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Using principal component analysis (PCA) in the investigation of aquifer storage and recovery (ASR) in Damascus Basin (Syria)

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Abstract A multivariate statistical technique was used to determine the major hydrochemical processes that control the groundwater quality variations during ''aquifer storage and recovery'' (ASR) operations. Principal component analysis (PCA) was applied on chemical and isotopic data of 83 groundwater samples. It has allowed the reduction of the 14 variables to four significant PCs (factors F) that explain 82.2 % of the total variance. F1 (54.1 %) represents the mineralization and groundwater nitrates pollution, whereas F2 (13.2 %) exhibits the isotopic enrichment by evaporation of waters components. Factor score of F1–F2 contains four groups: Injected freshwater (II) distinguished by low mineralization. Native groundwater characterized by relatively high mineralization. Mixing groundwater has intermediate mineralization. The shifting of native groundwater to injected water is related to the mixing ratio between these two components. Thus, the quality of groundwater samples situated close to the injected water is enhanced. Surface water (QF) is enriched by $\delta^{18}O$ and δ^2H compared to the other components due to evaporation. The spatial pattern of iso-factor scores maps is generally similar to the pattern of EC, CI^- , NO_3^- , and mixing ratio maps performed for the same experiment dates. Consequently, under the applied conditions of ASR process, the effective diameter is 250 m and 3 months of recovery time after complete injection. The results of this study clearly demonstrate the usefulness of multivariate statistical analysis as (PCA) in the ASR process investigation.

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Keywords Artificial recharge - Principal component analysis (PCA) · Groundwater quality · Drinking water · Syria

Introduction

Artificial recharge is regarded as groundwater management technique and generally termed ''Management of Aquifer Recharge'' (MAR). This term describes the augmentation of groundwater resources by artificial means including various forms of surface infiltration and direct well injection into the aquifers, stored there and recovered to meet the requirements of increasing water demand. It is widespread technique applied to cover different aspects of water management such as groundwater decline, saltwater intrusion and wastewater treatment (Dillon et al. [2006](#page-13-0); Guillaume and Xanthoulis [1996](#page-13-0)) and restore the groundwater balance (Dillon [2005\)](#page-13-0). In arid and semiarid areas, rainwater harvesting is considered as MAR method and experienced for thousands of years (Gal et al. [2002](#page-13-0)). Different engineering systems such as check dams and gravel pits are used to increase groundwater recharge in water scarcity areas (Martin-Rosales et al. [2007](#page-14-0)). Groundwater artificial recharge is used to manage seawater intrusion (Shammas [2008\)](#page-14-0). Common recharge techniques consists of riverbank filtration to improving source water quality were carried out in many countries (Ray et al. [2002;](#page-14-0) Regli et al. [2003](#page-14-0)).

Artificial storage and recovery (ASR) is considered as one of the MAR methods and applied in the worldwide at specially designed wells as open wells which are appropriate for very high injection rate for storage and recovery for reuse at demand. ASR is generally applied for different time scales, of which seasonal storage is the most

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frequently used to cover the water shortage during low season (Herczeg et al. [2004](#page-14-0)). Many studies have dealt with the feasibility of ASR (Pyne [1998](#page-14-0); Pavelic et al. [2002](#page-14-0)). The major source of water recharge used in aquifer storage and recovery (ASR) experiments is generally of good quality water particularly for potable water replenishment (Dillon [2005\)](#page-13-0). However, potable water or safe drinking water as defined by WHO Guidelines (WHO [2008,](#page-14-0) [2011](#page-14-0)) is good quality water suitable for human consumption (drinking, food and personal hygiene) and does not represent any significant risk to human health. Generally, the injection of high-quality water into adequate aquifers is used to enhance quantity and quality of groundwater reservoir. MAR processes are considered as a tool for sustainable water resources management to challenge many economic and environmental problems, e.g., mitigation the effects of climate changes (floods and droughts; Pyne [1995\)](#page-14-0). Generally, in arid and semiarid areas, water demand surpasses the water amount from renewable resources (e.g., the Middle East and Gulf region), and the excess water produced in desalination operation is used in ASR to restore groundwater reservoir and improve water quality (Mukhopadhyay and Al-Sulaimi [1998\)](#page-14-0).

Geochemical and isotope techniques are used in the investigation of MAR processes (Clark et al. [2004\)](#page-13-0). The environmental isotope methods are used in ASR when there is no significant geochemical difference between native groundwater and injected water (contrast) during injection and recovery processes. Consequently, environmental stable isotopes such as $\delta^{18}O$ and δ^2H are commonly used to identify mixing processes in the aquifers (Muir and Coplan [1981](#page-14-0); Ma and Spalding [1996](#page-14-0); Dillon et al. [2002](#page-13-0); Le Gal La Salle et al. [2005](#page-14-0)). Abou Zakhem and Hafez ([2012\)](#page-13-0) have used chemical and stable isotope techniques to investigate the effectiveness of ASR, compute the mixing ratios between water components (injected and native groundwater) and delineate its effective diameter and recovery time.

The aim of this study is using multivariate statistical technique in the investigation of artificial recharge to determine the major hydrochemical processes that control the variations of groundwater quality during ASR, to examine the correlations matrix among different variables and delineate the temporal and spatial variations of artificial recharge efficiency in the studied area. The multivariate statistical has the advantage over single tracers' analysis. Principal component analysis (PCA) was applied on the data from the same experimental site of ASR and will be compared with the interpretation of previous results.

Study area description

The lower part of Damascus Basin or the so-called Barada and Awaj basins is mainly formed by Damascus Oasis of about 1200 km^2 . The mean elevation of the Oasis is 650 m above sea level (a.s.l.). Damascus city is located in the northwestern part of the Oasis, while Oteibeh and Hijaneh ephemeral lakes are situated in the lower Damascus Basin and constitute the natural drainage of Barada and Awaj rivers, respectively (Fig. [1;](#page-2-0) Abou Zakhem and Hafez [2012](#page-13-0)). The study is under arid and semiarid Mediterranean climate type, which is characterized by rainy cold winters (November–May) and hot dry summer (June–September), with warm relatively short spring and autumn seasons (Soumi and Chayeb [1989\)](#page-14-0). The average annual precipitation is between 221 mm measured in Mazzeh and 136 mm in Damascus international airport gauges (Homsi et al. [1989](#page-14-0)). Damascus Oasis is a quaternary depression filled with detrital and lacustrine deposits, which are varied from pebble, gravel and conglomerates in the western part to fine sand and silty soils in the center of Damascus Oasis and change to loam, silt and clay of recent lacustrine formations surrounding Oteibeh and Hijaneh lakes (Ponikarov [1966](#page-14-0)). The catchment area of Barada and Awaj rivers is situated in highlands of the Anti-Lebanon Range in the northwest of the study area. These rivers are discharged during flood period into Oteibeh and Hijaneh, respectively. They form the major elements of the natural drainage system in Damascus Basin. The Barada River is mainly supplied by both Barada Spring $(3.12 \text{ m}^3/\text{s})$ and Fijeh Cretaceous Karstic Spring $(7.7 \text{ m}^3/\text{s})$ main annual discharges (Selkhozpromexport [1986](#page-14-0)). Figeh Spring is mainly used to supply Damascus city for drinking and domestic purposes. Quaternary alluvial aquifer system of 400–450 m thickness is mainly composed of pebble, gravel, sand, silt and clay lacustrine deposits. This aquifer compiles alluvial sediments complex including several sub-aquifers (multi layers) which are differentiated by clay content. The aquifer is generally assumed to be unconfined, becoming locally semi-confined related to the clay intercalations. Alluvial–proluvial mainly of loam and clay sediments constitutes the top of the aquifer (10–20 m). This aquifer is characterized by hydraulic conductivity varying between 3.7 and 142 m/day, while the transmissivity ranges from 165 to 3700 m²/day (Selkhozpromexport [1986\)](#page-14-0). The aquifer system is dramatically being exploited throughout the several thousands of wells drilled within the Damascus Oasis $(>=25,000)$ for drinking and irrigation purposes. Groundwater levels are continually dropped caused by over-exploitation of groundwater mainly for irrigation.

Fig. 1 (Above) Study location map. (Below) Groundwater level (March 2006) and sampling wells location map

During the last three decades, the groundwater levels have dropped to 40–50 m depth in the west and 100–120 m depth east of Damascus Oasis. Generally, the groundwater flow paths are from west to east. This direction of groundwater movement is parallel to that of the Barada and Aawaj rivers flow. The hydraulic head decreases from 650 m a.s.l. in the west to less than 500 m a.s.l. in the vicinity of Oteibeh Lake with hydraulic gradient of 0.004 ± 0.002 (Abou Zakhem and Hafez [2001\)](#page-13-0). The quality of groundwater is good (TDS < 0.9 g/l); however, socioeconomic and agricultural activities caused high pollution load since the 1990s. Consequently, the nitrate concentrations surpass the maximum concentration limit of Syrian standard (MCL: 50 mg/l) for drinking water in many areas (Abou Zakhem and Hafez [2009](#page-13-0), [2015](#page-13-0)).

Aquifer storage and recovery application

Aquifer storage and recovery (ASR) is applied in Mazraha pumping field during two hydrological cycles (2007 and 2008). Mazraha field is located at $36^{\circ}18'E$ and $33^{\circ}31'N$, and it includes 24 pumping wells used for drinking water supply (Fig. [1\)](#page-2-0). Groundwater level map (Fig. [1\)](#page-2-0) shows that flow directions are from NW to SE controlled by average gradient of 0.003. Subsequent to the sever decreases in groundwater level during the late 1990s, two open wells (I and II) were drilled and specially designed for very high injection rate. These wells are characterized by large diameter of about 3 and 30 m depth, 50 cm of perforated screen casing combined with 25 cm of gravel horde to

enhance infiltration and 6 aeration tubes of 30 m depth to avoid gas clogging during injection processes (Fig. 2). The aquifer is mainly composed of conglomerate, gravel and sand alternated with clay lenses overlaid by 8–10 m of loam and clay. The scheme of the ASR operation is to inject the excess of Fijeh springwater directly through the injection wells (I and II) during flood (February–May), for later extraction via 24 pumping wells during the dry season (July–October). Consequently, the ASR contributes to restore groundwater reservoir and enhance water quality, since water quality of Fijeh Spring is good. The injection process started in 6 March and finished in May 7, 2007 (62 days; Fig. [3\)](#page-4-0). The average injection rate was 7400 m³/day corresponding to the total injected water (TIW) of 461.8×10^3 m³. Pumping rate was about 19,500 m³ /day, between June 18 and August 31, 2007, and decreases gradually to $11,600 \text{ m}^3/\text{day}$ until March 5, 2008, forming a total pumped water (TPW) of 4.1 \times 10⁶ m³ during 261 days. The injection stage was between March 8 and April 23, 2008 (47 days), with constant injection rate of 8300 m³/day corresponding to 391.8 \times 10³ m³ of TIW. Pumping phase of about $10,137 \text{ m}^3/\text{day}$, between May 31 and June 2008, it increases to $15,500 \text{ m}^3/\text{day}$ in July and decreased gradually to $11,177 \text{ m}^3/\text{day}$ until December 2008 (215 days) forming 2.7×10^6 m³ of TPW. The TIW during ASR experiments represents only about 10–15 % of the TPW in Mazraha field (Abou Zakhem and Hafez [2012](#page-13-0)).

Sampling and analyses methodology

Thirteen groundwater samples were selected from active wells based on hydrogeology conditions (P2, P4, P10, P12, P16, P19, P20, P21, P25, Z1, LS, 9T and 11T), one injection well (II), and one sample representing surface water of Mazraha canal (QR) situated at 500 m north of injection center was sampled every 15 days during injection phase and once per month for recovery period during two hydrological cycles (2006–2007 and 2007–2008; Fig. [1](#page-2-0)). Groundwater samples were taken from active wells pumped continuously or pumped for a significant amount of time to get a representative sample. The samples were collected in 1- and 0.5-l new polyethylene bottles for chemical and isotopic analyses, respectively. All bottles had been rinsed three times with groundwater before filling it to capacity and then labeled accordingly. Prior to analysis in the laboratory, the samples were stored at a temperature below 4 °C . Physical parameters such as electrical conductivity (EC), temperature, pH and alkalinity were measured in the field. Stable isotope as $(\delta^{18}O \text{ and } \delta^2H)$ were analyzed using Finnigan Mat Mass Spectrometer DELTA $_{\text{plus}}$, after applying CO₂—water equilibration standard method (Epstein and Mayeda [1953\)](#page-13-0) with mea-Fig. 2 Schematic injection well (Abou Zakhem and Hafez [2012\)](#page-13-0) surement accuracy of ± 0.1 and ± 1 ‰, respectively. The

Inj.: Injection, Pum.: Pumping, Rec. V: Recovery volume, Rec. M: Recovery Mass.

Fig. 3 Water balance of artificial recharge (Abou Zakhem and Hafez [2012](#page-13-0))

tritium (^{3}H) values of water samples were measured by a liquid scintillation counter (Quantulus 1220) after electrolysis enrichment with an accuracy of ± 0.5 TU. Chemical analyses of major ions as Ca^{2+} , Mg^{2+} , Na^{+} , K^{+} , Cl^{-} , SO_4^2 , HCO_3^- and NO_3^- were analyzed using Chromatograph (Dionex 120). The analytical precision for the measurements of cations and anions was based on the ionic balance error (IBE). Generally, the precision of chemical analysis was about ± 5 %. Chemical and isotopic analyses were carried out in the laboratories of geology department (AECS). Reference materials and quality analysis are under the control of the National Quality Control Office in cooperation with the International Atomic Energy Agency (IAEA).

Principal component analysis (PCA)

PCA is considered as multivariate statistical technique that analyzes the data table including several observations and inter-correlated quantitative variables (I rows of observations and J columns of variables). The purpose of this analyzes is to extract the significant information from the data table and transform it to new orthogonal set called principal components (PC). The new representation condenses the data by maintaining only the high significant information, simplify the explanation and analyze the composition and structure of the data table (Abdi and Williams [2010](#page-13-0)).

Standardization of the original variables tends to remove the influence of measurement units rendering data dimensionless as expressed below:

$$
Z_i = (X_i - \mu) / \sigma \tag{1}
$$

where Z_i is the standardized value, X_i is the original value of parameter i, μ is the mean, and σ is the standard deviation of the data set (Davis [1973;](#page-13-0) Hamzaoui-Azaza et al. [2009](#page-14-0)).

The PCA method converts an original set of variables into a new orthogonal and uncorrelated set of principal components. Precisely, it decomposes the original matrix X (I observations, J variables) into scores and loadings matrices, as follows:

$$
X = TP' + E \tag{2}
$$

where each column of the matrix X is standardized; $T(I, I)$ J) represents the matrix of J principal components scores (each column of matrix T refers to a principal components); P' refers to the transpose of original data; and E, the residual matrix. Generally, the first principal components R are selected with high percentage of the total variance in the original matrix X ($R < J$), and thus, data reduction is realized (Iyer et al. [2003\)](#page-14-0). The PCs with eigenvalues ≥ 1 are rotated using varimax rotation to augment the participation of high significant variables and reduce the less significant variables based on there loadings and to generate a simple components structure (Kleinbaum et al. [1988](#page-14-0); Al-Tamir [2008](#page-13-0)).

Correlation matrix compiles the linear correlations coefficients $(-1 \text{ to } 1)$ between variables. High correlation coefficients reflect the significance of the relationship between two parameters. A positive coefficient demonstrates harmony and similarity between the correlated parameters. A negative coefficient exhibits opposite characters and evolution of these variables (Hamzaoui-Azaza et al. [2009\)](#page-14-0).

Results and discussion

Groundwater hydrochemistry

Chemical water quality is the major factor that determines its adequacy for drinking and public health purposes by comparing the water analysis with guideline values. The analytical results of physical and chemical parameters (EC, pH, T, Ca²⁺, Mg²⁺, Na⁺, K⁺, Cl⁻, SO₄²⁻, HCO₃⁻ and $NO₃⁻$) of water samples were matched with the guideline values of World Health Organization (WHO [2004,](#page-14-0) [2011\)](#page-14-0) and the Syrian standard (Ministry of Environment [1994\)](#page-14-0) for drinking water. Physical and chemical parameters, field measurements and isotopic data with mean (m) and standard deviation (SD), are presented in Table 1. Native groundwater has generally low mineralization, where water mineralization is defined as the water salinity measured by the concentrations of dissolved solid in water (cations and anions) or by electrical conductivity (EC), whereas the water is not valid for drinking purposes caused by the high nitrate concentration (mean \pm standard deviation; 68.7 ± 10.5 mg/l) which surpass the maximum concentration limit (MCL: 50 mg/l) in Syrian standards for drinking water (Abou Zakhem and Hafez [2009\)](#page-13-0). Native groundwater is characterized by relatively high EC and nitrate concentration $843 \pm 59 \,\mu\text{S/cm}$ and 68.7 ± 10.5 mg/l compared to the injected freshwater which has low EC and nitrate concentration 338 ± 28 μ S/ cm and 7.6 ± 0.8 mg/l, respectively. Consequently, the quality of mixed groundwater subsequent to injection operation is proportionally enhanced according to the mixing rate. Furthermore, the nitrate concentration has decreased below MCL of 50 mg/l $(38.8 \pm 16.5 \text{ mg/l})$ during the efficient period of ASR. Groundwater mean (m) temperatures and standard deviation (SD) values are 17.8 and $1 \degree C$, respectively, whereas injected water $m \pm SD$ are 14.9 \pm 1.1 °C and mixing groundwater are 17.4 \pm 1.3 °C. Groundwater mean pH and standard deviation values are 7.2 \pm 0.2, while injected water mean pH and standard deviation are 7.8 \pm 0.3. This relatively high pH is explained by the importance of recent waters derived from the fast flow component of Fijeh karstic spring during the flood period (Al-Charideh [2012](#page-13-0); UN-ESCWA and BGR [2013](#page-14-0)), characterized by low mineralization and moderate alkaline water, whereas native groundwater is characterized by relatively lower pH related to high mineral concentrations. This difference is further emphasized by PCAs.

Injected water and native groundwater have similar calcium bicarbonate $(Ca-HCO₃)$ category. Consequently, mixing groundwater has the same calcium bicarbonate type as well. The abundance of the major ions in native and mixing groundwater is in the following order: Ca^{2+} > $Mg^{2+} > Na^{+} > K^{+} = HCO_3^{-} > Cl^{-} > SO_4{}^{2-}$ $Mg^{2+} > Na^{+} > K^{+} = HCO_3^{-} > Cl^{-} > SO_4{}^{2-}$ $Mg^{2+} > Na^{+} > K^{+} = HCO_3^{-} > Cl^{-} > SO_4{}^{2-}$ (Fig. 4). Generally, the concentrations of cations and anions of groundwater samples are below the maximum acceptable level for dinking waters.

Principal component analysis (PCA)

PCA method allows reducing a large number of variables (measured physical parameters, major and minor elements in water samples), to smaller number of new orthogonal

Number of samples: (a) $n = 36$; (b) $n = 4$; (c) $n = 40$; and (d) $n = 3$

analyses $(m \pm SD)$

Fig. 4 Box plot of major elements (mini. 25 %, mean 75 % and maxi.) of mixing groundwater samples

factors illustrating the correlation matrix, without missing much information (Jackson [1991;](#page-14-0) Meglen [1992](#page-14-0); Cloutier et al. [2008\)](#page-13-0). PCA enable to generate eigenvectors of a variance covariance or a correlation matrix from a raw matrix (I observations and J variables; Hamzaoui-Azaza et al. [2009;](#page-14-0) Davis [1986](#page-13-0)).

PCA was applied to physico-chemical and isotopic data of 80 groundwater and 3 surface water samples from Mazraha station to extract the principal factors corresponding to the different processes that control water chemistry and sources of variation in the data during artificial recharge operations. Consequently, the 14 original variables produce 14 orthogonal factors called principal component (PC). As shown in Table [3,](#page-7-0) factor loadings control the contribution of the original variables in the PC, whereas factor scores measure the converted observations.

This technique was extensively used in hydrogeochemical characteristics and groundwater quality studies (Helena et al. [2000;](#page-14-0) Adams et al. [2001;](#page-13-0) Stamatis et al. [2011](#page-14-0)). XLSTAT software was used for PCA, and the 14 variables are temperature (*T*), pH, EC, δ^{18} O, δ^{2} H, ³H, Ca²⁺, Mg²⁺, K^+ , Na⁺, Cl⁻, SO₄²⁻, NO₃⁻ and HCO₃⁻.

Correlation matrix

PCA generates a correlation matrix including correlation coefficient (r) between variables. Correlation coefficient of 0.5 and above is considered significant and highlighted in bold (Table [2](#page-7-0)). High positive correlation coefficient is observed among HCO_3^- and Ca^{2+} ($r = 0.53$), Mg^{2+} and

 SO_4^{2-} ($r = 0.73$), Na⁺ and Cl⁻ gives ($r = 0.87$) and Na⁺ and SO_4^{2-} ($r = 0.81$). The positive and strong correlation among EC and Ca^{2+} , Mg^{2+} , Na^{+} , Cl^{-} and SO_4^{2-} is expected. These relationships reflect the lithological formation of the aquifer. Nevertheless, the correlations of NO_3 ⁻ with EC give ($r = 0.91$), Ca^{2+} ($r = 0.71$) and pH $(r = -0.56)$ which indicate to anthropogenic pollution.

The PCA was carried out by a diagonalization of the correlation matrix, so the problems arising from different measurement scales and numerical ranges of the original variables are avoided, since all variables are automatically auto-scaled to mean zero and variance unit.

Factor loadings

Table [3](#page-7-0) summarizes the PCA results including the loadings (participation of the original variables in the new ones) and the eigenvalue of each PC. The amount of variance (i.e., information) covered by each PC (also shown in Table [3\)](#page-7-0) depends on the relative value of its eigenvalue with respect to the total sum of eigenvalues. There are several criteria to identify the number of PCs (factors) to be retained in order to understand the underlying data structure (Jackson [1991](#page-14-0)). A plot of variance % of factors was used, which shows a change of slope after the fourth eigenvalue. Four factors were chosen based on their percentage contribution on the total cumulative variance (82.16 %; (Fig. [5\)](#page-7-0). Table [3](#page-7-0) includes factor loadings of 14 variables on these four PCs.

F1 explains 54.11 % of the variance and provide a good correlation with variables Ca^{2+} , Mg^{2+} , $Na^{\bar{+}}$, Cl^{-} , SO_4^{2-} ,

Table 2 Correlation matrix of 14 isotopic and physico-chemical variables

	\boldsymbol{T}	pH	EC	$\delta^{18}O$	$\delta^2 H$	$\rm ^3H$	Ca^{2+}	Mg^{2+}	$\rm K^+$	$Na+$	Cl^{-}	SO_4^{2-}	NO ₃	HCO ₃
\boldsymbol{T}	1													
pH	-0.63	1												
EC	0.60	-0.56	1											
$\delta^{18}O$	$0.00\,$	-0.03	0.39	$\mathbf{1}$										
$\delta^2 H$	-0.06	0.08	0.22	0.69	1									
$\rm ^3H$	0.04	-0.04	0.19	-0.06	-0.06	1								
Ca^{2+}	0.36	-0.27	0.80	0.34	0.23	0.25	1							
$\rm Mg^{2+}$	0.58	-0.51	0.92	0.21	0.12	0.29	0.78	1						
$\rm K^+$	0.24	-0.01	0.46	0.23	0.20	0.07	0.42	0.38	1					
$Na+$	0.54	-0.49	0.93	0.38	0.23	0.18	0.74	0.91	0.49	1				
Cl^-	0.54	-0.46	0.92	0.32	0.21	0.25	0.76	0.87	0.49	0.87	-1			
SO_4^2 ⁻	0.39	-0.34	0.78	0.41	0.32	0.22	0.62	0.73	0.37	0.81	0.74	1		
NO ₃	0.57	-0.56	0.91	0.25	0.11	0.19	0.71	0.88	0.36	0.88	0.93	0.69	1	
HCO ₃	0.56	-0.41	0.72	0.34	0.21	0.02	0.53	0.62	0.45	0.69	0.63	0.50	0.62	1

Bold numbers highlight significance above 0.5

Table 3 Loadings of 14 isotopic and physico-chemical variables

	F1	F ₂	F3	F4
τ	0.632	-0.439	-0.353	0.044
pH	-0.552	0.478	0.399	0.334
EC	0.978	-0.025	-0.014	-0.022
$\delta^{18}O$	0.401	0.763	-0.198	-0.237
$\delta^2 H$	0.264	0.820	-0.140	-0.226
$\rm{^{3}H}$	0.223	-0.154	0.808	-0.269
Ca^{2+}	0.810	0.109	0.233	0.037
$\rm Mg^{2+}$	0.928	-0.156	0.126	-0.052
K^+	0.511	0.249	0.129	0.724
$Na+$	0.954	0.014	0.017	0.020
Cl^{-}	0.937	-0.021	0.107	0.024
SO_4^2 ⁻	0.812	0.165	0.115	-0.140
NO_3^-	0.919	-0.154	0.019	-0.056
HCO ₃	0.751	0.028	-0.274	0.221
Eigenvalue	7.58	1.85	1.18	0.89
$%$ variance	54.11	13.21	8.45	6.39
Cumulative %	54.11	67.32	75.77	82.16

Bold numbers highlight significance above 0.5

 $NO₃⁻$, HCO₃⁻, T and EC, and negatively with pH as demonstrated by factor loadings (Table 3). This factor demonstrates groundwater nitrates pollution correlated to high Ca^{2+} , Mg^{2+} , Na^{+} , Cl^{-} , SO_4^{2-} , HCO_3^- concentrations, high EC and relatively low pH values. F2 expresses 13.21 % of the total variance and is mainly correlated with δ^{18} O and δ^{2} H, and negatively with temperatures. Based on

Fig. 5 Plot of variance % of factors

factor loadings, this factor exhibits the isotopic enrichments by evaporation of waters components. F3 explains 8.45 % of the variance and contributed by the variable as ³H. This factor reveals the transit time of groundwater and the tritium age of Fijeh springwater. F4 explains 6.39 % of the variance and involves variables as K^+ . The potassium is usually used in the detergent and can be considered as indicator of industrial pollution. This factor divulges the anthropogenic contamination. Factor loadings of F1 and F2 (67.3%) of 14 variables are shown in Fig. [6.](#page-8-0)

Factor scores

Factor score connected to F1 and F2 includes four water groups (Fig. [7](#page-9-0)):

Fig. 6 Factor loadings (14 variables) of F1 and F2 (67.3 %)

Group 1 has negative correlation with F1 and represents samples of injected water from Fijeh Spring (II). This group is distinguished by high pH and low $NO₃⁻, Ca²⁺,$ Mg^{2+} , Na⁺, Cl⁻, SO₄²⁻, HCO₃⁻ concentrations and low EC and T values. The water of this group is Ca-HCO₃ type.

Group 2 has positive correlation with F1 and involves samples of native groundwater before injection processes (P12, P16, 9T, LS and Z1). It is characterized by relatively low pH and high NO_3^- , Ca^{2+} , Mg^{2+} , Na^+ , Cl^- , SO_4^{2-} , $HCO₃⁻$ concentrations and high EC and T values. The water of this group is $Ca-HCO₃$ type.

Group 3 is related to F1 and represents mixing groundwater between native groundwater (group 2) and injected water (group 1). The shifting of native groundwater to injected water is related to the mixing ratio between these two components. Thus, the groundwater quality of the samples located toward the injected water samples is improved. This group is characterized by

relatively high pH and low $NO₃⁻, Ca²⁺, Mg²⁺, Na⁺, Cl⁻,$ SO_4^2 ⁻, HCO₃⁻ concentrations and low EC and T values. The water of mixing groundwater is $Ca-HCO₃$ type as well.

Group 4 is positively correlated to F2, where samples of surface water (QF) are enriched by $\delta^{18}O$ and δ^2H compared to native groundwater and injected water. The isotopic enrichment is considered as a strong indicator of evaporation effect on surface water that is clearly explained by F2. Consequently, surface water samples are located above the mixing line (F1) between these two components indicating no direct hydraulic connection between surface water and groundwater.

Factor scores F1 distribution maps

To examine the spatial variation of multivariate integrated physico-chemical parameters and for a better interpretation and display of the PCA results in space and time, iso-factor

Fig. 7 Factor scores (83 samples) of F1 and F2 (67.3 %)

maps were established. These maps were drawn using the values of factor scores (F1) of the observations (2 injections and 10 monitoring wells) between March 19 and July 31, 2007.

Samples with low negative factor scores correlated to F1 represent injected water from Fijeh Spring (wells I and II). These samples are characterized by high pH and low $NO₃⁻, Ca²⁺, Mg²⁺, Na⁺, Cl⁻, SO₄²⁻, HCO₃⁻ concentra$ tions and low EC and T values. The water of these samples is $Ca-HCO₃$ type. Samples with high positive factor scores correlated to F1 represent native groundwater before injection processes (P12, P16, 9T, LS and Z1). These samples are characterized by relatively low pH and high $NO₃⁻, Ca²⁺, Mg²⁺, Na⁺, Cl⁻, SO₄²⁻, HCO₃⁻ concentra$ tions and high EC and T values. The water of these samples is Ca-HCO₃ type.

Mixing groundwater between native groundwater and injected water is related to F1. The shifting of native groundwater to injected water depends on the mixing ratio between these two components and expressed by factor scores. These samples are characterized by relatively high pH and low NO_3^- , Ca^{2+} , Mg^{2+} , Na^+ , Cl^- , $SO_4{}^{2-}$, $HCO_3{}^$ concentrations and low EC and T values, while the samples with positive factor scores are characterized by relatively low pH and high NO_3^- , Ca^{2+} , Mg^{2+} , Na^+ , Cl^- , SO_4^{2-} , $HCO₃⁻$ concentrations and high EC and T values. The water of mixing samples is $Ca-HCO₃$ type as well.

To examine the spatial and temporal effectiveness of the artificial recharge operation by multivariate integrated parameters, we constructed iso-factor scores distribution maps of F1 during the injection and after the complete achievement of the injection process.

Factor scores distribution map of F1 (19-3-2007)

Iso-factor scores distribution map of F1 was performed after 13 days of injection (March 19, 2007; Fig. [8\)](#page-10-0). Positive anomalies (high factor scores) in the maps mean high

EC, NO₃⁻, Ca²⁺, Mg²⁺, Na⁺, Cl⁻, SO₄²⁻ and HCO₃⁻ concentrations, while negative anomalies (low factor scores) mean the opposite (i.e., low values of these chemical parameters). Figure [11](#page-12-0) shows that the factor scores tend to remarkably decrease down-gradient toward the injection center and very low factor scores are observed near two wells (I and II). The groundwater quality of the samples (P10, P21 and P25) having negative factor scores is enhanced. It is noteworthy that the spatial pattern observed in Fig. 8 is generally similar with the patterns of EC, CI^- , NO_3^- , and mixing ratio maps performed in March 19, 2007 (i.e., Fig. [9](#page-11-0) spatial distribution of EC value maps is performed for several dates). The mixing ratios between first end member (injected water) and second end member (native groundwater) are computed based on chloride mass balance which is considered as conservative element (Herczeg and Edmunds [2000](#page-14-0); Pavelic et al. [2005](#page-14-0); Abou Zakhem and Hafez [2012](#page-13-0)). Chemical equilibrium and the mixing ratios are computed using Hydrowin software.

Factor scores distribution map of F1 (15-5-2007)

Iso-factor scores distribution map of F1 was constructed after 62 days of injection in May 15, 2007 (Fig. [10\)](#page-11-0). Low factor scores in the maps mean low values of EC, $NO₃⁻$, Ca^{2+} , Mg^{2+} , Na^{+} , Cl^{-} , SO_4^{2-} and HCO_3^- concentrations. Figure [10](#page-11-0) shows the propagation of freshwater in the groundwater. Therefore, it is clear that negative factor scores cover almost the major part of the maps including I, II, P2, P10, P12, P20, P21, P19 and P25. Consequently, the groundwater quality of these samples is improved and

demonstrates the efficiency of the aquifer storage and recovery process.

Factor scores distribution map of F1 (26-6-2007)

Iso-factor scores map of F1 was performed after 50 days of the completion of injection process (Fig. [11\)](#page-12-0). According to this map, the groundwater dilutions plume progress to the south toward P2 and P21 including I, II P25, P10, P20 and P19 with negative factor scores. Thus, groundwater quality of these samples is improved by mixing and dilution process.

Factor scores distribution of F1 (31-7-2007)

After 85 days of complete injection process, iso-factor scores map of F1 was constructed (Fig. [12\)](#page-12-0). This map shows that the dilution plume centre is shifted to P2 including P19 and P21 characterized by negative factor scores. The displacement of the dilution plume is completely related to the hydrodynamic setting of the native groundwater and the application of artificial recharge operation conditions. It is clear that the spatial pattern of iso-factor scores maps observed in Figs. 8, [10](#page-11-0), [11](#page-12-0) and [12](#page-12-0) is generally similar with the patterns of EC (Fig. 10), Cl⁻, $NO₃⁻$ and mixing ratio maps, performed for the same experiment dates (Abou Zakhem and Hafez [2012](#page-13-0)). Consequently, under the applied conditions of aquifer storage and recovery process such as injected water amount, groundwater flow direction and its gradient and the boundary conditions of the aquifer, the estimated effective diameter is about 250 m.

Fig. 9 Spatial and temporal variation of EC value maps (Abou Zakhem and Hafez [2012\)](#page-13-0)

Iso-factor scores maps performed at several time intervals approve that the efficient time appropriate for recovery phase of the ASR is about 3 months subsequent to injection stage. Comparing with previous interpretation based on a single tracer. The mixing ratios were based on the chloride mass balance. The advantage of this new interpretation is

mainly dealing with multiples chemical and isotopic tracers in addition to physical parameters, where the mixing between native and injected water is expressed by F1. This interpretation is based on maximizing the important information and reducing the less important (e.g., the stable isotopes cannot be used as relevant tracers for mixing computation). In this case, PCA marginalizes the stable isotopes represented by F2 to differentiate between surface water and groundwater. Thus, the new factor score mapping is more substantial rigorous and global resuming all single trace maps. The results of this study clearly demonstrate the usefulness of multivariate statistical analysis as PCA in the interpretations of aquifer storage and recovery (ASR) process.

Conclusions

Principal component analysis (PCA) provide useful information not available at first glance. PCA was applied on chemical and isotopic data of 80 groundwater and 3 surface water samples (observations) and allowed the reduction of

the 14 variables to four significant PCs (factors F) that explain 82.2 % of the variance (i.e., information) of the original data set.

F1 (54.1 %) provide a positive correlation with Ca^{2+} , Mg^{2+} , Na⁺, Cl⁻, SO₄²⁻, NO₃⁻, HCO₃⁻ T, and EC, and negative correlation with pH. This factor demonstrates groundwater mineralization and nitrates pollution, whereas F2 (13.2 %) is related to $\delta^{18}O$, δ^2H and negatively with temperatures. This factor exhibits the isotopic enrichments by evaporation of waters components. F3 (8.5 %) contributed by $3H$ and reveals the transit time or the tritium age of groundwater. F4 (8.4 %) involve variables as K^+ related to the anthropogenic contamination.

Factor score of observations values linked to F1 and F2 includes four groups: Group 1 has negative correlation with F1 and represents injected water samples from Fijeh Spring (I and II). This group is distinguished by high pH and low mineralization, EC and T values. Group 2 has positive correlation with F1 and involves samples of native groundwater before injection processes. It is characterized by relatively low pH and high mineralization, EC and T values. Group 3 is related to F1 and represents mixing groundwater between injected water (group 1) and native groundwater (group 2). The shifting of native groundwater to injected water is related to the mixing ratio between these two components. Accordingly, the quality of groundwater samples placed close to the injected water is enhanced. Group 4 is positively correlated to F2 representing the surface water (QF). The isotopic enrichment by δ^{18} O and δ^2 H is considered as a strong indicator of evaporation effect on surface water that is clearly explained by F2. Consequently, surface water samples are located above the mixing line (F1) between these two components indicating no direct hydraulic connection between surface water and groundwater.

It is clear that the spatial pattern of iso-factor scores maps is generally similar with the patterns of EC , Cl^- , $NO₃⁻$ and mixing ratio maps performed for the same experiment dates. Consequently, under the applied conditions of aquifer storage and recovery process, the effective diameter is 250 m and 3 months suitable for recovery time after complete injection process.

The results of this study clearly demonstrate the usefulness of multivariate statistical analysis as (PCA) in the investigations of aquifer storage and recovery (ASR) process. Such approach is recommended as a helpful tool for sustainable water quality management and demonstrates the efficiency of the artificial recharge methods.

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