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Correlation of Selected Molecular Properties and Recovery Values in Volatile
 Organic Compounds Analysis: Comparison of Two Water Matrices

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10 Abstract

11 12 This study investigates if certain molecular properties can influence the recovery of 18 volatile organic 13 compounds (VOCs) in water under the applied analytical conditions, a purge and trap gas chromatographic 14 method with mass spectrometric detection (P&T GC MS). Statistical and quantitative structure property 15 relationship (QSPR) analyses were applied to find correlations between molecular parameters and analytical 16 recoveries in two different water matrices (clean water (CW) produced in the laboratory and natural 17 groundwater (GW)) at two different concentration levels. At 1µg/L, most compounds had higher recoveries in 18 CW than in GW, whereas at 15µg/L, the recoveries were higher in the GW matrix. Polarity number and 19 hydrophilic factor were correlated with the recovery differences at both concentration levels in GW. Polarity 20 was significant in the distinction of recovery differences for CW and GW matrices at low concentration, while 21 air diffusivity had an acceptable correlation with recovery differences for both matrices at the higher 22 concentration. Further correlation of the recoveries themselves with the molecular properties was made by 23 multivariate linear regression (MLR) resulting in a QSPR model. This was only possible for GW at the low 24 1 µg/L concentration. Partial least square analysis indicated that hydrophilic factor, polarity, and molecular 25 weight were the most important properties investigated. No significant correlation was found in CW matrix or in 26 the higher concentration level in GW matrix, which implies that the most significant properties might only be 27 relevant for the VOCs recovery at low concentrations and only if the matrix contains other water constituents 28 (DOC, salt).

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Keywords: VOC; Purge & Trap-GC/MS; recovery; water matrix; stepwise multiple linear regression; validation

1. Introduction

35 Volatile organic compounds (VOCs) may be present in different types of water. In surface water they 36 mostly originate from traffic pollution (benzene, toluene, ethylbenzene and xylene, e.g. BTEX compounds) and 37 agricultural or industrial pollution (i.e., chlorobenzenes, 1,2-dichloroethane, etc.), whereas in drinking water 38 they are mostly related to disinfection practices (i.e., trihalomethanes). In groundwater, their presence is usually 39 related to pollution caused by waste leakage or accidental pollution. The importance of their presence is 40 underlined in several publications.¹⁻⁴ These compounds are widely used as solvents, cosmetic products, fuels, 41 furnishings, etc. They can easily end up in wastewaters and may pose a risk to the environment. Some VOCs are 42 regulated by the Water Framework Directive⁵, but not all⁶. A year long survey in 1998/99 by Nikolau et al. 43 showed that in rivers and lake waters in Greece, VOCs concentrations were usually of the order of several ug/L 44 or lower. In wastewaters their concentrations were higher for some compounds (reaching tens or even hundreds 45 of $\mu g/L$).

Nowadays, VOC analysis in water is routine. Gas chromatography is mostly applied in combination with
 different extraction techniques: head space, purge/trap and solid phase micro-extraction. Different types of
 detectors can be used, such as flame ionisation, electron capture or mass spectrometric detection.⁸ The
 technique used in this research - purge and trap gas chromatographic method with mass spectrometric detection

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50 (P&T GC/MS), is recognised as highly sensitive and reliable for very low concentration ranges of various 51 compounds.6 52 A survey of the available literature shows an abundance of data on the precision and accuracy of the 53 different methods. However, almost no attempts have been made to compare recovery values for different 54 matrices. One exception is a paper by Barco-Bonilla et al.⁴ By comparing calibration curve ratios for different 55 wastewater effluents and liquid chromatography/mass spectrometry grade (LC/MS) water (ratios were within 56 the range of 0.8-1.2 for water spiked with 1 and 5 μ g/L), they showed that there were no matrix effects. For this 57 purpose, purge and trap (P&T) was coupled with gas chromatography (GC) triple quadruple mass spectrometry 58 and was used for the analysis of a membrane bioreactor, extended aeration, maturation pond and anaerobic 59 pond. In Standard Methods for the Examination of Water and Wastewater for Volatile Organic Compounds (Method 6200) published in 1998⁹ one can find data related to recovery values at a concentration level of 60 61 $0.5 \,\mu$ g/L in reagent water. They are within the range of 85-110% for the compounds-of-interest, with relative 62 standard deviations (RSD) in the range 6-10%. For reagent and raw water, recovery values can be found in EPA 63 524.2.¹⁰ (higher values for raw water for almost all analytes). Concentration levels are different and depend on 64 the water type. 65 66 Comparison of recovery values for different matrices is important from the point of view of analytical 67 efficiency. Changes in the recovery due to the matrix are influenced by water constituents and are not a means 68 for controlling the accuracy and precision of a method. Recovery is the amount of a compound that reaches the 69 GC for analysis relative to the amount that was originally present in the sample. The ideal recovery value is 70 100%. However, acceptable values range from 70-130% for most analytes in gas chromatography methods. In 71 P&T analysis it is essential to vaporise the substances and partition them into the gas phase. The vapour 72 pressure, solubility and extraction temperature affect the procedure. Once the procedure is established, we 73 assume that the influence of the water matrix, if present, can be understood as a possibility to alter the transfer 74 of the analyte into the gaseous phase due to interactions with the water constituents, while other analytical 75 conditions are kept constant as defined by the method. The aim of this study was to investigate if certain 76 molecular properties could influence recovery in the analysis of 18 volatile organic compounds (VOC) in two 77 different water matrices (i.e., clean water (CW) produced in the laboratory, and natural groundwater (GW)) at 78 two different concentration levels (1 and 15 μ g/L)). Statistical analysis followed by quantitative structure 79 property relationship analysis (QSPR) was applied. In general, QSPR models present relationships between 80 properties of series of molecules and their structural characteristics (derived either from experiments or 81 theoretically). Correlation of the selected molecular parameters and the recovery (assumed as a property under 82 the given analytical conditions) was not studied with the aim to predict recovery but rather to find out if any of 83 the selected molecular parameters could influence recovery under the applied analytical conditions. 84

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2. Materials and methods

2.1. Water matrices

The clean water matrix (CW) was produced in the laboratory by LABCONCO (WaterPro RO/PS Station,
Kansas City, USA) system (water of ASTM (American Society for Testing and Materials) Type I quality). It
was then boiled for 15 min and purged for 1 hour by nitrogen. It was stored in a glass bottle and prepared daily.
Dissolved organic carbon (DOC) content was less than the practical quantitation limit of 0.5 mg/L, and
electrical conductivity was 0.055 µS/cm. Measurements of DOC were performed after sample acidification with
concentrated hydrochloric acid to pH=2, according to method SRPS ISO 8245:2007.¹¹ pH measurement was
carried out on a WTW InoLab (Weilheim in Oberbayern, Germany) portable instrument.

98 The natural groundwater (GW) matrix was taken from the Danube riparian area (25-30 m depth) free from
99 VOC. Its characteristics were: 3.6 mg/L DOC, pH 7.02, electrical conductivity 770±101 μS/cm, dry residue of
100 415 mg/L (at 105°C, according to standard method 2540 B⁹).

2.2. Chemicals and reagents

Standards with concentrations of 2000 μg/mL in methanol as Volatile Organic Compounds Mix 7 (Dr.
Ehrenstorfer GmbH. Germany) (chloroform, 1,1,1-trichloroethane (1,1,1-TCE), 1,2-dichloroethane (1,2-DCE),
benzene, trichloroethene, bromodichloromethane (BDCM), dibromochloromethane (DBCM), bromoform, 1,4dichlorobenzene), Volatile Organic Compounds Mix 8 (Supelco) (chlorobenzene, ethylbenzene, o-xylene, mxylene, p-xylene, tetrachloroethene, toluene, 1,2-dichlorobenzene), vinyl chloride (Supelco) and internal

standard (IS) fluorobenzene (Supelco) were used to prepare working solutions. Dilution was done with 99.9%
 methanol (J.T. Baker, Avantor Performance Materials B.V., Deventer, The Netherlands).

2.3. Purge and trap conditions

A Tekmar Dohrmann 3100 Sample Concentrator with Vocarb 3000 trap (Carbopack B, 10 cm/Carboxen-1000, 6 cm/Carboxen 1001, 1 cm, *Supelco*, Sigma-Aldrich Co., St. Louis, USA) was used. 5mL aliquots of sample were dispensed into the 5 mL purging device with a gas tight syringe. The sample was purged with a stream of helium at 37.4 mL/min for 11 min at ambient temperature.

Before each sample analysis, the purge and trap system was baked (270°C for 3 min). This was followed by
 blank analysis of CW. After sample loading, desorption by heating the Vocarb 3000 trap was carried out at
 250°C for 2 min. The injector was set to split mode (30:1).

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2.4. Chromatographic and MSD conditions

124 An Agilent (USA) Gas chromatograph (GC) 7890A with 5975C Mass selective detector (MSD) was used. 125 The inlet temperature was 110°C. An Agilent J&W Scientific DB-5MS column (30 m x 0.25 mm x 0.25 µm) 126 was used. The oven temperature was set at 35°C (5 min) and raised to 100°C at 15°C/min and to 225°C at 127 25°C/min for 3 min, with a total run time of 20 min. Electron ionization was carried out at 70 eV, the source 128 temperature and GC interface were set at 230°C and 280°C respectively. Acquisition was performed in selected 129 ion monitoring/ scanning mode from 50-300 amu (SIM/SCAN). Quantification was carried out based on the 130 target ions given in Table 1. Compound confirmation was performed simultaneously based on the qualifier ions 131 (Table 1). The ions were selected from among the total ion chromatograms obtained by full mass range scans. 132 The GC carrier gas was helium of 99.999% purity, supplied by Messer (Novi Sad, Serbia).

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Table 1. Target ions and qualifiers in GC/MS analysis of VOCs

Compounds	Target ion	Qualifier ions	
Vinyl chloride	62	64	
Chloroform	83	85	
1,1,1 - TCE	97	99	
1, 2-DCE	62	98	
Benzene	78	77	
Trichloroethene	95	130, 132	
BDCM	83	127, 129	
Toluene	91	92	
DBCM	129	208, 173	
Tetrachloroethene	166	168, 129	
Chlorobenzene	112	77	
Ethylbenzene	91	106	
m+p- xylene	106	91	
o- xylene	106	91	
Bromoform	173	171, 252	
1,2-dichlorobenzene	146	148, 252	
1,4-dichlorobenzene	146	148, 252	
Fluorobenzene (IS)	96	97	

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2.5. Calibration and quantification

Calibration curves were obtained by spiking CW matrix with 17 VOCs mixture in methanol by gas-tight syringe, in accordance with the instructions given in Standard Methods $6200B^9$, with further sample processing of the samples as explained in Standard Methods $6200B^9$. 9 point calibration curves were made using concentrations of 0.4, 0.5, 1.0, 2.0, 5.0, 8.0, 10, 14 and 25 µg/L and for vinyl chloride within the range of 0.2-25 µg/L. The internal standard concentration was 10 µg/L. For the curves in the 0.4-25 µg/L concentration range, the coefficients of determination (R^2) were within the range of 0.991-0.998.

As a part of routine quality control, a calibration verification standard (CVS) from the same source and a laboratory control standard (LCS) from an independent source were used for verification. A criterion of $\pm 15\%$ of difference in comparison to the initial calibration was accepted in accordance with EPA 8000B.¹². The

148 obtained results indicated that all the measured values were within the range of $\pm 15\%$ of the expected value.

149RSDs for 50 measurements of the calibration verification standard (4 μ g/L CVS) were collected and for all150compounds were lower than 10%. LCS measurements ranged from 70% to130% of the expected concentration,151in accordance with EPA 8000B.¹²

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2.6. Method performance

EPA Methods 5030B ¹³ and 8260B ¹⁴ were used to develop an internal laboratory procedure for the analysis. Method detection limits were determined along the guidelines given by Glase et al. ¹⁵ (analyte added in a concentration which is 1-5 times the estimated method detection level (MDL)). The MDL was determined based on the RSD of 6 measurements of both spiked CW and GW matrix at a concentration level of $0.4\mu g/L$, except for vinyl chloride ($0.2\mu g/L$). The practical limit of quantitation (PQL) was calculated as 5 x MDL according to EPA Method 8260B.¹⁴ Repeatability for this concentration level was determined as a RSD of 6 consecutive measurements.

Precision was assessed as the RSD of recovery values determined for four samples analysed in 6 series in duplicates: clean water matrix at concentrations of 1 μ g/L and 15 μ g/L (CW₁ and CW₁₅) and groundwater matrix at the same concentrations (GW₁ and GW₁₅). One series consisted of a blank sample (non- spiked CW) and duplicates of the four mentioned samples. Additionally, the groundwater matrix (GW) was checked for the presence of VOC.

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2.7. Recovery comparison

The recovery comparison for the two water matrices was carried out by statistical analysis of 18 VOCs for
 replicates at both concentration levels, i.e. 1 and 15 μg/L. The concentration levels selected are similar to the
 VOC levels defined as relevant environmental quality standards for surface waters (Directive 2008/105/EC)¹⁶.
 Nine molecular parameters (descriptors) (water solubility (S, mg/L), Henry's law constant (HLC, atm-

177 178 m^{3} /mol), octanol-water partition coefficient (K_{ow}), air (D_{i.a}) and water diffusivity at 25°C (D_{i.w}) taken from USEPA,¹⁷ as well as molecular mass (MW), polarity number (Pol), hydrophilic factor (Hy) and molar 179 refractivity calculated by Dragon¹⁸) were selected for the study. Descriptor values are given in Supplementary 180 181 Table S2. The influence of molecular specific parameters on differences in recovery, i.e. its median value, was 182 analysed by finding correlations between differences in recovery for the following: two concentrations in clean 183 water (CW₁₋₁₅), two concentrations in groundwater (GW₁₋₁₅), a lower concentration in two different matrices (CW-GW)₁ and a higher concentration in two different matrices (GW-CW)₁₅ for compounds 1-18. In addition, 184 185 the recovery values themselves were correlated with the molecular descriptors.

186 In order to visualize similarities and differences in water matrices, hierarchical clustering analysis (HCA) 187 was used (Ward's linkage method). Furthermore, linear regression followed by stepwise multiple linear regression (MLR) (forward selection method) was applied^{19, 20}. QSPR models for median recovery values based 188 189 on a few deliberately selected explanatory variables (molecular descriptors) were constructed. The data were 190 organized in matrices X(18x9) where the rows represented the 18 investigated compounds (VOC), and the 191 columns corresponded to the 9 molecular parameters . The independent variables were the 9 selected descriptors 192 while the dependent variables were the observed median values of recovery for 12 measurements of each of the 193 18 VOCs. Additionally, the partial least squares (PLS) method was used since it is a well-known approach for the analysis of multidimensional data sets.²¹⁻²⁵ 194

All the calculations were carried out by STATISTICA v. 10.0.²⁷ The data were mean-centred (subtracting the mean and dividing by the standard deviation²⁶) before any statistical operation in order to prevent the highly abundant components from dominating in the final result over the components present in much smaller quantities.

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3. Results and discussion

3.1. Method performance

All the recovery values at the $0.4 \ \mu g/L$ concentration level ($0.2 \ \mu g/L$ for vinyl chloride) were within the acceptable range (70-130%), with the exceptions of chloroform and toluene, due to the laboratory contamination. This contamination affected only the low $0.4 \ \mu g/L$ concentration level. This caused the

208 determined MDL and PQL values for these analytes to be slightly higher (i.e. PQLs of 1.60 and 1.06 μ g/L 209 respectively). Generally speaking, the calculated MDL and PQL values were within the range found in Standard Methods⁹ and EPA Method 8260B¹⁴. In the case of BTEX, chlorobenzenes, chloroform, toluene, 210 211 trichloroethylene and tetrachloroethylene, the values were somewhat higher. Details on the method performance 212 and corresponding values from the literature are given in supplementary material S1. Based on the calculation 213 procedure explained in 2.6, different MDLs were obtained for different matrices. The MDL values in CW were 214 within the range of 0.027-0.32 μ g/L, while for GW the range was somewhat narrower (0.026-0.18 μ g/L), and 215 for most of the compounds the MDLs were lower (except vinyl chloride, 1,2-DCE and 1,1,1-TCE where there 216 were no differences between the samples), whereas for benzene it was higher. Repeatabilities for the $0.4 \,\mu g/L$ 217 concentration level (0.2 μ g/L for vinyl chloride) were determined as the RSD of 6 consecutive measurements 218 and were within the range of 1.2-13.2%.

It should be noted that the values from table S1 (given for evaluation of method performance) were not the basis for further recovery comparison at concentration levels of 1 and 15µg/L. The recovery values for concentrations of 1 and 15µg/L in different matrices were collected and further evaluated by QSPR. Average values for 12 replicates of each sample were in the range of 83-123% for CW matrix and in the range of 72-108% in GW matrix. Method precision at these concentrations was within the range of RSD 1.0-8.7% for CW matrix and within the range of RSD 2.0-9.4% for GW matrix.

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3.2. Comparison of recoveries

230 Box plots of recovery values for the 12 replicates of each substance at concentration levels of 1 and 15µg/L 231 and both matrices grouped by compound are given in Figure 1 and Figure 2. A parallel presentation of all the 232 performed measurements allows comparison of recovery values of individual samples. Boxes themselves and 233 their width indicate the degree of dispersion and skewness in the data. Central tendency (median) is connected 234 by line in order to facilitate comparison of the results obtained in different experimental conditions. In addition, 235 the quartiles are given as an indication of the dispersion of the recovery values. More precisely, the bottom and 236 the top of the box are the lower (Q_1) and upper (Q_3) quartiles, respectively. The ends of the whiskers represent 237 1.5 interquartile ranges (IQR= Q_3 - Q_1). The values below the lower whiskers (Q_1 -1.5 IQR) and above the upper 238 whiskers (Q_3 +1.5 IQR) are outliers. Consequently, extreme outliers are values beyond 3 IQR from Q_1 and Q_3 . 239 The asymmetrical position of the median in the boxes in Figures 1 and 2 is an indication that the recovery data 240 do not follow a normal distribution (most probably due to the limited number of data points) and hence 241 convenient descriptive statistics, such as mean and standard deviation, are not appropriate. 242



Fig 1. Box plot of recovery data for 12 replicates of 18 VOCs for clean water matrix at concentration levels of 246 $1 \mu g/L$ (CW₁) and 15 $\mu g/L$ (CW₁₅) (a) and for groundwater matrix at concentration levels of 1 $\mu g/L$ (GW₁) and 247 15 µg/L (GW₁₅) (b).



Fig 2. Box plot of recovery data for 12 replicates of 18 VOCs for for both clean and groundwater matrices at the concentration level of $1 \mu g/L$ (a) and $15 \mu g/L$ (b).

In clean water matrix (Figure 1a) for 1,2-dichloroethane, benzene, trichloroethylene and bromoform,
 recoveries are higher at the higher concentration level, while for vinyl chloride, 1,1,1-trichloroethane,
 bromodichloromethane, toluene, chlorodibromomethane, tetrachloroethylene, chlorobenzene, ethylbenzene, 1,2 dichlorobenzene and 1,4-dichlorobenzene, they are lower at the higher concentration level.
 (Figure 1b), all the compounds have higher recoveries at the higher concentration level.

Comparison of recovery values at the low concentration level for different matrices (Figure 2a) showed that
 the significant differences were not observed for benzene, trichloroethylene and bromoform, while for all the
 other compounds the values were higher in CW than in GW. In contrast, at the higher concentration level
 (Figure 2b), recoveries were higher in GW.

In all further consideration the medians were used, since they are not influenced by extreme outliers in thedata set, which is not the case with arithmetic mean values.

272 273 HCA was performed in order to visualize similarities (clusters formation) and differences (linkage distances 274 between clusters) of the recovery values for the two matrices and two concentration levels. The resulting 275 dendrogram is presented in Figure 3. The y-axis represents the corresponding linkage distances between the two 276 objects or clusters which are merged. The dendrogram reveals two distinct clusters: cluster A with recoveries 277 related to GW₁ and cluster B that included recoveries for CW₁ together with one sub-cluster with recoveries 278 observed in matrices spiked with higher concentration levels (GW₁₅ and CW₁₅). One can hypothesize that the 279 recovery values for GW_1 might be influenced by other factors than in the other three cases since they are 280 grouped in two separate clusters- A and B. Furthermore, the difference between the two matrices at the higher 281 concentration level is lower than at the lower concentration level of $1 \mu g/L$. 282



Fig 3. Dendrogram of two water matrices at two concentration levels (GW_1 - groundwater matrix with the 1µg/L concentration level, GW_{15} - groundwater matrix with the 15µg/L concentration level, CW_1 - clean water matrix with the 1µg/L concentration level, CW_{15} - clean water matrix with the 15µg/L concentration level) in the space of 18 VOC compounds obtained by the Ward linkage method using Euclidean distance

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3.2.1. Influence of chemical-specific molecular parameters on recovery

290 291 The differences evident in the recovery values obtained for the two matrices as well as the two concentration 292 levels may result from certain molecular characteristics that define the general behaviour of the compounds 293 during analysis, assuming that all the other analytical conditions are held constant and defined by the method. 294 These different intrinsic properties of molecules are usually expressed by various molecular descriptors. Their 295 selection is governed by general knowledge of the features that influence the transfer from water to air and 296 hence have a decisive influence on analytical recoveries while all the other analytical conditions are kept 297 constant. For example, transfer from water to air might be influenced by interactions between analytes and water 298 constituents. Our assumption is that water DOC is not extracted, nor purged nor injected into the GC during 299 P&T GC/MS analysis, since it is known from the literature that in natural groundwater a negligible portion of 300 the DOC is usually volatile. This *a priori* excludes the kind of influence by sorptive interaction of DOC with the 301 relevant instrument surfaces such that the only possible interactions are those of solutes and constituents in the

302 water phase. Correlations were therefore used as an exploratory tool to indirectly assess if the selected molecular 303 properties might influence analytical recoveries under controlled analytical conditions.

304 The influence of chemical molecular descriptors on recovery values was investigated for different matrices, 305 as well as for the different concentration levels for each matrix. Initially, the relationship between the 306 differences in recovery for the different concentrations in clean water (CW₁₋₁₅), for the different concentrations 307 in groundwater (GW₁₋₁₅), for the lower concentration in the two different matrices (CW-GW)₁ and for the higher 308 concentration in the two different matrices (GW-CW)₁₅ for VOCs and the selected chemical-specific parameters 309 was examined. The intention was to find out which parameter describes the observed differences with the best 310 fit. According to Pearson's correlation (Table 2), linear correlations for the observed differences in recovery and molecular descriptors appear in a wide range, but not higher than 0.68. According to Bevington ²⁸ the limiting 311 312 value for linear-correlation coefficients for an acceptable correlation of the 18 compounds at a probability level 313 of 0.05 is a minimum 0.468. Correlations exceeding this limit are bolded in Table 2 and are significant (high 314 enough for acceptable linear correlation). Compared to other molecular descriptors, polarity number (Pol) and 315 hydrophilic factor (Hy) had a significant linear correlation coefficient for the difference observed at the two 316 concentration levels in GW matrix. Pol is significant for the difference observed between CW and GW matrix at 317 the lower concentration level (bold in Table 2). At the higher concentration level, differences between the CW 318 and GW matrix showed significant linear correlation with air diffusivity (D_{ia}).

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320 **Table 2.** Significance of molecular specific parameters for recovery differences between two matrices and two concentration levels expressed by Pearson's correlation (p < 0.05, N=18).

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Recovery difference	S (mg/L)	log Kow	$\begin{array}{c} D_{i,a} \\ (cm^2\!/\!s) \end{array}$	$\begin{array}{c} D_{i,w} \ (cm^2/s) \end{array}$	MW	Pol	Ну	molar refractivity	HLC (atm- m ³ /mol)
CW ₁₋₁₅	0.2251	-0.1768	-0.2485	0.1754	-0.1490	-0.0672	-0.0763	-0.1065	-0.1059
GW ₁₋₁₅	-0.4063	0.3954	0.3499	-0.2330	-0.4185	0.5713	-0.6843	0.3802	-0.0030
$(CW-GW)_1$	-0.3489	0.4364	0.1945	-0.1774	-0.2437	0.4831	-0.3691	0.3545	-0.0209
(GW-CW) ₁₅	0.3603	-0.0589	-0.4974	0.3589	-0.2752	-0.0532	-0.0053	-0.0408	0.2821

323 S- water solubility, mg/L; HLC- Henry's law constant, atm- m^3 /mol; K_{ow}- octanol-water partition coefficient; 324 D_{i a}- air and D_{i w}- water diffusivity at 25°C taken from USEPA,¹⁷ MW- molecular mass; Pol- polarity number;

325 Hy- hydrophilic factor

 $Q^2 = 1 - \Sigma (Y_{pred} - Y_{exp})^2 / \Sigma (Y_{exp} - Y_{mean})^2$

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Subsequently, an attempt was made to correlate not the differences in recoveries for certain samples, but the
 recovery values themselves and more than one molecular descriptor by MLR. The goal was not to predict the
 recovery values but to investigate which among the selected molecular descriptors might influence them and
 under what conditions.

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Statistical parameters for the models calculated by stepwise regression on standardized data are shown in Table 333 3. The developed models were internally validated. The cross-validated regression coefficient (Q^2) is defined as:

(1)

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where Y_{pred} , Y_{exp} and Y_{mean} are the predicted, experimental, and mean values of the target property (recovery).

Q² higher than 0.5 is a necessary condition for the model to have predictive power, but still does not
 automatically imply high predictability.²⁹ Additional important cross-validation parameters accounting for a
 good estimate of the real predictive error of the model are PRESS (predictive residual error sum of squares),
 SSY (sum of squares of deviation of the dependent variable values from their mean) and PRESS/SSY that
 should be smaller than 0.4.³⁰ In Table 3, Model 1 shows the best statistics, although it is a borderline case for the
 PRESS/SSY ratio. Model 3 failed because it does not fulfil the statistical criteria (Q² > 0.5). The performance of
 model 4 cannot be accepted as it is over-parameterized (the ratio n/descriptors is >5).

Table 3. Internal validation statistics¹ for obtained QSPR models

	Molecular descriptor included in model	Multiple - R	Q²	Adjusted - R ²	F	р	PRESS	PRESS/SS Y	Model No.
CW	MW, Pol	0.7589	0.5760	0.5194	10.1869	0.0016	473.4190	0.424042	1
\mathbf{GW}_1	D _{i, a} , Pol	0.7484	0.5601	0.5014	9.5481	0.0021	491.1586	0.439931	2

GW ₁₅	S, Pol	0.5721	0.3274	0.2377	3.6500	0.0511	174.1626	0.672648	3
CW ₁₅	S, HLC, D _{i, a} , MW, Pol	0.8893	0.7909	0.7038	9.0776	0.0009	74.8705	0.289164	4

¹ R-correlation coefficients R^2_{adj} - adjusted square of the correlation coefficient ,p- significance level (set to $p \le 0.05$), and Fisher test for significance of the equation (F-test). The acceptance level for the individual 351 independent variable was set to 95% significance level.

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353 In Figure 4 the best constructed stepwise regression model describing recovery for the tested compounds in

- 354 GW_1 (model 1) is presented.
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Fig 4. Observed vs. predicted values of Recovery according to QSPR Model 1 from Table 3 (1 – vinyl chloride; 359 2 – chloroform; 3 – 1,1,1-trichloroethane; 4 – 1,2-dichloroethane; 5 – benzene; 6 – toluene; 7 – 360 bromodichloromethane; 8 – tetrachloroethylene; 9 – chlorodibromomethane; 10 – trichloerethylene; 11 – 361 chlorobenzene; 12 - ethylbenzene; 13 - m-xylene; 14 - p-xylene; 15 - o-Xylene; 16 - bromoform; 17 - 1,2-362 dichlorobenzene; 18 - 1,4-dichlorobenzene) with indicated 95% confidence interval band to the regression line. 363

364 Recovery values for compounds 2 (chloroform), 5 (benzene), 16 (bromoform) and 17, (1.2-dichlorbenzene))are 365 not well described by the model proposed (see Figure 3). Sorting the values for each descriptor in increasing 366 order one could find that these substances may be grouped only by HLC (all of them belong to the subgroup 367 which has values from 5.4 10^{-4} to 5.6 10^{-3} atm-m³/mol). However, six more substances belong to that group (o-368 xylene, chlorobenzene, 1,4-dichlorobenzene, bromodichloromethane, 1,2-dichloroethane and

369 chlorodibromomethane, numbered 15, 11, 18, 7, 4 and 9 respectively) but fit well with the proposed model. The 370 rest of the compounds have higher HLC values up to $2.7 \ 10^{-2} \text{ atm-m}^3/\text{mol}$.

371

372 MLR can be a good method for data analysis in cases when the descriptors are few in number, not significantly 373 collinear and if the equation has a good fit. However, if the number of descriptors is too large (in our case more 374 than 3 for 18 substances, like in model 4) MLR becomes inappropriate. Therefore we applied PLS regression. 375 The difference between MLR and PLS is that PLS fits the model simultaneously for all descriptors, while MLR 376 fits descriptors separately.³¹ On the other hand, comparison between MLR and PLS models is more abstract 377 which makes it difficult to understand and interpret. PLS regression is an extension of the multiple linear 378 regression model. The main purpose is to build a linear model transforming the original variables into the new 379 orthogonal variables, maximizing the description of a covariance between X and Y^{26} . 380

381 It was only possible to construct a PLS model for GW_1 , which is in accordance with findings related to results 382 presented in Figure 3 .Statistical data for the PLS model are given in Table 4. 63.33% of the sum of squares of 383 the dependent variables is explained by two PLS components. Figure 5 shows the observed vs the predicted

- 384 recovery values according to the PLS regression model for the tested compounds in GW₁. Compared to Figure
- 385 4, Figure 5 does not seem very much improved particularly for compounds 5 (benzene), 9
- 386 (chlorodibromomethane), 16 (bromoform) and 17 (1,2-dichlorobenzene). Improvement was achieved for 387 compounds 4 (1,2-dichloroethane) and 10 (trichloerethylene).
- 388
- 389 Table 4. Statistical data for PLS model related to GW1

390							
		R ² X(Cumul.)	Eigenvalues	R ² Y(Cumul.)	Q ² (Cumul.)	Significance	Iteration
	1	0.4254	3.5865	0.5517	0.4080	S	1
	2	0.6522	1.6407	0.6333	0.1449	NS	1

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Fig 5. Observed vs. predicted Recovery values for PLS model obtained for groundwater at concentration level 395 of 1µg/L (GW1) (1 - vinyl chloride; 2 - chloroform; 3 - 1,1,1-trichloroethane; 4 - 1,2-dichloroethane; 5 -396 benzene; 6 - toluene; 7 - bromodichloromethane; 8 - tetrachloroethylene; 9- chlorodibromomethane; 10 -397 trichloerethylene; 11 - chlorobenzene; 12 - ethylbenzene; 13 - m-xylene; 14 - p-xylene; 15 - o-xylene; 16 bromoform; 17 - 1,2-dichlorobenzene; 18 - 1,4-dichlorobenzene) with indicated 95% confidence interval band 398 399 to the regression line

400 401

402 In order to find out the significance of molecular specific parameters in the PLS model the variable importance

403 for the projection (VIP) is given in Fig 6. This allows quick identification of the explanatory variables

404 (descriptors) that contribute most to the model. The variables with higher VIP scores (e.g. power) are considered 405 more relevant.

- 406
- 407



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Fig 6. The VIP scores for molecular descriptors used in the PLS calculation for water matrix GW1

411 Thus the results obtained by PLS calculation show that the hydrophilic factor has the highest VIP score followed 412 by polarity, molecular weight and Di,a which were already indicated by the MLR.

413

414 Based on these results one can conclude that recovery values can only be correlated with some of the selected 415 molecular properties in the case of groundwater which contains DOC, analyte at a low concentration level 416 $(1\mu g/L)$ and salts (GW₁). A possible explanation might be related to the potential for physical or chemical 417 interaction of analytes and water constituents to the extent that could influence the recovery in an indirect way. 418 In the case of CW there is most probably no possibility for such interactions to occur so it is only logical that the 419 selected molecular properties have no influence on the recovery under the analytical conditions applied. As for 420 the groundwater matrix at higher concentrations of analytes (GW_{15}) and the same DOC and salt concentrations 421 as the low concentration level of spiked VOCs (GW_1), one possible reason for the absence of correlation with 422 the selected descriptors and recoveries might be a limited capacity for relevant interactions to alter the transfer 423 of analyte into the gas phase. Thus the influence might become negligible in comparison to GW₁. Testing this 424 hypothesis would require further work, at more concentration levels and in different natural matrices with 425 varying amounts of DOC amount and other characteristics, which is beyond the scope of this current work.

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Conclusions 4.

430 Statistical analysis of the dataset showed that the recovery values for VOCs have different behaviours depending 431 on the matrix type and concentration level. Comparison of the recovery values at the low concentration level 432 $(1\mu g/L)$ showed that for the majority of compounds the recovery values are higher in CW than in GW matrix. In 433 contrast, at the higher concentration level (15µg/L), recovery values are higher in GW matrix.

434

435 Using QSPR analysis, polarity number and hydrophilic factor were found to have an acceptable linear

- 436 correlation with differences between the recovery values at two concentration levels in GW matrix. Polarity was
- 437 significant in the distinction of recovery differences for CW and GW matrix at the low concentration level,

438 while air diffusivity had an acceptable value in the distinction of recovery differences for the two matrices at the 439 higher concentration.

440

441 Correlation between the molecular properties and the recovery values by MLR resulted in a QSPR model for the 442 groundwater matrix at the low VOC concentration. The QSPR includes molecular weight and substance polarity 443 as molecular parameters. This result was further supported by PLS, although with somewhat different results for 444 potentially important molecular parameters based on VIP scores: hydrophilic factor > polarity> molecular

445 weight. No significant correlation was found in either CW matrix or in GW matrix at the higher concentration

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level. This implies that theese most significant properties of VOC molecules might be relevant for the recovery
only at low VOC concentration levels and only if the matrix contains other water constituents (i.e. DOC, salts).

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References

- 458 1. S. Lacorte, L. Olivella, M. Rosell, M. Figueras, A. Ginebreda and D. Barcelo, *Chromatographia*, 2002, 56, 739.
- 460 2. E. Martinez, S. Lacorte, I. Llobet, P. Vianna and D. Barcelo, J. Chromatogr A., 2002, 959, 181.
- 461 3. A. Ikem, J. Food Compos. Anal., 2010, 23, 70.
- 462 4. N. Barco-Bonilla, P. Plaza- Bolaňos, J.L. Fernandez-Moreno, R. Romero-Gonzales, A.G. Frenich and J.L.
 463 Martinez Vidal, *Anal. Bioanal. Chem.*, 2011, 400, 3537.
- 464 5. WFD, Water Framework Directive 2000/60/EC, OJ L 327/1, 2000.
- 465 6. M. Rosell, S. Lacorte, A. Ginebreda and D. Barcelo, J. Chromatogr. A, 2003, 995,171.
- 7. Nikolau A.D., Golfinopoulos S.K., Kostopoulou M.N., Kolokythas G.A., lekkas T.D. (2002) Water Research
 36, 2883-2890.
- 8. P. Lepom, B. Brown, G. Hanke, R. Loos, P. Quevauviller and J. Wollgast, J. Chromatogr. A, 2009, 1216, 302.
- 470 9. AWWA-APHA-WEF Standard Methods for the Examination of Water and Wastewater, 20th ed., American
 471 Public Health Association/American Water Works Association/Water Environment Federation, Washington,
 472 DC, 1998. Volatile organic compounds (6200 B), pp 6-25
- 473 10. USEPA Method 524.2 Measurement of pureable organic compounds in water by capillary column gas
 474 chromatography/mass spectrometry, Revision 4, 1992.
- 475 11.SRPS ISO 8245:2007 Guidelines for determination of total organic carbon (TOC) and dissolved organic
 476 carbon (DOC) in water, 2007.
- 477 12. USEPA Method 8000B Determinative chromatographic separations, Revision 2, 1996b.
- 478 13. USEPA Method 5030B Purge-and-trap for aqueous samples, Revision 2, 1996a.
- 479 14. USEPA EPA Method 8260B Volatile organic compounds by gas chromatography/mass spectrometry
- 480 (GC/MS), Revision 2, 1996c.
- 481 15.A.A. Glase, D.L. Foerst, G.D. McKee, S.A. Quave and W.L. Budde WL, Environ. Sci. Tech., 1981, 15, 1426.
- 482 16. Directive 2008/105/EC, OJ L348/84, 24.12.2008
- 483 17. USEPA Supplemental guidance for developing soil screening levels for superfund sites. United States
 484 Environmental Protection Agency Solid Waste and Emergency Response, 2002.
- 485 <u>http://www.epa.gov/superfund/health/conmedia/soil/pdfs/part_5.pdf</u>. Accessed 15 June 2012.
- 18. Talete srl DRAGON (Software for Molecular Descriptor Calculation) Version 6.0 2011 <u>http://www.talete.mi.it</u>, 2011.
- 488 19. D. Cox, J. Royal Stat. Society: Series B, 1972, 34, 187.
- 489 20. S. Bergante, G. Facciotto and G. Minotta, *Central Europe Journal of Biology*, 2010, 5, 522.
- 490 21. D. Massart, B.Vandeginste, L. Buydens, S. DeJong, P. Lewi and J. Weyers-Verbeke, *Handbook of Chemometrics and Qualimetrics: Part A*, Elsevier: Amsterdam, 1997.
- 492 22. H. Martens and T. Naes, *Multivariate calibration by data compression*. In: Williams PC, Norris K (eds)
 493 Near-infrared Technology in Agricultural and Food Industries, American Association of Cereal Chemists: St.
 494 Paul, Minnesota, 1987.
- 495 23. H. Martens and T. Naes, *Multivariate Calibration*. John Wiley & Sons: New York, 1989.
- 496 24. H. Wold, *Soft modeling: the basic design and some extensions*. In: Jöreskog K, Wold H (eds) Systems Under
 497 Indirect Observation: Causality, Structure, Prediction, North-Holland: Amsterdam, 1981.
- 498 25. S. Wold, H. Martens and H. Wold, *The Multivariate Calibration Problem in Chemistry Solved by the PLS* 499 *Method.* Lecture Notes in Mathematics. Springer Verlag: Heidelberg, 1983.
- 500 26. P. Geladi, B. R. Kowalski, Analytica Chimica Acta, Volume 185, 1986, Pages 1-17
- 501 27. StatSoft Inc. STATISTICA (data analysis software system), version 10. <u>www.statsoft.com</u>, 2011, Tulsa, OK
 502 USA
- 503 28. P. R. Bevington, *Data Reduction and Error Analysis for the Physical Sciences*, McGraw Hill Book Co., New York, 1969.

- 505 29. A. Golbraikh and A. Tropsha, J. Mol. Graph. Model, 2002, 20, 269.
- 506 30. S. Wold, Mol. Inform., 1991, 10, 191.
- 507 31. T. Dijkstra, J. Econometrics, 1983, 22, 67.