

1 *Technical Note*

## 2 **AgroTutor: A Mobile Phone Application Supporting** 3 **Agricultural Sustainable Intensification**

4 **Juan Carlos Laso Bayas<sup>1\*</sup>, Andrea Gardeazabal<sup>2</sup>, Mathias Karner<sup>1</sup>, Christian Folberth<sup>1</sup>, Luis**  
5 **Vargas<sup>2</sup>, Rastislav Skalský<sup>1,3</sup>, Juraj Balkovič<sup>1,4</sup>, Anto Subash<sup>1</sup>, Moemen Saad<sup>1</sup>, Sylvain Delerce<sup>5</sup>,**  
6 **Jesús Crespo Cuaresma<sup>6,7,8,9</sup>, Jaroslava Hlouskova<sup>1,10,11</sup>, Janet Molina-Maturano<sup>1,12</sup>, Linda See<sup>1</sup>,**  
7 **Steffen Fritz<sup>1</sup>, Michael Obersteiner<sup>1</sup> and Bram Govaerts<sup>2</sup>**

8 <sup>1</sup> Ecosystems Services and Management Program, International Institute for Applied Systems Analysis  
9 (IIASA), Laxenburg, Austria

10 <sup>2</sup> International Maize and Wheat Improvement Center (CIMMYT), Mexico City, Mexico

11 <sup>3</sup> National Agricultural and Food Centre, Soil Science and Conservation Research Institute, Bratislava, Slovak  
12 Republic

13 <sup>4</sup> Department of Soil Science, Faculty of Natural Sciences, Comenius University in Bratislava, Bratislava,  
14 Slovak Republic

15 <sup>5</sup> Decision and Policy Analysis (DAPA), International Center for Tropical Agriculture (CIAT), Cali, Colombia

16 <sup>6</sup> World Population Program, International Institute for Applied Systems Analysis (IIASA), Laxenburg,  
17 Austria

18 <sup>7</sup> Department of Economics, Vienna University of Economics and Business (WU), Vienna, Austria

19 <sup>8</sup> Wittgenstein Centre for Demography and Global Human Capital (WIC), Vienna, Austria

20 <sup>9</sup> Austrian Institute of Economic Research (WIFO), Vienna, Austria

21 <sup>10</sup> Institute for Advanced Studies, Vienna, Austria

22 <sup>11</sup> Thompson Rivers University, Kamloops, BC, Canada

23 <sup>12</sup> Economics and Management of Natural Resources, Department of Agricultural Economics, Ghent  
24 University, Ghent, Belgium

25  
26 \* Correspondence: [lasobaya@iiasa.ac.at](mailto:lasobaya@iiasa.ac.at);

27 **Abstract:** Traditional agricultural extension services rely on extension workers, especially in  
28 countries with large agricultural areas. In order to increase adoption of sustainable agriculture, the  
29 recommendations given by such services must be adapted to local conditions and be provided in a  
30 timely manner. The AgroTutor mobile application was built to provide highly specific and timely  
31 agricultural recommendations to farmers across Mexico and complement the work of extension  
32 agents. At the same time, AgroTutor provides direct contributions to the United Nations Sustainable  
33 Development Goals, either by advancing their implementation or providing local data systems to  
34 measure and monitor specific indicators such as the proportion of agricultural area under  
35 productive and sustainable agriculture. The application is freely available and allows farmers to  
36 geo-locate and register plots and the crops grown there, using the phone's in-built GPS, or  
37 alternatively, on top of very high-resolution imagery. Once a crop and some basic data such as  
38 planting date and cultivar type have been registered, the app provides targeted information such as  
39 weather, potential and historical yield, financial benchmarking information, data-driven  
40 recommendations as well as commodity price forecasts. Farmers are also encouraged to contribute  
41 in-situ information, e.g., soils, management, and yield data. The information can then be used by  
42 crop models, which, in turn, would send tailored results back to the farmers. Initial feedback from  
43 farmers and extension agents has already improved some of the app's characteristics. More  
44 enhancements are planned for inclusion in the future to increase the app's function as a decision  
45 support tool.

46 **Keywords:** volunteered geographic information; agricultural intensification; sustainability; smart  
47 farming; citizen science; SDGs; decision support tool

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## 49 1. Introduction

50 Sustainable agricultural intensification responds to the need for increased food production and  
51 improved ecosystem services while land availability is limited [1,2]. For agricultural inputs like  
52 fertilizers or pest control methods to be effective, the recommendations for their use need to be  
53 targeted to local conditions. Excessive input use due to generalized agricultural recommendations,  
54 apart from having damaging spillover effects in surrounding ecosystems [3], also increase the costs  
55 incurred by the farmer. Technology and knowledge for making agriculture more effective and  
56 environmentally friendly already exist [4]; however, across the world, farmers with low incomes do  
57 not tend to benefit from agricultural innovations and techniques [5]. Agricultural extension services  
58 are one of the most common responses for promoting and transmitting such knowledge, although in  
59 developing countries, agricultural extension services are usually over-stretched. Widespread mobile  
60 smartphone technology is one potential vehicle for promoting sustainable agricultural intensification,  
61 supporting existing agricultural extension services. The expected expansion of mobile phone use<sup>1</sup>,  
62 especially smartphones<sup>2</sup>, as well as the rise in crowdsourcing [6,7], have paved the way for involving  
63 farmers in a new information-rich economy, providing them data directly, while at the same time  
64 asking them to improve this information through locally-sourced knowledge inputs. For the  
65 International Maize and Wheat Improvement Center (CIMMYT) in Mexico, monitoring, evaluation  
66 and learning efforts are an integral part of innovation systems. Extension agents generate summary  
67 diagnostics in the existing data collection system, but farmers have been requesting feedback that is  
68 more detailed, benchmarking and more continuous access to recommendations than individual  
69 extension agents can provide in the current system. As a result, CIMMYT has identified the need to  
70 develop an openly available mobile app that draws upon the vast experience and information  
71 collected over the years, but also provides some of the functionality that can be found in proprietary  
72 apps. Working with the International Institute for Applied Systems Analysis (IIASA), a prototype  
73 smartphone application called AgroTutor was developed. The application is aimed at providing  
74 specific and timely agricultural recommendations for farmers across Mexico. It allows farmers to  
75 query information for crops at specific locations, providing targeted agricultural and benchmarking  
76 information. Here we present the technology behind AgroTutor and the information provided to its  
77 users, as well as initial feedback provided by farmers and extension agents on the application. We  
78 also briefly discuss how AgroTutor contributes to the implementation and measurement of several  
79 of the United Nations Sustainable Development Goals (SDGs).

## 80 2. The AgroTutor mobile app

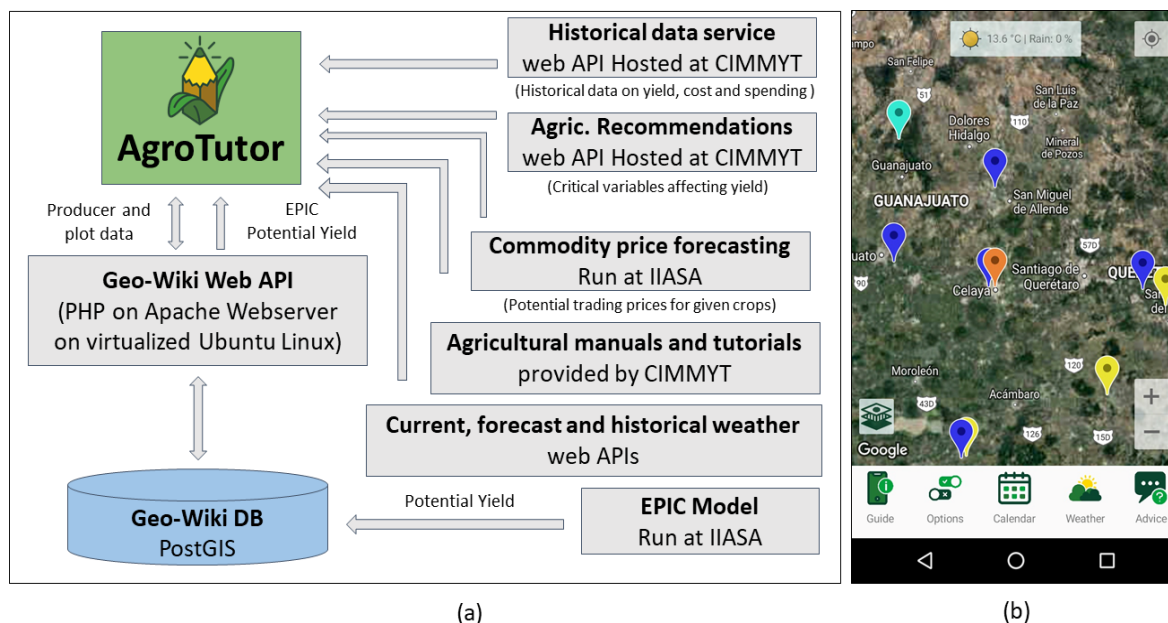
81 Figure 1(a) provides a general overview of the application's current structure including the  
82 different components while Figure 1(b) shows a screenshot of the main landing screen of AgroTutor.

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<sup>1</sup> <https://www.statista.com/statistics/274774/forecast-of-mobile-phone-users-worldwide/>

<sup>2</sup> <https://www.statista.com/statistics/203734/global-smartphone-penetration-per-capita-since-2005/>

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**Figure 1.** (a) AgroTutor data structure with its components and (b) a display of the main screen of the mobile application. Geo-Wiki is the current server at IIASA where data are stored and processed.

87 The current version of the app is built using Xamarin.Forms®, with Android and iOS versions.  
88 For building the architecture of the app, a modularity platform called Prism was chosen with a  
89 “Model – View – View – Model” (MVVM) approach [8]. The app is developed in two languages –  
90 English and Spanish – and allows for easy translation into other languages.

### 91 2.1. Information collected from the farmers

92 AgroTutor relies on the use of geo-location, allowing a farmer to register the position of a given  
93 plot, although a farmer can tap anywhere on the map display to query information or to register a  
94 plot. To store a plot in AgroTutor, a name is required as well as the planting date. If the crop is maize,  
95 the cultivar characteristics are also requested from the farmer (Table 1).  
96

97 **Table 1.** Maturity classes and climate suitability options available in AgroTutor with their  
98 corresponding growing degree days (GDD) / Potential Heat Units (PHU) and base temperature ( $T_{base}$ )  
99 for maize (Adapted from Ruiz et al. (1998) and Capristo et al. (2007)).

Maturity class	PHU/GDD [°C]	Climate suitability	$T_{base}$	$T_{base} + 100$ [°C]
Early	1680	Cold	4	101
Mid-early	1890	Temperate/subtropical	7	102
Intermediate	2100	Tropical	9	103
Mid-late	2310	Hybrid	10	104
Late	2520			105
				106
				107

108 AgroTutor can also store cropping activities for each plot with their specific details, e.g., costs,  
109 dosage, price, etc. A list of available activities in AgroTutor can be found in Supplementary Table 1.

### 110 2.2. Information provided to the farmers

111 When pressing the Advice button (lower right corner, Figure 1(b)) and selecting a location and  
112 a crop of interest, a set of benchmarking information is displayed. The location and crop queried  
113 can then be stored as an agricultural plot, where the benchmarking information previously seen will also

114 be shown but, additionally, all the stored plot information as well as weather information and a new  
 115 menu with several actions will become available. Possible actions include adding cropping activities,  
 116 adding videos and pictures, visualizing the crop calendar with plot activities and delineating the plot.  
 117 The latter can be added using high-resolution imagery (from Google Maps) as a backdrop when  
 118 online but also while offline, once a map layer has been downloaded<sup>3</sup>.

#### 119 2.2.1. Weather information

120 AgroTutor incorporates three types of weather information including: 1) current conditions, 2)  
 121 a 10-day forecast, and 3) historical data. The information is specific to the queried location or field  
 122 plot. The data are obtained from the API of aWhere<sup>4</sup>. The parameters shown include, amongst others,  
 123 precipitation (mm), relative humidity (%), temperature (°C), and solar radiation (Wm<sup>-2</sup>).

#### 124 2.2.2. Historic yield potential

125 Based on the geo-location of the field and associated data (i.e., soil<sup>5</sup>, climate [11], topography [12]  
 126 and growing season [13]), selected representative cultivars based on Ruiz *et al* (1998), and water  
 127 management (rain-fed or fully irrigated) farmers are provided with historic, non-nutrient and pest-  
 128 limited yield potential estimates as a benchmark, derived *a priori* from crop model outputs for the  
 129 time period 1980-2010. The results then provide information about the maximum yield potentially  
 130 attained, but also the variability of these results for the selected location. Crop model simulations  
 131 have been carried out with the well-established field-scale model, Environmental Policy Integrated  
 132 Climate (EPIC) [14], within a spatial computational framework running the model for specific  
 133 climate, soil, topography, and growing seasons in each pixel [15]. The yield potential provided is  
 134 targeted to the farmer's plot locations but is currently limited to maize, although in the future, more  
 135 crops could be considered if enough information is obtained from the field, e.g., using data entered  
 136 by farmers into the application.

#### 137 2.2.3. Benchmarking local information

138 Historical income, costs, profit and yield shown in AgroTutor are based on geo-tagged historical  
 139 data from 197,714 experimental, demonstrative and pioneer farmers' plots across Mexico. A map of  
 140 CIMMYT's support infrastructure is shown in Supplementary Figure 1 [16]. Since the historical data  
 141 are geo-tagged, the information displayed is targeted to registered plots, crops and products,  
 142 allowing the farmer to compare their own production costs, yields and profit against those from  
 143 nearby fields to showcase their chances of improvement using real cases.

#### 144 2.2.4. Windows of opportunity

145 Windows of opportunity are suggested 'best times' for executing agronomic management  
 146 activities on the selected plot, based on known responses of a given cultivar type (Table 1). Currently,  
 147 the system suggests the optimal nitrogen fertilizer split application for maize to the farmer, which  
 148 allows for more efficient fertilizer use by minimizing losses compared to applying the full rate at  
 149 planting [17–19]. This approach typically consists of a smaller start-up fertilizer dose applied at  
 150 planting and a second larger dose once the crop enters a phase of rapid linear biomass accumulation,  
 151 reached once the crop has accumulated 18-25% of the temperature requirement for maturing or

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<sup>3</sup> Layer is extracted from the "Mosaico Nacional" service available at  
<http://WMS.SIAP.gob.mx/MosaicoNacional/>

<sup>4</sup> <https://developer.awhere.com/>

<sup>5</sup> Soil profiles from the "Información Nacional sobre Perfiles de Suelo v1.2" service, produced by the Mexican National Institute for Statistics and Geography (INEGI), available at:  
[http://www3.inegi.org.mx/contenidos/temas/mapas/edafologia/metadatos/ok\\_suelosesp.pdf](http://www3.inegi.org.mx/contenidos/temas/mapas/edafologia/metadatos/ok_suelosesp.pdf)

152 growing degree days (GDD) [20]. The windows of opportunity are then shown to the farmer in the  
153 AgroTutor calendar.

#### 154 2.2.5. Recommended agricultural practices

155 The recommended practices are derived from analyses done by the International Center for  
156 Tropical Agriculture (CIAT-Colombia) on field data already collected by CIMMYT (see  
157 Supplementary Figure 1). Using machine learning algorithms, management practices related to  
158 higher yields are identified and subsequently shown as recommended practices for the area. The  
159 methodology used is based on previous studies on rice [21] and perennial crops [22] where random  
160 forest-based algorithms are used to assess the relevance of a set of predictor variables in explaining  
161 the output yield variability. The most frequent parameters highlighted for optimization are, e.g.,  
162 cultivar sown, levels of fertilization, sowing density and some weather parameters. An example for  
163 the ranking of variables according to their importance as determined by the random forest model in  
164 Guanajuato is shown in Supplementary Figure 2.

#### 165 2.2.6. Commodity price forecasting

166 This module provides farmers with direct information regarding projected prices at trading  
167 hubs for commodities such as wheat and maize, with predictions from 1 month to 12 months in  
168 advance. The method used to obtain forecasts of agricultural commodities is based on the estimation  
169 of an array of multivariate time series models making use of climatic, financial and macroeconomic  
170 variables, as well as market fundamentals corresponding to the particular commodity whose price is  
171 predicted (Crespo Cuaresma et al., 2018; Cuaresma Crespo et al., 2017). The estimated models are  
172 validated making use of out-of-sample forecasting exercises with respect to historical data and both  
173 loss- and profit-based performance measures. Supplementary Figure 3 shows a comparison of actual  
174 and predicted corn price with confidence intervals for the year 2016.

#### 175 2.2.7. Communication, data recording, accessibility, and user experience

176 Farmers can receive relevant messages on an ad hoc but targeted basis, e.g., training sessions in  
177 the area, or messages linked to motivation to participate, e.g., encouraging farmers to add  
178 information to the app. Features that allow farmers to document their plots, crops and practices using  
179 photographs and video are also included in the application. Documentation and messaging should  
180 also encourage farmer-to-farmer and farmer-to-expert discussions similar to that in the Digital Green  
181 project [23]. Since overall user experience and accessibility are crucial for technology uptake, the use  
182 of visual language, i.e., icons with few colors, simplified benchmarking information graphics, and  
183 direct access to information directly from the main map was emphasized during all stages of  
184 development of AgroTutor. Additionally, ensuring that basic smartphones (using the Android  
185 operating system) could employ the app was prioritized, since these are most common in the target  
186 rural agricultural areas. Based on smartphone penetration in Mexico<sup>6</sup>, it was determined that  
187 AgroTutor should be available for Android versions 19 (4.4 KitKat) and later. Finally, since  
188 AgroTutor is the result of a non-profit collaboration, all information provided, including  
189 benchmarking and geo-located points of service, is free of charge.

190

### 191 3. Preliminary tests and farmers willingness to adopt

192 To promote AgroTutor amongst farmers, CIMMYT is preparing a deployment strategy that will  
193 include advertisement strategies as well as incentives for technicians and farmers alike. Nevertheless,  
194 initial feedback was already provided by 27 farmers and field technicians (extension agents). Some  
195 of the comments were recorded in four interviews and a workshop lead by a co-author (JMM) that  
196 took place in the CIMMYT offices in Guanajuato on February (n=10), while others were sent back via

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<sup>6</sup> <http://gs.statcounter.com/android-version-market-share/mobile-tablet/mexico/#monthly-201904-201904-bar>

197 email after downloading the application (n=5) in September 2019, and the rest (n=12) were recorded  
 198 on a two-day visit by CIMMYT personnel to farmers in 2 areas: Moroleón and Purísima del Rincón,  
 199 in the state of Guanajuato, Mexico in October 2019. The feedback was provided for earlier iterations  
 200 of the application, so some of the issues mentioned were already incorporated into AgroTutor, e.g.,  
 201 weather inconsistencies, app not responsive, bugs, app crashing, map not loading, etc. IIASA and  
 202 CIMMYT are currently updating the application and including new features based on the feedback  
 203 provided. Table 2 summarizes the comments received.

204 **Table 2.** Direct feedback for AgroTutor received from farmers and field technicians in Guanajuato

Positive characteristics	Challenges and suggestions
<ul style="list-style-type: none"> <li>• Nice interface</li> <li>• Good design</li> <li>• Friendly icons/design</li> <li>• Great characteristics such as windows of opportunity, weather, parcel delineation and registration, producer profile, benchmarking information</li> <li>• Would highly recommend it</li> <li>• Good that no individual farmer characteristics or information are shown</li> <li>• Very useful for programming fertilization</li> <li>• High useful for programming irrigation</li> </ul>	<ul style="list-style-type: none"> <li>• Loading map problems and storing parcel problems</li> <li>• Location problems, cascading problems - weather, benchmarking</li> <li>• Historical weather problems /high data load requested</li> <li>• Some activity fields are repetitive, some units are not fully matching field measurements (e.g., kg/bags)</li> <li>• Too broad commercialization activity concept: Allow individual activities, e.g., transport, contract agriculture.</li> <li>• Pest and disease activity: Allow for broader set of options, not only insecticide or fungicide</li> <li>• Some bugs and crashes happening with benchmarking data. Slow rendering sometimes. Slow loading.</li> <li>• General flow and look: to fill in forms use an arrow rather than Enter. Add font size change (older farmers)</li> <li>• Share back with testers the changes or modifications made</li> <li>• Calendar - synchronize with telephone calendar for alerts</li> <li>• Allow the possibility of downloading reports</li> <li>• Employ more customization to engage user - to welcome the producer</li> <li>• Add guides to identify weeds, pests, and nutritional deficiencies - pictures/labels</li> <li>• Add tips section</li> <li>• Add activity: Soil identification (characteristics)</li> <li>• Add fruit trees</li> <li>• Maturity type is confusing. Better explanation needed</li> <li>• Allow exporting parcel location - map with coordinates</li> <li>• Add selling points with local prices</li> <li>• Add a calculator allowing conversion of units and transformations</li> </ul>

205 Additionally, to investigate the farmers' willingness to adopt the AgroTutor app, a survey based  
 206 on the Unified Theory of Acceptance and Use of Technology (UTAUT) [24–26] was conducted as part

207 of a separate study. Although the full set of results is soon to be published, preliminary findings  
 208 show that intentions to adopt the app were predicted by the app's perceived usefulness, how farmers  
 209 believe that technical infrastructure exists and by the expectation of the farmers using the app to  
 210 acquire new knowledge. These findings are useful for providing relevant feedback from the field to  
 211 the app designers and developers as well as providing additional insights into ways of encouraging  
 212 adoption.

#### 213 **4. AgroTutor and the Sustainable Development Goals: Current and potential contributions**

214 AgroTutor was built to promote sustainable intensification of agricultural production,  
 215 contributing directly to achieve food security, end hunger and to improve the nutritional status of  
 216 farmers while at the same time increasing productivity of the land. The recommendations provided  
 217 by the app on the use of pest and disease control, which help to ensure overuse of agrochemicals does  
 218 not happen, will protect and regenerate the soil, and avoid contamination of the water table, restoring  
 219 peri-agricultural ecosystems and protecting the surrounding environment. Recently, AgroTutor was  
 220 mapped as a citizen science project that could potentially contribute to two SDG Tier III indicators  
 221 [27], i.e., 2.4.1 Proportion of agricultural area under productive and sustainable agriculture, and 12.3.1  
 222 (b) Food waste index<sup>7</sup>. Recommendations of harvest and post-harvest management, which are being  
 223 incorporated into the app and which are already accessible via links to CIMMYT's manuals and best  
 224 practices, will contribute to avoidance of food waste and enhance financial gains for smallholder  
 225 farmers.

226 Furthermore, AgroTutor has the potential to contribute to other SDGs as the range of services is  
 227 expanded and refined based on stakeholder feedback in an iterative process. For example, since the  
 228 app already allows for the delineation of parcels, it can potentially be adapted to contribute to  
 229 initiatives such as FAO's Voluntary Guidelines for the Responsible Governance of Tenure [28]. This  
 230 would support a crowdsourcing approach for collecting tenure-related details by farmer  
 231 communities, thereby contributing to the monitoring of indicator 5.a (b) Share of women among  
 232 owners or rights-bearers of agricultural land, by type of tenure [27]. Similarly, these features, together  
 233 with farmer characteristics, can also potentially support the indicator 4.4.1 Proportion of youth and  
 234 adults with information and communications technology (ICT) skills [27]. The youth participation in  
 235 agriculture is a pressing challenge worldwide and specifically in Guanajuato, Mexico, where the pilot  
 236 test of the app showed an average farmer age of 65, whereas close to 40% of producers in Mexico are  
 237 between 46 and 60 years old [29].

238 By providing free access to high-quality recommendations to small farmers who cannot afford  
 239 consultancy services, thus helping them to reduce costs and increase profits, AgroTutor provides a  
 240 direct contribution to ending poverty (SDG 1). Also, the financial recommendations, such as expected  
 241 commodity price forecasts, have the potential to empower farmers, especially female farmers, who  
 242 may exploit this information to achieve higher profits for the sale of their produce and thereby  
 243 enhance their self-reliance and the budget management of their families. Finally, the provision of  
 244 environmentally friendly recommendations, e.g., guidance on how to perform conservation  
 245 agricultural practices that store soil moisture, will help farmers to cope better with impacts from  
 246 climate change such as recurring droughts, thereby strengthening their resilience and adaptive  
 247 capacity to climate-related hazards and natural disasters (SDG 13, Target 13.1). Supplementary Table  
 248 2 provides a summary of the actual and potential contributions of AgroTutor to the SDGs.

#### 249 **5. AgroTutor and crowdsourcing with small farmers**

250 This technical note has outlined AgroTutor, a mobile application developed to support  
 251 sustainable agricultural intensification. The application was built with the understanding that small  
 252 and medium farmers require decision support tools to strive for sustainable intensification that  
 253 translates into higher profits for their families while having the lowest possible environmental

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<sup>7</sup> Note that since the publication of [27], Tier III indicators no longer exist and both 2.4.1 and 12.3.1(b) are now Tier II indicators.

254 impact. Deployment of the app is currently under way in Guanajuato, Mexico, but a refinement of  
 255 features as well as an understanding of how much information farmers can contribute, will depend  
 256 on the additional feedback elicited by CIMMYT technicians after some seasons of testing in the field.

257 Even though the potential of citizen science data for monitoring the SDG indicators is being  
 258 recognized [27], potential synergies and trade-offs arise when focusing on small farmers and  
 259 agriculture [30]. When dealing with smallholders, satisfaction of basic needs is a priority, which may  
 260 sometimes put them at odds with being promoters of sustainable development [31]. Moreover, it is  
 261 very important to maintain privacy standards for crowdsourcing applications, especially in  
 262 agriculture, so as to avoid potential issues or additional barriers that may hinder participation [32] or  
 263 even set participants at risk by revealing critical information to the public. Hence, citizen science  
 264 applications should not only focus on obtaining the data but in managing it fairly and not harm nor  
 265 infringe upon any of the SDGs. In the best case, these applications should support smallholder basic  
 266 needs and promote inclusiveness. The design of suitable, low-cost and simple technological solutions,  
 267 which leaves no one behind, should be the core of agricultural citizen science projects and apps such  
 268 as AgroTutor.  
 269

270 **Supplementary Materials:** Table S1: List of agronomic management tasks and activities currently available in  
 271 AgroTutor, Table S2: Potential and current contributions of AgroTutor to SDGs, Figure S1: CIMMYT support  
 272 infrastructure across Mexico, showing research platforms, modules, extension and impact areas. More  
 273 information at <http://bem.cimmyt.org/Inicio/Default.aspx> (in Spanish), Figure S2: Output ranking of variable  
 274 importance from the random forest model in Guanajuato - rainfed maize, Figure S3: Actual and forecasted corn  
 275 prices with 1 and 2 standard deviations upper and lower (UCL, LCL) confidence intervals. Model forecasts  
 276 shown for the year 2016.

277 **Author Contributions:** Juan Carlos Laso Bayas, Andrea Gardeazabal, Bram Govaerts, Luis Vargas, Michael  
 278 Obersteiner and Steffen Fritz conceptualized the research; Mathias Karner, Luis Vargas and Christian Folberth  
 279 organized the database; Mathias Karner, Moemen Saad and Anto Subash developed the application; Christian  
 280 Folberth, Rastislav Skalský, Juraj Balkovič, Sylvain Delerce, Jesús Crespo Cuaresma and Jaroslava Hlouskova  
 281 contributed with data and analysis for the different app modules. Janet Molina-Maturano and Luis Vargas  
 282 contributed with the field feedback summaries. Janet Molina and Linda See contributed with additional insights  
 283 into the SDG contributions of AgroTutor. Linda See edited the manuscript. All authors wrote different sections  
 284 of the manuscript and contributed to manuscript revision. All authors read and approved the submitted version.

285 **Funding:** The app development by IIASA was funded by the CGIAR Research Programs on Wheat and Maize,  
 286 which includes contributions from Australia, the Bill & Melinda Gates Foundation, Canada, France, India, Japan,  
 287 Korea, New Zealand, Norway, Sweden, Switzerland, the United Kingdom, the World Bank, funding agencies of  
 288 Australia, the United Kingdom (DFID), USA (USAID), and China. Additionally, the conceptualization and  
 289 testing process was part of MasAgro Productor and MasAgro Guanajuato, supported by the SAGARPA  
 290 (Secretaría de Agricultura, Ganadería, Desarrollo Rural, Pesca y Alimentación de Mexico) and the state  
 291 government of Guanajuato. Any opinions, findings, conclusion, or recommendations expressed in this  
 292 publication are those of the authors and do not necessarily reflect the view of the donors mentioned previously.

293 **Acknowledgments:** This work was possible thanks to the support of Erick Ortiz from CIMMYT and his field  
 294 team who helped to conceptualize and test AgroTutor in Mexico. We would like to thank local Mexican farmers  
 295 who are currently helping us by testing the application and providing feedback. Thanks to Dr. Nele Verhulst for  
 296 her valuable insights, edits and recommendations on the manuscript. Thanks to Alicia Speratti for further editing  
 297 the paper. We would also like to acknowledge the anonymous reviewers that greatly contributed to the  
 298 manuscript improvement.

299 **Conflicts of Interest:** The authors declare that the research was conducted in the absence of any commercial or  
 300 financial relationships that could be construed as a potential conflict of interest.

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