance, utility, and productivity, were  
257 reported at most twice. [Developing country studies focused primarily on the generic use of mobile phones  
258 and targeted more basic phones and features such as text messaging and voice calling. On the other hand,  
259 developed country studies assessed smartphones and mobile applications with enhanced capabilities and  
260 computational power.](#)

261  
262 **Application of behavioral theories.** Less than half (43%) of the reviewed studies utilized  
263 behavioral/psychological theories to unravel MPT adoption among farmers. Of these, the Technology  
264 Acceptance Model (TAM) was the most commonly applied (18.8%), followed by the Unified Theory of  
265 Acceptance and Use of Technology (UTAUT) (13.2 %) in the original or extended version. 22.6% of studies  
266 applied other theories, as in Table 2. [Both developed and developing country studies used similar theories  
267 to assess the adoption of MPTs among farmers.](#)

268 [Table 2]

### 269 ***The methodological quality of studies on farmer MPT adoption***

270 Nearly half (41.5%) of the studies in this systematic review were rated as having “medium quality,” 24.5%  
271 were ranked as “low quality, and only 34% were considered to be of “high quality.” Generally, the low  
272 quality of studies was attributable to various methodological constraints ranging from (1) poorly defined  
273 research problems, objectives, and research questions, (2) missing information on sampling strategy,  
274 sample size, representativeness, response rates, and potential sources of bias, (3) missing information on  
275 measurement of constructs and critical indicators, data analysis, (4) inadequate substantiation of research  
276 findings using data to (5) the use of unstandardized or un-prettested measurement tools to (see Table 5  
277 and supplementary table2).

278 [Table 5]

279

### 280 ***3.2 Farmers' perception of different MPTs and their associated usage***

281 Farmers' perceptions regarding various MPTs were examined by 8% of the studies. Farmers rated phone  
282 features or apps related to herd management and crop protection as extremely useful. Regarding herd  
283 management, [Michels, Bonke, and Musshoff \(2019\)](#) reported that farmers found features facilitating  
284 reproduction management, feeding and milking control, data gathering, and animal health as the most  
285 useful. For crop protection, farmers considered apps that provide weather information, enable pest  
286 scouting and infestation forecast, facilitate making a product choice and documentation, and offer

287 recommendations to fertilizer quantities to be highly relevant (Michels, Bonke, & Musshoff, 2020).  
288 Similarly, (S. P. Thar et al., 2021) found that 85% of the farmers in their study said the agricultural apps  
289 they had used were useful. Regarding their information preferences, farmers attached high relative  
290 importance to information on crop production technologies, diseases and pests, weather forecasts,  
291 market prices, agricultural inputs, and disaster early warnings such as floods (Michels, Bonke, et al., 2020;  
292 S. P. Thar et al., 2021). Furthermore, farmers valued extension advice and information about their social  
293 relations (Rahman & Fadol, 2013; Siwel Yohakim Nyamba & Mlozi, 2012).  
294 Related to MPT usage, farmers reported utilizing their MPTs for both work-related and non-work-related  
295 purposes. However, the review revealed differences in the types of MPTs and the purposes for which  
296 farmers used MPTs between developed and developing countries. While farmers in developed countries  
297 often utilized more complex apps that required the internet to search for information, market their  
298 products, and keep records, farmers in developing regions often used basic phone features like calling and  
299 SMS apps. Furthermore, the farmers in developing countries primarily utilize their mobile phones to  
300 maintain social relations among families and friends and seldom for acquiring agricultural information.

301

### 302 **3.3 Types of variables**

303 All extracted variables were grouped as either descriptive, latent, or dependent variables based on  
304 Kamrath et al. (2019). Within the context of this systematic review, the dependent variable is farmers'  
305 MPT adoption. It was operationalized using several adoption indicators, for instance, behavioral intention,  
306 attitude, and perception. Explaining farmers' MPT adoption variation is of great research interest. Often  
307 such variation is explained by independent (latent) variables, otherwise conceptualized as predictors or  
308 antecedents (Hair, Black, Babin, & Anderson, 2019; Leary, 2012). The first category of farmers' MPT  
309 adoption predictors featured in the reviewed studies includes latent variables. For example, perceived  
310 ease of use, trust, social influence, and perceived behavioral control were described using the constructs  
311 of various well-established theories, including UTAUT and TAM. The second factors influencing the  
312 adoption of MPTs among farmers, known as descriptive variables, relate to farmers' and farm  
313 characteristics and include gender, marital status, age, and income (Kamrath et al., 2019). Some variables  
314 were assigned to more than one group based on how they were examined in the studies. For example,  
315 constraints, knowledge, and mobile phone ownership were considered dependent and descriptive, while  
316 attitude and intention qualified as dependent and latent variables. Thus, these variables' percentages vary  
317 based on the total number of studies reporting them in each category.

318 Figure 2 summarizes the main categories and percentages of variables extracted.

319 [Figure 2]

### 320 **3.4 Variation in the measurement of MPT adoption**

321 The included studies evaluated similar MPT adoption indicators, albeit with different wordings, for  
322 instance, rate of use or extent of use. In total, twelve indicators of adoption were reported across the  
323 various studies. These indicators were farmers' generic use of MPTs (30%), Behavioral intention (21%),  
324 mobile phone ownership (32%), attitude towards MPT usage (11%), adoption (9%), frequency of use (9%),  
325 constraints to adoption (9%), use intensity (8%), perception (8%), willingness to adopt (9%), willingness to  
326 pay (2%), and knowledge (2%). Only one study assessed both the use and nonuse of MPTs among farmers  
327 (Krell et al., 2020).

328 Although similar variables were assessed, studies showed significant heterogeneity in measuring  
329 indicators. Of the eleven studies that assessed behavioral intention, six applied five-point Likert scales, two  
330 applied seven-point Likert scales, two measured intention as a binary variable eliciting yes or no responses,  
331 and one did not specify how adoption intention was measured. Adoption was measured using a five-point  
332 Likert scale (n=1) and as a dummy variable (n=4). Similarly, constraints were assessed as a dummy variable  
333 (n=2) and using four-point Likert scales (n=1). Attitude was measured using five-point (n=5) and four-point  
334 (n=1) Likert scales. Use of MPT was measured as a discrete choice variable (n=8), the number of phone

335 transactions made (n=1), using Likert scales (n=5), and the number of sim cards owned (n=1). Use intensity  
336 was measured using three-point (n=1) and five-point (n=2) Likert scales and as the number of phone  
337 activities carried out (n=2). Frequency of use was assessed using five (n=1) and six-point Likert scales and  
338 as a discrete choice variable (n=2).

339 Further heterogeneity was observed in the wording and number of items per construct among the studies  
340 utilizing Likert scales. For example, Landmann, Lagerkvist, and Otter (2021) and Mwalupaso et al. (2019)  
341 measured behavioral intention to use MPT using two items; Beza et al. (2018) and Molina-Maturano et al.  
342 (2021) used three items while Omar, Yap, Ho, and Keling (2021) and Victor et al. (2021) used five items  
343 (see Table 3).

344 [Table 3]

345

### 346 **3.5 Determinants of MPT adoption among farmers**

347 Based on past reviews (Olum, Gellynck, Juvinal, Ongeng, & De Steur, 2019; Oluwamayokun, Anoma, &  
348 Ammar, 2022), five categories of factors driving farmers' MPT adoption emerged from the reviewed  
349 studies, as described in the following paragraphs. Table 4 summarizes the significant determinants of  
350 farmers' MPT adoption.

351 **Socio-demographic factors** refer to farm, farmer, or household characteristics (Olum et al., 2019). Among  
352 these factors, education was the most examined (66%), followed by age (62%), gender (47%), income  
353 (n=21), farm size (n=19), marital status (n=12), farming experience (n=10), and household size (n=8). While  
354 education was found to influence MPT adoption positively in 94% of the 35 studies that examined the  
355 effect of education, the remaining 6% reported its insignificant influence on farmer MPT. The reported  
356 effect of the rest of the sociodemographic factors on farmer MPT adoption remained mixed and  
357 inconclusive (see Table 4).

358 [Table 4]

#### 359 **Technological and mobile phone-use-related factors.**

360 Perceived advantage was examined by 23% of studies, followed by perceived usefulness (17%), perceived  
361 ease of use (13.2%), performance expectancy (8%), effort expectancy (6%), and perceived cost (6%). Of  
362 these technological factors, performance expectancy, effort expectancy, and perceived advantage  
363 positively influenced MPT adoption among farmers. However, the findings regarding the influence of  
364 perceived usefulness, perceived cost, and perceived ease of use varied among studies, with both positive  
365 and negative effects being reported.

366 Among the mobile-use-related factors assessed, mobile phone ownership and duration of ownership  
367 (32%), experience (9%), the number of mobile phones owned (9%), and farmer's skills and aptitude (n=8%)  
368 were all found to have a positive effect on the adoption and use of MPT among farmers.

#### 369 **Psychological and behavioral factors.**

370 Several factors dealing with farmers' psychological state and subjective evaluation of MPTs featured in the  
371 reviewed studies. They included subjective norm (23%), personal innovativeness (21%), attitude (17%),  
372 intention (9%), trust (6%), perceived reliability of mobile phone services (6%), emotions (4%), perception  
373 (4%), information seeking behavior (2%), habit (2%), self-efficacy (2%) and behavioral control (2%). These  
374 factors were reported to positively and significantly influence farmers' MPT adoption.

375 **Biophysical factors** such as operational, on-farm natural, and physical characteristics influence farmers'  
376 adoption decisions (Olum et al., 2019). The influence of diversification (8%), land ownership/tenancy (6%),  
377 farm type (8%), farmers' total landholding (4%), and area under cultivation (4%) were assessed. While  
378 diversification and land ownership were found to have insignificant influences on farmers' MPT adoption,  
379 the total landholding and farm type were reported to have negative and positive impacts, respectively.  
380 The area under cultivation yielded mixed effects on MPT adoption.

#### 381 **Constraints and institutional factors**



382 Nearly all studies (85%) reported on the constraints and institutional factors determining farmers'  
383 adoption and usage of MPTs. Of these, knowledge of Apps (6%), information awareness (4%), participation  
384 in credit (4%) and training programs (4%), type, quality, and complexity of agricultural information (6%),  
385 assets (4%), farmers' network capacity, and facilitating conditions influenced farmers' MPT adoption  
386 positively. In contrast, distance from the market center or nearest town (11%), poor quality mobile phones  
387 (battery), poor internet and mobile network connectivity (9%), high cost of internet services (6%), lack of  
388 digital knowledge and skills (40%), lack of interest (2%), and language barrier (28%) were found to have  
389 negative effects on farmers' MPT adoption. The influence of group membership (15%) and access to  
390 electricity (11%) remained mixed and inconclusive. *In addition to the above-mentioned factors, some*  
391 *farmers reported health-related issues such as hearing problems contributing to the failure to use mobile*  
392 *technologies.*

393 *Understanding that these constraints vary between developed and developing countries is essential. For*  
394 *example, network failure, language barrier, poverty, high prices of MPTs, limited digital aptitude and skills,*  
395 *lack of electricity, and limited access to mobile phone accessories were critical for MPT adoption in*  
396 *developing countries. However, the high cost of internet services and computer literacy were more crucial*  
397 *for developed countries, while the rest of the factors overlapped between developed and developing*  
398 *country contexts.*

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#### 404 **4 Discussion**

405 The 53 articles included in this systematic review comprise an active and growing body of research on  
406 farmers' adoption of MPTs, highlighting the main determinants, constraints methods and measures of  
407 adoption, *farmers' perception of the different MPTs*, and the methodological quality of studies in this field.  
408 These are discussed in the following sub-sections.

##### 409 **4.1 Distribution of studies**

410 This systematic review shows a significant focus of farmers' MPT adoption studies in countries on the  
411 African and Asian continents, mainly Nigeria, Ghana, Tanzania, Ethiopia, and India. This result reflects  
412 those of Olum et al. (2019), who, in their systematic review, also found that most studies on farmers'  
413 adoption of agricultural innovations targeted developing countries. This focus on developing countries  
414 may be due to the United Nation's 2030 agenda for Sustainable development call for actions to ameliorate  
415 the technological capabilities of all industrial sectors, including agriculture, especially in developing  
416 countries characterized mainly by inequality and low adoption of innovation (Stafford-Smith et al., 2017).  
417 As (Pangaribowo & Gerber, 2016), new platform technologies, including MPTs, provide avenues for  
418 combating various risks and uncertainties often faced by farmers before, during, and after agricultural  
419 production, consequently boosting their production. Such avenues include providing market and price  
420 information, knowledge sharing, crop insurance, improved market access, and decision-making tools,  
421 including what and when to grow. Thus, redirecting technological innovations from developed to  
422 developing countries could promote food security and nutrition by counteracting the complex and  
423 evolving challenges of the global food system (Juma & Yee-cheong, 2005), attesting to sustainable  
424 development goal (SDG) 2 on zero hunger.

425 The reviewed body of literature points to a growing interest among researchers and development  
426 practitioners in this particular topic, as evidenced by the yearly increase in publications on MPT adoption  
427 among farmers. Moreover, it affirms the progressive recognition of the role of MPT in enhancing  
428 agricultural productivity and contributing to sustainable development (Mugambiwa & Tirivangasi, 2017).

429

430 **4.2 Heterogeneity in the measurement of farmers' MPT adoption indicators.**

431 As noted by Clark and Watson (2019), the quality of real-world decisions made based on psychological  
432 measurements depends mainly on the construct validity of the measures. Although common indicators of  
433 MPT adoption were reported across the reviewed studies, significant variation was observed in the  
434 measurement of these indicators or constructs in terms of the scales used, the number and wording of  
435 items, and the measure employed. The observed heterogeneity in measures of MPT adoption might  
436 significantly compromise the validity of the research instruments used to collect data. Therefore, such  
437 discrepancies must be considered when comparing studies. From a practical point of view, we suggest (1)  
438 caution when interpreting findings from such studies, (2) harmonization or standardization of adoption  
439 indicators and their measures as well as instruments, and (3) future research might be needed to conduct  
440 studies on the impact of the methodological differences in the assessment of MPT adoption to gain insight  
441 on the extent to which the call for caution is warranted.

442 In accordance with past literature, the current study revealed a sub-optimal use of behavioral theories and  
443 models in the design of studies seeking to understand farmers' adoption of MPT. (Michie, Johnston,  
444 Francis, Hardeman, & Eccles, 2008) observed that choosing appropriate theories is complex and often  
445 results in most interventions seeking to change behavior being devised without reference to theory  
446 (Davies, Walker, & Grimshaw, 2010; Prestwich et al., 2014). Previous research, however, underscores the  
447 relevance of solid theoretical foundations when designing behavioral change programs to improve their  
448 success and allow synergistic effects (Dombrowski et al., 2012).

449 This review also found that studies do not employ a typical "standardized" or identical theory or model to  
450 examine MPT adoption and usage. The application of diverse models makes comparing studies difficult  
451 and prevents quantitative meta-analysis. While it may be hard to avoid heterogeneity in theories (and  
452 methods), we suggest that studies might at least pay more attention to theoretical underpinnings while  
453 considering the specificities of the different contexts to see which theoretical model fits best

454

455 **4.3 The role of determinants of MPT adoption**

456 **Socio-demographic factors.** Most studies (94%) reported the positive influence of education on farmers'  
457 decisions to adopt MPTs. This finding is consistent with that of Mittal and Mehar (2016), who found a  
458 positive relationship between the use of mobile phones and farmers' level of education. Several authors  
459 posit that knowledge and analytical capacity are prerequisites for adopting innovations in agricultural  
460 systems (Olum et al., 2019; S. P. Thar et al., 2021). While people with little or no education can perform  
461 basic phone functions like voice calling, operating smartphones with several features require a specific  
462 skillset (Islam & Gronlund, 2011). The significant role of education underscores the need for governmental,  
463 non-governmental, and private sector agencies to provide relevant training and technical education for  
464 farmers through capacity-building programs to increase their awareness of using MPTs for agriculture.  
465 Girma and Kelil (2021) suggest awareness campaigns organized by telecommunication companies, mobile  
466 app developers, and the agricultural department as possible avenues to advance adult education and  
467 enhance farmers' competence through tailored training on the use of mobile applications (Abdullahi,  
468 Oladele, & Akinyemi, 2021)

469 The somewhat surprising positive effect of age on MPT adoption reported in some studies could be  
470 attributed to the farming experience through which farmers realize the benefits and need for innovation  
471 (Kabirigi, Sekabira, Sun, & Hermans, 2022b). Consistent with extant literature, younger farmers in the  
472 included studies showed more willingness to adopt mobile agricultural apps or use their phones for  
473 agricultural purposes. Seemingly, younger farmers anticipate benefits from utilizing innovation in the long  
474 term. Therefore, they tend to be more radical and innovative concerning new technologies (Chellappan &  
475 Sudha, 2015). (Kabirigi et al., 2022b) recommend youth involvement through strategies that enable easy  
476 access to smartphones for successful agricultural digitalization and possible transfer of digital skills to the  
477 older farmers through the younger ones.

478 Our review highlighted a gender divide in MPT adoption among farmers in developed and developing  
479 countries. Similarly, past studies point to male dominance in technology adoption (Obisesan, 2014),  
480 suggesting differential adoption rates between men and women (Aduwo et al., 2019). This gender  
481 difference in adoption rates, especially in developing countries, might be attributed to gender inequality  
482 in the access to information, physical and financial resources, and societal norms (Radovic-Markovic, Kabir,  
483 & Jovicic, 2020; Theis, Lefore, Meizen-Dick, & Bryan, 2018). The prevalent gender gap in farmer adoption  
484 of MPT necessitates relevant stakeholders, including developers and researchers, to understand the  
485 factors associated with these differences. Understanding such factors would increase the uptake of MPTs  
486 among farmers, ensure the positive impacts of MPT are sustained, and contribute to SDG 5 on gender  
487 equality.

488 **Technological and mobile phone-use-related factors.** The growing popularity of mobile phones is partly  
489 due to their relative ease of use. The user's perception regarding the ease of use of a given technology  
490 determines the technology's adoption (Kabbiri et al., 2018). James (2009) notes that for users to accept  
491 and use MPTs, they must find their operations easy to learn. Similarly, performance expectancy, effort  
492 expectancy, and perceived advantage were reported to positively influence MPT adoption among farmers.  
493 Therefore, as also suggested by Kabbiri et al. (2018), it is vital for all actors within the mobile phone  
494 industry and the agricultural sector to consider farmers' perceptions relating to ease of use. For instance,  
495 application developers should consider designing mobile applications that are easy to learn and use for  
496 low-literacy farmers and suited to the farmers' context.

497 Perceived usefulness is the subjective expectation of an individual that using new technology will enhance  
498 their performance (Davis, 1993), and past studies allude to its positive effect on technology adoption  
499 (Venkatesh & Davis, 1996; Venkatesh & Morris, 2000). Similarly, perceived usefulness positively influenced  
500 farmers' MPTs adoption, generally among the reviewed studies. Most farmers believed that using a mobile  
501 phone could boost their agricultural performance. Therefore, farmer-oriented governmental and non-  
502 governmental agencies may find it valuable to collaborate with mobile phone companies to develop MPTs  
503 and applications that target farmers' specific needs, and enhance their daily operations. When such  
504 applications are considered relevant by the farming communities, their adoption and agricultural  
505 productivity will increase.

506 According to Omar et al. (2021), the perceived cost related to MPTs are severalfold, namely; initial cost  
507 (e.g., acquisition of MPT), usage cost (e.g., buying credit, data, and financial transaction fees), maintenance  
508 cost (e.g., app and internet subscription, repair, and accessories replacement); cost of building  
509 relationships with other stakeholders along the agricultural value chain and the costs concomitant with  
510 the time and effort taken to collect data and assess options to decision-making. Such costs were found to  
511 influence farmers' MPT adoption negatively generally. Therefore, efforts directed towards subsidizing the  
512 cost of MPTs by the government and telecom companies could facilitate MPT adoption among farmers,  
513 especially those in low-income countries. In line with the effect of costs, some exceptions were reported.  
514 For instance, Omar et al. (2021) found that farmers in Sarawak adopted a gain-oriented approach that  
515 motivated them to adopt apps in expectation of higher values regardless of the potential costs.  
516 Additionally, it is presumed that the MPT usage experience shapes farmers' perceptions of benefits and  
517 costs. This finding highlights the possible role of training and sensitization in creating awareness among  
518 farmers of the benefits of MPT.

519 The triple hurdle approach, proposed by (Asravor et al., 2021), postulates the ownership decision as the  
520 first critical hurdle toward MPT usage that generates the willingness and ability to explore MPT products  
521 and their functionalities. Mobile phone ownership and ownership duration were found to positively  
522 influence farmers' decisions to use MPTs in agriculture. Extended ownership periods probably allow  
523 farmers to witness firsthand the economic gains and cultural transformations associated with MPTs  
524 (Asravor et al., 2021). That mobile phone ownership increases usage by farmers is possibly a result of  
525 increased product knowledge (Osadebamwen & Ele, 2015). Thus, mobile phone developers, extension

526 workers, and development practitioners might consider it worthwhile to influence the decision to own a  
527 mobile phone by addressing constraints such as costs and creating awareness of the benefits of MPTs.

528 ***Psychological and behavioral factors.***

529 The studies reviewed reported a positive effect of social influence on farmers' intention to adopt MPTs,  
530 and previous reviews on technology adoption support this finding (Kamrath et al., 2019; Olum et al., 2019).  
531 Most farmers would adopt MPTs and other innovations if their family members or other farmers in their  
532 networks approved of and used them. Furthermore, subjective norms influenced people's intention to  
533 adopt digital financial services and M-payment in a collective society positively (Hussain, Mollik, Johns, &  
534 Rahman, 2019; Martins, Oliveira, & Popovic, 2014). Landmann et al. (2021) recommend using village-wide  
535 field demonstrations to enhance MPT usage rates since farmers may need initial guidance when  
536 introducing MPTs. Furthermore, field demonstrations might provide a great way to encourage village-wide  
537 adoption of MPTs because farmers value the judgment of their networks. Thus, programs seeking to  
538 promote MPT use in agriculture might target cooperative groups or farmer-based organizations as entry  
539 points to gain support for new MPT innovations and encourage widespread adoption. (Asravor et al., 2021)  
540 noted that such groups enhance farmers' knowledge, awareness, and capacity and contribute to the  
541 decision to use MPTs.

542 Self-efficacy, a person's belief in their ability to exercise control over specific tasks (Bandura, 1977), was  
543 found to influence farmers' adoption intentions regarding MPT positively. This finding underscores the  
544 role of individual optimism in one's ability and motivation for MPT usage (Landmann et al., 2021). It  
545 suggests the need for tailored training and skilling programs that meet farmers' mobile use needs and  
546 boost their confidence.

547 ***Constraints and institutional factors.*** The transfer of innovations and technology uptake require solid  
548 support from financial providers, private sector-led, and government institutions (Olum et al., 2019).  
549 Farmers' network capacity influenced MPT adoption decisions positively in both developed (e.g., Italy) and  
550 developing countries. Thus, institutional approaches could target farmer associations when introducing  
551 new MPTs. [For countries in developing regions, for example, Uganda, government policies could be](#)  
552 [directed towards reviving and supporting farmer cooperatives to facilitate the uptake of technologies,](#)  
553 [including MPTs.](#) Access to credit also had a positive influence on MPT adoption. Thus, micro-finance  
554 institutions and governments could consider providing financial support through subsidies or soft loans to  
555 facilitate farmers' uptake of MPT (Girma & Kelil, 2021; Sikundla et al., 2018).

556 Furthermore, information awareness positively affected farmers' intention and adoption of MPTs. Mwangi  
557 and Kariuki (2015) posit information awareness as a necessary precursor to farmers' adoption of  
558 technology. Yet, information asymmetries remain a significant obstacle to technology adoption,  
559 predominantly in developing countries (Nakasone et al., 2014). Without prior knowledge of available  
560 MPTs, many farmers fail to implement and benefit from them, despite having positive intentions (Misaki  
561 et al., 2018). Cognizant of the possibility that poorly packaged information can lower MPT adoption  
562 (Kabunga, Dubois, & Qaim, 2012; Mwangi & Kariuki, 2015), the information provided to farmers  
563 concerning or through MPTs must be timely, accurate, relevant, consistent, and from a trusted source to  
564 promote adoption (Olum et al., 2019).

565 According to Anjum (2015), incorporating extant social and psychological barriers in the design stage might  
566 promote farmers' MPT adoption. Our study highlighted the critical role of language barrier in limiting  
567 access and utilization of MPTs by local farmers, especially in countries where English is not an official  
568 language. Using foreign languages in a local context might lead to misinterpretation (Asenso-Okyere &  
569 Mekonnen, 2012) and misuse of the information provided. Thus, ethical and language consideration is  
570 critical to MPT companies in the initial design stage when developing mobile applications and content.  
571 Misaki et al. (2018) suggest incorporating local languages relevant to the cultural context in the MPT design  
572 process to promote the adoption and scalability of mobile phone use in agriculture. [From a managerial](#)  
573 [perspective, it may be cumbersome and even unrealistic to perhaps translate MPTs into all local languages,](#)

574 especially in an environment of many dialects typical of many developing countries in Africa and Asia.  
575 However, the benefits of translation still outweigh its downsides. Therefore, we suggest that the  
576 translation of MPT services and features into local languages could target leading (commonly) spoken  
577 languages for a start. Thus, even though some minority groups might be left out, the transmission of  
578 information, adoption, and utilization of MPTs could be significantly improved among the target  
579 populations.

580 Venkatesh et al. (2003) define the extent to which a person believes that organizational and technical  
581 structures are in place to support the use of a system as facilitating conditions. They constitute perceived  
582 environmental barriers or enablers that influence an individual's perception of the ease or difficulty of  
583 performing a task (Teo, 2010). According to Owusu, Yankson & Frimpong (2017), facilitating conditions  
584 such as resource availability, reliable electricity, good internet, mobile networks, skills, and technical  
585 infrastructure could play a key role in MPT adoption among farmers. This finding has important  
586 implications for developing countries often plagued by inadequate infrastructure, including epileptic  
587 power supplies, poor internet and network connectivity, and weak broadband signals (Falola & Adewumi,  
588 2011; Ogunniyi & Ojebuyi, 2016). It raises intriguing questions regarding the nature and extent of support  
589 different stakeholders can provide to address such challenges. We propose that governments and non-  
590 governmental organizations orient their development efforts towards upgrading rural infrastructure to  
591 facilitate farmers' MPT adoption for agricultural use (Owusu, Yankson, & Frimpong, 2017). Moreover,  
592 there is a need for policy reforms that deliberately provide an enabling environment to support mobile  
593 companies and farmers, thus encouraging MPT adoption and usage (Narine, Harder, & Roberts, 2019).  
594 Such reforms require effective public-private partnership and collaboration by stakeholders within the  
595 agricultural and mobile industry to implement appropriate policies (Mgbenka, Mbah, & Ezeano, 2015).

## 596 597 **5 Strengths and Limitations of the study**

598 Our study has its strengths in (1) using the well-established PRISMA guidelines and a review protocol to  
599 conduct a systematic literature review of farmers' adoption of MPT and ensure that the process can be  
600 replicated (2) inclusion of studies from both developing and developed countries perspective, and (3) the  
601 assessment of the methodological quality of included studies using the Mixed Method Appraisal Tool (4)  
602 enhancing understanding of the drivers of farmers' adoption of MPTs, measures, theories, and adoption  
603 indicators, and provides direction for future research.

604 Despite the robustness of our review, however, we acknowledge some limitations. First, we restricted our  
605 search to the web of science electronic database. Although this is not unprecedented in our field of  
606 research (see Inwood & Dale, 2019; Kamrath et al., 2019; Olum et al., 2019) and was supplemented by  
607 citation and reference list search, we recognize that some relevant studies not indexed by the Web of  
608 Science could have been missed in the literature search. Moreover, we considered only full-text articles  
609 published in English and could have left articles published in other languages and not electronically  
610 available. Although the relatively small sample size of included studies indicates that this growing research  
611 field is in its nascent stages, we recommend carefully interpreting our findings.

612 Secondly, the heterogeneity of MPT outcome measures and variables reported across the included studies  
613 due to the lack of standardized tools, methods, and reporting made it impossible to conduct a reliable  
614 statistical meta-analysis which would have otherwise enabled the calculation of the effect sizes of various  
615 determinants of MPT adoption (Wely, 2014).

616 Thirdly, only 34% of the included studies were ranked as "high quality, with nearly 50% of studies not  
617 providing details on how various constructs and indicators of MPT adoption were measured. Moreover,  
618 only reporting significant relationships with missing information on effect sizes constitutes publication bias  
619 (Hirschauer et al., 2016). Petticrew and Roberts (2006) noted that such targeted reporting complicates the  
620 comparison of significant to non-significant assessments. Therefore, interpretation of our findings should

621 factor in this publication bias. As Kamrath et al. (2019) suggested, researchers should aim to enhance  
622 transparency in research by providing information on the concepts and constructs they measure.

623

## 624 **6 Conclusions**

625 The current work was designed to systematically examine farmers' MPT adoption studies and appraise  
626 their methodological quality. Specifically, it sought to (a) analyze and synthesize the key determinants and  
627 constraints of MPT adoption among farmers; (b) evaluate the various measures, methods, and theories  
628 employed by studies to understand MPT adoption; (c) assess farmers' perception regarding different MPTs  
629 and their associated usage, (d) appraise study quality and (e) identify gaps and areas for future research  
630 regarding MPT adoption among farmers.

631 Our systematic review highlights a rapidly growing body of MPT adoption literature, especially targeting  
632 developing countries. However, these studies oriented towards MPT usage in developing countries target  
633 more basic phones and technologies such as SMS, unlike the developed country studies focusing on more  
634 advanced technologies exploring the usage of smartphones and applications.

635 The articles included in the review point to the positive role of factors such as perceived ease of use,  
636 perceived usefulness, farmers' innovativeness, mastery goal orientation, perceived advantage, education,  
637 farmer's network capacity, credit access, mobile phone ownership, and ownership duration on the  
638 decision to adopt MPTs. The importance of the determinants of farmer MPT adoption was shown to differ  
639 between regions. For instance, education, age, family size, and constraints such as the high cost associated  
640 with MPT acquisition and usage, language barrier, and inadequate infrastructure were identified as the  
641 main determinants of MPT adoption among farmers in developing countries. However, Computer literacy,  
642 weather variability, and price volatility were recognized as the primary challenges hindering MPT adoption  
643 among farmers in developed countries (e.g., Germany). Understanding these contextual differences might  
644 be beneficial in need-specific tailoring of solutions to promote MPT among farmers in developed and  
645 developing countries.

646 All of the studies examined were cross-sectional and mainly applied quantitative descriptive methods,  
647 followed by the mixed methods approaches, then qualitative methods.

648 The reviewed articles featured a variety of MPT adoption indicators and outcome measures ranging from  
649 mobile phone ownership, intention to use, generic MPT usage, adoption, perception, attitude, knowledge  
650 intensity, frequency of use, constraints, willingness to pay, and willingness to adopt. Similarly, the included  
651 studies varied significantly in measuring indicators, with some soliciting yes/no responses and others using  
652 Likert scales ranging from 3 to 7-point scales, revealing heterogeneity in the measurement of adoption  
653 among farmers. Heterogeneity was also found in the number and wording of items used to measure the  
654 same construct of a given theory to examine MPT adoption among farmers.

655 The generally low quality of the studies included in this review suggests a dire need to improve study  
656 designs, methods, and measurements to capture evidence on farmers' MPT adoption more satisfactorily.  
657 Similar to the suggestion by Parikh et al. (2022), enhancing the quality of research and filling the knowledge  
658 gap will be crucial to influencing MPT adoption among farmers and transforming agricultural productivity.  
659 We, therefore, propose the following

### 660 ***Methodological recommendations;***

- 661 • Complement the cross-sectional designs with longitudinal studies to predict MPT adoption  
662 behavior and beliefs over time (Omar et al., 2021).
- 663 • Future studies should consider transparently stating their methodology, including sampling  
664 methods, data collection and analysis, measurement of different variables, and publicly availing  
665 their research tools (Diaz et al., 2021).
- 666 • Develop and disseminate pre-tested and validated standardized tools to facilitate harmonized  
667 data collection on mobile phone technologies across diverse farmer profiles and settings.

- 668 • As also proposed by Parikh et al. (2022), augment quantitative and qualitative methods for a more  
669 holistic assessment of the status of MPT adoption.
- 670 • Future studies should identify straightforward research questions and objectives, use appropriate  
671 methodologies and validated tools, and base interpretation of results on the data obtained.
- 672 • Farmers' preferences for different MPTs and applications should be further investigated using a  
673 choice experimental design.
- 674 • Address the lack of theoretical underpinnings in MPT adoption, and if possible, consider the  
675 context in which these theories could best be applied.

676 ***Future research recommendations***

- 677 • Extend the scope of MPT adoption research to evaluate both users and non-users in a single study  
678 design, to compare similarities and differences in perceptions, and the reason for non-adoption  
679 (Michels, Bonke, & Musshoff, 2019).
- 680 • Furthermore, studies should investigate why most farmers do not use mobile agricultural apps  
681 despite perceiving them as applicable (Michels et al., 2020). Relatedly, identify the work-specific  
682 barriers preventing the growing number of farmers who already own a smartphone from using it  
683 for agricultural purposes (Landmann, Lagerkvist, and Otter, 2020).
- 684 • Examine farmers' perceptions regarding the impacts of MPT adoption on their social networks and  
685 agricultural outcomes.
- 686 • Investigate the possibility and effectiveness of mobile data collection, especially among  
687 smallholders.
- 688 • More studies are still needed to compare and contrast the inconclusive effects of various  
689 determinants on MPT adoption.
- 690 • Measure MPT adoption patterns among farmers using established theories, such as diffusion of  
691 innovation theory (Michels, Bonke, & Musshoff, 2019).
- 692 • Extend the scope of studies examining farmers' willingness - to - Pay for agricultural applications  
693 beyond soliciting yes/No responses and assess the extent of their desire to understand the value  
694 farmers attribute to mobile applications (Bonke et al., 2018)
- 695 • Examine the relationship between farmers' identity and their engagement with agricultural  
696 (smartphone) applications (Kenny & Regan, 2021).
- 697 • Direct efforts towards understanding farmers' lack of trust in agricultural Apps and MPTs.
- 698 • Assess the apparently strong link between farmer network participation and farm advisor use with  
699 the higher level of agricultural app adoption (Schulz et al., 2021).
- 700 • More research is required to integrate MPTs with traditional farmer extension and training  
701 methods.
- 702 • Assess the role of government in farmers' adoption of MPTs and related services.
- 703 • Examine how the importance of various determinants on farmer MPT adoption change over time  
704 or in context
- 705 • Conduct studies on the impact of the methodological differences in assessing MPT adoption.
- 706 • Extend the scope of MPT adoption studies beyond the farmer level, and include traders, suppliers,  
707 distributors, government organizations, and respective boards of agriculture in one study design  
708 to provide a different yet holistic perspective on the adoption of MPTs (Nabhani et al., 2016)

709 ***Policy and development recommendations***

- 710 • Heed to the intersection between gender and MPT adoption among farmers when designing or  
711 implementing ICT programs (Molina-Maturano et al., 2021; Owusu, Yankson, & Frimpong, 2017)
- 712 • Direct efforts towards addressing farmers' lack of trust in agricultural Apps and MPTs.
- 713 • Consider social dynamics within farming communities, such as sharing and borrowing phones,  
714 when designing MPT-related policies or programs.

- 715 • Orient efforts towards educating farmers on the benefits of mobile apps and offering tailored
- 716 training to enhance their competence in mobile phone usage (Abdullahi, Oladele, & Akinyemi,
- 717 2021; Nabhani et al., 2016; Okoroji, Lees & Lucock, 2021)
- 718 • Enhance farmers' positive intention toward MPT usage through subsidization of costs associated
- 719 with MPT adoption and use (Nyamba & Mlozi, 2012; Okoroji, Lees & Lucock, 2021)
- 720 • To unlock the full potential of MPTs in developing countries, consider interventions like rural
- 721 electrification and solar phone charging stations (Obong, Mugonola, & Paul Phillips, 2018)
- 722 • Incorporate mobile phones in the traditional extension systems, and consider providing
- 723 information to farming households via MPTs.
- 724 • Focus on implementing public sensitization to create awareness regarding various MPTs.
- 725 • Make MPT services more customizable to the local context, and allow for translation into local
- 726 languages to facilitate use by low-literacy farmers (Kabirigi et al., 2022).
- 727 • Target the youth when designing and introducing new MPT innovations.
- 728 • Formative evaluation is necessary to ensure farmers' feedback is factored into the rollout of MPT
- 729 programs
- 730 • Pay attention to the social differences between developed and developing countries during MPT
- 731 designs, implementation, and further research. Similarly, create relevant and quality content
- 732 when designing MPTs for farmers.
- 733 • To boost adoption and support for the use of MPTs, mobile app developers and public and private
- 734 sector players may consider co-designing applications and mobile-related programs with farmers.
- 735 • Together with the government, mobile phone developers could consider subsidizing the internet,
- 736 airtime, and other services to facilitate farmers' use of MPTs in agriculture (Sikundla, Mushunje, &
- 737 Akinyemi, 2018). As a further step, telecommunication companies should extend and strengthen
- 738 their network coverage and enhance the quality of their products to allow MPT adoption among
- 739 farmers.

740

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744

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1075 **Biographical endnote for authors**

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Table 1: The PCC Framework

	Definition
<b>Population</b> The target population is <b>farmers</b> .	For this review, a “farmer” refers to anyone engaged in agriculture, raising livestock, fish, or growing crops for food, raw materials, or other commercial reasons and owns or hires the production land.
<b>Concept</b> The fundamental concept of this review is <b>the adoption of mobile phone technology</b> .	To serve the aim of this review, “adoption” was taken to mean the acceptance, integration, and use of an innovation/ technology (MPT for our case), Mobile phone technology encompasses all mobile phones, smartphones, and applications
<b>Context</b> Global	This review targets studies in both developed and developing countries.
<b>*PCC= Population, Concept and Context framework (Hannes &amp; Macaitis, 2012; Ng et al., 2014) was utilized to define the key components of this review.</b>	

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Table 2: Types of MPTs and behavioral theories reported in the included studies

	n	References
<b>Theories/models</b>		
TAM	10	(Kabbiri et al., 2018; Kacharo, Mvena, & Sife, 2018; Landmann, Lagerkvist and Otter, 2020 ; Nabhani et al., 2016; Sikundla, Mushunje, & Akinyemi, 2018; Verma & Sinha, 2018; Okoroji, Lees & Lucock 2021)
UTAUT	7	(Beza et al., 2018; Michels, Bonke, and Musshoff, 2020; Molina-Maturano et al., 2021; Mwalukasa et al., 2018; Narine, Harder, & Roberts, 2019; Omar et al., 2021)
Other theories	11	(Abdullahi, Oladele, & Yusuf, 2019; Bonke et al., 2018; Folitse et al., 2018; Kabirigi et al., 2022; Krell et al., 2020)
<b>Type of MPT</b>		
Mobile (smart) phone	19	(Abdullahi, Oladele & Akinyemi, 2021; Folitse et al., 2018; Hoang & Drysdale, 2021; Kabirigi et al., 2022; Kenny & Regan, 2021; Lubua & Kyobe, 2019; Otene, Ezihe, & Torgenga, 2017)
Decisions Support Applications (DSA)	14	(Bonke et al., 2018; Krell et al., 2020; <a href="#">Michels, Bonke, &amp; Musshoff, 2019</a> ; <a href="#">Michels, Bonke, &amp; Musshoff, 2020</a> ; Molina-Maturano et al., 2021; Nikam, Kumar, & Kingsly, 2021; Owusu, Yankson, & Frimpong, 2017; Thar et al., 2021)
Text messaging applications (SMS)	13	(Abdullahi, Oladele, & Yusuf, 2019; Aricat & Ling, 2017; Islam & Gronlund, 2011; Kaske et al., 2017; Khan et al., 2019; Narine, Harder, & Roberts, 2019 )
Social media applications	7	(Abdullahi, Oladele, & Yusuf, 2019; Aricat & Ling, 2017; Khan et al., 2019; Krell et al., 2020) (Aricat & Ling, 2017; Kaske et al., 2017; Khan et al., 2019)
News/information applications	7	(Abdullahi, Oladele, & Yusuf, 2019; Asravor, Boakye, & Essuman, 2021 ; Khan et al., 2020)
Others	8	

Comments: Other theories refer to theories that were mentioned by at most two studies, and they include affordance theory, graph theory, goal-orientation theory, Goal-based behavioral model (MGB); Theory of reasoned action (TRA), Integrative Approach to Models of Technology Adoption (IAMTA); household behavior theory, theory of information and communication technology for development, social cognitive theory, Actor-Network theory, and model of personal computer use. TAM = Technology acceptance model; UTAUT = Unified Theory of acceptance and use of technology. DSA identified through this review include crop protection apps, Agri-info, and advisory apps; market information systems, livestock management apps; camera function apps; and disease and weather forecast apps.  
 n is the number of studies in each category in the table

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Table 3: A table of reported indicators of MPT adoption and their measurement

	<b>Adoption indicator</b>	<b>Measurement</b>	<b>Reference</b>
1	Mobile phone ownership	<ul style="list-style-type: none"> <li>Dummy variable requiring Yes/No responses (Y=1 for households that own and use the mobile phones and Y=0 otherwise);</li> <li>Discrete choice responses: owns a phone, has access, no access</li> </ul>	(Abdullahi, Oladele, & Yusuf, 2019; Aricat & Ling, 2017; Asravor, Boakye, & Essuman, 2021; Folitse et al., 2018; Islam & Gronlund, 2011; Animashaun et al., 2014; Krell et al., 2020; Michels, Bonke, & Musshoff, 2020)
2	Behavioral intention	<ul style="list-style-type: none"> <li>Likert scales (5-point, 7-point)</li> <li>Dummy variable ((1 = 'I intend to MPT to communicate with extension officers,' and 0 = 'I do not intend to MPT to communicate with extension officers.'))</li> </ul>	(Aldosari et al., 2017; Beza et al., 2018; Landmann, Lagerkvist and Otter, 2020; Molina-Maturano et al., 2021; Mwalupaso et al., 2019; Nabhani et al., 2016; Narine, Harder, & Roberts, 2019 ; Omar et al., 2021; Otene, Ezihe, & Torgenga, 2017; Verma & Sinha, 2018; Okoroji, Lees & Lucock 2021)
3	Generic MPT use	<ul style="list-style-type: none"> <li>Likert scales (5-point)</li> <li>Number of mobile phone activities, e.g., dialed call, received call, sent short message service (SMS), received SMS, listening to the radio);</li> <li>number of mobile phones and sim cards</li> <li>Dummy variable (1: user; 2: non-user; 1 = farmer used mobile phone, 2 = otherwise, 1= adopted, 0 if otherwise)</li> </ul>	(Abdullahi, Oladele, & Akinyemi, 2021; Aldosari et al., 2017; Folitse et al., 2018; Girma & Kelil, 2021; Kabirigi et al., 2022; Kacharo, Mvena, & Sife, 2019; Kaske et al., 2017; Kassem et al., 2020; Mwalukasa et al., 2018; Mwalupaso et al., 2019; Mwantimwa, 2017; Nyamba & Mlozi, 2012)
4	Use intensity	<ul style="list-style-type: none"> <li>Likert Scales (3-point, 4-point, 5-point, 6-point)</li> <li>Number of transactions (sending/receiving money, saving, paying bills, calls, SMS I a day, week, or month, (daily, weekly, and monthly)</li> </ul>	(Abdullahi, Oladele, & Yusuf, 2019; Asravor, Boakye, & Essuman, 2021; Filippini et al., 2020; Hoang & Drysdale, 2021; Islam & Gronlund, 2011; AL; Lubua & Kyobe, 2019; Michels et al., 2020; Ogunniyi & Ojebuyi, 2016; Otene, Ezihe, & Torgenga, 2017)
5	Use frequency	<ul style="list-style-type: none"> <li>Likert scale (5-point, 6-point)</li> <li>Discrete choices: daily, weekly, monthly</li> </ul>	(Filippini et al., 2020; Islam & Gronlund, 2011; Lubua & Kyobe, 2019; Michels et al., 2020; Mwantimwa, 2017)
6	Adoption	<ul style="list-style-type: none"> <li>Dummy variable (adoption= 1, 0 otherwise; 1 = adoption of crop protection smartphone app; 0 = non-adoption; 1=adopter, 0= non-adopter); Yes/No responses</li> </ul>	(Alam et al., 2018; Kabbiri et al., 2018; Michels, Bonke, & Musshoff, 2020; Nikam, Kumar, & Kingsly, 2021 ; Owusu, Yankson, & Frimpong, 2017; Sikundla, Mushunje, & Akinyemi, 2018 )
7	Attitude	<ul style="list-style-type: none"> <li>Likert Scale (5-point)</li> <li>Likert scale (4-point, 5-point)</li> </ul>	(Abdullahi, Oladele, & Akinyemi, 2021; Kassem et al., 2020; Schulz et al., 2021; Tanrikulu & Ozturan, 2012; Thar et al., 2020)
8	Perception	<ul style="list-style-type: none"> <li>Likert Scale (5-point)</li> </ul>	(Bonke et al., 2018; Kabirigi et al., 2022; Kenny & Regan, 2021; Lubua & Kyobe, 2019)

9	Knowledge	• Discrete choice (True or False)	(Abdullahi, Oladele, & Akinyemi, 2021)
10	Willingness to Adopt	• Likert scale (5-point)	(Diaz et al., 2021)
11	Constraints	• Likert scale (4-point) • Yes/No responses	(Abdullahi, Oladele, & Akinyemi, 2021; Kabirigi et al., 2022; Kenny & Regan, 2021; Mwantimwa, 2019)
12	Willingness to pay	• Likert Scale (5-point); • Yes/No responses	(Bonke et al., 2018)

Note; the references above are only for studies that indicated how the indicators were measured. It does not reflect the total number of studies that measured these indicators, and caution is advised when interpreting these.

Remarks: **Sample statement to measure behavioral intention** "I am planning to use a smartphone to generate agricultural knowledge" (scale 1 = strongly disagree to 5 = strongly agree) by Landmann, Lagerkvist, and Otter (2020) and "I plan to use or continue using mobile phone apps frequently that provide agronomic information" (scale 1= totally disagree to 7= totally agree) by Molina-Maturano et al. (2021). Note: the construct behavioral intention was measured using different number of items. For example, 3 items (Beza et al., 2018; Molina-Maturano et al., 2021); 2 items Landmann, Lagerkvist and Otter (2020), and 5 items (Omar et al., 2021; Okoroji, Lees & Lucock 2021). **Measurement of behavioral intention** as a binary variable where 1 = 'I intend to use text messaging to communicate with extension officers,' and 0 = 'I do not intend to use text messaging to communicate with extension officers.' Taken from Narine, Harder, & Roberts (2019). **Measurement of willingness-to-pay:** "Farmers were asked if they were willing to pay a yearly fee for a crop protection app which included all of the functions they perceive as useful (responses: Yes/No)" (Bonke et al., 2018); **Measurement of mobile phone adoption:** (MPA = 1, 0 otherwise) where farmers that use mobile phone for calls, SMS or both in agricultural marketing are considered adopter, whereas those that have never used mobile phone services (i.e. calls and SMS or both) for agricultural marketing are considered as non-adopters" taken from Sikundla, Mushunje, & Akinyemi (2018). **Measurement of Mobile phone ownership:** "1, if the farmer has a smartphone; 0 otherwise 1, if the farmer has a mobile phone; 0 otherwise" (Michels, Bonke, & Musshoff, 2020). The majority of studies did not provide the measurement items that were used to measure various constructs and indicators.

Table 4: A table of determinants of farmers' MPT adoption

Category of determinants	Significant variables	Effects	References
<b>Socio-demographic variables</b>	Age	Inconclusive (+/-)	(Animashaun et al., 2014; Filippini et al., 2020; Folitse et al., 2018; Hoang, 2020; Hoang & Drysdale, 2021; Islam & Gronlund, 2011; Kabirigi et al., 2022; Kacharo, Mvena, & Sife, 2018; Khan et al., 2020; Krell et al., 2020; Mwalukasa et al., 2018; Mwantimwa, 2019; Thar et al., 2020)
	Farming experience	Inconclusive (+/-)	(Folitse et al., 2018; Kaske et al., 2017; Mwalupaso et al., 2019)
	Gender	Inconclusive (+/-)	(Filippini et al., 2020; Folitse et al., 2018; Hoang, 2020; Animashaun et al., 2014; Mwalukasa et al., 2018; Mwantimwa, 2017; Otene, Ezihe, & Torgenga, 2017; Sikundla, Mushunje, & Akinyemi, 2018; Thar et al., 2020)
	Marital status	Inconclusive (+/-)	(Animashaun et al., 2014; Mwalukasa et al., 2018; Mwalupaso et al., 2019; Mwantimwa, 2017; Otene, Ezihe, & Torgenga, 2017)
	Farm size	Inconclusive (+/-)	(Folitse et al., 2018; Girma & Kelil, 2021; Hoang & Drysdale, 2021; Kacharo, Mvena, & Sife, 2018; Michels et al., 2020; Mwalukasa et al., 2018; Narine, Harder, & Roberts, 2019; Otene, Ezihe, & Torgenga, 2017; Tanrikulu & Ozturan, 2012)
	Income	Inconclusive (+/-)	(Girma & Kelil, 2021; Hoang, 2020; Hoang & Drysdale, 2021; Kacharo, Mvena, & Sife, 2018; Kaske et al., 2017; Mwantimwa, 2017; Otene, Ezihe, & Torgenga, 2017; Sikundla, Mushunje, & Akinyemi, 2018; Tanrikulu & Ozturan, 2012)
	Education	Inconclusive (+/l)	(Girma & Kelil, 2021; Hoang, 2020; Hoang & Drysdale, 2021; Animashaun et al., 2014; Khan et al., 2020; Krell et al., 2020; Michels, Bonke, & Musshoff, 2020; Michels et al., 2020; Mwalukasa et al., 2018; Nikam, Kumar, & Kingsly, 2021; Otene, Ezihe, & Torgenga, 2017; S. P. Thar et al., 2020)
	Number of dependents	Positive (+)	(Islam & Gronlund, 2011)
	Household size	Inconclusive (+/-)	(Animashaun et al., 2014)
<b>Technological factors and Mobile use related attributes</b>	Duration of (MP ownership)	Positive	(Kacharo, Mvena, & Sife, 2018; Kaske et al., 2017; Khan et al., 2020)
	Farmers' skill and aptitude	Positive (+)	(Khan et al., 2020)

	Number of MP	Positive (+)	(Diaz et al., 2021; Kaske et al., 2017; Khan et al., 2020; Mwalukasa et al., 2018)
	Experience	Positive (+)	(Beza et al., 2018; Folitse et al., 2018; Kacharo, Mvena, & Sife, 2018; Kaske et al., 2017; Mwalukasa et al., 2018; Narine, Harder, & Roberts, 2019; Owusu, Yankson, & Frimpong, 2017; Tanrikulu & Ozturan, 2012)
	Mobile phone ownership	Positive (+)	(Aldosari et al., 2017; Aricat & Ling, 2017; Asravor, Boakye, & Essuman, 2021; Folitse et al., 2018; Islam & Gronlund, 2011; Khan et al., 2019; Krell et al., 2020; Nyamba & Mlozi, 2012)
	Perceived ease of use	Inconclusive (+/-)	(Kabbiri et al., 2018; Landmann, Lagerkvist and Otter, 2020 ; <a href="#">Michels, Bonke, &amp; Musshoff, 2019</a> ; Nabhani et al., 2016)
	Perceived usefulness	Inconclusive (+/-)	(Diaz et al., 2021; Kabbiri et al., 2018; Kaske et al., 2017; Verma & Sinha, 2018; Okoroji, Lees & Lucock 2021)
	Performance expectancy	Positive (+)	( <a href="#">Michels, Bonke, &amp; Musshoff, 2020</a> ; Narine, Harder, & Roberts, 2019; Omar et al., 2021)
	Effort expectancy	Positive (+)	( <a href="#">Michels, Bonke, &amp; Musshoff, 2020</a> ; Narine, Harder, & Roberts, 2019; Omar et al., 2021)
	Perceived cost	Inconclusive (+/-)	(Diaz et al., 2021; Mwalukasa et al., 2018; Omar et al., 2021)
	Perceived advantage	Positive (+)	(Kabbiri et al., 2018; Kaske et al., 2017)
<b>Psychological/behavioral factors</b>	Attitude/	Positive (+)	(Abdullahi, Oladele, & Akinyemi, 2021; Kaske et al., 2017; Landmann, Lagerkvist and Otter, 2020; Mwalupaso et al., 2019; Tanrikulu & Ozturan, 2012; Thar et al., 2020; Verma & Sinha, 2018)
	Self-efficacy	Positive (+)	( Landmann, Lagerkvist and Otter, 2020)
	Social influence	Positive (+)	(Beza et al., 2018; Diaz et al., 2021; <a href="#">Michels, Bonke, &amp; Musshoff, 2020</a> ; Molina-Maturano et al., 2021; Nabhani et al., 2016; Narine, Harder, & Roberts, 2019 ; Omar et al., 2021; Otene, Ezihe, & Torgenga, 2017; Verma & Sinha, 2018; Okoroji, Lees & Lucock , 2021)
	Intention to use	Positive (+)	(Diaz et al., 2021; <a href="#">Michels, Bonke, &amp; Musshoff, 2020</a> ; <a href="#">Michels et al., 2020</a> ; Omar et al., 2021; Okoroji, Lees & Lucock 2021)
	Information need and seeking behavior	Positive (+)	(Kacharo, Mvena, & Sife, 2018)
	Perceived innovativeness	Negative (-)	(Beza et al., 2018; Bonke et al., 2018; Diaz et al., 2021; <a href="#">Michels, Bonke, &amp; Musshoff, 2019</a> ; Molina-Maturano et al., 2021)
	Emotion	Positive (+)	(Landmann, Lagerkvist and Otter, 2020)
	Perception	Positive (+)	(Abdullahi, Oladele, & Akinyemi, 2021; Bonke et al., 2018; Kaske et al., 2017; Khan et al., 2019; Owusu, Yankson, & Frimpong, 2017; Thar et al., 2020)
	Habit	Positive (+)	(Beza et al., 2018; Landmann, Lagerkvist and Otter, 2020)
	Perceived reliability of mobile phone services	Positive (+)	(Kacharo, Mvena, & Sife, 2018; Kaske et al., 2017)
	Trust	Positive (+)	(Beza et al., 2018; Kassem et al., 2020; Molina-Maturano et al., 2021; Okoroji, Lees & Lucock ,2021)

<b>Constraints and Institutional factors</b>	Information awareness	Positive (+)	(Diaz et al., 2021; Folitse et al., 2018; Kassem et al., 2020; Otene, Ezihe, & Torgenga, 2017; Okoroji, Lees & Lucock, 2021)
	Knowledge of Apps	Positive (+)	(Abdullahi, Oladele, & Akinyemi, 2021; Michels, Bonke, & Musshoff, 2019)
	Type and quality of agricultural information	Positive (+)	(Nyamba & Mlozi, 2012)
	Access to electricity	Inconclusive (+/-)	Hoang, 2020; Hoang & Drysdale, 2021)
	Distance from market center	Negative (-)	(Filippini et al., 2020; Folitse et al., 2018; Hoang, 2020; Animashaun et al., 2014; Kacharo, Mvena, & Sife, 2018; Mwalupaso et al., 2019; Thar et al., 2020)
	Distance from the nearest town	Negative (-)	(Mwalupaso et al., 2019; Thar et al., 2020)
	Assets	Positive (+)	(Krell et al., 2020)
	Group membership	Inconclusive (+/-)	(Hoang, 2020; Krell et al., 2020; Mwalupaso et al., 2019)
	Network capacity of the farmer	Positive (+)	(Filippini et al., 2020)
	Credit access/credit program participation	Positive (+)	(Girma & Kelil, 2021; Hoang, 2020)
	Facilitating conditions	Positive (+)	(Beza et al., 2018; Michels, Bonke, & Musshoff, 2020; Molina-Maturano et al., 2021; Ogunniyi & Ojebuyi, 2016; Omar et al., 2021)
	Constraints	Negative	(Falola & Adewumi, 2012; Kacharo, Mvena, & Sife, 2018; Kaske et al., 2017; Kenny & Regan, 2021; Krell et al., 2020; Mwantimwa, 2019; Nyamba & Mlozi, 2012; Ogunniyi & Ojebuyi, 2016; Otene, Ezihe, & Torgenga, 2017; Obong, Mugonola & Paul Phillips, 2018; Thar et al., 2020; Wyche & Steinfield, 2015)
<b>Biophysical factors</b>	Landholding	Negative	(Nikam, Kumar, & Kingsly, 2021)
	Area under cultivation	Inconclusive (+/-)	(Mwalukasa et al., 2018; Nikam, Kumar, & Kingsly, 2021)
	Diversification	Inconclusive	(Animashaun et al., 2014; Bonke et al., 2018; Kabirigi et al., 2022; Kassem et al., 2020; Thar et al., 2020)
	Farm type	Positive	(Kabirigi et al., 2022)

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Table 5: A table of characteristics and quality of the included studies

	Categories reported	Percentage of studies in each category
<b>Publication year</b>	2011	1.8%
	2012	7.5%
	2013	1.8%
	2014	1.8%
	2015	1.8%
	2016	3.7%
	2017	11.3%
	2018	16.9%
	2019	15.0%
	2020	15.0%
	2021	22.6%
	2022	1.8%
<b>Geographic distribution of studies</b>	Africa	53%
	Asia	28%
	Europe	13%
	America	4%
	Australia	1.8%
<b>Setting</b>	Rural	47.0%
	Urban	6.0%
	Peri-urban/Rural	4.0%
	Not specified	43.0%
<b>Type of farming activity</b>	Crop production	21.0%
	Livestock production	17.0%
	Mixed farming	11.0%
	Not specified	51.0%
<b>Scale of production</b>	Small scale farmer	30.0%
	Medium Scale farmer	2.0%
	Large scale farmer	4.0%
	Not specified	64.0%
<b>Research method and design</b>	Quantitative descriptive	75.4%
	Qualitative	5.6%
	Mixed methods	19.0%
	Cross-sectional design	100%
<b>MMAT Criteria met</b>	0% (-)	11.3%
	40% (**)	13.2%
	60% (***)	41.5%
	80% (****)	26.5%
	100% (*****)	7.5%

1217  
 1218  
 1219  
 1220

Note: Africa includes South Africa, Nigeria, Kenya, Uganda, Ghana, Ethiopia, Tanzania, Zambia; Asia includes India, Myanmar, Kingdom of Saudi Arabia, Vietnam, Bangladesh, Pakistan, and Malaysia; Europe has Germany, Turkey, Italy; North America includes Mexico; Oceania includes Australia; South America includes Trinidad.



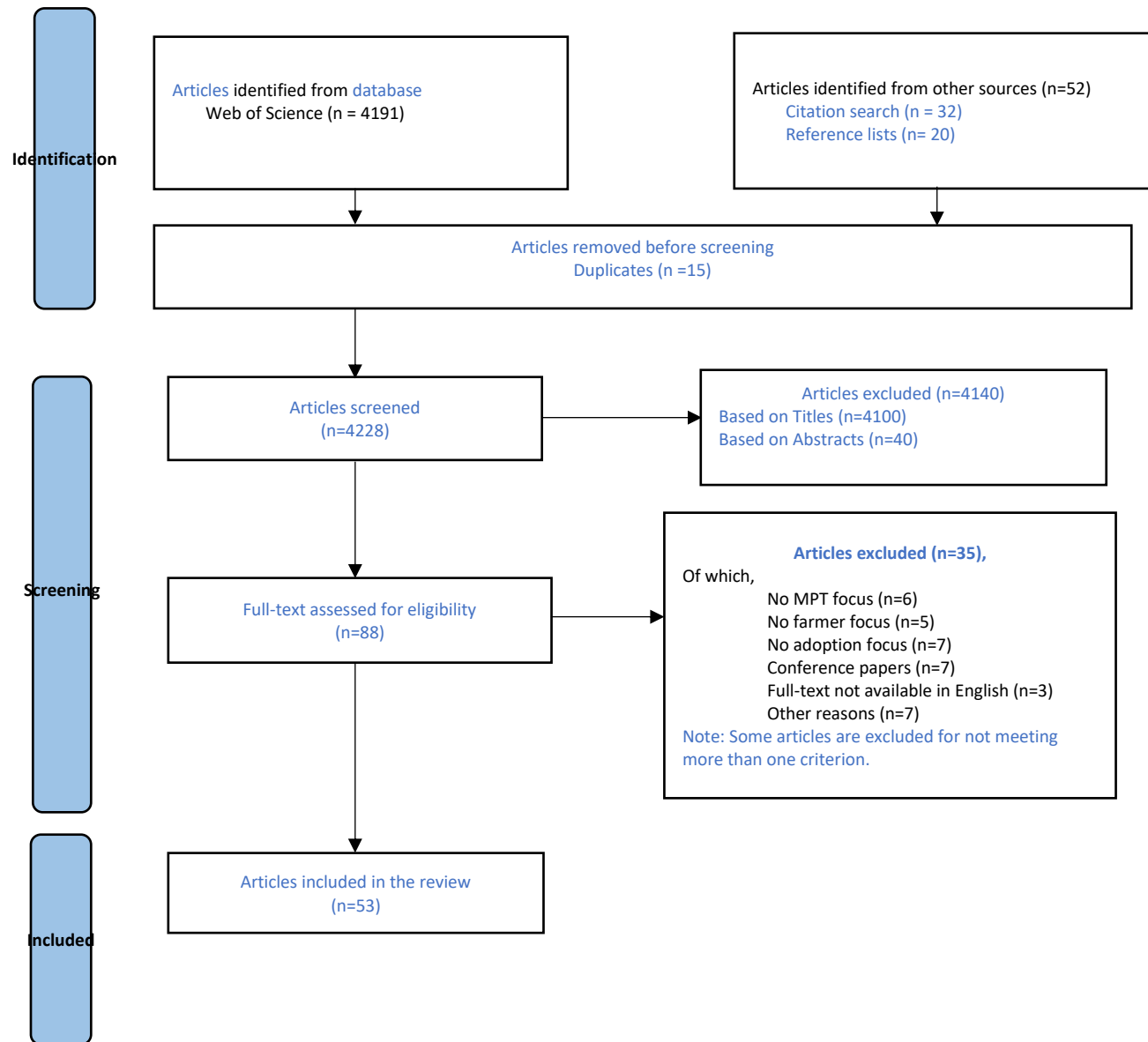


Figure 1: A PRISMA flow chart of the detailed screening strategy for identified studies

Remark: Studies were only included in the systematic review if they met all inclusion criteria. In total, 53 articles met the eligibility criteria.



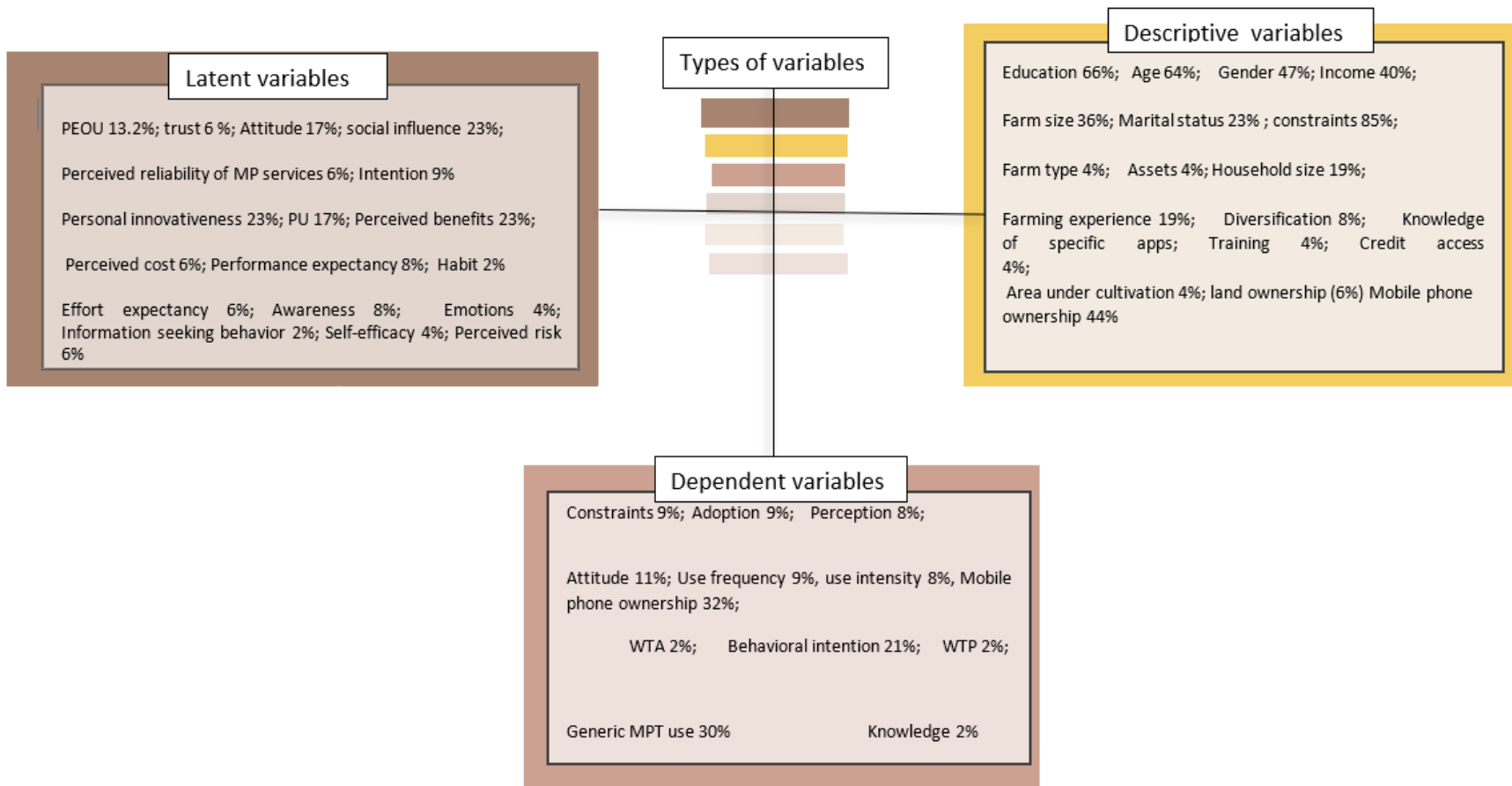


Figure 2: Type of variables

Note: Intention and attitude were reported as latent variables predicting MPT adoption and an outcome measure (dependent variable). Likewise, mobile phone ownership and constraints were both descriptive and dependent variables. Percentages for the variables in Figure 2 were calculated as a percentage of the number of included studies that measured them divided by the total number of articles included.

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