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A quantitative review and meta-analysis on phytoscreening applied to aquifers contaminated by chlorinated ethenes

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## Science of the Total Environment

# A quantitative review and meta-analysis on phytoscreening applied to aquifers contaminated by chlorinated ethenes --Manuscript Draft--

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Abstract:	Applications and acceptance of phytoscreening, i.e., the use of trees as screening tools for underground contamination, are still limited in many countries due to the lack of awareness of application policies, the intrinsic qualitative nature of the technique, and the paucity of critical analyses on available data. To date, the conditions influencing the effectiveness of the technique have been descriptively discussed, yet rarely quantified. This review will contribute to filling this knowledge gap, shedding light on the most suitable approaches to apply phytoscreening. The focus was placed specifically on chlorinated ethene compounds since they are among the main organic contaminants in groundwater and have been the most studied in the field of phytoscreening. Chlorinated ethenes' behavior and biodegradation potential largely depend on their physicochemical properties as well as the hydrogeological features of the system in which they migrate. Besides, their fate and transport in surface ecosystems are still poorly understood. Here, phytoscreening data from sites contaminated by chlorinated ethenes were extracted from relevant literature to form a global-scale database. Data were statistically analyzed to identify the major drivers of variability in tree-cores concentration. Correlation between tree-core and groundwater concentration was quantified through Spearman's rank coefficients, whilst detectability potential was determined based on tree-cores showing non-detection of contaminants. The influence on such parameters of factors like contaminant phytovolatilization regulate correlation and detectability, respectively. Conditions increasing the correlation (e.g., sites with shallow and permeable aquifers) are recommended for phytoscreening applications aimed at mapping and monitoring contaminant plumes, whereas conditions increasing detectability (e.g., sampling tree-cores near ground level) are recommended to preliminary screen underground contamination in poorly investigated areas.
Response to Reviewers:	To the kind attention of Professor Jay Gan, Co Editor-in-chief, Science of the Total Environment Dear Professor Gan, Please find enclosed the revised manuscript 'A quantitative review and meta-analysis on phytoscreening applied to aquifers contaminated by chlorinated ethenes' submitted by myself on behalf of all co-authors for publication in STOTEN. This is the second revision of the manuscript, after the careful editing from 4 reviewers and your suggestion for English improvements. The paper is now significantly improved grammatically and in the clarity of the

contents.
An account of the changes we made can be found on the "Revised manuscript with changes marked" file.
Thank you for your consideration of this paper, please feel free to contact me for any further information or request concerning this manuscript.
Sincerely,
Carlotta Leoncini

## A quantitative review and meta-analysis on phytoscreening applied to aquifers contaminated by chlorinated ethenes

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## To the kind attention of Professor Jay Gan,

Co Editor-in-chief, Science of the Total Environment

Dear Professor Gan,

Please find enclosed the revised manuscript 'A quantitative review and meta-analysis on phytoscreening applied to aquifers contaminated by chlorinated ethenes' submitted by myself on behalf of all co-authors for publication in STOTEN.

This is the second revision of the manuscript, after the careful editing from 4 reviewers and your suggestion on English improvements.

The paper is now significantly improved grammatically and in the clarity of the contents.

An account of the changes we made can be found on the "Revised manuscript with changes marked" file.

Thank you for your consideration of this paper, please feel free to contact me for any further information or request concerning this manuscript.

Sincerely,

Carlotta Leoncini

## 1 A quantitative review and meta-analysis on phytoscreening applied to

## 2 aquifers contaminated by chlorinated ethenes

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#### 7 ABSTRACT

Applications and acceptance of phytoscreening, i.e., the use of trees to screen for as screening tools for 8 9 underground contamination, are still limited in many countries due to the lack of awareness on of application policies, the intrinsic qualitative nature of the screening methodtechnique, and the lack-paucity of critical 10 analyseis on available data. To date, the conditions influencing the effectiveness of the technique have been 11 12 descriptively discussed, yet rarely quantified. This review will contribute to filling this knowledge gap, shadedding light on the most suitable field and intrinsic conditions approaches to apply phytoscreening 13 towards effective use of phytoscreening, with. The-a-focus was placed specifically on chlorinated ethene 14 compounds -contaminantssince they are. Chlorinated ethenes are among the main organic contaminants in 15 16 groundwater and have been the most studied in the field of phytoscreening. Chlorinated ethenes' Their behavior and their biodegradation potential largely depend largely on their intrinsic physicochemical 17 properties-of the contaminants, but also on as well as- the hydrogeological features of the system in which 18 19 they migrate. Besides, tTheir fate and transport in surface ecosystems are still poorly understood. Here, phytoscreening data from sites contaminated by chlorinated ethenes were extracted from relevant literature to 20 form a global-scale database. DThe data were statistically analyzed statistically to identify the major factors 21 22 drivers of ing the\_variability inof tree-corespollutant concentration in tree cores. CC orrelation between treecore and groundwater concentration was quantified through Spearman's rank correlation-coefficients, whilst 23 24 and\_detectability potential was determined based on tree-cores showing non-detection of contaminants. {The 25 influence on such parameters of factors like n such correlation of contaminant properties, hydrogeology, tree features, and sampling/analytical protocols was assessed. Attention was also given to tree-cores that showed 26 non-detection of contaminants to identify the conditions leading to undetectability. Results suggest that 27

28	factors influencing <u>controlling plant</u> uptake and <u>contaminant</u> phytovolatilization <del>processes</del> regulate,
29	respectively, the correlation between tree core and groundwater concentration and the degree of detectability
30	in tree cores, respectively. Conditions favoring higherincreasing the correlation (e.g., sites with shallow and
31	permeable aquifers) are optimal-recommended for phytoscreening applications aimed at delineating mapping
32	and monitoring contaminant plumes, whereasilst factors conditions favoring higher increasing detectability
33	(e.g., sampling tree-cores near ground level) are ideal-recommended for to preliminary screening of
34	underground contamination in underinvestigated poorly investigated areas.

35 KEYWORDS: groundwater, CEs, trees, quantitative phytoscreening

36 1. INTRODUCTION

Tree roots can carry contaminants dissolved in water through the <u>xylem-trunk\_up</u> to the <u>leaf sectorleaves</u>. This transport is <u>due-based toon</u> direct contact of the roots with water <u>occurring</u>-inside the porous medium <u>that of thesurrounds the root zonerhizosphere</u>. Water can occur\_-in different energy states: free moving gravity water in the saturated zone of the aquifer, <u>the so called referred to as groundwater</u> (gw), <u>or</u>; retention water (rw) subjected to suction and attached to soil particles as capillary or pellicular water in the unsaturated zone.

43 The use of plants in environmental hydrogeology has risengained increasing attention during the last decades 44 in academic research and consultant activity due to stimulating applicationve perspectives. In conjunction 45 parallel with the removal of contaminants removal by direct uptake and degradation (phytoremediation), 46 Vroblesky et al. (1999) demonstrated for the first time that headspace analysis of tree-cores allows toean 47 delineate shallow gw contamination by chlorinated ethenes (CEs) such as trichloroethene (TCE) and cis 1,2-48 dichlorethene (cDCE). Later, Sorek et al. (2008) termed the technique "phytoscreening" and defined it as a 49 simple, fast, non-invasive, and inexpensive screening for detecting subsurface contamination by volatile 50 organic compounds (VOCs). Since Vroblesky et al. (1999), several comparisons between subsurface (soil, 51 soil gas, and gw) and plant contamination were documented conducted, especially mostly by using with the 52 use of tree-cores but also with by using leaf and branch samples (e.g., -Holm & Rotard, 2011; Wilcox & 53 Johnson, 2016; Gopalakrishnan et al., 2007).

54 <u>Besides Besides</u> screening and monitoring contaminant concentration for plume tracking or natural 55 attenuation evaluation (e.g., Larsen et al., 2008), phytoscreening was used to assess soil vapor intrusions

(e.g., Wilson et al., 2017; Algreen et al. 2015) and to age-date contamination events through dendroecology 56 (Balouet et al., 2007). Phytoscreening potential applicability was demonstrated for VOCs (e.g., BTEX; 57 58 Wilson et al., 2013), perchlorate (e.g., Limmer et al., 2014), per- and polyfluoroalkyl substances (PFAS; Gobelius et al., 2017), or inorganic compounds like heavy metals (e.g., Algreen et al., 2014). However, -59 60 although CEs wereare the most frequently target encountered in reported applications of this 61 techniquephytoscreening. This review is focusedexamines on phytoscreening applications for CEs in groundwater, for which for which exist literature provides sufficient literature information for ato conduct a 62 63 quantitative meta-analysis. These compounds are indeed particularly responsive forto uptake by plants, being 64 relatively small and moderately hydrophobic (Burken & Schnoor, 1998). In addition to their persistence, 65 ubiquity, and toxicity (Pankow & Cherry, 1996), characterization and monitoring of these plumes require 66 advanced technologies and onerous-high funding, that which could be alleviated mitigated by integrating a 67 time-, and cost-effective technique like phytoscreening. The chance to determine the occurrence of 68 subsurface volatile contaminants through trees is also important for evaluating the risks to human health such 69 as potential ingestion and respiration from vapor intrusion into buildings, which are exposure pathways 70 potentially associated with plant uptake. CEs detection in trees is affected by several contaminant-specific 71 loss mechanisms (e.g., volatilization, phytodegradation) which may result in lowerreduce concentrations in 72 plantvegetal tissues relative as compared to gw concentrations. 73 Phytoscreening of contaminated gw was indeed considered an valuable ecohydrogeological application 74 (Cantonati et al., 2020). However, -but, to make the screening technique broadly applicable and accepted, it 75 is necessary to identify the main control-factors that drive the correlation between gw and tree-core 76 contaminant concentration as well as the contaminants' detectability potential in trees in order to make this

screening technique widely applicable and accepted, of trees. The The identification of such factors would allow maximizing the correlation and detection capability can be maximized and wouldby providinge directions onfor the the optimization and standardization of sampling, analysis, and data interpretation procedures. Several studies concluded that the technique is only qualitatively reliable due to the poor correlation observed between gw and tree-core concentration (Holm and Rotard, 2011; Larsen et al., 2008; Ottosen et al., 2018). Such poor correlationThis was attributed to a variety ofmultiple factors that come at into stakeplay when dealing with living organisms (trees) to signal the state of contamination of an

84	environmental matrix <u>such as (gw.) with which p P</u> lants are-indeed involved interact with gw in-through a	
85	complex partitioning mechanisms mediated by various chemical, biological, hydrological, and climatic	
86	factors. Some sSynthesis efforts have beenwere directed to the studyinvestigate-of the the limiting factors	
87	that limit the and application opportunities of this technique. For As a general example, Trapp (2007)	
88	proposed a complex-theoretical model for the prediction of chemical uptake in trees_,-based upon more than	
89	30 parameters, either of hydrogeological or ecological nature, so-thus addressing the complexity of	
90	quantitative phytoscreening.	
91	Our work This paper provides a systematic review of former literature on the main factors that likely affect:	
92	a) the correlation between CEs tree-core and gw concentration. Factors conditions that that determine	
93	constrain higher correlations can be viewed as are favorable to apply phytoscreening to monitor,	
94	quantitatively, CEs contamination severity and degradation or natural attenuation processes.	
95	b) the CEs detectability potential in trees. Factors that determine conditions that constrain a lower	
96	number of contaminantcontaminants undetections non-detections in trees can be viewed asare	
97	favorable to preliminary screen for suspected underground contaminations by CEs in	
98	underinvestigated poorly investigated areas.	
99	Several constraining factors were selected and ,-grouped as follows: 1) physicochemical properties of the	
100	contaminants (molecular weight, water solubility, volatility, partition coefficients); 2) hydrogeology (depth	
101	to water table, aquifer thickness, hydraulic conductivity); 3) tree identity and anatomy (genus and family,	
102	xylem structure, tree trunk diameter); 4) sampling and extraction methodology (height above ground of tree-	
103	core sampling, tree-core length, extraction method of the contaminants).	
104	2. MATERIALS AND METHODS	
105	2.1 Data source	
106	A systematic search for relevant studies of phytoscreening on CEs was conducted in Scopus in January 2021.	
107	The search string was the following:	
108	TITLE-ABS-KEY (phytoscreening OR (tree AND groundwater AND (trichloroethene OR perchloroethene	
109	OR dichloroethene OR "chlorinated ethenes"))).	

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The first database search yielded 64 references. To form a global-scale database of phytoscreening data on 110 111 CEs, only the references containing datasets of contaminated sites that met the following criteria were 112 selected: (1) sampling by tree trunk coring, (2) tree-core analysis of at least one compound among PCE, TCE, and cDCE, and (3) spatial proximity between a given tree and a borehole where gw concentration 113 114 analysis showed concentrations above detection limits (the distances between the locations of trees and 115 boreholes varied from ~1 to ~10 m, or greater in few cases). When needed, we created contaminant concentration contour maps to supplement the reported gw concentration data. A total of 7 articles were 116 117 identified reporting site datasets suitable for this study. The reference lists of these 7 articles were manually 118 searched for further studies containing relevant datasets, providing 1 positive result (a technical report). 119 Some of the final 8 selected documents contributed with more than one investigated site, providing 120 information on a total of 11 sites. A total of 267 tree-core samples (and respective gw samples) were 121 compiled in the global database, some of them reporting more than one compound concentration (Table 1). 122 The number of tree-core concentration data was 419 (see Supplementary material for the databaseset), 118 123 being below the analytical detection limit (ND data hereafter). ND data represent a small fraction for most 124 sites (below 15%; e.g., Struckhoff et al., 2005), whereas in a few sites they are almost the majority (e.g., Larsen et al., 2008; Cox, 2002). The observations above detection limits are distributed as follows: PCE - 43 125 126 observations, TCE - 194 observations, cDCE - 124 observations, and the sum of CEs - 58 observations 127 relatedpertaining to one study that did not indicate single compound concentrations (Wittlingerova et al., 128 2013). No data were compiled for VC because only Ottosen et al. (2018) were able to detect traces of VC in 129 trees in very specificunder specific environmental conditions. No spatial or temporal average concentrations 130 were included in the database except for one site (Nogales site, Arizona; Duncan et al., 2017) where in which 131 only an average PCE average gw concentration (2 µg/L) was reported provided. It-This site was included in 132 the database for its significance in terms of uniqueness-in the database: cores were sampled and extracted 133 with methanol from trees of 4 different families inhabiting in an arid environment with a high DWTdeep aquifer table (9-10 m b.g.l.), -cores from trees of 4 different families were sampled and extracted with 134 135 methanol, showing PCE concentration up to 500 µg/kg.

Data <u>for on</u> contaminant concentration in tree-core samples and gw samples were reported in two different types of units, i.e., mass/mass (typically  $\mu g/kg$ ) and mass/volume (typically  $\mu g/L$ ), respectively. No

conversion was performed from the mass/mass unit to the mass/volume unit for tree-cores. This was 138 139 considered acceptable since the focus was on the correlation between the concentration in different matrices 140 along with the detectability potential. The unfeasibility of the conversion is mostly due to the lack of information on sampled tree-core dry weights and volumes. Besides, wood-water partition coefficients of the 141 142 contaminants would also be needed for a reliable conversion. Very few studies estimated the latter and for a 143 limited small number of tree species (e.g., poplars in Baduru et al., 2008). The unfeasibility of suchlack of 144 conversion hindered the possibility of performing multivariate statistical analyses. Due to this limitation, a 145 meta-analysis of the influence of each factor was performed to analyze the database. It is worth noting that 146 the the results of such this analysis will be subject to an intrinsic uncertainty associated with processing each

147 factor as independent of one another and singular.

Information on the factors influencing the correlation between tree-cores and gw concentration as well as the potential for detectability in trees (i.e., the hydrogeological conditions of the underlying contaminated aquifers, tree identity and anatomy, and sampling and extraction protocols; Table 2) were retrieved from the 8 selected documents articles and associated to each of the 419 tree-core concentration data.

152 As for hydrogeological parameters, involving either the permeable porous medium (aquifer) or gw flowing inside it, we retrieved: depth to water table (DWT in m below ground level), intended as the distance 153 154 between ground level and the surface at water pressure equal to atmospheric pressure (information retrieved 155 for all the 419 tree concentration data); thickness of the saturated portion of the aquifer (b in m) intended as 156 the distance between the water table and the low permeability bottom of the aquifer (retrieved for 235 out of 157 419 tree concentration); bulk saturated hydraulic conductivity of the aquifer (K in m/s; 373 out of 419 tree 158 concentration). When K values of the aquifers were not specified (25% of the total datasetdatabase), ranges 159 of conductivities were inferred in agreement with the local description of the lithology (Freeze & Cherry, 160 1979).

With regards to tree identity and anatomy, we retained information on the genus (known for 398 out of 419 data) and on the tree diameter at breast height (DBH in cm;-for\_142 out of 419 tree concentrationdata). From theBased on the genus, we were also able to report-retrieved the correspondent xylem structure, intended as the distribution of pores and vessels among growth rings. The xylem, or sapwood, is the active portion of the trunk where water transport takes place. We considered three main wood-xylem types: coniferous, diffuseporous, and ring-porous (Panshin & de Zeeuw, 1970). Coniferous xylems are characterized by small cells used for water transport and structural support. Diffuse-porous xylems additionally contain large vessels that are randomly distributed throughout the wood, while ring-porous xylems have larger diameter vessels concentrated in the early-wood. Conifers and diffuse-porous trees tend to have deep functional xylems as well as low average conductivity due to small and short conduits. In contrast, most of the conductance in ring-porous species is isolated to the outermost annual growth ring that contains functional vessels (Bush et al., 2010; Cermak et al., 1992).

When available, the sampling and extraction protocols used to prepare tree-cores for analysis were retrieved: length of tree-core samples (L in cm; 369 out of 419 tree-concentrationdata), sampling height along on the trunk (H in cm above ground level; known for 398 out of 419 tree-concentrationdata), and extraction method (419 out of 419 tree-concentrationdata) including extraction from dry vials (Cox, 2002; Larsen et al., 2008; Struckhoff et al., 2005; Vroblesky et al., 1999, 2004), vials containing organic-free water (Wittlingerova et al., 2013), or methanol solutions (Duncan et al., 2017), Solid Phase Micro-Extraction (SPME; Holm & Rotard, 2011).

180 2.2 Statistical meta-analysis

181 The correlation between gw and tree-core concentration was quantified by Spearman's rank correlation 182 coefficient  $\rho$  (Journel and Deutsch, 1997), a widely used approach for assessing the relationship between 183 parameters when it is expected to be non-linear, as in highly skewed datasets. Indeed, both tree-core and gw concentrations represent highly skewed data sets, as normally regularly found in contaminated sites (Juang et 184 185 al., 2001). As an example, tree-core concentrations had a distribution with positive skewness of 3.17, 8.21, 186 and 6.85 for PCE, TCE, and cDCE, respectively. A useful property of  $\rho$  is that its value is invariant to any 187 monotonic transformation applied to the data (e.g., logarithmic transformation). Outliers were not removed 188 from the databaseset due to the lack of knowledge about the uncertainty associated with the measurements. It 189 is worth noting that  $\rho$  is insensitive to outliers, therefore, representing a robust statistic tool-in the 190 abovementioned circumstance for our database. The detectability potential of contaminants in trees was 191 quantified as the percentage of ND (ND%) data to the total number of observations.

- 192 Spearman's ρ and ND% were calculated separately for each compound (PCE, TCE, and cDCE) to assess the
- influence of contaminant-specific properties such as molecular weight  $(M_w)$ , water solubility  $(S_w)$ , Henry's

194 constant ( $H_c$ ), and octanol-water partition coefficient (log  $K_{ow}$ ). In the case of the dataset of Wittlingerova et 195 al. (2013) reporting only sums of CEs (PCE, TCE, cDCE, tDCE, 1,1-DCE, and VC),  $\rho$  and ND% were 196 calculated on the sums.

- 197 <u>EThe factors of in</u> Table 2 were then <u>splitdivided</u> into intervals and Spearman's  $\rho$  were derived for the 198 concentration data associated withwithin each interval. For continuous factors (e.g., aquifer properties or tree 199 diameter), we determined discrete intervals based on medians and percentiles associated with each factor to 200 have a similar number of observations within each interval. Only concentration data above detection limits 201 were included in the countconsidered. In the case of discrete factors (e.g., tree species, extraction method), 202 only values associated with a minimum of 10 observations were considered in the statistical analysis, except 203 for one single case where only 8 observations were associated with extraction with methanol (Duncan et al., 204 2017). The ND% was determined within each of the aforementioned intervals to assess the influence of the 205 different factors on detectability. To avoid biases associated with trees that could be growing above more 206 dilute contamination areas, the ND% was calculated only when the concentration in gw was above 11  $\mu$ g/L. 207 This threshold was determined as the 5<sup>th</sup> percentile of gw concentration data. The final number of ND data 208 was 104 out of 118 ND tree-core data. It is noteworthy that in some cases, such as when processing 209 hydrogeological parameters like b and K, uncertainty on the results may be uncertain should be considered 210 since somecertain parameters do not appreciably vary spatially across a specific site.
- Results were interpreted in terms of high or low correlations (determining optimal factors conditions to characterize the contamination) and high or low detectability potential (determining optimal factors conditions to screen gw contamination).
- 214 3. RESULTS AND DISCUSSION
- 215 3.1 Contaminant properties
- The correlation between the concentration in tree-cores and gw was statistically meaningful (*p*-value≤0.05)
  for all four series (PCE, TCE, cDCE, and the sum of CEs).
- The  $\rho$  values indicate a low variability among CEs in terms of correlation between tree-core and gw concentration (Figure 1), with slightly lower values for higher chlorinated compounds PCE and TCE ( $\rho$  of 0.37 and 0.34, respectively) compared to the lower chlorinated cDCE ( $\rho$  of 0.41). On the other hand, ND% is widelygreatly different differed between higher chlorinated compounds PCE and TCE (12 and 23%,

respectively) and cDCE (47%) with the first two performing better in terms of detectability potential. It is noteworthy that the highest correlation coefficient (0.63) and lowest ND% value (0%) were found for the sum of CEs (reported by Wittlingerova et al., 2013). This could be explained by contaminant transformation processes taking place either in the rhizosphere or in the xylem (Newman and Reynolds, 2004) that would negatively affect the correlation of the single compounds. <u>However, t</u>This result is <u>although</u> associated with only one site.

228 The physicochemical properties of CEs (Table 3) likely influence the behavior of each compound in trees. Larsen et al. (2008) observed a better correlation for cDCE compared to TCE and PCE which was attributed 229 230 to the higher volatility (higher H<sub>c</sub>) of the latters possibly causing a higher loss through the bark (Vroblesky et 231 al., 1999). According to Limmer & Burken (2015), contaminant concentrations decreased with increasing 232 volatility, due either to volatilization from the roots, bark, or subsurface. Limmer & Burken (2015) reported 233 a decrease in contaminant concentrations with increasing volatility, attributing it to volatilization from either 234 the bark, the roots, or the subsurface. The results of oOur study agrees are consistent with these findings, 235 with PCE and TCE showing a slightly lower  $\rho$  compared to cDCE. In addition,  $\pm$ the correlation potential of 236 PCE and TCE may also be hindered by a their lower tendency for to plant uptake into trees compared to 237 cDCE, which is driven by their higher  $M_w$  and  $K_{ow}$ , and lower  $S_w$ , and higher  $K_{ow}$ . Uptake from tree roots 238 was indeed reported to be favored for compounds with low  $M_w$  (Baduru et al., 2008)-due to their higher 239 tendency to diffuse (Baduru et al., 2008), whereas higher Sw would possibly favor contaminant dissolution in 240 water and consequent tree uptake. Besides,

241 Hhigh sorption compounds (log K<sub>ow</sub>>3) were reported to were shown to have a higher tendencytend to be 242 absorbed\_primarily by root surfaces, resulting in less translocation within treesto the xylem (Schnoor et al., 243 1995). Similarly, a study by Dettenmaier et al. (2009) indicated that highly-high-hydrophilic compounds are 244 most likely to be uptakenabsorbed by plant roots and translocated transferred to the xylem. We may ean also 245 assumespeculate that once in the treexylem, higher log Kow compounds likely tend to get be absorbed in by 246 the xylem tissues, resulting in a prolonged accumulation in the tree and thus -and a higher detectability potential. PCE and TCE higher log Kow-can indee This could explain their the higher detectability potential 247 248 of PCE and TCE (lower ND%) compared to cDCE. Besides Moreover, PCE and TCE their higher H<sub>c</sub> can aid 249 the analytical detection when using headspace methods. Concurrently, the On the other hand, cDCE-lower Formatted: English (United States)

log  $K_{ow}$  and <u>lower-M<sub>w</sub> of cDCE may\_respectively</u>, hinder accumulation and favor contaminant loss <del>out</del> of<u>through</u> the bark <u>despite its low H<sub>c</sub>-despite the low H<sub>e</sub>-(Baduru et al., 2008), resulting in an overall lower detectability potential. The extremely rare detection of VC confirms the role of H<sub>c</sub>, M<sub>w</sub>, and log K<sub>ow</sub> in contaminant detectability in trees.</u>

254 3.2 Hydrogeology and aquifer parameters

255 Three intervals were considered for DWT: DWT<1, 1≤DWT<3, and DWT≥3 m b.g.l.- Concentration data 256 shows a statistically meaningfulsignificant correlation value (*p*-value≤0.05; Figure 2) in the three cases. A 257 slightly decreasing correlation with increasing DWT is-was observed ( $\rho$ =0.63, 0.54, and 0.52 for DWT<1, 1≤DWT<3, and DWT≥3 m b.g.l., respectively). On the other hand, The higher ND% was seen within the 258 259 shallow interval showed the higher ND% (35%) eompared towhereas the inferior intervals had significantly 260 lower ND% (16% and 17%, respectively, for the medium and the deep interval for the medium and deep 261 interval, respectively). Duncan & Brusseau (2018) assessed for the first time how DWT could affect the 262 correlation between VOCs concentration in tree tissues and gw (based on 100 measurements). They observed 263 a higher correlation in samples from sites with a DWT<4 m, concluding that a low thickness of the 264 unsaturated zone significantly affects phytoscreening efficiency. Despite the overall consistency of our 265 results with the cited literature (decreasing  $\rho$  with increasing DWT),  $\rho$  shows small differences among DWT 266 intervals, suggesting a low influence of this factor on the degree of correlation. The difference with Duncan 267 & Brusseau (2018) may lie in the use of different correlation coefficients and interval divisions. In our 268 analysis p was chosen due to the non-linear distribution of the concentration dataset while Duncan & 269 Brusseau (2018) assessed the correlation through Pearson's coefficient  $(r^2)$  thus assuming linearity of the 270 dataset. Besides, at sites with a more substantial vadose zone, mineralization of CEs can occur before 271 translocation of the contaminant in the tree due to more hypoxic conditions (Bradley & Chapelle, 2011), 272 leading to lower correlation potential for deeper aquifers. Similar findings have been reported by Wilson et 273 al. (2013) for BTEX translocation in trees. At the same time, when DWT is lower, volatilization loss of CEs 274 is promoted at the interface between the saturated and the vadose zone, being the thickness and water content 275 of the latter more subject to atmospheric variations (Pankow & Cherry, 1996). As tree roots are usually 276 located at this interface, the enhanced volatilization of CES-CEs could induce a lowering indecrease their 277 detectability potential (higher ND% for the shallow interval). At the same time, at sites with a more

substantial vadose zone, mineralization of CEs can occur prior to translocation of the contaminant in the tree
 due to more hypoxic conditions (Bradley & Chapelle, 2011), leading to higher correlation potential for
 shallower aquifers. Similar findings have been reported by Wilson et al. (2013) for BTEX translocation in
 trees.

Aquifer thicknesses were ranked-clustered into two intervals: b≤3.5 m. Data associated with lower 282 283 aquifer thickness b show –a very high positive correlation ( $\rho$ =0.71), whereas those associated with thicker 284 aquifershigher b have a significantly weaker correlation ( $\rho=0.30$ ). A reason for that This may be because that 285 CEs, in the majority of contaminant events, enter the subsoil as dense non-aqueous phase liquids (DNAPLs), 286 which tend to sink towards deeper sections of the aquifer, thus influencing the shape of dissolved 287 contaminant plumes (e.g., Parker et al., 2003). In particular, the sinking of DNAPLs-CEs could result in an 288 increased distance of the dissolved contaminant plume from the root zone in thicker aquifers. Notably, 289 aquifer thickness appears to have the highest influence on correlation compared to other factors. On the other 290 hand, a- lower ND% is associated with the thicker aquifer interval compared to the thinner intervalone (4% 291 and 19% for b>3.5 m and 19% for and b≤3.5 m, respectively). This result finds poor scientifichas poor 292 validations: we can speculate that thin aquifers have a lower geometrical probability of being intercepted by 293 tree roots than thick ones.

Two intervals of K were considered:  $K<1x10^{-5}$  and  $K\ge1x10^{-5}$  m/s. Concentrations referred to higher K values show a slightly higher correlation ( $\rho=0.73$ ) compared to those referred to lower conductivities ( $\rho=0.66$ ). These results suggest that K poorly affects the correlation between tree-core and gw concentration. On the other hand, the lower K interval includes 38% of ND whereas the higher K interval includes 18%. More permeable aquifers are therefore more suited in terms of detectability potential. A relatively higher permeability can indeed enhance the mobility of contaminants in the subsoil, likely favoring plant uptake, similarly to what happens when extracting gw from wells or soil gas from soil gas probes.

301 3.3 Tree identity and anatomy

The 22 tree genera of our data<u>base</u>set were <u>clustered bydivided according to their</u> families and <del>relative</del> xylem structure<u>s</u> (coniferous, diffuse-porous, ring-porous). Results show a significant positive correlation with most families (Figure 3), with  $\rho$  being highest for coniferous, i.e.\_-Pinaceae ( $\rho$ =0.86) and Cupressaceae ( $\rho$ =0.66). 305 These conifers also have moderately low ND% (24% and 30%, respectively for Pinaceae and Cupressaceae. 306 respectively). Consistently with our observations, Trapp et al. (2007) stated that conifers are best suited for 307 phytoscreening because they have a a-broad sapwood-xylem zone zone (the active portion of the stem), and transpire throughout the whole year, resulting in a continuous uptake of gw. Ring-porous Fagaceae 308 309 (primarily *Quercus*) presented a slightly positive correlation ( $\rho$ =0.39) and a high ND% (57%). The ring-310 porous structure likely promotes volatilization loss through the bark, -possibly affecting the observed low 311 correlation and detectability potential. Indeed, in ring porous trees, over 90% of water is transported in the 312 outermost growth ring in ring-porous trees whereas in diffuse-porous and coniferous trees water flow is more 313 equally distributed among rings (Ellmore and Ewers, 1986). Diffuse-porous Nyssaceae and Betulaceae show 314 a slightly lower  $\rho$  compared to conifersous trees (0.74 and 0.65, respectively), and a low ND% (29 and 25%, 315 respectively) although results for Nyssaceae must be taken with caution because they refer to one single 316 study site (Savannah River Site, USA; Vroblesky et al., 1999) where the aquifer was shallow (DWT<1 m 317 b.g.l.). On the other hand, diffuse-porous Salicaceae and Altingiaceae do not show a significant correlation 318 (p-value>0.05) and highly fluctuating ND% (very low for Salicaceae - 3% and very high for Altingiaceae -319 64%). The high variability among diffuse-porous families in terms of  $\rho$  and ND% could be associated with 320 different arrangements and sizes of the vessels regulating the conductivity of the sapwoodxylem, which in 321 turn can also vary with age. For example, Salicaceae (Salix and Populus), widely used in phytoscreening and 322 phytoremediation due to their fast growth, high uptake rates, and widespread occurrence in temperate 323 climates, showed the highestbest detectability potential but although no correlation between tree-core and gw 324 concentration. Besides, Negri et al. (2003) stated that Salicaceae are genetically predisposed to develop roots 325 extending to the water table at depths greater than 12 m b.g.l., thereby extending-widening their detectability 326 potential to deep aquifers. Altingiaceae (Liquidambar; present only in the study of Vroblesky et al., 1999) 327 also showed no correlation whilst a low detectability potential. This family was studied by Strycharz and 328 Newman (2009) in a greenhouse experiment where also Platanaceae and Salicaceae were involved. RThe 329 results showed that among the 3 families, Altingiaceae was the less recommended for phytoremediation 330 activities. Other diffuse-porous, like Betulaceae (Alnus and Betula), showed instead a correlation and 331 detectability potential comparable to coniferous. Lewis et al. (2015) calculated that a Betula pPendula can 332 accumulate similar quantities amounts of TCE as Populus trees due to its lack of heartwood (nonfunctioning 333 xylem) and homogeneous sapflow-xylem flow (Westhoff et al., 2008), making this species it an ideala 334 suitable candidate for phytoremediation and phytoscreening activities. Eventually, the low number 335 of Platanaceae (diffuse-porous), Ulmaceae (ring-porous), and Rosaceae (diffuse-porous) in the dataset database (Table 2) made unfeasiblehindered an analysis on these families. Even so, Limmer & Burken 336 337 (2015) showed that the genus Platanus (Platanaceae) had a high detectability potential, especially for PCE 338 gw contamination, if compared to non-Platanus trees. Their result was although associated with Platanus 339 trees growing primarily in areas with shallow gwroundwater. OppositelyConversely, our datadata associated 340 with on Platanaceae-trees in our dataset showed that 3 out of 4 times this family ese trees did not detect 341 contamination contaminants even with shallow gw (DWT<1 m b.g.l.; Savannah River Site, USA; Vroblesky 342 et al., 1999), although albeit in that particular site aquifer in that specific site (Savannah River Site, USA; 343 Vroblesky et al., 1999) K was very low (5.3x10<sup>-6</sup> m/s). Yung et al. (2017) pointed out that besides Populus 344 and Salix (Salicaceae, diffuse-porous) and Betula (Betulaceae, diffuse-porous), Quercus and Ulmus (Fagaceae and Ulmaceae, both ring-porous) are also efficient biomonitors of PCE and TCE contamination. 345 346 Notwithstanding the contrast with our results on Fagaceae, the 2 Ulmaceae trees in our dataset-database 347 (Carswell Golf Course, USA; Vroblesky et al., 2004) showed TCE concentrations above detection limits, with DWT at 1 and to 5 m b.g.l. and with an, -aquifer K of 7x10-5 m/s, and b of 0.9 m. This may suggest that 348 349 even among ring-porous trees a great variability in concentration results is expected.

350 Tree size (measured as DBH) in our database-database ranges between 18 and 102 cm and values were split into 3 intervals: DBH<25, 25≤DBH<40-em, and DBH≥40 cm. DSince data were filled only for 34% of the 351 352 database\_database (Table 2), so that results should must be taken considered with caution. The lower interval 353 shows a high positive correlation ( $\rho$ =0.79). The correlation decreases <u>significantly</u> in the medium interval 354 ( $\rho$ =0.47). The higher interval shows no significant correlation (*p*-value>0.05). ND% is instead comparable 355 among DBH intervals (10%, 8%, and 8%, respectively for DBH<25, 25≤DBH<40 cm, and DBH≥40 cm). 356 Several studies suggest that tree size has little effect on tree-core concentration (Limmer & Burken, 2015; 357 Wahyudi et al., 2012) while other studies demonstrated that diffusional loss in small trees (DBH of 2 cm) occurs at a rate 10 times higher than in trees with DBH 15 cm due to their greater surface area to volume 358 359 ratio that more quickly depletes the compound reservoir in the trunk (Schumacher et al., 2004; Struckhoff, 360 2003). This could explain the slightly higher ND% of smaller trees. In contrast, our results show that smaller

trees have greater efficiency in terms of quantitative analysis of a-gw contamination (high  $\rho$ ). In smaller trees, we could indeed expect less variability in concentration around and across the trunk due to a less compartmentalized flow in the sapwoodxylem. Also, in smaller trees it is highly probable to sample a consistent thickness of the total sapwoodxylem, resulting in a concentration that averages the radial variability. More variability is instead observed in larger trees where sampling direction has a strong influence on the concentration, thus influencing the correlation.

367 3.4 Sampling and analysis protocols

368 In our dataset\_database\_tree-core L ranges from 3.8 to 12.5 cm and was clusteredarranged into two intervals: 369  $L \le 6$  and L > 6 cm. Shorter cores do not show a significantly correlation between e with tree-core and gw 370 concentration (*p*-value>0.05) whereas longer cores have a high positive correlation ( $\rho$ =0.71; Figure 6). This 371 result finds agreement agrees with-in Ma & Burken (2003) and-the the USGS user guide published by 372 Vroblesky (2008), which suggesting a better correlation when the core samples are longer than  $\sim$ 7-8 cm. 373 Shorter cores may be also acceptable for ring-porous trees, in which water transport takes place mostly in the 374 outermost ring (Ellmore and Ewers, 1986). The detectability potential is lower for the longer coresthe higher 375 L interval interval (ND of 11% and 30%%-\_for L≤6 cm and 30% for and L>6 cm, respectively), likely 376 possibly-indicating that drilling longer tree-cores could promote diffusional loss out of the sample since 377 sampling employs requires relatively higher longer sampling timesperiods (tree-cores are usually cut in 378 smaller pieces before being put in the vials) that can possibly promote diffusional loss out of the sample.

379 The sSampling-elevation, H from the base of the trunk-(m a.g.l.), ranges between 50 and 900 cm a.g.l. and 380 was clusteredsplit into three intervals: H<99, 99<H≤120, and H>120 cm a.g.l.. The medium interval 381 99<H $\leq$ 120 cm shows a very high correlation between tree-core and gw concentration ( $\rho$ =0.84; Figure 6), 382 which decreases above 120 cm a.g.l. for higher H ( $\rho$ =0.42). Sampling height below 99 cm The lower interval 383 shows no significant correlation (p-value>0.05) although the data pertain to only one survey by Holm & 384 Rotard (2011). The reason for the high  $\rho$  value at medium sampling height H is still unknown. but wWe can 385 speculate that this is related to the attainment of an equilibrium of the contaminant inside the wood-air-water 386 partitioning system. On the other hand, a low number of ND% data was registered for the lowest lower H 387 interval (7%) whereas the higher intervals <u>99<H≤120 and H>120 cm a.g.l.</u> are associated with higher ND% 388 (24% and 33% for the medium and higher H interval<sub>37</sub> respectively). This result is consistent with the experiments of Ma & Burken (2003) where a higher TCE loss was observed higher up the trunk. Thus, a lower rate of diffusional loss from the bark may be expected for the lowers H intervals. On the other hand, Ottosen et al. (2018) sampled tree-cores just above the ground surface without distinguishing a clear advantage from this sampling strategy.

393 Our databaseset includes 4 extraction methods used for analysis (dry sample, organic-free water, methanol 394 solution, and SPME; Figure 6). The dry and the water extracted samples showed moderately high correlation 395 potential (p=0.55 and p=0.64, for dry and water extraction, respectively for dry and water extraction). However, dData from water extracted samples although-pertain to only one survey where concentration data 396 397 corresponds-relate to the sum of CEs (Wittlingerova et al., 2013). The high  $\rho$  could be associated either with 398 the extraction method or with the fact that the sum of CEs was considered (see section 3.1). Methanol 399 extracted samples, pertaining related to the study cases of Duncan et al. (2017), were collected with 400 unfavorable conditions in arid--and-hot environments (Nogales site, Park-Euclid, and Motorola 52<sup>ad</sup> 401 superfund site in Arizona, USA). The small number of tree-cores (8) sampled with this method can be the 402 reason for its associated inmay explain the observed non-significant correlation\_value-(p-value>0.05). 403 Besides, Teree-core concentrations sampled in particularly associated with arid environments are could be 404 possibly\_more\_likely\_a function of associated with vadose zone vapor phase concentration in the vadose zone 405 rather than gw concentration since tree roots would unlikely reach a DWT of 26 m b.g.l. as in the case of the 406 Park-Euclid site. In support of this hypothesis, The fact that we also observed that 5 out of 8 tree-core 407 methanol concentrations of PCE were higher than 300 µg/kg although despite being associated with very low 408 gw concentrations of PCE of  $2 \mu g/L$  (2  $\mu g/L$ ; i.e., lower than the gw concentration threshold we defined for 409 calculations on ND% calculation; See Section 2.2,-) further supports the hypothesis suggesting that trees 410 were absorbing -uptake of contaminants from a matrix different other than gw. This may have further 411 hindered correlation with gwin the case of methanol extraction. The analysis following SPME, associated 412 with only one site (Potsdam-Krampnitz military base, Germany; Holm & Rotard, 2011), also showed no significant correlation (p-value>0.05). Even so, tThe ND% was very low for SPME (7%) and as well as 413 414 water extracted samples (0%). However, since these methods were used in single study cases, although these 415 results may be associated withrelated to other site-specific conditions or sampling protocols-since these

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416	methods pertain to single study cases. In terms of ND%, tThe dry and methanol extraction showed-produced
417	comparable results in terms of ND%-(34%, and 33% for dry and methanol extraction, respectively).
418	4. SUMMARY AND CONCLUSIONS
419	The efficiency effectiveness of phytoscreening has been tested assessed through via a meta-analysis of
420	literature datato define determine the potential of trees to (a) the potential of trees to monitor groundwater
421	plumes of CEs, here expressed as a-the degree of correlation between tree-core and gw concentration, and (b)
422	the potential of trees to detect the occurrence of groundwater gw contamination events by CEs in poorly
423	underinvestigated areas, here-expressed as a-the percentage-rate of tree-cores that showed concentrations
424	below-above the detection limit in significantly contaminated areass in the occurrence of groundwater
425	contamination. To these aims, sSeveral factors possibly likely influencing correlation and detectability were
426	taken into accountconsidered. These factors included, namely the physicochemical properties of CEs, the
427	hydrogeological conditions of the underlying contaminated aquifers, tree $\underline{s}$ identity and anatomy, and
428	sampling and extraction protocols.
429	The correlation (quantitative quantitative monitoring potential) appears to beis higher when (1)-the
430	hydrogeological dynamics favor direct uptake of contaminated water, and $(2)$ -the concentration is
431	homogeneously distributed in the tree and the tree-core sample. Uptake-Direct tree uptake of contaminated

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432 gw-is favored for the lighter and more soluble cDCE .- and in the case of in the case of shallow water tables 433 (DWT<3 m b.g.l.), and in thin (b<3.5 m), and permeable aquifers (b<3.5 m;  $K \ge 1x10^{-5}$  m/s). The 434 homogeneity of concentration in the sapwood-xylem is likely higher for Pinaceae and Cupressaceae 435 (coniferous), due to their non-porous xylem, and in-for\_smaller diameter trees (DBH<25 cm), whereas homogeneity of concentration in the tree-core is enhancedfacilitated in the casewhen sampling of longer tree-436

437 cores (L>6 cm), and possibly at a sampling height along the stemon the trunk between  $99 \le H \le 120$  cm a.g.lr.

438 The detectability (qualitative screening potential) is higher when factor conditions favor accumulation in the

439 xylem and hinder volatilization loss through the bark. In these terms, PCE and TCE are more suited 440 compared to cDCE due to This is the case of PCE and TCE thanks to their higher sorption and weight. Low

442 height <u>onalong</u> the trunk (H<99 cm a.g.l.), and for shorter cores <u>due to reduced time of sampling (L $\leq$ 6 cm)</u>

to reduced time of sampling. In the case of Salicaceae, high uptake rates may compensate for 443 due

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volatilization loss was also inferred in the case of large-diameter trees (DBH≥40 cm), at a low sampling

444	volatilization losses, thus increasing detectability. EventuallyFinally, the process of contaminant extraction
445	has also an effect on detectability that seemappears to be maximized when using organic-free water
446	extraction and SPME.
447	Despite the clarifications provided by our meta-analysis, several factors and processes possibly influencing
448	phytoscreening effectiveness remain unexplored at a global scale as in the case of 1), among which: climatic
449	and meteorological conditions influencing-affecting uptake and loss from the tree; 2) porosity and volumetric
450	water content of the unsaturated zone influencing uptake and volatilization loss at the ground surface; 3)
451	organic content in saturated and unsaturated layers influencing sorption of CEs to the solid matrix; 4)
452	phytodegradation processes that may hinder correlation with $CE_{\underline{S}}$ concentration in gw; <u>5</u> ) radial distance to
453	boreholes likely affecting correlation between tree-core and gw concentration. It is therefore necessary to
454	conduct additional research in these areas to improve the applicability of Further research is needed in order
455	to fill this lack of knowledgethe technique.
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#### 620 FIGURES CAPTION

- 621 FIGURE 1. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of compounds series. Blank symbols
- 622 refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 624 hydrogeological parameters. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 625 FIGURE 3. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each family and correspondent
- k xylem structure. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 627 FIGURE 4. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each interval considered for tree
- 628 diameters. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 629 FIGURE 5. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each interval considered for
- 630 sampling and analysis protocols. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05

#### 631 TABLES CAPTION

- 632 TABLE 1. References used for statistical analysis of the database: geographical location, number of tree-core samples, and
- 633 correspondent detected compounds (Above Detection Limit: A.D.L.). Total number of data and Non-Detection% in the last
- 634 columns.
- 635 TABLE 2. Factors potentially affecting the effectiveness of phytoscreening of CEs in gw and relative descriptive
- 636 statistics. Selected intervals, relative number of observations, and relative sites per interval.
- 637 TABLE 3. Physico-chemical properties of the chlorinated ethenes: molecular weight (Mw), water solubility (Sw),
- 638 Henry's constant (H<sub>c</sub>), octanol-water partition coefficient (log K<sub>ow</sub>). Derived from Mackay et al. (2006)

**Field Code Changed** 

**Graphical Abstract** 



## HIGHLIGHTS

- Chlorinated ethenes groundwater contamination can be low cost screened by trees; •
- Phytoscreening is a quantitative method for shallow, thin, and permeable aquifers; Coniferous and diffuse-porous trees show high detectability potential; •
- •
- Sampling longer tree-cores at lower heights is preferable.

## 1 A quantitative review and meta-analysis on phytoscreening applied to

## 2 aquifers contaminated by chlorinated ethenes

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## 7 ABSTRACT

8 Applications and acceptance of phytoscreening, i.e., the use of trees as screening tools for underground 9 contamination, are still limited in many countries due to the lack of awareness of application policies, the 10 intrinsic qualitative nature of the technique, and the paucity of critical analyses on available data. To date, the 11 conditions influencing the effectiveness of the technique have been descriptively discussed, yet rarely 12 quantified. This review will contribute to filling this knowledge gap, shedding light on the most suitable approaches to apply phytoscreening. The focus was placed specifically on chlorinated ethene compounds since 13 they are among the main organic contaminants in groundwater and have been the most studied in the field of 14 phytoscreening. Chlorinated ethenes' behavior and biodegradation potential largely depend on their 15 16 physicochemical properties as well as the hydrogeological features of the system in which they migrate. Besides, their fate and transport in surface ecosystems are still poorly understood. Here, phytoscreening data 17 from sites contaminated by chlorinated ethenes were extracted from relevant literature to form a global-scale 18 database. Data were statistically analyzed to identify the major drivers of variability in tree-cores 19 20 concentration. Correlation between tree-core and groundwater concentration was quantified through 21 Spearman's rank coefficients, whilst detectability potential was determined based on tree-cores showing non-22 detection of contaminants. The influence on such parameters of factors like contaminant properties, 23 hydrogeology, tree features, and sampling/analytical protocols was assessed. Results suggest that factors 24 controlling plant uptake and contaminant phytovolatilization regulate correlation and detectability, respectively. Conditions increasing the correlation (e.g., sites with shallow and permeable aquifers) are 25 26 recommended for phytoscreening applications aimed at mapping and monitoring contaminant plumes, whereas

conditions increasing detectability (e.g., sampling tree-cores near ground level) are recommended to
preliminary screen underground contamination in poorly investigated areas.

29 **KEYWORDS:** groundwater, CEs, trees, quantitative phytoscreening

### 30 1. INTRODUCTION

Tree roots can carry contaminants dissolved in water through the trunk up to the leaves. This transport is based on direct contact of the roots with water inside the porous medium that surrounds the rhizosphere. Water can occur in different energy states: free moving gravity water in the saturated zone of the aquifer, referred to as groundwater (gw), or retention water (rw) subjected to suction and attached to soil particles as capillary or pellicular water in the unsaturated zone.

36 The use of plants in environmental hydrogeology has gained increasing attention during the last decades in 37 academic research and consultant activity due to stimulating application perspectives. In parallel with 38 contaminants removal by direct uptake and degradation (phytoremediation), Vroblesky et al. (1999) 39 demonstrated for the first time that headspace analysis of tree-cores allows to delineate shallow gw 40 contamination by chlorinated ethenes (CEs) such as trichloroethene (TCE) and cis 1,2-dichlorethene (cDCE). Later, Sorek et al. (2008) termed the technique "phytoscreening" and defined it as a simple, fast, non-invasive, 41 and inexpensive screening for detecting subsurface contamination by volatile organic compounds (VOCs). 42 Since Vroblesky et al. (1999), several comparisons between subsurface (soil, soil gas, and gw) and plant 43 44 contamination were conducted, mostly by using tree-cores but also by using leaf and branch samples (e.g., 45 Holm & Rotard, 2011; Wilcox & Johnson, 2016; Gopalakrishnan et al., 2007).

46 Besides screening and monitoring contaminant concentration for plume tracking or natural attenuation evaluation (e.g., Larsen et al., 2008), phytoscreening was used to assess soil vapor intrusions (e.g., Wilson et 47 48 al., 2017; Algreen et al. 2015) and age-date contamination events through dendroecology (Balouet et al., 2007). 49 Phytoscreening applicability was demonstrated for VOCs (e.g., BTEX; Wilson et al., 2013), perchlorate (e.g., 50 Limmer et al., 2014), per- and polyfluoroalkyl substances (PFAS; Gobelius et al., 2017), or inorganic compounds like heavy metals (e.g., Algreen et al., 2014). However, CEs were the most frequent target in 51 52 reported applications of phytoscreening. This review examines phytoscreening applications for CEs in 53 groundwater, for which exist sufficient literature information to conduct a quantitative meta-analysis. These compounds are indeed particularly responsive to uptake by plants, being relatively small and moderately 54

55 hydrophobic (Burken & Schnoor, 1998). In addition to their persistence, ubiquity, and toxicity (Pankow & Cherry, 1996), characterization and monitoring of these plumes require advanced technologies and high 56 57 funding, which could be mitigated by integrating a time-, and cost-effective technique like phytoscreening. The chance to determine the occurrence of subsurface volatile contaminants through trees is also important for 58 evaluating the risks to human health such as potential ingestion and respiration from vapor intrusion into 59 buildings, which are exposure pathways potentially associated with plant uptake. CEs detection in trees is 60 61 affected by several contaminant-specific loss mechanisms (e.g., volatilization, phytodegradation) which may 62 result in lower concentrations in plant tissues as compared to gw concentrations.

63 Phytoscreening of contaminated gw was considered a valuable ecohydrogeological application (Cantonati et 64 al., 2020). However, it is necessary to identify the main factors that drive the correlation between gw and tree-65 core contaminant concentration as well as contaminants' detectability potential in trees in order to make this 66 screening technique widely applicable and accepted,. The correlation and detection capability can be 67 maximized by providing directions on the standardization of sampling procedures. Several studies concluded 68 that the technique is only qualitatively reliable due to the poor correlation observed between gw and tree-core 69 concentration (Holm and Rotard, 2011; Larsen et al., 2008; Ottosen et al., 2018). This was attributed to multiple 70 factors that come into play when dealing with living organisms (trees) to signal the state of contamination of 71 an environmental matrix such as gw. Plants indeed interact with gw through complex partitioning mechanisms 72 mediated by various chemical, biological, hydrological, and climatic factors. Synthesis efforts were directed 73 to investigate the factors that limit the application opportunities of this technique. As a general example, Trapp 74 (2007) proposed a theoretical model for the prediction of chemical uptake in trees based upon more than 30 parameters, either of hydrogeological or ecological nature, thus addressing the complexity of quantitative 75 76 phytoscreening.

77 This paper provides a systematic review of former literature on the main factors that likely affect:

a) the correlation between CEs tree-core and gw concentration. Factors that determine higher correlations
 are favorable to apply phytoscreening to monitor, quantitatively, CEs contamination severity and
 degradation or natural attenuation processes.

3

b) the CEs detectability potential in trees. Factors that determine a lower number of contaminants nondetections in trees are favorable to preliminary screen for suspected underground contaminations by
CEs in poorly investigated areas.

Several factors were selected and grouped as follows: 1) physicochemical properties of the contaminants (molecular weight, water solubility, volatility, partition coefficients); 2) hydrogeology (depth to water table, aquifer thickness, hydraulic conductivity); 3) tree identity and anatomy (genus and family, xylem structure, tree trunk diameter); 4) sampling and extraction methodology (height above ground of tree-core sampling, tree-core length, extraction method of the contaminants).

89 2. MATERIALS AND METHODS

90 2.1 Data source

91 A systematic search for relevant studies of phytoscreening on CEs was conducted in Scopus in January 2021.

- 92 The search string was:
- 93 *TITLE-ABS-KEY* (phytoscreening OR (tree AND groundwater AND (trichloroethene OR perchloroethene
  94 OR dichloroethene OR "chlorinated ethenes"))).

95 The first database search yielded 64 references. To form a global-scale database of phytoscreening data on 96 CEs, only references containing datasets of contaminated sites that met the following criteria were selected: 1) 97 sampling by tree trunk coring, 2) tree-core analysis of at least one compound among PCE, TCE, and cDCE, 98 and 3) spatial proximity between a given tree and a borehole where gw concentration analysis showed 99 concentrations above detection limits (the distances between the locations of trees and boreholes varied from 100  $\sim 1$  to  $\sim 10$  m, or greater in few cases). When needed, we created contaminant concentration contour maps to supplement the reported gw concentration data. A total of 7 articles were identified reporting site datasets 101 suitable for this study. The reference lists of these 7 articles were manually searched for further studies 102 containing relevant datasets, providing 1 positive result (a technical report). Some of the final 8 selected 103 documents contributed with more than one investigated site, providing information on a total of 11 sites. A 104 105 total of 267 tree-core samples (and respective gw samples) were compiled in the global database, some of them reporting more than one compound concentration (Table 1). The number of tree-core concentration data was 106 419 (see Supplementary material for the database), 118 being below the analytical detection limit (ND data 107

108 hereafter). ND data represent a small fraction for most sites (below 15%; e.g., Struckhoff et al., 2005), whereas 109 in a few sites they are almost the majority (e.g., Larsen et al., 2008; Cox, 2002). The observations above 110 detection limits are distributed as follows: PCE - 43 observations, TCE - 194 observations, cDCE - 124 observations, and the sum of CEs - 58 observations related to one study that did not indicate single compound 111 concentrations (Wittlingerova et al., 2013). No data were compiled for VC because only Ottosen et al. (2018) 112 were able to detect traces of VC in trees under specific environmental conditions. No spatial or temporal 113 114 average concentrations were included in the database except for one site (Nogales site, Arizona; Duncan et al., 115 2017) in which only PCE average gw concentration (2  $\mu$ g/L) was provided. This site was included in the database for its significance in terms of uniqueness: cores were sampled and extracted with methanol from 116 117 trees of 4 different families inhabiting an arid environment with a deep aquifer table (9-10 m b.g.l.), showing 118 PCE concentration up to 500  $\mu$ g/kg.

Data on contaminant concentration in tree-core samples and gw samples were reported in two different types 119 120 of units, i.e., mass/mass (typically  $\mu g/kg$ ) and mass/volume (typically  $\mu g/L$ ), respectively. No conversion was performed from the mass/mass unit to the mass/volume unit for tree-cores. This was considered acceptable 121 122 since the focus was on the correlation between the concentration in different matrices along with the detectability potential. The unfeasibility of the conversion is mostly due to the lack of information on sampled 123 tree-core dry weights and volumes. Besides, wood-water partition coefficients of the contaminants would also 124 125 be needed for a reliable conversion. Very few studies estimated the latter for a small number of tree species 126 (e.g., poplars in Baduru et al., 2008). The lack of conversion hindered the possibility of performing multivariate 127 statistical analyses. Due to this limitation, a meta-analysis of the influence of each factor was performed to 128 analyze the database. It is worth noting that the results of this analysis will be subject to an intrinsic uncertainty 129 associated with processing each factor as independent of one another.

Information on the factors influencing the correlation between tree-cores and gw concentration as well as the potential for detectability in trees (i.e., the hydrogeological conditions of the underlying contaminated aquifers, tree identity and anatomy, and sampling and extraction protocols; Table 2) were retrieved from the 8 selected articles and associated to each of the 419 tree-core concentration data.

As for hydrogeological parameters, involving either the permeable porous medium (aquifer) or gw flowing
inside it, we retrieved: depth to water table (DWT in m below ground level), intended as the distance between

136 ground level and the surface at water pressure equal to atmospheric pressure (information retrieved for all the 419 tree concentration data); thickness of the saturated portion of the aquifer (b in m) intended as the distance 137 138 between the water table and the low permeability bottom of the aquifer (retrieved for 235 out of 419 tree 139 concentration); bulk saturated hydraulic conductivity of the aquifer (K in m/s; 373 out of 419 tree concentration). When K values of the aquifers were not specified (25% of the database), ranges of 140 conductivities were inferred in agreement with the local description of the lithology (Freeze & Cherry, 1979). 141 142 With regards to tree identity and anatomy, we retained information on the genus (398 out of 419 data) and on 143 the tree diameter at breast height (DBH in cm; 142 out of 419 data). Based on the genus, we retrieved the 144 correspondent xylem structure, intended as the distribution of pores and vessels among growth rings. The 145 xylem, or sapwood, is the active portion of the trunk where water transport takes place. We considered three xylem types: coniferous, diffuse-porous, and ring-porous (Panshin & de Zeeuw, 1970). Coniferous xylems are 146 147 characterized by small cells used for water transport and structural support. Diffuse-porous xylems additionally 148 contain large vessels that are randomly distributed throughout the wood, while ring-porous xylems have larger 149 diameter vessels concentrated in the earlywood. Conifers and diffuse-porous trees tend to have deep functional 150 xylems as well as low average conductivity due to small and short conduits. In contrast, most of the conductance in ring-porous species is isolated to the outermost annual growth ring that contains functional 151 152 vessels (Bush et al., 2010; Cermak et al., 1992).

When available, sampling and extraction protocols used to prepare tree-cores for analysis were retrieved: length of tree-core samples (L in cm; 369 out of 419 data), sampling height on the trunk (H in cm above ground level; 398 out of 419 data), and extraction method (419 out of 419 data) including extraction from dry vials (Cox, 2002; Larsen et al., 2008; Struckhoff et al., 2005; Vroblesky et al., 1999, 2004), vials containing organicfree water (Wittlingerova et al., 2013), or methanol solutions (Duncan et al., 2017), Solid Phase Micro-Extraction (SPME; Holm & Rotard, 2011).

159 2.2 Statistical meta-analysis

160 The correlation between gw and tree-core concentration was quantified by Spearman's rank coefficient ρ 161 (Journel and Deutsch, 1997), a widely used approach for assessing the relationship between parameters when 162 it is expected to be non-linear, as in highly skewed datasets. Indeed, both tree-core and gw concentrations 163 represent highly skewed data sets, as regularly found in contaminated sites (Juang et al., 2001). As an example, 164 tree-core concentrations had a distribution with positive skewness of 3.17, 8.21, and 6.85 for PCE, TCE, and cDCE, respectively. A useful property of  $\rho$  is that its value is invariant to any monotonic transformation applied 165 166 to the data (e.g., logarithmic transformation). Outliers were not removed from the database due to the lack of knowledge about the uncertainty associated with the measurements. It is worth noting that  $\rho$  is insensitive to 167 outliers, therefore representing a robust statistic tool for our database. The detectability potential of 168 contaminants in trees was quantified as the percentage of ND (ND%) data to the total number of observations. 169 Spearman's p and ND% were calculated separately for each compound (PCE, TCE, and cDCE) to assess the 170 171 influence of contaminant-specific properties such as molecular weight  $(M_w)$ , water solubility  $(S_w)$ , Henry's constant (H<sub>c</sub>), and octanol-water partition coefficient (log K<sub>ow</sub>). In the case of the dataset of Wittlingerova et 172 al. (2013) reporting only sums of CEs (PCE, TCE, cDCE, tDCE, 1,1-DCE, and VC), p and ND% were 173 174 calculated on the sums.

Factors in Table 2 were then split into intervals and Spearman's p were derived for the concentration data 175 176 within each interval. For continuous factors (e.g., aquifer properties or tree diameter), we determined discrete intervals based on medians and percentiles associated with each factor to have a similar number of observations 177 178 within each interval. Only concentration data above detection limits were considered. In the case of discrete 179 factors (e.g., tree species, extraction method), only values associated with a minimum of 10 observations were considered in the statistical analysis, except for one single case where only 8 observations were associated with 180 181 extraction with methanol (Duncan et al., 2017). The ND% was determined within each of the aforementioned 182 intervals to assess the influence of the different factors on detectability. To avoid biases associated with trees that could be growing above more dilute contamination areas, the ND% was calculated only when the 183 concentration in gw was above 11 µg/L. This threshold was determined as the 5<sup>th</sup> percentile of gw 184 185 concentration data. The final number of ND data was 104 out of 118 ND tree-core data. It is noteworthy that 186 in some cases, such as when processing hydrogeological parameters like b and K, results may be uncertain 187 since some parameters do not vary spatially across a specific site.

188 Results were interpreted in terms of high or low correlations (determining optimal factors conditions to
189 characterize the contamination) and high or low detectability potential (determining optimal factors conditions
190 to screen gw contamination).

## 191 3. RESULTS AND DISCUSSION

#### 192 3.1 Contaminant properties

193 The correlation between the concentration in tree-cores and gw was statistically meaningful (p-value $\leq 0.05$ ) 194 for all four series (PCE, TCE, cDCE, and the sum of CEs).

The p values indicate a low variability among CEs in terms of correlation between tree-core and gw 195 196 concentration (Figure 1), with slightly lower values for higher chlorinated compounds PCE and TCE (p of 0.37 197 and 0.34, respectively) compared to the lower chlorinated cDCE ( $\rho$  of 0.41). On the other hand, ND% widely 198 differed between higher chlorinated compounds PCE and TCE (12 and 23%, respectively) and cDCE (47%) 199 with the first two performing better in terms of detectability potential. It is noteworthy that the highest 200 correlation coefficient (0.63) and lowest ND% value (0%) were found for the sum of CEs (reported by 201 Wittlingerova et al., 2013). This could be explained by contaminant transformation processes taking place 202 either in the rhizosphere or in the xylem (Newman and Reynolds, 2004) that would negatively affect the 203 correlation of single compounds. However, this result is associated with only one site.

204 The physicochemical properties of CEs (Table 3) likely influence the behavior of each compound in trees. 205 Larsen et al. (2008) observed a better correlation for cDCE compared to TCE and PCE which was attributed 206 to the higher volatility (higher H<sub>c</sub>) of the latter possibly causing a higher loss through the bark (Vroblesky et 207 al., 1999). According to Limmer & Burken (2015), contaminant concentrations decreased with increasing 208 volatility, due either to volatilization from the roots, bark, or subsurface. The results of our study are consistent 209 with these findings, with PCE and TCE showing a slightly lower p compared to cDCE. In addition, the 210 correlation potential of PCE and TCE may be hindered by their lower tendency to plant uptake compared to cDCE, driven by their higher M<sub>w</sub> and K<sub>ow</sub>, and lower S<sub>w</sub>. Uptake from tree roots was indeed reported to be 211 favored for compounds with low  $M_w$  due to their higher tendency to diffuse (Baduru et al., 2008), whereas 212 213 higher S<sub>w</sub> would possibly favor contaminant dissolution in water and consequent tree uptake. Besides, high sorption compounds (log  $K_{ow}>3$ ) were reported to tend to be absorbed primarily by root surfaces, resulting in 214 215 less translocation to the xylem (Schnoor et al., 1995). Similarly, a study by Dettenmaier et al. (2009) indicated 216 that high-hydrophilic compounds are most likely absorbed by roots and transferred to the xylem. We may also 217 assume that once in the xylem, higher  $\log K_{ow}$  compounds likely tend to be absorbed by xylem tissues, resulting 218 in a prolonged accumulation in the tree and thus a higher detectability potential. This could explain the higher 219 detectability potential of PCE and TCE (lower ND%) compared to cDCE. Moreover, their higher H<sub>c</sub> can aid

the analytical detection when using headspace methods. Concurrently, the lower log  $K_{ow}$  and  $M_w$  of cDCE may hinder accumulation and favor contaminant loss through the bark despite its low H<sub>c</sub> (Baduru et al., 2008), resulting in an overall lower detectability potential. The extremely rare detection of VC confirms the role of H<sub>c</sub>, M<sub>w</sub>, and log K<sub>ow</sub> in contaminant detectability in trees.

3.2 Hydrogeology and aquifer parameters

Three intervals were considered for DWT: DWT<1, 1 ≤ DWT<3, and DWT≥3 m b.g.l. Concentration data show 225 226 a statistically significant correlation value (p-value ≤0.05; Figure 2) in the three cases. A slightly decreasing 227 correlation with increasing DWT was observed (p=0.63, 0.54, and 0.52 for DWT<1, 1≤DWT<3, and DWT≥3 228 m b.g.l., respectively). On the other hand, the shallow interval showed the higher ND% (35%) whereas the 229 inferior intervals had significantly lower ND% (16% and 17% for the medium and deep interval, respectively). 230 Duncan & Brusseau (2018) assessed for the first time how DWT could affect the correlation between VOCs 231 concentration in tree tissues and gw (based on 100 measurements). They observed a higher correlation in samples from sites with a DWT<4 m, concluding that a low thickness of the unsaturated zone significantly 232 233 affects phytoscreening efficiency. Despite the overall consistency of our results with the cited literature (decreasing p with increasing DWT), p shows small differences among DWT intervals, suggesting a low 234 235 influence of this factor on the degree of correlation. The difference with Duncan & Brusseau (2018) may lie 236 in the use of different correlation coefficients and interval divisions. In our analysis p was chosen due to the 237 non-linear distribution of the concentration dataset while Duncan & Brusseau (2018) assessed the correlation through Pearson's coefficient  $(r^2)$  thus assuming linearity of the dataset. Besides, at sites with a more 238 substantial vadose zone, mineralization of CEs can occur before translocation of the contaminant in the tree 239 240 due to more hypoxic conditions (Bradley & Chapelle, 2011), leading to lower correlation potential for deeper 241 aquifers. Similar findings have been reported by Wilson et al. (2013) for BTEX translocation in trees. At the same time, when DWT is lower, volatilization loss of CEs is promoted at the interface between the saturated 242 243 and the vadose zone, being the thickness and water content of the latter more subject to atmospheric variations 244 (Pankow & Cherry, 1996). As tree roots are usually located at this interface, the enhanced volatilization of 245 CEs could decrease their detectability potential (higher ND% for the shallow interval).

Aquifer thicknesses were clustered into two intervals:  $b \le 3.5$  and b > 3.5 m. Data associated with lower b show a very high positive correlation ( $\rho=0.71$ ), whereas those associated with higher b have a significantly weaker 248 correlation ( $\rho=0.30$ ). This may be because CEs, in the majority of contaminant events, enter the subsoil as 249 dense non-aqueous phase liquids (DNAPLs), which tend to sink towards deeper sections of the aquifer, thus 250 influencing the shape of dissolved contaminant plumes (e.g., Parker et al., 2003). In particular, the sinking of 251 CEs could result in an increased distance of the dissolved contaminant plume from the root zone in thicker aquifers. Notably, aquifer thickness appears to have the highest influence on correlation compared to other 252 factors. On the other hand, a lower ND% is associated with the thicker aquifer interval compared to the thinner 253 254 interval (4% and 19% for b>3.5 m and b $\leq$ 3.5 m, respectively). This result has poor validations: we can 255 speculate that thin aquifers have a lower geometrical probability of being intercepted by tree roots than thick 256 ones.

Two intervals of K were considered:  $K < 1x10^{-5}$  and  $K \ge 1x10^{-5}$  m/s. Concentrations referred to higher K values show a slightly higher correlation ( $\rho=0.73$ ) compared to those referred to lower conductivities ( $\rho=0.66$ ). These results suggest that K poorly affects the correlation between tree-core and gw concentration. On the other hand, the lower K interval includes 38% of ND whereas the higher K interval includes 18%. More permeable aquifers are therefore more suited in terms of detectability potential. A relatively higher permeability can indeed enhance the mobility of contaminants in the subsoil, likely favoring plant uptake, similarly to what happens when extracting gw from wells or soil gas from soil gas probes.

264 3.3 Tree identity and anatomy

The 22 tree genera of our database were clustered by families and xylem structures. Results show a significant 265 266 positive correlation with most families (Figure 3), with  $\rho$  being highest for coniferous, i.e. Pinaceae ( $\rho$ =0.86) and Cupressaceae (p=0.66). These conifers also have moderately low ND% (24% and 30%, for Pinaceae and 267 Cupressaceae, respectively). Consistently with our observations, Trapp et al. (2007) stated that conifers are 268 best suited for phytoscreening because they have a broad xylem zone, and transpire throughout the whole year, 269 resulting in a continuous uptake of gw. Ring-porous Fagaceae (primarily Quercus) presented a slightly positive 270 271 correlation ( $\rho=0.39$ ) and a high ND% (57%). The ring-porous structure likely promotes volatilization loss 272 through the bark, possibly affecting the observed low correlation and detectability potential. Indeed, over 90% 273 of water is transported in the outermost growth ring in ring-porous trees whereas in diffuse-porous and 274 coniferous trees water flow is more equally distributed among rings (Ellmore and Ewers, 1986). Diffuse275 porous Nyssaceae and Betulaceae show a slightly lower  $\rho$  compared to conifers (0.74 and 0.65, respectively), 276 and a low ND% (29 and 25%, respectively) although results for Nyssaceae must be taken with caution because 277 they refer to one single study site (Savannah River Site, USA; Vroblesky et al., 1999) where the aquifer was 278 shallow (DWT<1 m b.g.l.). On the other hand, diffuse-porous Salicaceae and Altingiaceae do not show a 279 significant correlation (p-value>0.05) and highly fluctuating ND% (very low for Salicaceae - 3% and very high for Altingiaceae – 64%). The high variability among diffuse-porous families in terms of  $\rho$  and ND% could 280 281 be associated with different arrangements and sizes of the vessels regulating the conductivity of the xylem, 282 which in turn can also vary with age. For example, Salicaceae (Salix and Populus), widely used in 283 phytoscreening and phytoremediation due to their fast growth, high uptake rates, and widespread occurrence 284 in temperate climates, showed the highest detectability potential although no correlation between tree-core and 285 gw concentration. Besides, Negri et al. (2003) stated that Salicaceae are genetically predisposed to develop 286 roots extending to the water table at depths greater than 12 m b.g.l., thereby widening their detectability 287 potential to deep aquifers. Altingiaceae (Liquidambar; present only in the study of Vroblesky et al., 1999) also 288 showed no correlation whilst a low detectability potential. This family was studied by Strycharz and Newman 289 (2009) in a greenhouse experiment where also Platanaceae and Salicaceae were involved. Results showed that 290 among the 3 families, Altingiaceae was the less recommended for phytoremediation activities. Other diffuse-291 porous, like Betulaceae (Alnus and Betula), showed instead a correlation and detectability potential comparable 292 to conifers. Lewis et al. (2015) calculated that Betula pendula can accumulate similar amounts of TCE as 293 *Populus* trees due to its lack of heartwood (nonfunctioning xylem) and homogeneous xylem flow (Westhoff 294 et al., 2008), making this species a suitable candidate for phytoremediation and phytoscreening activities. 295 Eventually, the low number of Platanaceae (diffuse-porous), Ulmaceae (ring-porous), and Rosaceae (diffuse-296 porous) in the database (Table 2) hindered an analysis on these families. Even so, Limmer & Burken (2015) 297 showed that the genus *Platanus* (Platanaceae) had a high detectability potential, especially for PCE gw 298 contamination, if compared to non-Platanus trees. Their result was although associated with Platanus trees 299 growing in areas with shallow gw. Conversely, our data on Platanaceae showed that 3 out of 4 times this family 300 did not detect contaminants even with shallow gw (DWT<1 m b.g.l.; Savannah River Site, USA; Vroblesky et al., 1999) albeit in that particular site aquifer K was low  $(5.3 \times 10^{-6} \text{ m/s})$ . Yung et al. (2017) pointed out that 301 302 besides Populus and Salix (Salicaceae, diffuse-porous) and Betula (Betulaceae, diffuse-porous), Ouercus and 303 *Ulmus* (Fagaceae and Ulmaceae, both ring-porous) are also efficient biomonitors of PCE and TCE 304 contamination. Notwithstanding the contrast with our results on Fagaceae, the 2 Ulmaceae trees in our database 305 (Carswell Golf Course, USA; Vroblesky et al., 2004) showed TCE concentrations above detection limits, with 306 DWT at 1 to 5 m b.g.l., aquifer K of  $7x10^{-5}$  m/s, and b of 0.9 m. This may suggest that even among ring-porous 307 trees a great variability in concentration results is expected.

Tree size (measured as DBH) in our database ranges between 18 and 102 cm and values were split into 3 308 309 intervals: DBH<25, 25 DBH<40, and DBH 240 cm. Since data were filled only for 34% of the database (Table 310 2), results must be considered with caution. The lower interval shows a high positive correlation ( $\rho=0.79$ ). The 311 correlation decreases significantly in the medium interval ( $\rho=0.47$ ). The higher interval shows no significant 312 correlation (*p*-value>0.05). ND% is instead comparable among DBH intervals (10%, 8%, and 8%, respectively for DBH<25, 25 \le DBH<40 cm, and DBH \le 40 cm). Several studies suggest that tree size has little effect on 313 tree-core concentration (Limmer & Burken, 2015; Wahyudi et al., 2012) while other studies demonstrated that 314 315 diffusional loss in small trees (DBH of 2 cm) occurs at a rate 10 times higher than in trees with DBH 15 cm due to their greater surface area to volume ratio that more quickly depletes the compound reservoir in the trunk 316 317 (Schumacher et al., 2004; Struckhoff, 2003). This could explain the slightly higher ND% of smaller trees. In contrast, our results show that smaller trees have greater efficiency in terms of quantitative analysis of gw 318 319 contamination (high  $\rho$ ). In smaller trees, we could indeed expect less variability in concentration around and 320 across the trunk due to a less compartmentalized flow in the xylem. Also, in smaller trees it is highly probable 321 to sample a consistent thickness of the total xylem, resulting in a concentration that averages the radial 322 variability. More variability is instead observed in larger trees where sampling direction has a strong influence 323 on the concentration, thus influencing the correlation.

324 3.4 Sampling and analysis protocols

In our database tree-core L ranges from 3.8 to 12.5 cm and was clustered into two intervals: L $\leq$ 6 and L>6 cm. Shorter cores do not show a significant correlation between tree-core and gw concentration (*p*-value>0.05) whereas longer cores have a high positive correlation ( $\rho$ =0.71; Figure 6). This result agrees with Ma & Burken (2003) and the USGS user guide published by Vroblesky (2008), which suggest a better correlation when the core samples are longer than ~7-8 cm. Shorter cores may be also acceptable for ring-porous trees, in which water transport takes place mostly in the outermost ring (Ellmore and Ewers, 1986). The detectability potential is lower for the higher L interval (ND of 11% and 30% for L $\leq$ 6 and L>6 cm, respectively), likely indicating that drilling longer tree-cores could promote diffusional loss out of the sample since sampling requires relatively longer periods (tree-cores are usually cut in small pieces before being put in the vials).

Sampling H from the base of the trunk ranges between 50 and 900 cm a.g.l. and was clustered into three 334 intervals: H<99, 99<H≤120, and H>120 cm a.g.l. The medium interval shows a very high correlation between 335 tree-core and gw concentration ( $\rho$ =0.84; Figure 6), which decreases for higher H ( $\rho$ =0.42). The lower interval 336 337 shows no significant correlation (p-value>0.05) although the data pertain to only one survey by Holm & Rotard 338 (2011). The reason for the high  $\rho$  value at medium H is still unknown. We can speculate that this is related to 339 the attainment of an equilibrium of the contaminant inside the wood-air-water partitioning system. On the other 340 hand, a low ND% was registered for the lower H interval (7%) whereas the higher intervals are associated with higher ND% (24% and 33% for the medium and higher H interval, respectively). This result is consistent with 341 342 the experiments of Ma & Burken (2003) where a higher TCE loss was observed higher up the trunk. Thus, a 343 lower rate of diffusional loss from the bark may be expected for the lower H intervals. On the other hand, Ottosen et al. (2018) sampled tree-cores just above the ground surface without distinguishing a clear advantage 344 345 from this sampling strategy.

Our database includes 4 extraction methods used for analysis (dry sample, organic-free water, methanol 346 347 solution, and SPME; Figure 6). The dry and the water extracted samples showed moderately high correlation 348 ( $\rho=0.55$  and  $\rho=0.64$  for dry and water extraction, respectively). However, data from water extracted samples 349 pertain to only one survey where concentration data relate to the sum of CEs (Wittlingerova et al., 2013). The 350 high p could be associated either with the extraction method or with the fact that the sum of CEs was considered 351 (see section 3.1). Methanol extracted samples, related to the study cases of Duncan et al. (2017), were collected with unfavorable conditions in arid-hot environments (Nogales site, Park-Euclid, and Motorola 52<sup>nd</sup> superfund 352 353 site in Arizona, USA). The small number of tree-cores (8) sampled with this method may explain the observed 354 non-significant correlation value (p-value>0.05). Tree-core concentrations associated with arid environments are possibly more a function of vadose zone vapor phase concentration than gw concentration since roots 355 would unlikely reach DWT of 26 m b.g.l. as in the case of the Park-Euclid site. The fact that 5 out of 8 tree-356 core concentrations of were higher than 300 µg/kg despite being associated with low gw concentrations of 357 PCE (2 µg/L; i.e., lower than the gw concentration threshold defined for ND% calculation; See Section 2.2) 358

further supports the hypothesis that trees were absorbing contaminants from a matrix other than gw. This may have hindered correlation with gw in the case of methanol extraction. The analysis following SPME, associated with only one site (Potsdam-Krampnitz military base, Germany; Holm & Rotard, 2011), also showed no significant correlation (*p*-value>0.05). Even so, the ND% was very low for SPME (7%) as well as water extracted samples (0%). However, since these methods were used in single study cases, these results may be related to other site-specific conditions or sampling protocols. In terms of ND%, the dry and methanol extraction produced comparable results (34%, and 33% for dry and methanol extraction, respectively).

## 366 4. SUMMARY AND CONCLUSIONS

The effectiveness of phytoscreening has been assessed via a meta-analysis of literature data to determine the potential of trees to a) monitor groundwater plumes of CEs, expressed as the degree of correlation between tree-core and gw concentration, and b) detect the occurrence of gw contamination events by CEs in poorly investigated areas, expressed as the rate of tree-cores that showed concentrations above the detection limit in significantly contaminated areas. Several factors likely influencing correlation and detectability were considered. These factors included physicochemical properties of CEs, hydrogeological conditions of the underlying contaminated aquifers, trees identity and anatomy, and sampling and extraction protocols.

The correlation (quantitative monitoring potential) is higher when the hydrogeological dynamics favor direct uptake of contaminated water and the concentration is homogeneously distributed in the tree and the tree-core sample. Direct tree uptake is favored for the lighter and more soluble cDCE and in the case of shallow (DWT<3 m b.g.l.), thin (b<3.5 m), and permeable aquifers (K $\ge$ 1x10<sup>-5</sup> m/s). The homogeneity of concentration in the xylem is higher for Pinaceae and Cupressaceae (conifers), due to their non-porous xylem, and for smaller diameter trees (DBH<25 cm) whereas homogeneity of concentration in the tree-core is enhanced when sampling longer tree-cores (L>6 cm) and at a height on the trunk between 99<H≤120 cm a.g.l.

The detectability (qualitative screening potential) is higher when factor conditions favor accumulation in the xylem and hinder volatilization loss through the bark. In these terms, PCE and TCE are more suited compared to cDCE due to their higher sorption and weight. Low volatilization loss was also inferred in the case of largediameter trees (DBH $\geq$ 40 cm), at a low height on the trunk (H<99 cm a.g.l.), and for shorter cores due to reduced time of sampling (L $\leq$ 6 cm). In the case of Salicaceae, high uptake rates may compensate for volatilization losses, thus increasing detectability. Finally, the process of contaminant extraction appears to be maximizedwhen using organic-free water extraction and SPME.

388 Despite the clarifications provided by our meta-analysis, several factors influencing phytoscreening effectiveness remain unexplored at a global scale as in the case of 1) climatic and meteorological conditions 389 affecting uptake and loss from the tree; 2) porosity and volumetric water content of the unsaturated zone 390 influencing uptake and volatilization loss at the ground surface; 3) organic content in saturated and unsaturated 391 392 layers influencing sorption of CEs to the solid matrix; 4) phytodegradation processes that may hinder 393 correlation with CEs concentration in gw; 5) radial distance to boreholes likely affecting correlation between 394 tree-core and gw concentration. It is therefore necessary to conduct additional research in these areas to 395 improve the applicability of the technique.

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## 541 FIGURES CAPTION

- 542 FIGURE 1. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of compounds series. Blank symbols
- 543 refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 544 FIGURE 2. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each interval considered for the
- 545 hydrogeological parameters. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 546 FIGURE 3. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each family and correspondent
- 547 xylem structure. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 548 FIGURE 4. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each interval considered for tree
- diameters. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05
- 550 FIGURE 5. Spearman's ρ (x-axis) and ND% (y-axis, values in inverse order) of each interval considered for
- sampling and analysis protocols. Blank symbols refer to as *p*-values>0.05. Solid symbols refer to as *p*-values≤0.05

## 552 TABLES CAPTION

- 553 TABLE 1. References used for statistical analysis of the database: geographical location, number of tree-core samples, and
- 554 correspondent detected compounds (Above Detection Limit: A.D.L.). Total number of data and Non-Detection% in the last
- columns.

- **TABLE 2.** Factors potentially affecting the effectiveness of phytoscreening of CEs in gw and relative descriptive statistics.
- **557** Selected intervals, relative number of observations, and relative sites per interval.
- 558 TABLE 3. Physico-chemical properties of the chlorinated ethenes: molecular weight (Mw), water solubility (Sw),
- Henry's constant (H<sub>c</sub>), octanol-water partition coefficient (log K<sub>ow</sub>). Derived from Mackay et al. (2006)

560

**TABLE 1.** References used for statistical analysis of the database: geographical location, number of tree-core samples, and correspondent detected compounds (Above Detection Limit: A.D.L.). Total number of data and Non-Detection% in the last columns.

Reference	Location	$\mathbf{n}^{\mathrm{o}}$ of tree-core	Contaminants	$\mathbf{n}^{\circ}$ of	ND	
		samples	A.D.L.	data	%	
Vroblesky et al. (1999)	Savannah River site (South Carolina,	86	TCE, cDCE	179	40	
	USA)					
Cox (2002)	Site SS-34N, McChord AFB	14	TCE	14	10	
	(Washington, USA)				0	
Vroblesky et al. (2004)	Carswell Golf Course (Texas, USA)	24	TCE	24	4	
	Air Force Plant PJKS (Colorado, USA)	9	TCE	9	0	
	Naval Weapons Station Charleston (South	10	TCE	10	0	
	Carolina, USA)					
Struckhoff & Burken (2005)	Front Street, Riverfront Superfund site	20	PCE	20	15	
	(Missouri, USA)					
Larsen et al. (2008)	North Bohemia Carcass Disposal Plant	17	PCE, TCE,	51	47	
	(Czech Republic)		cDCE			
Holm & Rotard (2011)	Former military base Potsdam-Krampnitz	23	TCE, cDCE	46	7	
	(Germany)					
Wittlinglerova et al. (2013)	North Bohemia Carcass Disposal Plant	58	Sum of CEs	58	0	
	(Czech Republic)					
Duncan et al. (2017)	Nogales site (Arizona, USA)	4	PCE	4	0	
	Park-Euclid (Arizona, USA)	1	PCE, TCE	2	0	
	Motorola 52 <sup>nd</sup> Street Superfund site	1	TCE	2	50	
	(Arizona, USA)					
8	11	267		419	28	

COMPOUND	M <sub>w</sub> [g/mol]	S <sub>w</sub> at 25°C [mg/L]	H <sub>c</sub> [at 17.5°C] [adim.]	log K <sub>ow</sub> [adim.]	
РСЕ	165.8	206	0.492	3.40	
ТСЕ	131.3	1118	0.265	2.61	
cDCE	96.9	3500	0.111	1.86	
VC	62.4	2700	0.811	1.46	

 $\label{eq:TABLE 3. Physico-chemical properties of the chlorinated ethenes: molecular weight (Mw), water solubility (S_w),$ 

Henry's constant (H<sub>c</sub>), octanol-water partition coefficient (log K<sub>ow</sub>). Derived from Mackay et al. (2006)

## **TABLE 2.** Factors potentially affecting the effectiveness of phytoscreening of CEs in gw

Торіс	Factors	Measuring unit	n° of available data (out of 419)	Min.	1st Quartile	Median	Mean	3rd quartile	Max.	Selected intervals	n° of observation per interval	n° of sites per interval
Hydrogeology	Depth to water table (DWT)	m b.g.l. (below ground level)	419	0.35	0.75	1	2.39	2.5	27	DWT<1, 1≤DWT<3, and DWT≥3	204, 138, and 77	3, 5, 9
	Average aquifer thickness (b)	m	235	0.9	3	3	3.29	4	6.6	b≤3.5, b>3.5	157 and 77	5,4
	Aquifer hydraulic conductivity (K)	m/s	373	4x10 <sup>-6</sup>	1x10 <sup>-5</sup>	1x10 <sup>-5</sup>	9x10 <sup>-5</sup>	1x10 <sup>-4</sup>	1x10- 3	K<1x10 <sup>-5</sup> , K≥1x10 <sup>-5</sup>	189 and 184	2, 9
Tree identity and anatomy	Family	e.g., Salicaceae	398							Cupressaceae, Pinaceae, Salicaceae, Nyssaceae, Betulaceae, Altingiaceae, Fagaceae, Platanaceae, Ulmaceae, Rosaceae	112, 71, 62, 14, 68, 18, 45, 4, 3, 2	2, 6, 7, 1, 3, 1, 6, 1, 1, 1
	Xylem structure	e.g., coniferous	398							Coniferous/Diffuse- porous/Ring-porous	183,169, 46	6, 9, 6
	Tree diameter at breast height (DBH)	cm	142	11	24	25	31.96	36	102	DBH<25, 25≤DBH<40 cm, and DBH≥40 cm	51, 66, 24	3, 5, 2
Sampling and extraction methodology	Trunk drilling length (L)	cm	369	3.8	6	6.8	6.92	6.8	12.5	L≤6 and L>6 cm	123, 245	6, 5
	Sampling height (H)	cm a.g.l. (above ground level)	398	50	100	150	129.1	150	900	H<99, 99 <h≤120, and H&gt;120 cm a.g.l.</h≤120, 	46, 123, 228	1, 3, 7
	Extraction method	e.g., methanol extraction	419							Dry/Water extraction/Methanol extraction/SPME	307, 8, 58, 46	7, 1, 3, 1





▲ Hydraulic conductivity (m/s)







■ Extraction method

Supplementary Material

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## **Declaration of interests**

 $\boxtimes$  The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

□The authors declare the following financial interests/personal relationships which may be considered as potential competing interests:

## **CRediT** author statement

Carlotta Leoncini: Conceptualization, Methodology, Formal Analysis, Investigation, Writing-Original draft
Maria Filippini: Conceptualization, Methodology, Writing-Review & Editing, Supervision
Juri Nascimbene: Conceptualization, Methodology, Writing-Review & Editing, Supervision
Alessandro Gargini: Conceptualization, Methodology, Writing-Review & Editing, Supervision