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COMPARASION OF CROP WATER STRESS INDEX (CWSI) AND WATER DEFICIT INDEX (WDI) BY USING REMOTE SENSING (RS)

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Abstract

Drought, water scarcity and climate changes are very important threats for agriculture on a global basis. Remote sensing (RS) is accepted as a technique to collect data and determine water stress indices. Water Stress Indices (WSI) are useful tools to prevent drought and determine irrigation scheduling. The water stress indices are primarily identified as the Crop Water Stress Index (CWSI) and the Water Deficit Index (WDI). The effect of soil background is major problem to establish CWSI especially during early growth stage measurements of canopy temperature (Ts). Hence, WDI is a better index when it comprised with CWSI because of Ts. CWSI and WDI can be determined by two different techniques. These are determined by using measured by using traditional components to collect data and estimated methods by applying RS components to collect necessary data. Estimated method has many advantages when this method compared with measured method. However, estimated method needs some RS components which are infrared gun (IR), sling psychrometer, Spectro radiometer. With the help of these tools, the necessary data are obtained and WDI is determined. By using Spectro radiometer vegetation indices are defined. Among the many vegetation indices, the Normalized Difference Vegetation Index (NDVI) is mostly used one. By using NDVI determination of vegetation cover is easy and accurate technique to establish WDI. Establishing these both stress indices with less fieldwork and by saving

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money, time and labor conveys the necessary information for agriculturists using remotely sensed data especially for large agricultural fields.

Key Words: Water Stress Index, Crop Water Stress Index, Water Deficit Index, Remote Sensing

INTRODUCTION

Drought is a main reason of limited agricultural productivity and of plant vield uncertainty throughout the world (Martinez-Fernandez et al., 2016). For this reason, agricultural drought is a very big threat that can be defined by shortage of water supply (Al-Solaimani et al., 2017). Water is one of the fundamental elements required to keep important functions of all living things in the past and coming years (Akay and Önder, 2016). The total water supply of the world is 1410 million km³ and unfortunately only 1.3% of it is fresh-water (USGS, 2006a). Rapidly growing population is causing the consumption of clean water resources to a certain extent (El-Shirbeny et al., 2015). Irrigation is the dominant user of fresh-water on a global basis (USGS, 2006b). Droughts and water scarcity are already a serious problem in many European countries (Giannerini and Genovesi, 2015). Moreover, growing global fear regarding water scarcity means that efficient water use is important (Gal et al., 2003) by improving water management and focusing plant stress in agricultural management as well (Xu et al., 2015). All the current trends show that water scarcity threaten humanity globally due to available freshwater amount and quality have significantly decreased in the last few decades (Silber et. al., 2015); therefore, water management is very important issue (Tanriverdi, 2005). Not only water scarcity and drought but also climate change is very important topic for agriculture productivity (Liu et al., 2017). Climate warming and changes in the extremes have eventually influenced vegetation dynamics in northern terrestrial ecosystems based on the modeling and observational studies (Piao et al., 2014). Determining exactly how climate change effects vegetation changes has received common attention in the past several decades (Yang et al., 2012). Vegetation growing is limited by water scarcity (Zhang et al., 2010). It is necessary to figure out the vegetation dynamic changes and responses to climate change to estimate the quality of ecosystem and maintain optimal ecosystem functioning (Mu et al., 2013). Monitoring climate changes is possible by using new technologies which are remote sensing (RS) and Normalized Difference Vegetation Index (NDVI) (Tucker, 1979; Nash et al., 2017).

Effective monitoring of water status by utilizing indices is required to efficiently maximize yield and quality of crop under today's competitive conditions of reduced water resources (Rud et al., 2014). Thus, water stress indices (WSI) are important to prevent drought and determine irrigation scheduling (Qui et al., 2000; Kacar, 2007; Sharda et al., 2013). Consequently, water stress indices may be decided to have a better water management. The water stress indices are primarily identified as the Crop Water Stress Index (CWSI) and the Water Deficit Index (WDI).

The objectives of this study were to explain importance of the water stress indices and determine them to have a better irrigation management with indicate the benefits of using RS. Withal compare these indices (CWSI and WDI) to result which one is better index to use in irrigation management.

REMOTE SENSING (RS)

Recent developments of Remote Sensing (RS) components have extended the practice of remote sensing in agriculture (Turner et al., 2011; Hoffmann et al., 2016a; Liu et al., 2017). The ability to automate data collection, and to process and perform such data has been significantly increased by RS. The detection and recording components to this order are known as remote sensors. These sensors are like photographic cameras, mechanical scanners and radar systems (Diker, 1998).

RS may have utilized to estimate the evapotranspiration (ET) (Jimenez et al., 2011) especially for large areas (Tanriverdi, 2003; Yang et al., 2017); by known estimation ET, RS supplies analysis of irrigation system performance, high productivity and it is significant guidance to managers and system designers as well (Mulla and Schepers, 1997). RS is verified to be useful in agriculture for stress classification, irrigation scheduling through canopy temperature monitoring, and yield prediction. Progress in thermal RS of ET were extremely summarized, including major reviews by Carlson (1986); Moran and Jackson (1991); Kustas and Anderson (2009). Petropoulos et al. (2009) provided a comprehensive and systematic review of the surface temperature-vegetation index (T_s-VI) space method. Presently, Wang and Dickinson (2012) served a comprehensive review of ET observation, modeling and climatic variability. Rud et al. (2014) stated that RS methods utilizing the thermal part of the spectrum can provide an indication of crop water status.

CROP WATER STRESS INDEX (CWSI)

With increasing water scarcity and drought impacts, water saving irrigation techniques are commonly used in agricultural fields to determine better water management (Belder et al., 2004; Uphoff et al., 2010; Zia et al., 2012; Li et al., 2014). Crop water deficit definition or water status monitoring is the base to proper irrigation scheduling (Xu et al., 2015). Therefore, crop water deficit definition techniques based on soil water condition, crop water potential, and leaf physiological parameters are of great worry (Yatapanage and So, 2001; Silva et al., 2007). The crop water stress index (CWSI) outputs are quite stable, proposing the feasibility of implementing this method for irrigation management (Rud et al., 2014; Bai et al., 2017). CWSI is able to notice stress 24-48 hrs. prior to the stress detection by visual observation (Kacira et al., 2002). CWSI is a technique that utilizes vapor pressure deficit (VPD) and measurements of the crop surface temperature (T_s) mostly by an infrared thermometer (IR) to establish the crop water status (Idso et al., 1977; Jackson et al., 1981; Peng et al., 2011). CWSI based on canopy temperature and meteorological terms following Idso et al. (1981) was used (Rud et al., 2014). After Idso et al. (1981) CWSI has been successfully applied in many different plants, such as wheat (Yuan et al., 2004), cotton (O'shaughnessy et al., 2011), maize (Romano et al., 2011), potato (Ramirez et al., 2016), bean (Erdem et al., 2006), and some vegetables (Cremona et al., 2004; Rud et al., 2014) or fruits (Paltineanu et al., 2009).

Air temperature (T_a) , T_c and VPD are necessary data to establish CWSI. The dry and wet bulb temperatures measure with an aspirated psychrometer at the level of 2 meters in the open field part. The mean T_a is determined from average of these measurements (Erdem et al., 2005). While other factors should be considered in order to attain accurate measurements of stress levels, T_e is the most important (Jackson et al., 1981). A crop's T_c can be measured remotely through the use of infrared thermometers (IR) or thermal scanners (Jackson, 1982). The mean VPD is computed as the average of calculated instant VPD utilizing the corresponding instant wet and dry bulb temperatures and the pyschrometer equation (Allen et al., 1998). Numerous early researches focused on T_{and} T_{differences} to determine the transpiration. The relationship between T_d and water stress is based on the statement that as a crop transpires, the evaporated water-cools the leaves lower the T_a. Jackson et al. (1981) confirmed that when a plant is transpiring fully, the T_s is 1 to 4 degrees cooler than the T_s and the CWSI comes 0. T will increase, as the crop becomes water stressed, the stomata close and transpiration decreases (Rud et al., 2014). As transpiration decreases, the T_s increases and can be 4 to 6 degrees warmer than the T_s (Jackson et al., 1981). Thus, an estimate measurement of water stress in crops is based on the inverse correlation between T_s and stomal opening (Fuchs, 1990). Garcia et al. (2000) declared that in the stressed plot, the average of T_s was 5 or 4 °C higher than the unstressed plot. When the plant is not transpiring anymore, the CWSI is 1 (El-Shirbeny et al., 2015). A primary impediment to the routine operation of IR in assessing crop water stress status and irrigation scheduling was overcome when Idso et al. (1981) developed an innovative CWSI that normalizes the stress-degree-day parameter (Idso et al., 1977; Jackson et al., 1977) for environmental variability. Under natural terms of evaporation, the condition of a given mass of air can be described by its temperature and vapor pressure (VP). The CWSI is an indicator of the relation transpiration rate occurring from a crop at the time of measurement using a measure of T_s and the VPD, which is a measurement of air dryness. A sample of VPD baselines is given in Figure 1. The atmospheric VPD (kPa) (Banerjee et al., 2012) was calculated with the relative humidity (RH, %) and temperature (T, °C) measured by the automatic weather station meanwhile of the image capturing.

$$VPD = 0.6108e^{(17.27T/T+237.3)}(1-RH)$$

Thus, VPD (the vapor pressure deficit of air at T_a (kPa)) is an essential calculation to determine VPD upper (maximum stress) and lower (minimum stress) baselines. Besides, it was used to calculate the lower limit of $T_s - T_a$ (dT_µ).

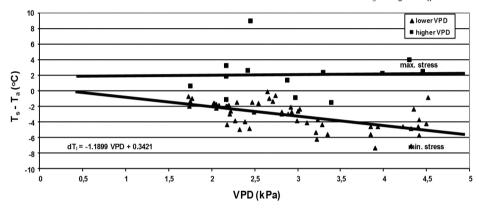


Figure 1. VPD upper and lower baselines for corn (Tanriverdi, 2003).

$$\begin{split} dT_{ll} &= Slope \; (VPD) + Intercept, \\ dT_{ul} &= Intercept + Slope \; (VP_{sat} \{T_a\} - VP_{sat} \{T_a + Intercept\}) \end{split}$$

These calculated parameters $(dT_1 \text{ and } dT_u)$ and the current difference in $T_s - T_a (dT)$ are utilized in the CWSI formula (Idso et al., 1981).

$$CWSI = (dT - dT_{\mu}) / (dT_{\mu} - dT_{\mu})$$

where:

dT; the current difference between canopy and air temperature °C. $(T_s - T_a)$ dT_{ll}; the baseline difference between canopy and air temperature of a well-watered, transpiring plant, °C. $(T_s - T_a)_{ll}$ dT_{ul}; the baseline difference between canopy and air temperature of a fully stressed, non-transpiring plant, °C. $(T_s - T_a)_{ul}$

WATER DEFICIT INDEX (WDI)

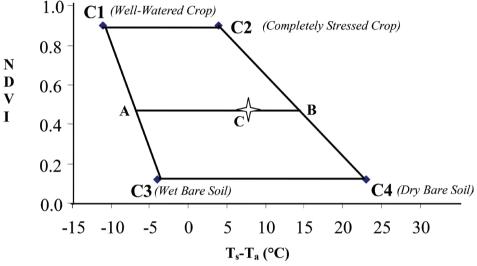
Water deficit index (WDI) is another WSI and it works the same way as CWSI. That is, 0 indicates no water stress and 1 indicates that water stress is at the maximum point (El-Shirbeny et al., 2015). The CWSI is functional technique in only full vegetation cover (Wang and Takahashi, 1999). Thus, the WDI used by Moran et al. (1994) addressed this limitation by proposing the WDI to explain a soil canopy water deficit status, was derived from the theory of CWSI presented by Jackson et al. (1981) (Colaizzi et al., 2003). WDI is noted a composite of both vegetation cover and soil background to measure T_s (Hoffmann et. al., 2016b).

Determining the vegetation cover is the most important part of preparing the WDI. Both estimated by using remotely sensed data and measured by using Light Bar (LB) vegetation cover (fraction, f) techniques can be utilized for determination of vegetation cover in agricultural fields. There are numerous techniques for estimating vegetation cover generally by utilizing Spectro radiometer, such as Normalized Difference Vegetation Index (NDVI), Soil Adjusted Vegetation Index (SAVI), Modified Soil Adjusted Vegetation Index (MSAVI), Visible Atmospherically Resistant Indices (VARI_{green}) and Vegetation Index (VIgreen) (Tanriverdi, 2003). NDVI, as an effective indicator to monitor vegetation and natural environment at regional and global scales, has been commonly utilized (Colaizzi et al., 2000; Beck et al., 2006; Fensholt et al., 2010; Fensholt and Proud, 2012; Gao et al., 2015; Tong et al., 2017), to search the effect of climate change on the vegetation growth and ecosystem structure and functions (Zhang et al., 2005; Park and Sohn, 2010). There is no doubt that vegetation growth depends upon a suitable hydro-thermal condition (Xu et al., 2011). Relations between NDVI and T_s are generally utilized to investigate the hydro-thermal limit on vegetation growth (Zhou et al., 2015). The vegetation cover is plotted versus the difference between measured T_s and T_a $(T_s - T_a)$ (°C) to determine WDI. There are three key techniques for determining the vegetation cover which are estimate by using RS components, measure by using light bar and theoretical. Figure 2 shows the VIT trapezoid by estimating vegetation cover by using NDVI.

The NDVI proposed by Rouse et al. (1973) for monitor vegetation is one of most widely used indices. NDVI is defined as the difference between the NearInfrared (NIR) and Visible (VIS, generally red) bands divided by their sum (Tucker, 1979; Liu et al., 2017). There has been substantial evidence that NDVI formulation can reduce atmospheric, Bidirectional Reflectance Distribution Function (BRDF) and other effects (van Leeuwen et al., 2006).

NDVI = (NIR - RED) / (NIR + RED)

As shown in figure 2, the VIT trapezoid offers lines that attach the four corners (C1, C2, C3, C4), in which the left line contains point A and describes the range of $(T_s - T_a)$ possible for full vegetation cover where water is not limiting. In like way, the right line includes point B and describes the possible range of $(T_s - T_a)$ for status of no available water. The range of $(T_s - T_a)$ possible for full vegetation cover is characterized by the top line between C1 and C2, which is also the range of $(T_s - T_a)$ possible for the CWSI. In figure 2, A, B, and C are formed on the plot of the WDI for a given amount of vegetation cover.



Source: Colaizzi et al. (2000)

Figure 2. VIT trapezoid for cotton in Arizona

Point A presents a non-limiting condition for ET; point B is the upper limit at which no water is available, and point C presents the actual measurement. The WDI is defined as the proportion of the distances of AC to AB. Using the VIT trapezoid, WDI is calculated,

$$WDI = \overline{AC} / \overline{AB}$$

Moran et al. (1994) introduce the origin and underlying assumptions of the equations on a theoretical basis, as well as the assumptions underlying the VIT trapezoid. The necessary meteorological data include T_a , VPD, R_s , and u (wind speed (m s⁻¹)) measured at height z, all of which are accessible from the weather station. With the measurement of actual $(T_s - T_a)$ and vegetation cover (Point C, figure 2), $(T_s - T_a)$ at the four corners of the VIT trapezoid have to also be identified.

CONCLUSION

Agricultural drought monitoring is of rising interest and more accurate and efficient calculation and assessment techniques, as well as early warning systems, are needed (Martinez-Fernandez et al., 2016). For this purpose, stress indices are important subject because index can warn to irrigators before 24-48 hours for irrigation need (Kacira et al., 2002). Moreover, rapid increase of global population, pollution of natural resources, global warming and climate change nowadays increases the pressure on limited water resources (Colak et al., 2014). Therefore, one of the primary important entries for agricultural sector is irrigation as accepted by many researchers. If accurately managed, obviously irrigation management plays a critical role in increasing crop output. Al-Solaimani et al. (2017) found that increased yield could be achieved with limited water if WDI was used. Moreover, Garen and Moore (2005) mentioned, known traditional methods were not easy techniques before RS was widely available to collect extensive spatial data sets on field, soils and vegetation. Also, RS has an advantage to collect data from large field if it is compare with traditional methods (Tanriverdi, 2003; Hoffmann et al., 2016b). Especially, RS technology allows capability for large scale field level or regional evaluation and monitoring of crop field cover, crops growing and status by remotely sensed data (Yang et al., 2017).

Many scientists using CWSI to quantify water stress and schedule irrigations should use early studies to determine a suitable non-water stressed baseline equation (Nielsen, 1990; Çolak et al., 2014). However, CWSI has some difficulties to measure crop surface temperature (T_s). WDI is accepted both soil and crop canopy temperatures as a surface temperature (T_s) (El-Shirbey et al., 2015). Contrarily, collected surface temperature data T_s had to symbolize the plant surface temperature to establish the CWSI. In this way, some of the data are not useful to determine the baselines of CWSI for early growing season because of partial vegetation cover. The CWSI is merely applicable to cases of full vegetation coverage, however Moran et al. (1994) developed WDI that allowed the index to be estimated for partial vegetation coverage as well. Under these circumstances, WDI was found to be a better index when compared with CWSI since T_s measurements were more reliable in determining the four corners of VIT.

As mentioned before, there are two different techniques (measured and estimated) to determine the vegetation cover values for WDI. A LB can be used to calculate the fraction (f) which is called as a measured vegetation cover technique to establish WDI. Calculated f can be plotted versus $T_s - T_a$ (°C) to determine the WDI like estimated technique. Clearly, f needs more fieldwork to represent vegetation cover when it comprised with estimated vegetation cover. Tanriverdi (2003) mentioned that statistically, vegetation indices have a reliable

relationship with measured vegetation cover (f). Thus, the review evidences the importance of utilizing vegetation indices, which are quick and easy to calculate. Contrarily, for measured vegetation cover, it is hard to complete the data set with the limited time (during noon under clear sky conditions) available (Payero et al., 2005). When estimated vegetation cover calculations are compared with measured vegetation cover calculation clearly estimated vegetation cover (vegetation indices) methods have many advantages especially in large agricultural fields (Rud et al., 2014). Among the many vegetation indices, NDVI is the most popular one to establish WDI. Many scientists are identified NDVI which has reliable and accurate outputs to represent vegetation cover from agricultural fields.

Daily data would also have ensured a better comprehension of VWC in the soil profile and would therefore be a better tool for improving irrigation management. Utilizing RS daily essential data can be collected even for large fields, but it should be very hard by using measured technique. Finally, this review stated that;

- Drought, water scarcity and climate changes are vital topics for agriculture.
- WDI is better water stress index when it compared with CWSI because of T_s measurements.
- RS conveys the necessary information to the irrigator with less fieldwork, and is useful in saving money, time and labor
- Among the many vegetation indices, NDVI is the best one to determine vegetation cover to established WDI.
- Estimated vegetation cover technique is useful if it is compare with measured technique.

REFERENCES

Akay, A., Önder, S. (2016). *Benefits of rain water usage in urban green areas for water saving*. 2nd International Conference on Science Ecology and Technology (ICONSETE, 2016) Barcelona / Spain. Pp:804-812.

Allen, R.G., Pereira, L.S., Raes, D., Smith, M. (1998). *Crop Evopotranspiration. FAO Irrigation and Drainage Paper* 56, Rome, pp 299.

Al-Solaimani, S.G., Alghabari, F., Ihsan, M.Z., Fahad, S. (2017). *Water deficit irrigation and nitrogen response of Sudan grass under arid land drip irrigation conditions*. Irrigation and Drainage, Published online in Wiley Online Library DOI: 10.1002/ird.2110.

Azzali A., Menenti M. (2000). *Mapping vegetation-soil complexes in southern Africa using temporal Fourier analysis of NOAA AVHRR NDVI data*. International Journal of Remote Sensing, 21: 973-996. doi: http://dx.doi.org/10.1080/014311600210380.

Bai J.J., Yu Y., Di L.P. (2017). Comparison between TVDI and CWSI for drought monitoring in the Guanzhong Plain, China. Journal of Integrative Agriculture, Vol. 16: 389-397.

Banerjee, A., de Fortier Smit A., Prozzi, J.A. (2012). *Modeling the effect of environmental factors on evaporative water loss in asphalt emulsions for chip seal applications*. Constr. Build. Mater., 27: 158-164.

Beck, P.S.A., Atzberger, C., Hogda, K.A. (2006). *Improved monitoring of vegetation dynamics at very high latitudes, a new method using MODIS NDVI*. Remote Sens. Environ. 2006, 100, 321–336.

Belder, P., Bouman, B.A.M., Cabangon, R., Guoan, L., Quilang, E.J.P., Yuanhua, L., Spiertz J.H.J., Tuong, T.P. (2004). *Effect of water-saving irrigation on rice yield and water use in typical lowland conditions in Asia*. Agric. Water Manag., 65: 193-210.

Carlson, T.N. (1986). *Regional scale estimation of surface moisture availability and thermal inertia using remote thermal measurements*. Remote Sens. Rev. 1:197–247.

Colaizzi, P.D., C.Y. Choi, P.M. Waller, E.M. Barnes, Clarke, T.R. (2000). *Determining irrigation management zones in precision agriculture using the water deficit index at high spatial resolutions*. 2000 ASAE Annual International Meeting. Pp. 1-19.

Colaizzi, P.D., Barnes, E.M., Clarke, T.R., Choi, C.Y. Waller, P.M. Haberland, J., Kostrzewski, M. (2003). *Water stress detection under high frequency sprinkler irrigation with water deficit index*. Journal of Irrigation and Drainage Engineering, Vol. 129(1): 36-43.

Cremona, M.V., Stützel, H., Kage, H. (2004). Irrigation scheduling of kohlrabi (Brassica oleracea var. gongylodes) using crop water stress index. Hortic. Sci., 39:276-279.

Colak, Y.B., Yazar, A., Colak, İ., Akça, H., Duraktekin, G. (2014). Evaluation of crop water stress index (CWSI) for eggplant under varying irrigation regimes using surface and subsurface drip systems. Available online at www.sciencedirect.com.

Diker, K. (1998). Use of geographic information management systems (GIMS) for nitrogen management. Ph. D. Thesis, Department of Chemical and Bioresource Engineering, Colorado State University, Spring 1998.

El-Shirbeny, M.A., Ali, A.M., Rashash, A., Badr, M.A. (2015). *Wheat yield response to water deficit under central pivot irrigation system using remote sensing techniques*. World Journal of Engineering and Technology, Vol. 3: 65-72.

Erdem, Y., Erdem, T., Orta, A.H., Okursoy, H. (2005). *Irrigation scheduling for watermelon with crop water stress index (CWSI)*. Journal Central European Agriculture, 6(4): 449-460.

Erdem, Y., Schirali, S., Erdem T., Kenar, D. (2006). *Determination of crop water stress index for irrigation scheduling of bean (Phaseolus vulgaris L.)*. Turk. J. Agric. For., 30: 195-202.

Fensholt, R., Sandholt, I., Proud, S.R., Stisen, S., Rasmussen, M.O. (2010). Assessment of MODIS sun-sensor geometry variations effect on observed NDVI using MSG SEVIRI geostationary data. International Journal of Remote Sensing, 31(23): 6163-6187.

Fensholt, R. Proud, S.R. (2012). *Evaluation of earth observation based global long term vegetation trends Comparing GIMMS and MODIS global NDVI time series*. Remote Sens. Environ. 119: 131–147.

Fuchs, M. (1990). *Infrared measurement of canopy temperature and detection of plant water-stress*. Theoretical Applied Climatology, 42: 253-261.

Gal, L.Y.P., Rieu, T., Fall, C. (2003). *Water pricing and sustainability of Self Governing Irrigation Schemes*. Irrigation and Drainage Systems. Vol. 17(3).

Garcia, A., Andre, R.G.B., Ferreira, T. do Paço, (2000). *Diurnal and seasonal variations of CWSI and non-water-stressed baseline with nectarine trees*. III International Symposium on Irrigation of Horticultural Crops, Vol. 2. DOI: 10.17660/ActaHortic.2000.537.49

Garen, C. D., Moore D.S. (2005). *Curve number hydrology in water quality modelling: uses, abuses, and future directions*. Journal of the American Water Resources Association. April, 377-388.

Gao, Y., Walker, J.P., Allahmoradi, M., Monerris A., Ryu, D., Jackson, T.J. (2015). *Optical Sensing of Vegetation Water Content: A Synthesis Study*. IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing. 1-9. DOI:10.1109/JSTARS.2015.2398034.

Giannerini, G. Genovesi, R. (2015). *The water saving with irriframe platform for thousands of Italian farms*. Journal of Agricultural Informatics. Vol. 6(4):49-55.

Hoffmann, H., Jensen, R., Thomsen, A., Nieto, H., Rasmussen, J. Friborg, T. (2016a). *Crop water stress maps for an entire growing season from visible and thermal UAV imagery*. Biogeosciences, 13: 6545-6563.

Hoffmann, H., Nieto, H., Jensen, R., Guzinski, R., Zarco-Tejada, P., Friborg, T. (2016b). *Estimating evaporation with thermal UAV data and two-source energy balance models*, Hydrol. Earth Syst. Sci., 20, 697–713,

Hou, W., Gao, J., Wu, S., Dai, E. (2015). *Interannual variations in growing-season NDVI and its correlation with climate variables in the Southwestern Karst region of China*. Remote Sensing, 7: 11105-11124.

Idso, S.B., Jackson, R.D. Reginato, R.J. (1977). *Remote sensing of crop yields*. Science, 196:19-25.

Idso, S.B., Jackson, R.D, Pinter, P.J., Reginato, R.J, Hatfield, J.L. (1981). *Normalizing the stress-degree-day parameter for environmental variability*. Agricultural Meteorology, 24:45-55.

Jackson, R.D., Reginato, R.J., Idso, S.B. (1977). Wheat canopy temperature: a practical tool for evaluating water requirements. Water Resources Research, 13: 651-656.

Jackson, R.D., Idso, S.B., Reginato, R.J. Pinter Jr, P.J.(1981). *Canopy temperatures as a crop water stress indicator*. Water Resources Research, Vol. 17: 1133-1138.

Jackson, R.D. (1982). *Canopy temperatures and crop water stress*. In Advances in Irrigation, D.I. Hillel, Editor, Academic Press, 1: 43-85.

Jimenez, C., Prigent, C., Mueller, B., Seneviratne, S.I., McCabe, M.F., Wood, E.F., Rossow W.B., Balsamo, G., Betts, A.K., Dirmeyer, P.A. (2011). *Global intercomparison of 12 land surface heat flux estimates*. Journal of Geophys. Res. Atmos., 116, DOI:10.1029/2010JD0145.

Kacar, M.M. (2007). Investigation Of Cotton Water Stress Index Variations Under Different Water And Fertilizer Systems. Çukurova Ünv. Fen Bilimleri Enst., Tarımsal Yapılar ve Sulama ABD, Yüksek Lisans Tezi.

Kacira, M., Ling, P.P., Short, T.H., (2002). *Establishing crop water stress index (CWSI) threshold values for early, non-contact detection of plant water stress*. Transactions of ASAE, Vol. 45(3): 775-780.

Kerr, J.T. Ostrovsky, M. (2003). From space to species: ecological applications for remote sensing. Trends in Ecology & Evolution, 18 (6): 299-305.

Kustas, W. Anderson, M. (2009). Advances in thermal infrared remote sensing for land surface modeling. Agric. for Meteorol. 149:2071-2081.

Li, B., Wang, T., Sun, J. (2014). Crop water stress index for off-season greenhouse green peppers in Liaoning, China. Int. J. Agric. Biol. Eng., 7: 28-35.

Liu, X., Ferguson, R.B., Zheng, H., Cao, Q., Tian, Y., Cao, W., Zhu, Y. (2017). Using an active-optical sensor to develop an optimal NDVI dynamic model for high-yield rice production. Sensors, 17, 672. DOI:10.3390/s17040672.

Martinez-Fernandez, J., Gonzalez-Zamora, A., Sanchez, N., Gumuzzio, A., Herrero-Jimenez, C.M. (2016). *Satellite soil moisture for agricultural drought monitoring: Assessment of the SMOS derived soil water deficit index*. Remote Sensing of Environment, 177: 277-286.

Moran, M.S., Jackson, R.D. (1991). Assessing the spatial distribution of evapotranspiration using remotely sensed inputs. J Environ. Qual. 20:725–737.

Moran, M.S., Clarke, T.R., Inoue, Y., Vidal, A. (1994) *Estimating crop water deficit using the relation between surface-air temperature and spectral vegetation index*. Remote Sensing Environment, Vol. 32: 125-141.

Mu, S.J., Yang, H.F., Li, J.L., Chen, Y.Z., Gang, C.C., Zhou, W., Ju, W.M. (2013). *Spatiotemporal dynamics of vegetation coverage and its relationship with climate factors in Inner Mongolia*, China. J Geogr Sci 23(2):231–246.

Mulla. D.J., Schepers. J.S., (1997). *Key processes and properties for site-specific soil and crop management*. Pp.1-18. In F.J. Pierce and E. J. Sadler (eds.) The State of Site-Specific Management for Agriculture. American Society of Agronomy, madison, WI.

Nash, M.S., Wickham, J., Christensen J., Wade, T. (2017). *Changes in Landscape Greenness and Climatic Factors over 25 Years (1989–2013) in the USA. Remote Sensing*, 9,295; DOI:10.3390/rs9030295.

Nielsen, D.C. (1990). Scheduling irrigations for soybeans with crop water stress index (CWSI). Field Crops Res., Vol. 23: 103-116.

O'shaughnessy, S., Evett, S., Colaizzi, P., Howell, T. (2011). Using radiation thermography and thermometry to evaluate crop water stress in soybean and cotton. Agric. Water Manag., 98: 1523-1535.

Paltineanu, C., Chitu E., Tanasescu, N. (2009). *Correlation between the crop water stress index and soil moisture content for apple in a loamy soil: A case study in southern Romania*. VI Int. Symp. Irri. Hort. Crops 889: 257-264.

Park, H.S., Sohn, B. (2010). Recent trends in changes of vegetation over East Asia coupled with temperature and rainfall variations. J. Geophys. Res., 115.

Payero, J.O., Neale, C.M.U., Wright J.L. (2005). *Non-water-stressed baselines for calculating crop water stress index (CWSI) for alfalfa and tall fescue grass*. Transactions of ASAE, 48(2): 653-661.

Peng, S., Yang, S., Xu, J., Gao, H. (2011). Field experiments on greenhouse gas emissions and nitrogen and phosphorus losses from rice paddy with efficient irrigation and drainage management. Sci. China Technol. Sci., 54: 1581-1587

Petropoulos, G., Carlson, T.N., Wooster, M.J., Islam, S. (2009). A review of T-s/VI remote sensing based methods for the retrieval of land surface energy fluxes and soil surface moisture. Prog Phys Geogr. 33:224-250.

Piao, S.L., Nan, H.J., Huntingford, C., Ciais, P., Friedingstein, P., Sitch, S., Peng, S.S., Ahlstrom, A., Canadell, J.G., Cong, N., Levis, S., Levy, P.E., Liu, L.L., Lomas, M.R., Mao, J.F., Myneni, R.B., Peylin, P., Poulter, B., Shi, X.Y., Yin, G.D., Viovy, N., Wang, T., Wang, X.H., Zaehle, S., Zeng, N., Zeng, Z.Z., Chen, A.P. (2014). *Evidence for a weakening relationship between interannual temperature variability and northern vegetation activity*. Nat. Commun. 5, 5018.

Ramirez, D.A., Yactayo, W., Rens, L.R., Rolando, J.L., Palacios, S., De Mendiburu, F., Mares, V., et al. (2016). *Defining biological thresholds associated to plant water status for monitoring water restriction effects: Stomatal conductance and photosynthesis recovery as key indicators in potato*. Agricultural Water Management, 177:369-378.

Rud, R., Cohen, Y., Alchanatis, V., Levi, A., Brikman, R., Shenderey, C., Heuer, B., Markovitch, T., Dar, Z., Rosen, C., Mulla, D., Nigon, T. (2014). *Crop water stress index derived from multi yearground and aerial thermal images as an indicator of potato water status*. Precision Agric., 15: 273-289.

Romano, G., Zia, S., Spreer, W., Sanchez, C., Cairns, J., Araus, J.L., Müller, J. (2011). Use of thermography for high throughput phenotyping of tropical maize adaptation in water stress. Comput. Electron. Agric., 79:67-74.

Rouse, J.W. Haas, R.H., Schell, J.A. and Deering, D.W., 1973. "Monitoring vegetation systems in the Great Plains with ERTS," in Proc. 3rd Earth Resour. Technol. Satell. Symp., Washington, DC, USA, 1973, pp. 309–317.

Sharda, V., Srivastava, P., Kalin, L., Asce, M., Ingram, K., Chelliah, M. (2013). *Development of community water deficit index: Drought-forecasting tool for small to mid-size communities of the southeastern United States.* Journal of hydrologic engineering, Vol. 18: 846-858.

Silber, A., Israeli, Y., Elingold, I., Levi, M., Levkovitch, I., Russo, D., Assouline, S. (2015). *Irrigation with desalinated water: A step toward increasing water saving and crop yields*. Water Resources Research. 51, 450-464.

Silva, M.D.A., Jifon, J.L., Da Silva J.A., Sharma, V. (2007). Use of physiological parameters as fast tools to screen for drought tolerance in sugarcane. Braz. J. Plant Physiol. 19:193-201.

Tanriverdi, C. (2003). Available water effects on water stress indices for irrigated corn grown in sandy soils. Ph. D. Thesis, Department of Chemical and Bioresource Engineering, Colorado State University, Fall 2003.

Tanriverdi, C. (2005). *Using TDR in the agricultural water management*. KSU, Journal of Science and Engineering, Vol. 8(2): 108-115.

Turner, D., Lucieer, A., Watson, C. (2011). *Development of an Unmanned Aerial Vehicle (UAV) for hyper resolution vineyard mapping based on visible, multispectral, and thermal imagery*, in: En Proceedings of 34th International Symposium on Remote Sensing of Environment, 1–4, available at: http://www.isprs.org/ proceedings/2011/ isrse-34/211104015Final00547.pdf (last access: 5 April 2016).

Tucker, C.J. (1979). *Red and Photographic Infrared Linear Combinations for Monitoring Vegetation*. Remote Sens. Environ., Vol. 8:127–150.

Tong, X., Brandt, M., Hiernaux, P., Herrmann, S., Tian, F., Prishchepov, A.V., Fensholt, R. (2017). *Revisiting the coupling between NDVI trends and cropland changes in the Sahel drylands: A case study in western Niger*. Remote Sensing of Environment, 191: 286-296.

USGS science for a changing world, (2006a). *Water science for school: how much water is there on earth?* http://ga.water.usgs.gov/edu/earthhowmuch.html Pp. 1-3.

USGS science for a changing world, (2006b). *Irrigation water use, from USGS water science*. http://ga.water.usgs.gov/edu/wuir.html Pp. 1-4.

Uphoff, N., Kassam, A., Harwood, R. (2010). *SRI as a methodology for raising crop and water productivity: productive adaptations in rice agronomy and irrigation water management.* Paddy Water Environ., 9: 3-11.

Van Leeuwen, W.J.D, Orr, B.J., Marsh, S.E., Herrmann, S.M. (2006). *Multi-sensor NDVI data continuity: uncertainties and implications for vegetation monitoring applications*. Remote Sens. Environ. 100(1):67-81.

Wang, Q., Takahashi, H. (1999). A land surface water deficit model for an arid and semiarid region: impact of desertification on the water deficit status in the loess plateau, *China*. Journal of Climate, 12(1):244-257.

Wang, K, Dickinson, R.E. (2012). A review of global terrestrial evapotranspiration: observation, modeling, climatology, and climatic variability. Rev Geophys 50, RG2005:1-54.

Xu, W.X., Gu, S., Zhao, X.Q., Xiao, J.S., Tang, Y.H., Fang, J.Y., Zhang, J., Jiang, S. (2011). *High positive correlation between soil temperature and NDVI from 1982 to 2006 in alpine meadow of the Three-River Source Region on the Qinghai-Tibetan Plateau*. Int. J. Appl. Earth Obs. Geoinf. 13(4):528–535.

Xu, Y., Yang J., Chen Y. (2015). *NDVI-based vegetation responses to climate change in an arid area of China*. Theor Appl Climatol. 1:213-222.

Xu, J., Lv, Y., Dalson, T., Yang, T., Wu, J. (2015). *Diagnosing Crop Water Stress of Rice using Infrared Thermal Imager under Water Deficit Condition*. International Journal of Agriculture & Biology, DOI: 10.17957/IJAB/15.0125.

Yang Y., Xu J.H., Hong Y.L., Lv G.H. (2012). *The dynamic of vegetation coverage and its response to climate factors in Inner Mongolia, China*. Stoch Environ. Res. Risk. Assess. 26:357–373.

Yang, Z., Wen-bin, W.U., Di, L., Ustundag, B. (2017). *Remote sensing for agricultural applications*. Journal of Integrative Agriculture, 16(2): 239-241.

Yatapanage, K.G., So, H.B. (2001). *The relationship between leaf water potential and stem diameter in sorghum*. Agron. J., 93: 1341-1343.

Yi, S.H., Zhou, Z.Y., Ren, S.L., Xu, M., Qin, Y., Chen, S.Y., Ye, B.S. (2011). *Effects of permafrost degradation on alpine grassland in a semi-arid basin on the Qinghai–Tibetan Plateau*. Environ. Res. Lett. 6 (4), 045103.

Yuan, G., Luo, Y., Sun, X., Tang, D. (2004). *Evaluation of a crop water stress index for detecting water stress in winter wheat in the North China Plain*. Agric. Water Manag., 64: 29-40.

Zhang, X.X., Ge, Q.S., Zheng, J.Y. (2005). *Impacts and lags of global warming on vegetation in Beijing for the last 50 years based on remotely sensed data and phonological information*. Chin. J. Ecol., 24:123–130.

Zhang X.Y., Goldberg M., Tarpley D., Friedl M.A., Morisette J., Kogan F., Yu, Y.Y. (2010). *Drought-induced vegetation stress in southwestern North America*. Environ. Res. Lett. 5:024008. doi:10.1088/1748-9326/5/2/024008.

Zhou, Z.Y., Yi, S.H., Chen, J.J., Ye, B.S., Sheng, Y., Wang, G.X., Ding, Y.J. (2015). *Responses of alpine grassland to climate warming and permafrost thawing in two basins with different precipitation regimes on the Qinghai–Tibetan Plateau*. Arct. Antarct. Alp. Res. 47(1):125–131.

Zia, S., Du, W., Spreer, W., Spohrer, K., He, X., Müller, J., (2012). Assessing crop water stress of winter wheat by thermography under different irrigation regimes in North China Plain. Int. J. Agric. Biol. Eng., 5: 24-34.

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