





Exploring Machine Learning Models in Predicting Irrigation Groundwater Quality Indices for Effective Decision Making in Medjerda River Basin, Tunisia

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OUTLINE



BACKGROUND





SDGs 6 :Clean Water and Sanitation

To achieve sustainable development goals (SDGs 6) related to the efficient use of water, it is crucial to improve water management, rationalize the water irrigation uses and improve the tools of groundwater quality assessment.



- The assessment of groundwater quality require a sampling protocol, laboratory analysis, and at a larger scale, testing and data management which increase the cost and study time of water quality assessment and affects the decision-making on water quality management planning.
- To cope with these issues, it is crucial to develop a powerful and cost-effective approach for quick and accurate assessment of irrigation water quality.

Machine Learning models have opted for non-physical tool that successfully predicting groundwater quality

The Lower valley (LV) of Medjerda basin is part of the largest watershed of Tunisia, where it supplies about half of the country's drinking water.

It is essentially agricultural, where irrigation water supply depends on surface water in conjunction with groundwater resources.

Despite the importance of groundwater in the Medjerda basin, there is currently **a huge lack of data regarding its quality** that undermines the ability of decision makers and users to manage it properly.

Few groundwater sampling campaigns and analyses were conducted, and they are **therefore insufficient to fill the existing data gap and to give a real time information** about suitability of groundwater use.

Thus, improving the water quality evaluation process based on **non-cost data** using an objective tool with flexibility in its decision-making capacity for water management and planning is essential in the LV of Medjerda basin.

STUDY CONTEXT





The focus of this study was to test the performance of the novel approach to predict the suitability of groundwater quality for irrigation purposes useful to support decision-making under uncertainty in water resource management.

to evaluate the effectiveness of machine learning (ML) models to predict the irrigation water quality parameters (IWQ): TDS, PS, SAR, ESP, and MAR using physico-chemical parameters as input variables



to evaluate the accuracy of the implemented models;

to analyse the uncertainty and sensitivity of the tested models



METHODOLOGY

RF

AdaBoost

11.57

27.55

-0.09

0.91

-0.02

0.04

0.56

0.74

0.19

0.69



Results of tree 1

Results of tree 2

RESULTS

1. Statistical Analysis



- Electrical conductivity (EC) is a more correlated input variable with the predicted parameters than pH and temperature.
- The lowest correlations between T, pH, and EC prove that these parameters are **separable and non-redundant** and, therefore, are useful for improving the predictive accuracy of machine learning

2. Implementation and Evaluation of Model

RESULTS









- SVR model has significant values of RBIAS and RMSE compared with the other models for predicting the TDS parameter
- The ANN, RF, and AdaBoost models revealed high accuracy in predicting the TDS parameter during the learning process
- The AdaBoost model had a good performance in predicting all the IWQs parameters.
- The four models perform satisfactory for the prediction of the sodium absorption ratio (SAR) and the percent exchangeable sodium (ESP)
- Random Forest and artificial neural network models were unable to predict the MAR parameter

2. Implementation and Evaluation of Model







 The Pearson's coefficient values range from 0.65 to 0.94 for TDS, PS, SAR, and ESP over ANN and SVR models.

RESULTS

- RMSE showed an unacceptable performance for all models for the simulation of the TDS and MAR parameters.
- RBIAS showed a lowest performance for the SVR model for the simulation of the TDS and MAR parameters

ANN and SVR models presented very close results during the two processes for the prediction of all IWQs parameters

3. Relationship between observed & simulated variables over IWQs parameters



 It identifies a better distribution on the X = Y line for the random forest for all models

RESULTS

- SVR model has the weakest performance in predicting PS and SAR parameter followed by ANN and the RF in predicting TDS, PS, and SAR
- The predicted values are very close to the observed values for the AdaBoost model except for the MAR parameter



AdaBoost model has the best performance in predicting all IWQ parameters

4.Generalization ability (GA) indices of the models

To have useful models to predict new data sets, while avoiding errors, it is necessary to test its generalization capability.



• The ANN model for TDS is overfitted while for all other models are underfitted.

RESULTS

 The GA indices of the random forest and AdaBoost models are weaker than the ANN and SVR models.



5. Model Uncertainty Analysis

Parameter	Error	SVR	ANN	RF	AdaBoo st
TDS (mg L ⁻¹)	E	412.48	-27.01	4.79	11.57
	CB (95%)	142.56	55.07	50.65	27.55
PS (meq L ⁻¹)	E	0.45	-0.27	0.21	-0.09
	CB (95%)	0.96	1.00	0.97	0.91
SAR (meq ^{0.5} L ^{-0.5})	E	0.04	-0.36	-0.01	-0.02
	CB (95%)	0.37	0.47	0.09	0.04
ESP (%)	E	-1.45	-1.31	0.13	0.56
	CB (95%)	1.89	1.69	1.14	0.74
MAR (%)	E	0.27	-0.05	-0.02	0.19
	CB (96%)	2.47	2.01	1.47	0.69

SVR model has the highest confidence bound values (95%), followed by the ANN, RF, and AdaBoost models

6. Sensitivity analysis results

The sensitivity of the model provides an overview of the impact of input variables on the output. This analysis is necessary to assess how the model acts according to shifts in input values.

The models are more sensitive to:

- EC followed by temperature and pH, respectively for predicting TDS and MAR;
- pH for predicting ESP parameter;
- EC followed by the pH and the temperature, respectively for predicting PS and SAR.

AdaBoost model is the most sensitive model







- This research presents an effective use of machine learning models in forecasting the irrigation groundwater quality indices through low-cost data; using only physicochemical parameters as input variables without decreasing the efficiency of the models.
- These findings can be used as decision support systems (DSS) tool for sustainable water management in LV of Medjerda basin.
- The traditional simulation modelling approaches are dependent on datasets that involve a large amount of unknown or unspecified input data and generally consist of high-cost timeconsuming processes.
- Therefore, setting up a DSS based on machine learning models will boost the efficient use of water and rationalize its use by all water stakeholders at watershed level.









Article

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Sponsors : SMART IWRM Medjerda "Improving Sustainable Groundwater Management of the Lower valley of Medjerda basin" is the Research & Development project funded by the PEER cycle 7 program (NAS_USAID) and led by the Higher School of Engineers of Medjez El Bab (ESIM) and the United States Geological Survey (USGS).









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