

# References

- Allen, R.G., Pereira, L.S., Raes, D., Smith, M., Ab, W., 1998. Crop evapotranspiration - guidelines for computing crop water requirements. FAO - Food Agric. Organ. United Nations Rome 1–15. <https://doi.org/10.1016/j.eja.2010.12.001>.
- API Reference [WWW Document], 2017. AccuWeather APIs (accessed 9.2.17). <https://developer.accuweather.com/accuweather-forecast-api/apis>.
- Chen, Q., Chen, T.Y., 1993. Estimation of river basin evapotranspiration over complex terrain using NOAA AVHRR data. *Acta Geogr. Sin.* 48 (1), 61–69.
- Cobaner, M., 2011. Evapotranspiration estimation by two different neuro-fuzzy inference systems. *J. Hydrol.* 398, 292–302. <https://doi.org/10.1016/j.jhydrol.2010.12.030>.
- da Cruz, M.A.A., Rodrigues, J.J.P.C., Al-Muhtadi, J., Korotaev, V., Albuquerque, V.H.C.,
- A reference model for internet of things middleware. 1 1. *IEEE Internet Things J.* 5. <https://doi.org/10.1109/JIOT.2018.2796561>.
- Davis, S.L., Dukes, M.D., 2010. Irrigation scheduling performance by evapotranspirationbased controllers. *Agric. Water Manage.* 98, 19–28. <https://doi.org/10.1016/j.agwat.2010.07.006>.
- Davis, S.L., Dukes, M.D., Miller, G.L., 2009. Landscape irrigation by evapotranspirationbased irrigation controllers under dry conditions in Southwest Florida. *Agric. Water Manage.* 96, 1828–1836. <https://doi.org/10.1016/j.agwat.2009.08.005>.
- Drucker, H., Burges, C.J., Kaufman, L., Smola, A., Vapnik, V., 1997. Support vector regression machines. *Adv. Neural Inf. Process. Syst.* 1, 155–161. <https://doi.org/10.1.1.10.4845>.
- G. o. I. NITI Aayog, 2015. Raising Agricultural Productivity and Making Farming Remunerative for Farmers [WWW Document] <http://niti.gov.in/content/raisingagricultural-productivity-and-making-farming-remunerative-farmers> (accessed 9.10.18).
- Hargreaves, George H., Samani, Zohrab A., 1985. Reference crop evapotranspiration from temperature. *Appl. Eng. Agric.* 1, 96–99. <https://doi.org/10.13031/2013.26773>.
- Gill, M.K., Asefa, T., Kemblowski, M.W., McKee, M., 2006. Soil moisture prediction using support vector machines. *J. Am. Water Resour. Assoc.* 42, 1033–1046. <https://doi.org/10.1111/j.1752-1688.2006.tb04512.x>.
- Goldstein, A., Fink, L., Meitin, A., Bohadana, S., Lutenberg, O., Ravid, G., 2017. Applying machine learning on sensor data for irrigation recommendations: revealing the agronomist’s tacit knowledge. *Precis. Agric.* 19, 421–444. <https://doi.org/10.1007/s11119-017-9527-4>.
- Gubbi, J., Buyya, R., Marusic, S., Palaniswami, M., 2013. Internet of things (IoT): a vision, architectural elements, and future directions. *Futur. Gener. Comput. Syst.* 29, 1645–1660. <https://doi.org/10.1016/j.future.2013.01.010>.
- Gutiérrez, J., Villa-medina, J.F., Nieto-Garibay, A., Porta-gándara, M.Á., Gutierrez, J., Villa-medina, J.F., Nieto-Garibay, A., Porta-Gandara, M.A., 2014. Automated irrigation system using a wireless sensor network and GPRS module. *IEEE Trans. Instrum. Meas.* 63, 166–176. <https://doi.org/10.1109/TIM.2013.2276487>.
- Hearst, M.A., Dumais, S.T., Osuna, E., Platt, J., Scholkopf, B., 1998. Support vector machines. *IEEE Intell. Syst. Appl.* 13, 18–28. <https://doi.org/10.1109/5254.708428>.
- 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. *Agric. Meteorol.* 24, 45–55. [https://doi.org/10.1016/0002-1571\(81\)90032-7](https://doi.org/10.1016/0002-1571(81)90032-7).
- Jaguey, J.G., Villa-Medina, J.F., Lopez-Guzman, A., Porta-Gandara, M.A., 2015. Smartphone irrigation sensor. *IEEE Sens. J.* 15, 5122–5127. <https://doi.org/10.1109/JSEN.2015.2435516>.
- Jones, J.W., Ritchie, J.T., 1990. Crop growth models. In: Hoffman, G.J., Howel, T.A., Solomon, K.H. (Eds.), *Management of Farm Irrigation System ASAE*, pp. 63–89.
- Shang, K.Z., Wang, S.G., Ma, Y.X., Zhou, Z.J., Wang, J.Y., Liu, H.L.Y., 2007. A scheme for calculating soil moisture content by using routine weather data. *Atmos. Chem. Phys.* 7, 5197–5206. <https://doi.org/10.1017/CBO9781107415324.004>.
- Kanungo, T., Mount, D.M., Netanyahu, N.S., Piatko, C.D., Silverman, R., Wu, A.Y., 2002. An efficient k-means clustering algorithm: analysis and implementation. *IEEE Trans. Pattern Anal. Mach. Intell.* 24, 881–892. <https://doi.org/10.1109/TPAMI.2002.1017616>.
- Mean Squared Error [WWW Document], 2018. Tutorvista (accessed 8.29.18). <https://math.tutorvista.com/statistics/mean-squared-error.html>.
- O’Shaughnessy, S.A., Evett, S.R., 2010. Canopy temperature based system effectively schedules and controls center pivot irrigation of cotton. *Agric. Water Manage.* 97, 1310–1316. <https://doi.org/10.1016/j.agwat.2010.03.012>.
- Ojha, T., Misra, S., Raghuvanshi, N.S., 2015. Wireless sensor networks for agriculture: the state-of-the-art in practice and future challenges. *Comput. Electron. Agric.* 118, 66–84. <https://doi.org/10.1016/j.compag.2015.08.011>.
- Pandey, V.S., Sharma, D., Shukla, A.K., Tyagi, S., 2017. A low-cost zigbee based temperature and humidity acquisition system for research and industrial applications. In: Dutta, C.R.K.S.D.K. (Ed.), *International Conference on Communication Computing and Networking*, pp. 379–385.
- Roopaai, M., Rad, P., Choo, K.R., Choo, R., 2017. Cloud of things in smart agriculture: intelligent irrigation monitoring by thermal imaging. *IEEE Cloud Comput.* 4, 10–15.
- Sharma, D., Shukla, A.K., Bhondekar, A.P., Ghanshyam, C., Ojha, A., 2016. A technical assessment of IOT for Indian agriculture sector. In: *IJCA Proc. Natl. Symp. Mod. Inf. Commun. Technol Digit. Indiapp.* 1–5.
- Theobald, C.M., 1974. Generalizations of mean square error applied to ridge regression. *J. R. Stat. Soc. Ser. B* 36, 103–106. <https://doi.org/10.2307/2984775>.
- United Nations, 2013. World population projected to reach 9.6 billion by 2050 [WWW

Document]. Dep. Econ. Soc. Aff. Popul. Div. United Nations (accessed 2.12.18).

<http://www.un.org/en/development/desa/news/population/un-report-worldpopulation-projected-to-reach-9-6-billion-by-2050.html>.

Viani, F., Bertolli, M., Salucci, M., Polo, A., 2017. Low-cost wireless monitoring and decision support for water saving in agriculture. *IEEE Sens. J.* 17, 4299–4309. <https://doi.org/10.1109/JSEN.2017.2705043>.

Wang, X., Yang, W., Wheaton, A., Cooley, N., Moran, B., 2010. Efficient registration of optical and IR images for automatic plant water stress assessment. *Comput. Electron. Agric.* 74, 230–237. <https://doi.org/10.1016/j.compag.2010.08.004>.

Weather API [WWW Document], 2012. Openweather (accessed 8.30.17). <https://openweathermap.org/api>.