

# Optical sensors for precision agriculture: a new look

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## ABSTRACT

The increasing human population added to the rural exodus has aggravated the pressure in the agricultural sector for greater production. Faced with this problem, research has increased to develop optical sensors for agriculture with the purpose of minimizing the effects of rural exodus, obtaining rapid information and promoting the rational use of natural resources. Optical sensors have a differential consisting of the ability to use the spectral signature of an attribute or part of it to gain information, often not obvious. This review provides recent advances in optical sensors as well as future challenges. The studies have shown the wide range of applicability of optical sensors in agriculture, from detection of weeds to identification of soil fertility, which favors management in different areas of agriculture. The main limitation of optical sensors used in agriculture in most of the world has been the cost of purchasing the devices, especially in poor countries, so one of the future challenges is the reduction of final prices paid by consumers.

**Keywords:** Smartphone; weed; hydric stress; pathogen detection; soil fertility.

## 1. INTRODUCTION

The growth of the world population implies an increase in food demand. With natural resources, such as limited freshwater and fertilizers, the implementation of initiatives aimed at incrementing a productive and efficient use of natural resources is needed. In this context, several scientific efforts have been made to multiply agricultural production. The sensor-based information system is one of these efforts, being one of the bases of precision agriculture and of fundamental applicability for agricultural monitoring and decision making oriented towards greater production and efficiency [1].

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38 For precision farming, knowledge about soil attributes, the health of developing plants and  
39 the quality of fruits and grains harvested are extremely important. In view of this, several  
40 types of sensors have been researched and developed, either to monitor soil attributes such  
41 as moisture, salinity, conductivity and fertility; monitor environmental conditions such as  
42 precipitation, solar radiation and relative humidity; or monitor plant attributes such as  
43 chlorophyll content, nitrogen requirement, water stress, among others [2].

44 Among the different types of sensors, optical sensors have a differential aspect compared to  
45 others, which is the ability to use the spectral signature of an attribute or part of it. To do this,  
46 every optical sensor has the ability to measure reflectance or use the reflectance property  
47 for information. This ability to differentiate, for example, the state of a normal plant from one  
48 with some problem, be it water deficit or lack of some nutrient, such as nitrogen [3].

49 Thus, to carry out the present study, we undertook a bibliographic review aiming at to seek  
50 for the uses of optical sensors in precision agriculture, presenting future advances and  
51 challenges.

## 52 **2. MATERIAL AND METHODS**

53

54 The method proposed for this study was based on the review of publications related to the  
55 applicability of optical sensors in precision agriculture, presenting future advances and  
56 challenges in the exploration of agricultural activity in a global way. According to the  
57 objective of the study, the review comprised five stages: i) establishment of the theme and  
58 selection of the research question; ii) establishment of inclusion and exclusion criteria; iii)  
59 definition of the information to be extracted from the selected articles; iv) analysis and  
60 interpretation of results and v) presentation of knowledge review and synthesis. Considering  
61 the specificity of the topic, the methodology used and the main results were used as  
62 parameters for the definition of the information to be extracted from the selected  
63 publications.

64 The inclusion criteria of the papers used were: publications between 2003 and 2018, which  
65 portrayed the subject matter of global use in agriculture; and that addressed the key words  
66 and expressions like smartphone, weed control, water stress, pathogen detection and soil  
67 fertility.

68 For the analysis of the data, a thorough reading of the selected papers was carried out, in  
69 order to verify the adherence and consistency to the focus of this research. The ideas were  
70 grouped by similarity so as to compose a narrative synthesis of the results and discussion of  
71 the information related to the study.

## 72 **3. RESULTS AND DISCUSSION**

73

74

### **Applicability of optical sensors in agriculture**

#### **Irrigation**

76 The scarcity of water in various areas of the world and the increase in the cost of use leads  
77 to the need for proper use of this. Therefore, knowing the right moment to irrigate and  
78 quantity is grounded for rational use.

79 The use of optical sensors such as thermographic, multispectral and hyperspectral cameras  
80 is being studied by many researchers to monitor the canopy, identify water stress of plants  
81 and estimate the stomatal conduct to assist in irrigation planning.

82 The use of thermal imaging obtained by thermographic camera, was evaluated by [4] as a  
83 potential for irrigation management by serving as a water stress indicator for a commercial  
84 42ha orchard located in Murcia, Spain. The results showed that thermal imaging is a  
85 valuable tool for decision making regarding the timing of orchard irrigation.

86 In this perspective, [5] evaluated the use of thermal imagery to assess water stress in  
87 soybean and cotton crops in Texas, USA. [6] studied the use of thermographic camera for  
88 the detection of water stress in citrus and persimmon trees in Valencia, Spain. [7] evaluated  
89 the use of a thermal camera to determine water stress in vines in the town of Lleda, Spain.  
90 [8] studied the use of VANT to detect water stress in orange and tangerine cultivars using  
91 hyperspectral and thermal images in Seville, Spain. These mentioned papers have allowed  
92 to draw the conclusion that the use of thermal imaging is an efficient tool to identify the water  
93 stress of crops and guide the management of irrigation

94 Multispectral cameras and thermal cameras on board unmanned aerial vehicles (UAV) were  
95 used by [9] in the cultivation of apple trees for the detection of water stress in the trees.  
96 Captured images allowed water stress to be detected at the individual tree level in order to  
97 allow localized management of irrigation.

98 All the researches show great applicability of multispectral, thermographic and hyperspectral  
99 cameras to identify plants in water stress. To achieve this result, complex image processing  
100 was developed and good performance computers were required.

101 These studies must be improved so that they can get into the hands of producers, since the  
102 results are still dependent on the laboratory environment.

### 103 **Management of nitrogen fertilization**

104 The chlorophyll is the most important pigment of the leaf and some of the most important of  
105 the plant, since it is through the manages to capture the sunlight and to use it as energy  
106 source. In order to quantify the amount of chlorophyll in the leaf, it is possible to estimate the  
107 lack of sufficient amount of nitrogen in the plant, indicating the need for nitrogen fertilization  
108 or not [10].

109 Nitrogen is one of the most influential nutrients in plant development, being a limiting  
110 element of production. Due to this characteristic, it is intensively used in productive crops,  
111 aiming the crop to reach its maximum potential [11]. However, if used in excess of the cost  
112 of production also leads to contamination of water resources due to leaching and  
113 evaporation [10].

114 Commercial optical sensors such as the GreenSeeker and Minolta SPAD-502 are based on  
115 NIR and SPAD Analysis of Soil Plants. With the NDVI, as measured by GreenSeeker, it is  
116 possible to estimate the nitrogen fertilization for the crop according to the desired  
117 productivity, with the SPAD as measured by the Minolta SPAD-502, the amount of  
118 chlorophyll in the plant is estimated and thus it is possible to identify the state of health, as  
119 well as to recommend nitrogen fertilization.

120 Yara N-Sensor is another sensor also used for nitrogen fertilization. It is based on spectral  
121 reflection in specific bands related to the chlorophyll and biomass content of the cultures.

122 The CropCircle optical sensor makes readings of up to 6 spectral bands covering blue,  
123 green, red, near-red and near-infrared. With the combination of these bands it is possible to  
124 estimate different vegetation indices, among them NDVI [12].

125 [11] have constructed to prototype optical sensor to measure NDVI aiming at low production  
126 cost. They set up an experiment with corn and wheat to verify the calibration and  
127 performance of the prototype with the GreenSeeker commercial sensor. Their results  
128 showed that prototype is a useful sensor to measure NDVI and by means of this estimate of  
129 nitrogen fertilization. The performance and accuracy are lower than the GreenSeeker, due to  
130 the low cost of the prototype, but it does not disturb the farmer who uses it.

131 [13] and [14] have developed very similar surveys with commercial geraniums. They verified  
132 the performance of the GreenSeeker and Minolta SPAD-502 sensors in the identification of  
133 nitrogen concentration in two geranium cultivars. NDVI and SPAD measures are possible to  
134 identify changes in the nitrogen concentration state, but they point out that research must  
135 correlate these variations with the necessary dose of nitrogen to be applied in the  
136 geraniums.

137 [15] used the CropCircle sensor to measure the Meris Terrestrial chlorophyll index (MTCI) of  
138 corn crops in Brazil submitted to different treatments of nitrogen fertilization. With the MTCI  
139 data and the correlation with the nitrogen dose used in each treatment, they created an  
140 algorithm to estimate the application rate of nitrogen in corn.

141 [16] evaluated the performance of the NDVI sensor prototype developed by [11] and the  
142 SPAD-502 sensor in the identification of the nitrogen concentration in Gaillardia. The results  
143 indicate that both sensors can be used to identify the nitrogen concentration of this flower,  
144 as long as the sampling time is not short. [16] point out that in order to develop fertilization  
145 guidelines it is necessary to investigate further the different production practices and  
146 additional cultivars with the measured NDVI and SPAD values.

147 The studies indicate that there is a field of research to develop algorithms that estimate the  
148 nitrogen dose to be applied in different commercial cultivars according to the value of SPAD  
149 or NDVI measured, or other index. GreenSeeker, for example, uses algorithm that  
150 recommends only dose to be applied to grains. Therefore, there are a variety of agricultural  
151 species still to be studied.

## 152 **Chemical properties of soil**

153 Studies show that the number of ions in the soil and organic matter affect the reflectance,  
154 absorption or transmittance of electromagnetic waves by the soil. This fact may be  
155 interesting for the use of optical sensors as a measure of soil chemical properties [17].

156 [18] used a mobile NIR spectrophotometer to map the surface layer of organic farms and to  
157 study the correlation among the spectral data with the results of the laboratory analysis for  
158 P, K, Mg, soil organic matter (OM), N and pH. For the local calibrations, the best results  
159 were pH, N-total, MO, K-total and Mg-total, with  $r^2$ : 0.71, 0.69, 0.61, 0.55, 0.53, respectively;  
160 therefore, showing correlation between NIR spectral data of the soil with the chemical  
161 properties of this soil. However, they concluded that the correlation between the spectra and  
162 the parameters was location dependent, and this would make it difficult to develop general  
163 calibration models.

164 [19] developed a prototype using NIR spectrophotometer to map soil reflectance and  
165 correlate with chemical parameters. The results of an initial study indicated that the locally  
166 weighted regression analysis was able to predict moisture, C-total, N-total and pH, with  $r^2$ :  
167 0.82, 0.87, 0.86 and 0.72, respectively. The experimental unit produced data with a high  
168 level of repeatability, thus showing soil patterns related to NIR spectral reflectance.

169 **Detection of pathogens in plants**

170 Studies in the literature show that plants after being attacked by pathogens suffer damage  
171 that causes changes in the rate of transpiration and flow of water throughout the plant or in  
172 organs. This leads to increased temperature in localized parts of the plant, such as leaves  
173 [20, 21].

174 [22] studied the applicability of the multispectral camera and thermographic camera for the  
175 detection of Huanglongbing disease in citrus trees. The experiment was carried out in the  
176 experimental field of citrus of the University of Florida, USA. Their results conclude that  
177 using the band of the visible and thermal infrared as input characteristics, the overall  
178 average classification accuracy of 87%, with 89% specificity and 85% sensitivity, could be  
179 achieved to classify trees with leaves infected by Huanglongbing. The support vector  
180 machine model was used for identification.

181 [23] used a multispectral camera coupled to UAV to diagnose citrus trees affected by  
182 Greening's disease, based on spectroscopy. For this, the data generated from the  
183 processing of six spectral bands and seven vegetation indices derived from these bands,  
184 among them the NIR / R (near infrared / red), were used in the classification algorithm.  
185 Among the indexes analyzed, NIR / R showed a better significant difference between  
186 healthy trees and infected plants. The authors concluded that the processing of multispectral  
187 images taken at low altitudes is reliable in the detection of Greening disease (the  
188 classification reached an accuracy of 85%), being a tool that could reduce the production  
189 costs of the citrus crop due to the rapid identification of the disease.

190 **Apps for smartphone**

191 Smartphones are a device that in addition to presenting a fast processing system also a  
192 camera feature, being an interesting platform for image processing. In light of this, work has  
193 been developed using the images captured by the RGB camera to create applications for  
194 precision agriculture.

195 [10] created an application called SmartSPAD responsible for estimating the SPAD of corn  
196 plants by means of contact image obtained by the camera of smartphones. Its application is  
197 based on two models of SPAD prediction from the corn leaf image: neural network model,  
198 and the multivariate linear model. For the validation of SmartSPAD, the SPAD values  
199 measured by it were compared with the SPAD values measured by the Minolta SPAD-502  
200 device, used as standard. The validation  $r^2$  values were 0.88 and 0.72 and the mean square  
201 error was 4.03 and 5.96 for neural network and linear model, respectively. The application  
202 proved to be a good estimator of SPAD values at a low cost.

203 [24] have created a ground classification sensor based on smartphone application. The  
204 sensor is formed by external optical support and a smartphone application. The support is  
205 formed of two external lenses and a shading cover, since the classification application is  
206 based on the linear discriminant analysis model. The Munsell color card was used as the soil  
207 classification standard. The results reached by the authors show that the sensor had hits  
208 above 90% for all soil samples evaluated.

209 A similar research to the work of [24] was also developed by [25]. The latter authors also  
210 applied an application for Android smartphones with the aim of classifying **soil** in relation to  
211 Munsell color card through RGB images. Their results were obtained in controlled lighting  
212 environments and showed that the ratings by the application were good and acceptable in a  
213 controlled lighting environment.

214

## Future Challenges Regarding Optical Sensors

215 The maximum nitrogen fixation by plants, in the traditional form of fertilization, is around  
216 50%, with the world average being 33%. This is due to several factors, either by leaching,  
217 evaporation and / or plant losses [11] Thus, of all the nitrogen fertilization used in the world  
218 for agricultural production, an average of 67% is wasted.

219 The use of commercial optical sensors with GreenSeeker, Yara N-Sensor, CropCircle and  
220 SPAD-502 promotes improved fixation rate, but these sensors are expensive and not very  
221 accessible to many farmers, especially in developing countries. These countries correspond  
222 to about 70% of the nitrogen consumption for fertilization in the world [11].

223 According to [26], it is realistic to expect crops on the farms of the future to be managed  
224 plant by plant. This approach will require the collection and analysis of massive data on a  
225 scale not considered today and the need for stationary or mobile sensors that can measure  
226 individual plant characteristics in real time.

227 Real-time point-to-point sampling is possible today but at a very high cost. And due to cost,  
228 sampling in a productive area is done with few points, which decreases the accuracy of the  
229 final result, and inefficient becomes the whole set.

230 The acquisition cost of a thermographic, multispectral and hyperspectral camera is high,  
231 especially in countries not benefited by the local currency. This makes it difficult for many  
232 research centers around the world to reduce research development in many areas that could  
233 leverage technology to improve their research and make new discoveries [21].

234 Given the current context, it will be future challenges to develop low-cost optical sensors and  
235 make them as accessible as possible to the producer and the research centers. That these  
236 sensors promote the improvement of the nitrogen fixation in different agricultural crops and  
237 that they can monitor in real time the plant or the homogeneous set of these, facilitating the  
238 management at the varied rate.

239 Another challenge will be to develop optical sensors that all steps of image capture,  
240 processing and final result take place on the same equipment. This will facilitate the  
241 immersion of this technology in the field.

242 Table 1 presents a summary table of the most studied research fields with emphasis on the  
243 use of optical sensors for the monitoring of agricultural crops and agricultural processes.

244 **Table 1. More developed research to study the use of optical sensors for the**  
245 **monitoring of crops and agricultural processes**

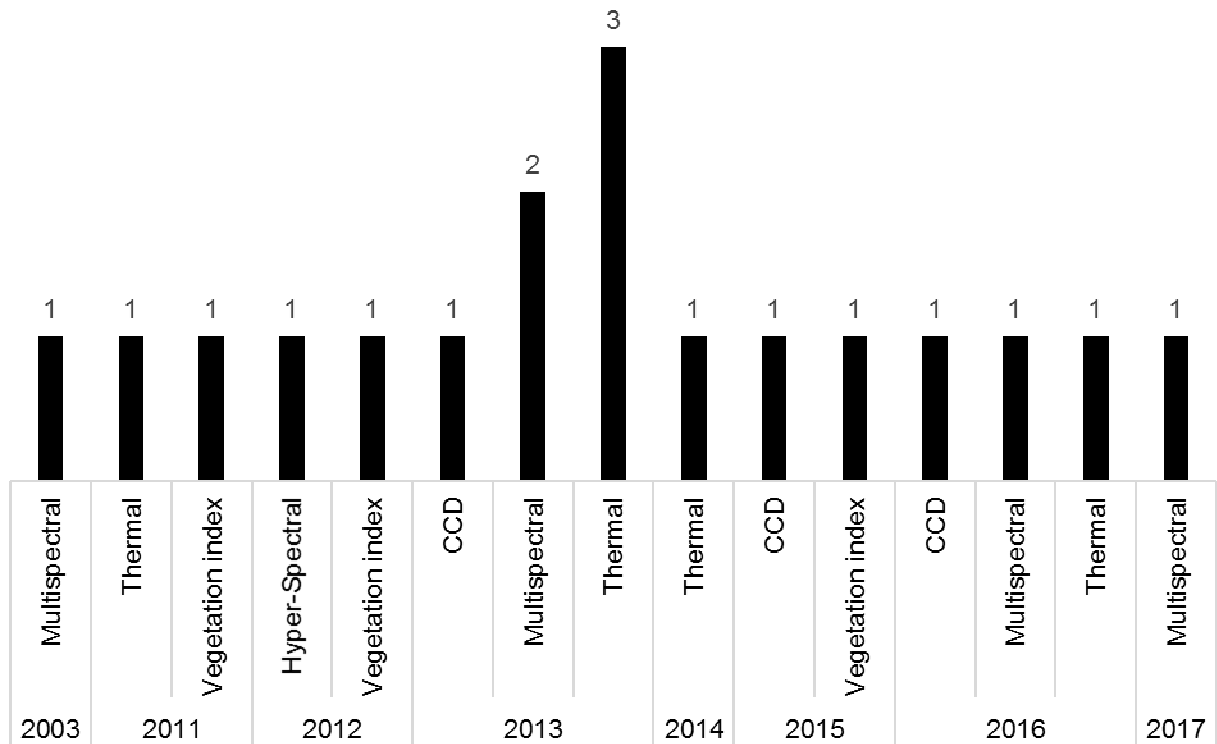
Country	Product	Optical sensor feature	Reference
Spain	Water stress in almond, apricot, peach, lemon and orange	Thermal	[4]
USA	Water stress in cotton	Thermal	[5]
Spain	Water stress on persimmon and citrus trees	Thermal	[6]

Spain	Water stress in the vine	Thermal	[7]
Spain	Water stress in orange and tangerine feet	Hyper-Spectral and thermal	[8]
France	Water stress in apple trees	Thermal	[9]
USA	SPAD reading application	CCD	[10]
USA/ Mexico	Management of nitrogen fertilization in maize	NDVI reader	[11]
Brazil	Management of nitrogen fertilization in maize	MTCI Reader	[15]
USA	Management of nitrogen fertilization in Gaillardia	NDVI/SPAD Reader	[16]
USA	Chemical properties of soil	Multispectral NIR	[18]
USA	Chemical properties of soil	Multispectral NIR	[19]
USA	Huanglongbing on citrus trees	Thermal	[22]
USA	Huanglongbing on citrus trees	Multispectral	[23]
China	Application to sort soil	CCD/lenses	[24]
Spain	Application to sort soil	CCD	[25]
Greece	Identification of <i>Silybum marianum</i>	Multispectral	[27]
Spain	Identification of <i>Sorghum halepense</i>	Multispectral and RGB	[28]

246

247 In analyzing Table 1, as well as the various literature cited in this study, it is noteworthy that  
 248 the USA followed by Spain is the country that presents the most published study on the use  
 249 of optical sensors in various areas of agriculture, from identification of the soil chemical  
 250 properties, as well as classification, identification of diseases.

251 Regarding the period of publication, of the total of works analyzed, one was published in  
 252 2003, two in 2011, two in 2012, six in 2013, one in 2014, two in 2015, three in 2016 and one  
 253 in 2017. In 33.3% were thermal, 5.6% hyperspectral, 16.7% charge-coupled device (CCD),  
 254 27.8% multispectral and 16.7% studied reading sensors of vegetation indices. The year and  
 255 type of publication are shown in Figure 1.



256

257 **Fig. 1. Number of publications by type and year.**

258 **4. CONCLUSION**

259 The studies developed and presented show the great applicability of optical sensors as a  
 260 precision agriculture tool from identification of water stress and weeds to nitrogen fertilization  
 261 management in crops.

262 The main limitation of an optical sensor used in agriculture in most parts of the world is the  
 263 cost of purchasing the devices, especially in poor countries where agriculture is the basis of  
 264 the economy. therefore, a future challenge will be to develop efficient sensors with low  
 265 acquisition cost.

266 Image processing for precision farming is a very effective information method, however, the  
 267 results are not immediate and you need a computer that performs well to get them.  
 268 smartphones have combined processor and camera into one device. due to this feature, the  
 269 smartphone has proven to be very useful for digital image processing. the trend is for  
 270 processing to become better, given that every day better smartphones, in terms of processor  
 271 and camera are launched with cost-effectiveness.

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