

Optical sensors for precision agriculture: an outlook ~~new look~~

Lucas de Arruda Viana^{1*}, Deborah Campos Tomaz², Rodrigo Nogueira Martins³, Jorge Tadeu Fim Rosas⁴, Fernando Ferreira Lima dos Santos⁵, Marcelo Fagundes Portes⁶

ABSTRACT

The ~~increasing growing~~ 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 ~~more productive~~ 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 ~~to the use~~ of optical sensors ~~used~~ in agriculture in most ~~parts~~ 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~~ ~~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~~ ~~augment~~ 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].

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

Among the different types of sensors, optical sensors have a differential aspect compared to others, which is the ability to use the spectral signature of an attribute or part of it. To do this, every optical sensor has the ability to measure reflectance or use the reflectance property

for information. This ability to differentiate, for example, the state of a normal plant from one with some problem, be it water deficit or lack of some nutrient, such as nitrogen [3].

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

2. MATERIAL AND METHODS

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

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

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

3. RESULTS AND DISCUSSION

3.1. Applicability of optical sensors in agriculture

3.1.1. Irrigation

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

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

The use of thermal imaging obtained by thermographic camera, was evaluated by ~~González-Dugo et al.~~ [4] as a potential for irrigation management by serving as a water stress indicator for a commercial 42_ha orchard located in Murcia, Spain. The results showed that thermal imaging ~~is-was~~ a valuable tool for decision making regarding the timing of orchard irrigation.

In this perspective, ~~O'shaughnessy et al.~~ [5] evaluated the use of thermal imagery to assess water stress in soybean and cotton crops in Texas ~~in~~, USA. ~~Ballester et al.~~ [6] studied the use of thermographic camera for the detection of water stress in citrus and persimmon trees

Formatted: Justified

in Valencia, in Spain. Bellvert et al. [7] evaluated the use of a thermal camera to determine water stress in vines in the town of Lleda in, Spain. Zarco-Tejada et al. [8] studied the use of VANT to detect water stress in orange and tangerine cultivars using hyperspectral and thermal images in Seville in, Spain. These above-mentioned papers have allowed to draw the conclusion that the use of thermal imaging is an efficient tool to identify the water stress of crops and guide the management of irrigation

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

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

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

3.1.2. Management of nitrogen fertilization

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

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

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

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

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

Crain et al. [11] have constructed to-a prototype of optical sensor to measure NDVI aiming at low production cost. They set up an experiment with corn and wheat to verify the calibration and performance of the prototype with the GreenSeeker commercial sensor. Their results showed that the prototype is-was a useful sensor to measure NDVI and by means of this estimate of nitrogen fertilization. The performance and accuracy are lower than those of the

Comment [O1]: The statement is not true may be because it was not well formulated.

Comment [O2]: To read and rewrite, eventually.

Comment [O3]: Give the meaning in full and put the abbreviation in the parentheses

Comment [O4]: Give the meaning in full and put the abbreviation in the parentheses

119 GreenSeeker, due to the low cost of the prototype, but it does not disturb the farmer who
120 uses it.

121 | Wang et al. [13] and Wang et al. [14] have developed very similar surveys with commercial
122 geraniums. They verified the performance of the GreenSeeker and Minolta SPAD-502
123 sensors in the identification of nitrogen concentration in two geranium cultivars. NDVI and
124 SPAD measures are possible to identify changes in the nitrogen concentration state, but
125 | they pointed out that research must correlate these variations with the necessary dose of
126 nitrogen to be applied in the geraniums.

127 | Shiratsuchi et al. [15] has used the CropCircle sensor to measure the Meris Terrestrial
128 chlorophyll index (MTCI) of corn crops in Brazil submitted to different treatments of nitrogen
129 fertilization. With the MTCI data and the correlation with the nitrogen dose used in each
130 treatment, they have created an algorithm to estimate the application rate of nitrogen in corn.
131 Dunn et al. [16] has evaluated the performance of the NDVI sensor prototype developed by
132 Crain et al. [11] and the SPAD-502 sensor in the identification of the nitrogen concentration
133 in Gaillardia. The results indicate that both sensors can be used to identify the nitrogen
134 concentration of this flower, as long as the sampling time is not short. Dunn et al. [16]
135 pointed out that in order to develop fertilization guidelines it is necessary to further
136 investigate further the different production practices and additional cultivars with the
137 measured NDVI and SPAD measured values.
138 The studies indicate that there is a field of research to develop algorithms that estimate the
139 nitrogen dose to be applied in different commercial cultivars according to the value of SPAD
140 or NDVI measured, or other index. GreenSeeker, for example, uses algorithm that
141 recommends only dose to be applied to grains. Therefore, there are a variety of agricultural
142 species still to be studied.

143 **3.1.3. Chemical properties of soil**

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

147 | Schirrmann et al. [18] has used a mobile NIR spectrophotometer to map the surface layer of
148 organic farms and to study the correlation among the spectral data with the results of the
149 laboratory analysis for P, K, Mg, soil organic matter (OM), N and pH. For the local
150 calibrations, the best results were pH, N-total, MO, K-total and Mg-total, with representing
151 0.71, 0.69, 0.61, 0.55, 0.53, respectively; therefore, showing correlation between NIR
152 spectral data of the soil with the chemical properties of this soil. However, they concluded
153 that the correlation between the spectra and the parameters was location dependent, and
154 this would make it difficult to develop general calibration models.

155 | Christy et al. [19] has developed a prototype using NIR spectrophotometer to map soil
156 reflectance and correlate with chemical parameters. The results of an initial study indicated
157 that the locally weighted regression analysis was able to predict moisture, C-total, N-total
158 and pH, with representing 0.82, 0.87, 0.86 and 0.72, respectively. The experimental unit
159 produced data with a high level of repeatability, thus showing soil patterns related to NIR
160 spectral reflectance.

161 **3.1.4. Detection of pathogens in plants**

162 Studies in the literature show that plants after being attacked by pathogens suffer damage
163 that causes changes in the rate of transpiration and flow of water throughout the plant or in

Formatted: Normal, Justified

Formatted: Font: (Default) Helvetica

Formatted: Font: (Default) Helvetica

Formatted: Normal, Justified, Space After:
12 pt

Formatted: Normal, Justified

Formatted: Space Before: 12 pt

164 organs. This leads to increased temperature in localized parts of the plant, such as leaves
165 [20, 21].

166 [Sankaran et al.](#) [22] have studied the applicability of the multispectral camera and
167 thermographic camera for the detection of Huanglongbing disease in citrus trees. The
168 experiment was carried out in the experimental field of citrus of the University of Florida in
169 USA. Their results concluded that using the band of the visible and thermal infrared as input
170 characteristics, the overall average classification accuracy of 87%, with 89% specificity and
171 85% sensitivity, could be achieved to classify trees with leaves infected by Huanglongbing.
172 The support vector machine model was used for identification.

Formatted: Normal, Justified, Space After:
12 pt

Formatted: Font: (Default) Helvetica

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

182 **3.1.4. Apps for smartphone**

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

Comment [O5]: To make sure that the sentence
makes sense.

187 [Vesal et al.](#) [10] created an application called SmartSPAD responsible for estimating the
188 SPAD of corn plants by means of contact image obtained by the camera of smartphones. Its
189 application is based on two models of SPAD prediction from the corn leaf image: neural
190 network model, and the multivariate linear model. For the validation of SmartSPAD, the
191 SPAD values measured by it were compared with the SPAD values measured by the
192 Minolta SPAD-502 device, used as standard. The validation r^2 values were 0.88 and 0.72
193 and the mean square error was 4.03 and 5.96 for neural network and linear model,
194 respectively. The application proved to be a good estimator of SPAD values at a low cost.

Formatted: Normal, Justified, Space After:
12 pt

Formatted: Font: (Default) Helvetica

195 [Han et al.](#) [24] have created a ground classification sensor based on smartphone
196 application. The sensor is formed by external optical support and a smartphone application.
197 The support is formed of two external lenses and a shading cover, since the classification
198 application is based on the linear discriminant analysis model. The Munsell color card was
199 used as the soil classification standard. The results reached by the authors showed
200 the sensor had hits above 90% for all soil samples evaluated.

Formatted: Font: (Default) Helvetica

201 A similar research to the work of [Han et al.](#) [24] was also developed by [Mulla](#) [25]. The latter
202 authors also applied an application for Android smartphones with the aim of classifying **soil**
203 in relation to Munsell color card through RGB images. Their results were obtained in
204 controlled lighting environments and showed that the ratings by the application were good
205 and acceptable in a controlled lighting environment.

206 **3.2. Future Challenges Regarding Optical Sensors**

Formatted: Justified

207 The maximum nitrogen fixation by plants, in the traditional form of fertilization, is around
 208 50%, with the world average being 33%. This is due to several factors, either by leaching,
 209 evaporation and / or plant losses [11]. Thus, of all the nitrogen fertilization used in the world
 210 for agricultural production, an average of 67% is wasted.

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

215 According to [Mulla](#) [26], it is realistic to expect crops ~~on-in~~ the farms ~~of-in~~ the future to be
 216 managed plant by plant. This approach will require the collection and analysis of massive
 217 data on a scale not considered today and the need for stationary or mobile sensors that can
 218 measure individual plant characteristics in real time.

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

222 The acquisition cost of a thermographic, multispectral and hyperspectral camera is high,
 223 especially in countries not benefited by the local currency. This makes it difficult for many
 224 research centers around the world to ~~reduce-carry out~~ research and development in many
 225 areas that could leverage technology to improve their research and make new discoveries
 226 [21].

227 Table 1 presents summarizes most studied research fields with emphasis on the use of
 228 optical sensors for the monitoring of agricultural crops and agricultural processes.

229 **Table 1. More developed research on the use of optical sensors for the monitoring of**
 230 **crops and agricultural processes**

<u>Country</u>	<u>Product</u>	<u>Optical sensor feature</u>	<u>Reference</u>
<u>Spain</u>	<u>Water stress in almond, apricot, peach, lemon and orange</u>	<u>Thermal</u>	<u>[4]</u>
<u>USA</u>	<u>Water stress in cotton</u>	<u>Thermal</u>	<u>[5]</u>
<u>Spain</u>	<u>Water stress on persimmon and citrus trees</u>	<u>Thermal</u>	<u>[6]</u>
<u>Spain</u>	<u>Water stress in the vine</u>	<u>Thermal</u>	<u>[7]</u>
<u>Spain</u>	<u>Water stress in orange and tangerine feet</u>	<u>Hyper-Spectral and thermal</u>	<u>[8]</u>
<u>France</u>	<u>Water stress in apple trees</u>	<u>Thermal</u>	<u>[9]</u>
<u>USA</u>	<u>SPAD reading application</u>	<u>CCD</u>	<u>[10]</u>
<u>USA/ Mexico</u>	<u>Management of nitrogen</u>	<u>NDVI reader</u>	<u>[11]</u>

	<u>fertilization in maize</u>		
<u>Brazil</u>	<u>Management of nitrogen fertilization in maize</u>	<u>MTCI Reader</u>	<u>[15]</u>
<u>USA</u>	<u>Management of nitrogen fertilization in Gaillardia</u>	<u>NDVI/SPAD Reader</u>	<u>[16]</u>
<u>USA</u>	<u>Chemical properties of soil</u>	<u>Multispectral NIR</u>	<u>[18]</u>
<u>USA</u>	<u>Chemical properties of soil</u>	<u>Multispectral NIR</u>	<u>[19]</u>
<u>USA</u>	<u>Huanglongbing on citrus trees</u>	<u>Thermal</u>	<u>[22]</u>
<u>USA</u>	<u>Huanglongbing on citrus trees</u>	<u>Multispectral</u>	<u>[23]</u>
<u>China</u>	<u>Application to sort soil</u>	<u>CCD/lenses</u>	<u>[24]</u>
<u>Spain</u>	<u>Application to sort soil</u>	<u>CCD</u>	<u>[25]</u>
<u>Greece</u>	<u>Identification of <i>Silybum marianum</i></u>	<u>Multispectral</u>	<u>[27]</u>
<u>Spain</u>	<u>Identification of <i>Sorghum halepense</i></u>	<u>Multispectral and RGB</u>	<u>[28]</u>

In analyzing Table 1 as well as the various literature cited in this study, it is noteworthy that the USA followed by Spain are the countries that present the most published study on the use of optical sensors in various areas of agriculture, including the identification of the soil chemical properties, as well as the classification and identification of diseases.

Comment [O6]: Is it true?

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

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

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

~~Table 1. More developed research to study the use of optical sensors for the 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]

Formatted Table
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left, Don't keep with next
Formatted: Left
Formatted: Left
Formatted: Space After: 0 pt, Line spacing: single
Formatted: Left
Formatted: Left
Formatted: Left, Don't keep with next
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left
Formatted: Left

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

Comment [07]: Is it true?

Regarding the period of publication, of the The total of 18 works analyzed published from 2003 to 2017 including, one was published in 2003, two in 2011, two in 2012, six in 2013, one in 2014, two in 2015, three in 2016, and one in 2017. About 33.3% were thermal, 5.6% hyperspectral, 16.7% charge-coupled device (CCD), 27.8% multispectral and 16.7% studied reading sensors of vegetation indices. The year and type of publication are shown in Figure 1.

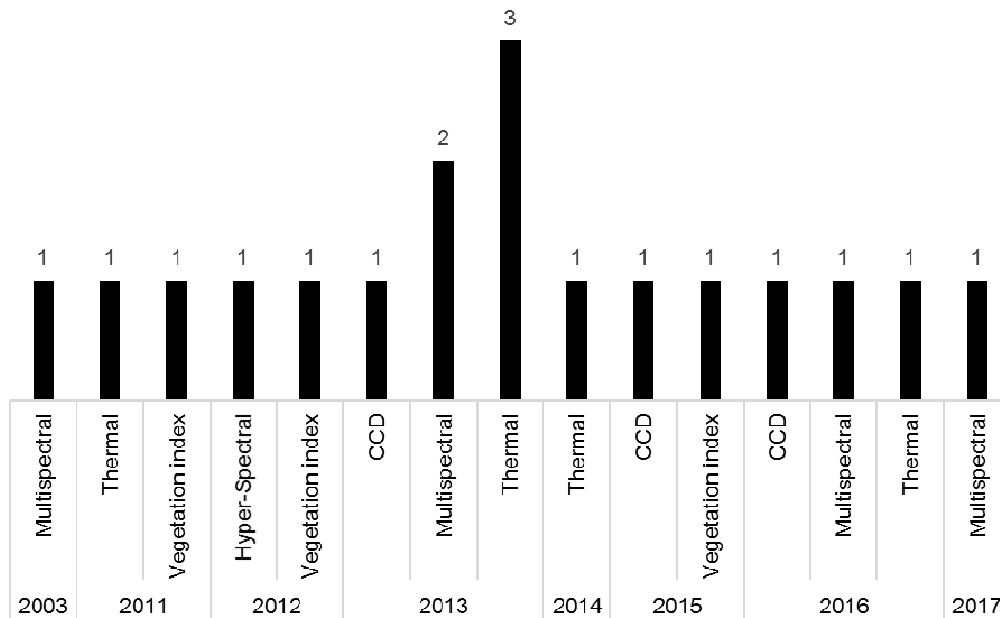


Fig. 1. Number of publications by type and year.

4. CONCLUSION

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

The main limitation to the use of an optical sensor used in agriculture in most parts of the world is the cost of purchasing the devices, especially in poor countries where agriculture is the basis of the economy. Therefore, a future challenge is will the be to develop production of cost effective-efficient sensors, with low acquisition cost.

Image processing for precision farming is a very effective information method, however, the results are not immediate and you need a computer that performs well to get them.

smartphones have combined processor and camera into one device. ~~due~~Due to this feature, the smartphone has proven to be very useful for digital image processing. ~~the~~The trend is for processing to become better, given that every day better smartphones, in terms of processor and camera are launched with cost-effectiveness.

ACKNOWLEDGEMENTS

The authors are grateful to the Federal University of Viçosa - UFV for the academic support and availability of laboratories and the coordination for the improvement of higher level personnel-CAPES for the financial support.

REFERENCES

1. Rehman, A.-U.; Abbasi, A.-Z.; Islam, N.; Shaikh, Z.-A. A review of wireless sensors and networks' applications in agriculture. *Computer Standards & Interfaces*, v.2014,36(2): 263-270, 2014. DOI: <https://doi.org/10.1016/j.csi.2011.03.004>
2. OJHA, T.; MISRA, S.; RAGHUWANSHI, N. S. Wireless sensor networks for agriculture: The state-of-the-art in practice and future challenges. *Computers and Electronics in Agriculture*, v.118: 66-84, 2015. DOI: <https://doi.org/10.1016/j.compag.2015.08.011>
3. LING, C.; LIU, H.; JU, H.; ZHANG, H.; YOU, J.; LI, W. A Study on Spectral Signature Analysis of Wetland Vegetation Based on Ground Imaging Spectrum Data. *Journal of Physics: Conference Series*, v. 910, 8p, 2017. DOI: <https://doi.org/10.1088/1742-6596/910/1/012045>
4. GONZÁLEZ-DUGO, V.; ZARCO-TEJADA, P. J.; NICOLÁS, E.; NORTES, P. A.; ALARCÓN, J. J.; INTRIGLIOLO, D. S.; FERERES, E. Using high resolution UAV thermal imagery to assess the variability in the water status of five fruit tree species within a commercial orchard. *Precision Agriculture*, v. 14(6): 660-678, 2013. DOI: <https://doi.org/10.1007/s11119-013-9322-9>
5. O'SHAUGHNESSY, S.A.; EVETT, S.R.; COLAIZZI, P.D.; HOWELL, T.A. Using radiation thermography and thermometry to evaluate crop water stress in soybean and cotton. *Agricultural Water Management*, v.98(10): 1523-1535, 2011. DOI: <https://doi.org/10.1016/j.agwat.2011.05.005>
6. BALLESTER, C.; JIMÉNEZ-BELLO, M.A.; CASTEL, J.R.; INTRIGLIOLO, D.S. Usefulness of thermography for plant water stress detection in citrus and persimmon trees. *Agricultural and Forest Meteorology*, v.168:120-129, 2013. DOI: <https://doi.org/10.1016/j.agrformet.2012.08.005>
7. BELLVERT, J.; ZARCO-TEJADA, J.; GIRONA J.; FERERES, E. Mapping crop water stress index in a 'Pinot-noir' vineyard: comparing ground measurements with thermal remote sensing imagery from an unmanned aerial vehicle. *Precision Agriculture*, v.15(4): 361-376, 2014. DOI: <https://doi.org/10.1007/s11119-013-9334-5>
8. ZARCO-TEJADA P.J.; GONZÁLEZ-DUGO, V.; BERNI, J.A.J. Fluorescence, temperature and narrow-band indices acquired from a UAV platform for water stress detection using a micro-hyperspectral imager and a thermal camera. *Remote Sensing of Environment*, v.117: 322-337, 2012. DOI: <https://doi.org/10.1016/j.rse.2011.10.007>

Comment [O8]: Journal names must be abbreviated according to Chemical Abstracts

Comment [O9]: To be rewritten in conformance with the Instruction For Authors of the Journal. Or to a published paper like Santos et al.; JEAI, 34(4): 1-12, 2019; Article no. JEAI. 48446

Formatted: Highlight

Formatted: Highlight

313 9. GÓMEZ-CANDÓN, D.; VIRLET, N.; LABBÉ, S.; JOLIVOT, A.; REGNARD, J. Field
314 phenotyping of water stress at tree scale by UAV-sensed imagery: new insights for thermal
315 acquisition and calibration. *Precision Agriculture*, v.17(6): 786-80, 2016. DOI:
316 <https://doi.org/10.1007/s11119-016-9449-6>

317 10. VESALI, F.; OMID, M.; KALEITA, A.; MOBLI, H. Development of an android app to
318 estimate chlorophyll content of corn leaves based on contact imaging. *Computers and*
319 *Electronics in Agriculture*, v.116: 211-220, 2015. DOI:
320 <https://doi.org/10.1016/j.compag.2015.06.012>

321 11. CRAIN, J.; ORTIZ-MONASTERIO, I.; RAUN, B. Evaluation of a Reduced Cost Active
322 NDVI Sensor for Crop Nutrient Management. *Journal of Sensors*, v.2012, ID 582028, 10
323 páginas, 2012. DOI: <http://dx.doi.org/10.1155/2012/582028>

324 12. CAO, Q.; MIAO, Y.; WANG, H.; SHANYUHUANG, S.; SHANSHANCHENG, S.; KHOSLA,
325 R.; JIANG, R. Non-destructive estimation of rice plant nitrogen status with Crop Circle
326 multispectral active canopy sensor. *Field Crops Research*, v.154: 133-144, 2013. DOI:
327 <https://doi.org/10.1016/j.fcr.2013.08.005>

328 13. WANG, Y.; DUNN, B.L.; ARNALL, D.B. Assessing nitrogen status in potted geranium
329 through discriminant analysis of ground-based spectral reflectance data. *HortScience*, v.47:
330 343-348, 2012a.

331 14. WANG, Y.; DUNN, B.L.; ARNALL, D.B.; MAO, P. Use of an active canopy sensor and
332 SPAD chlorophyll meter to quantify geranium nitrogen status. *HortScience*, v.47: 45-50,
333 2012b.

334 15. SHIRATSUCHI, L. S.; VILELA, M. F.; FERGUSON, R. B.; SHANAHAN, J. F.;
335 ADAMCHUK, V. I.; RESENDE, A. V.; HURTADO, S. C.; CORAZZA, E. J. Desenvolvimento
336 de um algoritmo baseado em sensores ativos de dossel para recomendação da adubação
337 nitrogenada em taxas variáveis. In: INAMASU, R. Y.; NAIME, J. M.; RESENDE, A. V.;
338 BASSOI, L.H.; BERNARDI, A. C. C. *Agricultura de precisão: um novo olhar*. São Carlos:
339 Embrapa Instrumentação, p. 184-188, 2011.

340 16. DUNN, B. L.; SHRESTHA, A.; GOAD, C.; KHODDAMZADEH, A. A. Use of optical
341 sensors to monitor Gaillardia Foug. nitrogen status. *Journal of Applied Horticulture*, v.17(3):
342 181-185, 2015.

343 17. ADAMCHUK, V.I.; HUMMEL, J. W.; MORGAN, M.T.; UPADHYAYA, S. K. On-the-go soil
344 sensors for precision agriculture. *Computers and Electronics in Agriculture*, v.44(1): 71-91,
345 2004. DOI: <https://doi.org/10.1016/j.compag.2004.03.002>

346 18. SCHIRRMANN, M.; GEBBERS, R.; KRAMER, E. Performance of Automated Near-
347 Infrared Reflectance Spectrometry for Continuous in Situ Mapping of Soil Fertility at Field
348 Scale. *Vadose Zone Journal*, v.12(4): 14p, 2013. DOI: <https://doi.org/10.2136/vzj2012.0199>

349 19. CHRISTY, C.D.; DRUMMOND, P.; LAIRD, D. A. An on-the-go spectral reflectance
350 sensor for soil. ASAE Annual Meeting, nº 031044, 2003.

351 20. MAHLEIN, A.K. Plant Disease Detection by Imaging Sensors - Parallels and Specific
352 Demands for Precision Agriculture and Plant Phenotyping. *APS Journals*, v.100(2): 241-251,
353 2016.

354 21. VIANA, L. A.; ZAMBOLIM, L.; SOUSA, T. V.; TOMAZ, B. C. Potential use of thermal
355 camera coupled in UAV for culture monitoring. *Brazilian Journal of Biosystems Engineering*,
356 v.12(3): 286-298, 2018. DOI: <http://dx.doi.org/10.18011/bioeng2018v12n3p286-298>

357 22. SANKARAN, S.; MAJA, J. M.; BUCHANON, S.; EHSANI, R. Detecção de Huanglongbing
358 (Citrus Greening) usando técnicas visíveis, Near Infrared e Thermal Imaging. *Sensors*,
359 v.13(2): 2117-2130, 2013. DOI: <https://doi.org/10.3390/s130202117>

360 23. GARCIA-RUIZ, F.; SANKARAN, S.; MAJA, J. M.; LEE, W. S.; RASMUSSEN, J.;
361 EHSANI, R. Comparison of two aerial imaging platforms for identification of Huanglongbing-
362 infected citrus trees. *Computers and Electronics in Agriculture*, v.91: 106-115, 2013. DOI:
363 <https://doi.org/10.1016/j.compag.2012.12.002>

364 24. HAN, P.; DONG, D.; ZHAO, X.; JIAO, L.; LANG, Y. A smartphone-based soil color
365 sensor: For soil type classification. *Computers and Electronics in Agriculture*, v.123: 232-
366 241, 2016. DOI: <https://doi.org/10.1016/j.compag.2016.02.024>

367 25. GÓMEZ-ROBLEDÓ, L.; LÓPEZ-RUIZ, N.; MELGOSA, M.; PALMA, A. J.; CAPITÁN-
368 VALLVEY, J. F.; SÁNCHEZ-MARAÑÓN, M. Using the mobile phone as Munsell soil-colour
369 sensor: An experiment under controlled illumination conditions. *Computers and Electronics*
370 *in Agriculture*, v.99: 200-208, 2013. DOI: <https://doi.org/10.1016/j.compag.2013.10.002>

371 26. MULLA, D. J. Twenty five years of remote sensing in precision agriculture: Key
372 advances and remaining knowledge gaps. *Biosystems Engineering*, v.114(4): 358-371,
373 2013. DOI: <https://doi.org/10.1094/PDIS-03-15-0340-FE>

374 27. PANTAZI, X. E.; TAMOURIDOU, A. A.; ALEXANDRIDIS, T. K.; LAGOPODI, A. L.;
375 KASHEFI, J.; MOSHOU, D. Evaluation of hierarchical self-organising maps for weed
376 mapping using UAS multispectral imagery. *Computers and Electronics in Agriculture*, v.139:
377 224-230, 2017. DOI: <https://doi.org/10.1016/j.compag.2017.05.026>

378 28. LÓPEZ-GRANADOS, F.; TORRES-SÁNCHEZ, J.; CASTRO, A.; SERRANO-PÉREZ, A.
379 MESAS-CARRASCOSA FJ.; PEÑA, JM. Object-based early monitoring of a grass weed in a
380 grass crop using high resolution UAV imagery. *Agronomy for Sustainable Development*,
381 12p, 2016. DOI: <https://doi.org/10.1007/s13593-016-0405-7>