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Thornton, S.F. orcid.org/0000-0002-0235-1600, Spence, M.J., Bottrell, S.H. et al. (1 more author) (2024) Natural attenuation of dissolved petroleum fuel constituents in a fractured Chalk aquifer: Contaminant mass balance with probabilistic analysis. Quarterly Journal of Engineering Geology and Hydrogeology. ISSN 1470-9236

https://doi.org/10.1144/qjegh2023-116

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Steven F. Thornton, Michael J. Spence, Simon H. Bottrell & Keith H. Spence

Natural attenuation of dissolved petroleum fuel constituents in a fractured Chalk aquifer: Contaminant mass balance with probabilistic analysis

Steven F. Thornton [1,*], Michael J. Spence [1,3], Simon H. Bottrell [2] and Keith H. Spence [2,4]

[1] Groundwater Protection and Restoration Group, Department of Civil and Structural Engineering, Mappin Street, University of Sheffield, Sheffield, S1 3JD, UK

[2] School of Earth and Environment, University of Leeds, Leeds, LS2 9JT, UK

[3] Present address : British Geological Survey, Environmental Science Centre, Keyworth, Nottingham, NG12 5GG, UK

[4]. Present address : Environment Agency, Alchemy, Bessemer Road, Welwyn Garden City, Hertfordshire, AL7 1HE, UK

[*] Author for correspondence; ORCID : https://orcid.org/0000-0002-0235-1600; email: s.f.thornton@sheffield.ac.uk

Abstract

A plume-scale mass balance is developed to assess the natural attenuation (NA) of dissolved organic contaminants in fractured, dual porosity aquifers. This methodology can be used to evaluate contaminant distribution within the aquifer, plume source term, contaminant biodegradation and plume status. The approach is illustrated for a site on the UK Upper Chalk aquifer impacted by petroleum fuel containing MTBE and TAME. Variability in site investigation data and uncertainty in the mass balance was assessed using probabilistic analysis. The analysis shows that BTEX compounds are biodegraded primarily by denitrification and sulphate reduction in the aquifer, with an equivalent plume-scale first-order biodegradation rate of 0.49 year^{-1}. Other biodegradation processes are less important. Sorption contributes to hydrocarbon attenuation in the aquifer but is less important for MTBE and TAME. Uncertainty in the plume source term and site hydrogeological parameters had the greatest effect on the mass balance. The probabilistic analysis enabled the most likely long-term composition of the plume source term to be deduced and provided a site-specific estimate of contaminant mass flux for the prediction of plume development. The mass balance methodology provides a novel approach to improve NA assessments for petroleum hydrocarbons and other organic contaminants in these aquifer settings.

Monitored natural attenuation (MNA) is an established risk-based management approach for groundwater contaminated with petroleum hydrocarbons (Wiedemeier *et al.* 1999; Rivet and Thornton, 2008). Technical guidance is available to assist the implementation of MNA (e.g. Wiedemeier *et al.* 1995; 1999; Environment Agency, 2000; ASTM, 2016), but this is difficult to apply at sites with complex source histories and typically provides an indirect estimate of contaminant mass loss using rate constants obtained from limited groundwater monitoring (Wilson *et al.* 2004; Hosseini, 2009; Thornton *et al.* 2013). Rate constants derived in this way may be in error or neglect the variability in contaminant concentration (Ravi *et al.* 1997; Environment Agency, 2002; Hosseini, 2009). However, mass balances can provide significant improvements in process understanding and assessment of contaminant fate in plumes (e.g. Hunkeler *et al.* 1999; King *et al.* 1999; Thornton *et al.* 2001). Furthermore, the performance assessment of natural attenuation (NA) commonly uses deterministic analysis to predict plume behaviour and risk to receptors (Ravi *et al.* 1997; Wiedemeier *et al.* 1999; Hosseini, 2009; Zhang, 2010). This seldom considers the variability in environmental data and in many cases site investigation for MNA assessments does not yield the extensive data set required to evaluate uncertainty in predictions (McNab and Dooher, 1998; Environment Agency, 2001; Liang *et al.* 2011; Beck and Mann, 2012). Sources of uncertainty include knowledge of representative parameter values, the underlying conceptual model and observation data (USEPA, 1999; Environment Agency, 2001*a,b*; Anderson *et al.* 2007; Winter and Tartakovsky, 2008; Zhang, 2010; Liang *et al.* 2011; Beck and Mann, 2012; Middlemis and Peeters, 2018; La Cecilia *et al.* 2020). Deterministic analysis may also result in overly conservative predictions with higher costs to achieve remediation objectives (Anderson *et al.* 2007; Liang *et al.* 2011; Rivera-Velasquez *et al.* 2013). Instead, performance assessment and evaluations of contaminant fate and transport should include probabilistic analysis to account for incomplete conceptual understanding of the system interactions and uncertainty in parameter values describing processes (Woodward, 1995; Environment Agency, 2001*a*; Bredehoeft, 2003; 2005; Bolster *et al.* 2009; Beck and Mann, 2012). This approach is often incorporated within environmental risk assessments (e.g. Anderson *et al.* 2007; Ho *et al.* 2007; Winter and Tartakovsky, 2008; Liang, 2009; Zhang, 2010; Zhang *et al.* 2010; Rodak and Silliman, 2012; Yang et al, 2018; Gonzalez *et al.* 2020; La Cecilia *et al.* 2020; Troldborg *et al.* 2022), but has seldom

been used to interpret the NA of plumes at field-scale (Thornton *et al.* 2001; Hosseini, 2009; Liang, 2009; O'Malley and Vesselinov, 2014; Xie *et al.* 2016).

This paper presents a mass balance methodology integrated with probabilistic analysis for the field-scale performance assessment of petroleum hydrocarbon NA in fractured, dual-porosity aquifers. The methodology is based on a conceptual model which describes the distribution and attenuation of potentially biodegradable contaminants in a dual porosity aquifer system, and enables the calculation of mass balances for contaminant plumes in groundwater using data obtained from a typical site investigation and groundwater monitoring study for MNA. It is illustrated for a petroleum fuel-release site on the UK Upper Chalk aquifer. The Chalk is the most important aquifer in the UK, accounting for 60% of groundwater used in England and Wales (Allen *et al.* 1997; MacDonald and Allen, 2001). It is a soft, white limestone (~99% $CaCO_3$) that forms a deep consolidated, fractured dual porosity aquifer, with minor marl and flint layers but very low mineral oxide (MnO_2, $Fe(OH)_3$) content (Bottrell *et al.* 2010). The aquifer matrix has a high primary porosity, typically 30-45% in the Upper Chalk (Jones and Robbins, 1999), but low effective permeability (Allen et al 1997). The matrix pore water has a long residence time compared with the fracture water and within the saturated zone can be considered essentially immobile storage relative to the timescale of plume development investigated here (Younger and Elliot, 1995; Bloomfield, 1996; Allen *et al.* 1997). The matrix is dissected by fractures, which contribute minor secondary porosity (typically only 1%), but which have very high permeability and may contribute 99% of the flow (Price, 1996; Allen *et al.* 1997; MacDonald and Allen, 2001). Groundwater flow and solute transport in the system occurs primarily in the fractures, which have high transmissivity, whereas the low transmissivity matrix provides storage for the bulk of the groundwater (Bloomfield, 1996; Allen *et al.* 1997; MacDonald and Allen, 2001). Solute exchange between the matrix and fractures occurs by molecular diffusion (Younger and Elliot, 1995).

The purpose of this paper is to evaluate the mass balance methodology in this contaminated aquifer setting, considering dual-porosity effects on contaminant distribution and attenuation, and the

contribution of parameter uncertainty in predictions of NA performance. The specific objectives were to:

a) Undertake a plume-scale mass balance to evaluate the contaminant distribution and NA of dissolved organic contaminants in groundwater at field-scale for the site under study;

b) Evaluate the effect of variability in parameter inputs on the mass balance for the site and identify key parameters responsible for uncertainty in the performance prediction for NA, using probabilistic methods.

Site description and investigation history

The site is a former retail filling station located on the Upper Chalk aquifer in southern England (Figure 1). Petroleum fuel was accidently released from an underground storage tank (UST) in February 1999. This has contaminated the saturated zone of the aquifer beneath the site with light non-aqueous phase liquid (LNAPL), which has migrated through vertical fractures up to 15m below the water table, in response to the pressure head developed by the connected vertical height of LNAPL in fractures (Wealthall *et al.* 2002). The theoretical basis for this behaviour is well understood (Mercer and Cohen, 1990; Newell *et al.* 1995; Hardisty *et al*, 1998) and relatively small volumes of LNAPL within vertical or sub-vertical fractures can produce significant LNAPL pressure heads that result in penetration into the saturated zone (Hardisty *et al.* 1998, 2003; Thornton and Wealthall, 2008; CL:AIRE, 2014). Dissolution of this residual LNAPL has contaminated the groundwater with petroleum hydrocarbons, including benzene, toluene, m/p-xylene, o-xylene (BTEX), the fuel additives MTBE, TAME, tert-butyl alcohol (TBA) and other aromatic compounds (Wealthall *et al.* 2001; Spence *et al.* 2005*a*).

The aquifer is unconfined in the vicinity of the site and overlain by ~7 m of drift deposits. The mean water table is ca. 20-22 m below ground level. The groundwater flow direction is south-easterly, approximately parallel to the main highway adjacent to the filling station (Spence *et al.* 2005*a*) (Figure 1). A network of single screen (~9 m screen length) monitoring wells (MW7, MW9, MW10, MW11, MW13, MW22 and MW24) and multilevel groundwater sampling (MLS) wells (MW14, MW15

MW16, MW17, MW18, MW19, MW20 and MW23) which allow level-discrete sampling over the vertical interval of the plume (Einarson and Cherry, 2002; Thornton and Wealthall, 2008) were installed at the site (Figure 1). An MLS, MW14, was installed upgradient of the site to characterise background groundwater quality. Similarly, MW23 is a MLS installed in the saturated zone underneath the former leaking UST. All monitoring wells are referenced relative to this datum (as meters below forecourt level, mbfl) using ordnance survey data.

Groundwater quality surveys of dissolved contaminant distribution were completed at intervals of 1-3 months after the fuel release. These show that the locations of MW10 and MW11 are uncontaminated and not on the plume flow path. Three comprehensive surveys of groundwater chemistry were completed for this study on all monitoring wells at annual intervals between 2001 and 2003.

Methods

Site investigation

The site investigation undertaken at the site is described in Wealthall *et al*. (2002). The methods used for groundwater sampling and analysis are explained in Spence *et al.* (2005*a*). Cored boreholes were drilled in the saturated zone at the locations of MW15, MW16, MW17, MW20 and MW23 using rotary water flush methods, whereas rock core from the unsaturated zone at MW23 was recovered using percussion drilling techniques (Spence *et al.* 2005*a,b*).

Vertical profiles of solute distribution were obtained from the MLS completions after installation of a Continuous Multichannel Tubing (CMT) system, containing seven monitoring ports fitted with 10 cm screens (Einarson and Cherry, 2002; Thornton and Wealthall, 2008). The MLS completions were installed to a depth of 55m, with monitoring intervals between 1-2 m in the plume wells (MW15, MW16, MW17, MW19 and MW23) and up to 7.6 m in the upgradient well (MW14).

Groundwater samples for chemical analysis were collected according to Spence *et al.* (2005*a*). Dissolved hydrocarbon, MTBE, TAME and TBA concentrations were determined using the method of

Dewsbury *et al.* (2003) and other chemical parameters were determined using standard methods (Spence *et al.* 2005*a*). Rock cores were sampled anaerobically at intervals of ~1m for pore water chemistry, as described by Spence *et al.* (2005*b*). This analysis also provides an estimate of the saturated porosity of samples over the same interval as the pore water chemistry. The chemical composition of pore water samples was determined using the methods previously indicated. Rock core from MW20 (uncontaminated location) was sampled at 1m intervals and analyzed for solid phase organic carbon content using standard methods (van Reeuwijk, 1987).

Mass balance

Conceptual model and assumptions

A conceptual model to estimate the contaminant mass balance in the saturated zone was developed to assess the fate of dissolved phase contaminants released to groundwater from dissolution of residual LNAPL trapped in fractures below the water table. The model is presented graphically in Figure 2 and the calculation of the different terms in the model is described in detail below. In essence the model considers saturated zone processes only, so the net release is the total spill mass (M_{spill}) less any product recovered from the unsaturated zone (e.g. by soil vapour extraction, SVE).

In the saturated zone contaminant is initially present as a separate non-aqueous phase that dissolves into flowing water present in fractures ($M_{SZ\ fract\ dissolved}$). This input over the history of the plume is estimated from groundwater monitoring well data, within the term $M_{SZ\ dissolved\ released}$. The current mass of dissolved contaminants in the saturated zone ($M_{SZ\ dissolved}$) is also estimated using the monitoring well data. Dissolved contaminants in the fracture water may biodegrade (included in the term $M_{SZ\ degraded}$) or diffuse into pore-water within the Chalk matrix ($M_{residual\ ED}$). Similarly, electron acceptors within the aquifer matrix bounded by the plume that have been consumed by biodegradation in the fractures is also included ($M_{residual\ EA}$). Biodegradation within fracture waters is assumed to be driven by electron acceptors advected into the source zone from upgradient ($M_{advected\ EA}$) and electron acceptors stored in the matrix pore-water within the plume volume that diffuse into fracture waters as biodegradation lowers EA concentrations in the fracture waters ($M_{diffusion\ EA}$). Biodegradation is

considered to only occur within the fractures, as pore throat sizes in the Chalk are typically too small for microbes to access (e.g. Frederickson *et al.* 1997; Bottrell *et al.* 2000, Rebata-Landa and Santamarina, 2006; Thornton *et al.* 2006; Bartlett *et al.* 2010). Contaminant that has diffused into the Chalk matrix pore-water is therefore treated as an immobile reservoir, part of which may partition by sorption to the small amount of organic matter present in the Chalk matrix ($M_{SZ\,sorbed}$). The overall mass of contaminant biodegraded in the saturated zone is represented bv $M_{SZ\,biodegraded}$.

The relevant input terms governing the distribution of dissolved phase fractions are :

$$M_{SZ\,dissolved\,released} = M_{SZ\,dissolved} + M_{SZ\,fract\,dissolved} + M_{SZ\,sorbed} + M_{SZ\,biodegraded} \tag{1}$$

where $M_{SZ\,dissolved\,released}$ is mass dissolved from the trapped LNAPL to form the plume, and $M_{SZ\,dissolved}$, $M_{SZ\,fract\,dissolved}$, $M_{SZ\,sorbed}$ and $M_{SZ\,biodegraded}$ represent dissolved mass present in the aquifer matrix, fractures, sorbed to the aquifer matrix and biodegraded in the groundwater, respectively.

$M_{SZ\,biodegraded}$ includes contributions from the following sources:

$$M_{SZ\,biodegraded} = M_{advected\,EA} + M_{residual\,EA} + M_{diffusion\,EA} \tag{2}$$

where $M_{advected\,EA}$ represents biodegradation due to consumption of aqueous electron acceptors (O_2, NO_3, SO_4) supplied by advection of uncontaminated groundwater into the plume, $M_{residual\,EA}$ reflects biodegradation due to Mn(IV)/Fe(III) oxide-reduction, SO_4-reduction and methanogenesis within the plume, and $M_{diffusion\,EA}$ reflects biodegradation due to consumption of aqueous electron acceptors originally present in the aquifer matrix bounded by the plume. Biodegradation reactions are assumed to be instantaneous for petroleum hydrocarbons (Wiedemeier *et al.* 1999).

The plume is represented as a box model, in which reactions describing contaminant attenuation are coupled with transport processes to determine the spatial and temporal distribution of mass in the

system. Following Anderson *et al.* (2007), the plume width, depth and length are assumed to be respectively defined by the distance between contaminated monitoring wells transverse to the flow path, depth of dissolved contaminants in MLS boreholes close to the source, and maximum downgradient migration of dissolved contaminants, determined from the MLS installations. Monitoring data show that the groundwater flow direction is consistent over the reference period of the plume (December 1999-July 2003).

Inputs in the mass balance are estimated from measurements of groundwater hydrochemistry, aquifer hydrogeology, geochemical properties and other site data. The groundwater chemistry of MW7 and MW13 is assumed to represent the potential plume source composition. This is justified as the MLS profiles show that contaminants are not present in the saturated zone below the former leaking tank (at the location of MW23) and MW7, 16m down hydraulic gradient of MW23, contains the highest concentrations of dissolved hydrocarbons in groundwater at the site.

Plume source term input

This input was evaluated using monthly time-series data for contaminant concentrations in MW7 (Figure 3) and MW13 (Figure 4). These monitoring wells have the highest dissolved contaminant concentrations and are immediately down hydraulic gradient of the UST location (MW23). The total mass of dissolved contaminants released into the plume over this period, $M_{SZ\ dissolved\ released}$, was obtained from the following relationship:

$$M_{SZ\ dissolved\ released} = C_{source} \text{ x Thickness }_{K\ interval} \text{ x W x K x I x T} \tag{3}$$

where C_{source} (M/L^3) is the concentration of each contaminant in MW7 or MW13, W is the plume width (L), I is the hydraulic gradient (L/L), K is the aquifer hydraulic conductivity (L/T), Thickness $_{K\ interval}$ is the packer test interval (aquifer thickness) used in the hydraulic tests to determine the aquifer hydraulic conductivity at successive depths (L), and T is the time period of the analysis (December 1999 – July 2003).

To estimate the contaminant mass flux from MW7 or MW13, contaminant concentrations were assumed to be consistent over the vertical interval of aquifer monitored by each well, to the full depth of the plume. This is reasonable as these monitoring wells have 9m screens and vertical mixing of groundwater is expected within the casing, leading to homogenisation of vertical contaminant concentration gradients across the screen (Puls and Paul, 1997; Schreiber and Bahr, 1999). The temporal variation in dissolved contaminant concentrations in MW7 and MW13 likely results from groundwater intermittently accessing fresh LNAPL trapped in fractures in the unsaturated zone and an increased flux of dissolved contaminants through a zone of more intensive fracturing in the shallow saturated zone (deduced from rock core fracture logs and tracer tests), due to seasonal variation in recharge and fluctuating water table (Hardisty *et al,* 1998; Bottrell *et al*, 2010). This is a characteristic feature of the Chalk aquifer properties and hydrogeology (Allen *et al.* 1997). There is a significant variation in aquifer vertical hydraulic conductivity, which will determine the vertical variation in contaminant mass flux over the depth of the plume. The hydraulic conductivity profile for MW15 (Supplementary Figure 1), the MLS borehole closest to MW7 and MW13 (Figure 1), was used in this calculation. The plume width, W, is assumed to be the distance between contaminated monitoring wells, MW13 and MW19, transverse to the plume flow path.

Dissolved contaminant mass

Inputs of $M_{SZ\,dissolved}$, and $M_{SZ\,fract\,dissolved}$ were estimated using x-z contour plots of contaminant concentrations obtained from the June 2002 groundwater monitoring survey (Figure 4). The contaminant concentrations in 2002 were greater than in the 2001 and 2003 surveys, representing the maximum input for the mass balance. Due to different porosity terms, $M_{SZ\,fract\,dissolved}$ was estimated using:

$$M_{SZ\,fract\,dissolved} = \varepsilon_f \int C_{residualED}\, \partial V \qquad (4)$$

where ε_f is the fracture porosity (dimensionless), $C_{residual\ ED}$ is the mean residual electron donor (organic contaminant) concentration for a given contour interval (M/L^3) and V is the volume of aquifer contained within the contour interval (L^3). A fracture porosity value of 0.01 was used in the calculations for $M_{SZ\ fract\ dissolved}$, as typical for the Upper Chalk aquifer (Price, 1996). Values of $M_{SZ\ dissolved}$ were estimated using:

$$M_{SZ\,dissolved} = \varepsilon_m \left(1 - \varepsilon_f\right) \int C_{residualED} \partial V \tag{5}$$

where ε_m is the matrix porosity (dimensionless) and all other terms are as for $M_{SZ\ fract\ dissolved}$. Values of ε_m were obtained from analyses (n=17) of rock core collected below the location of the former UST at MW23 (Spence *et al.* 2005*b*). Only ε_m values for the monitored depth of the plume (20-40m) were used.

Sorbed contaminant mass

Sorption of hydrophobic organic compounds such as petroleum hydrocarbons to the aquifer matrix can be approximated by the following expression (Schwarzenbach *et al.* 1993):

$$S = K_d C \tag{6}$$

where S is sorbed solute concentration ($M_{solute}/M_{sorbent}$), C is dissolved solute concentration ($M_{solute}/L^3_{solution}$) and K_d is the equilibrium distribution coefficient ($L^3_{solution}/M_{sorbent}$). K_d can be used to estimate a retardation factor, R_f, describing the effect of sorption on solute transport in the porous medium (Fetter, 1999):

$$R_f = 1 + \frac{\rho_b K_d}{\varepsilon} \tag{7}$$

where ρ_d is the bulk density of the porous medium (M/L^3) and ε is the porosity (dimensionless). The contribution of fracture porosity was considered by replacing ε with (ε_m - ε_f) in Eq. 7. The sorbed mass of a solute can then be estimated according to (King *et al.* 1999):

$$\text{Sorbed mass} = \text{Dissolved mass} \times (R_f - 1) \qquad (8)$$

Eq. 8 was used to estimate $M_{SZ\ sorbed}$ for each contaminant in the plume, using estimates of $M_{SZ\ dissolved}$ (Eq. 5) and R_f (Eq. 7). It was assumed that sorption occurs after diffusion into the matrix and is negligible for fracture surfaces. This is justified since contaminants diffuse from the fractures into the aquifer matrix, which has a significantly greater surface area. There is excellent agreement between vertical profiles of organic contaminants in matrix pore water from rock cores and fracture water chemistry in MLS wells installed in the same borehole (Spence *et al.* 1995b). This suggests that there is chemical equilibrium between contaminant concentrations in the fracture water and matrix pore water, supporting the use of a linear sorption model in the calculations. Values of K_d for contaminants in the plume were estimated indirectly by two methods using empirical relationships from the literature, described below.

For low solute concentrations, K_d, considered in terms of the organic carbon-normalised distribution coefficient, K_{oc}, can be expressed as a linear function of the fraction of particulate organic carbon (f_{oc}) on the sorbent (aquifer matrix) using the relationship (Schwartzenbach *et al.* 1993):

$$K_d = K_{oc} \times f_{oc} \qquad (9)$$

where K_{oc} has units of L^3/M and f_{oc} is dimensionless. Published values of K_{oc} were used for organic compounds (Fetter, 1999). However, for 1,2,4-trimethylbenzene (1,2,4-TMB) and 1,3,5-trimethylbenzene (1,3,5-TMB) K_{oc} was estimated using the general relationship (Schwarzenbach and Westall, 1981):

$$\log K_{oc} = a \log K_{ow} + b \tag{10}$$

where K_{ow} is the octanol-water partition coefficient for the organic compound, and a and b are regression coefficients. Values of 1.01 and –0.72 were respectively used for the coefficients a and b in Eq. 10, based on the empirical expression derived for aromatic hydrocarbons in Schwartzenbach *et al.* (1993, p.274, Eq. 11-22a), to estimate K_{oc} for 1,2,4-TMB and 1,3,5-TMB. The K_{oc} estimates were then used to calculate K_d values for these two compounds using Eq. 10. The f_{oc} values used to estimate organic contaminant K_d were obtained from chemical analysis of uncontaminated rock core collected from MW20 (Figure 1) (Bottrell *et al*, 2010).

Biodegraded contaminant mass

Estimates of $M_{advected\ EA}$ were obtained by comparing profiles of dissolved O_2, NO_3 and SO_4 in the upstream (uncontaminated) MLS (MW14) with MW15, the MLS immediately downgradient of the plume source (Figure 1). Concentrations of dissolved O_2 and NO_3 are below detection limits in MW15 and dissolved SO_4 is depleted, relative to MW14 on each annual survey (Spence et al, 2005a). This indicates complete consumption of O_2 and NO_3 supplied to the plume by advection, with partial consumption of SO_4 in the plume. The contribution of each electron acceptor in the budget for $M_{advected\ EA}$ was quantified using the following expression:

$$M_{advected\ EA} = C_{background\ EA} \times Thickness_{K\ interval} \times W \times K \times I \times T \tag{11}$$

where $C_{background\ EA}$ is the concentration (M/L^3) of O_2, NO_3 and SO_4 in uncontaminated groundwater (MW14) and the other terms are as in Eq. 3. Dissolved electron acceptor concentrations obtained from two monthly and three annual groundwater quality surveys of MW14 between 2000 and 2003 were used in the calculation of this input. It is assumed that concentrations of O_2, NO_3 and SO_4 in groundwater entering the plume at the location of MW15 are identical to those in the profile at MW14 over the plume history. There is good agreement between profiles of these electron acceptors obtained

from the annual surveys of MW14, suggesting that this is reasonable (Spence et al, 2005a). The hydraulic conductivity profile for MW15 (Supplementary Figure 1) was used to calculate the mass flux of each electron acceptor entering the rear of the plume; this input is then a function of the aquifer K-profile in MW15 and the electron acceptor profile in MW14. The electron acceptor mass flux profile obtained for the monitoring intervals in MW14 was then integrated using the trapezoid rule over the full depth (~15m) of the plume profile at MW15.

Estimates of $M_{residual\ EA}$ are based on the residual concentration of Mn(II), Fe(II) and CH_4 in the plume, which are higher than those in the uncontaminated groundwater at MW14. This input was determined by integration of the contoured x-z concentration plots for the dissolved Mn(II), Fe(II) and CH_4 distribution in 2002, using the volume of aquifer contained within a contoured interval and a mean solute concentration for that contour interval, according to:

$$M_{residualEA} = \varepsilon_m \int C_{residualEA} \partial V \qquad (12)$$

where $C_{residual\ EA}$ is the mean concentration of the residual species for a given contour interval (M/L^3). Although SO_4-reduction occurs in the plume, SO_4 was never depleted (Spence et al, 2005a) and this residual SO_4 must be considered in the estimation of $M_{SZ\ biodegraded}$. The depletion of SO_4 in groundwater is strongly correlated with the distribution of BTEX (Spence *et al.* 2005*a*) and $M_{residual\ EA}$ for SO_4 was therefore estimated with Eq. 12 by integration of the SO_4 contour plot over the area occupied by the benzene plume in 2002. This year corresponds to the maximum spatial distribution of dissolved BTEX in the aquifer.

$M_{diffusion\ EA}$ is the contribution to $M_{SZ\ biodegraded}$ from the consumption of aqueous electron acceptors (O_2, NO_3 and SO_4) formerly present in the uncontaminated aquifer matrix but now depleted by biodegradation in the plume. These electron acceptors diffuse from the matrix to support biodegradation in the fractures. This is supported by the excellent agreement between matrix pore

water and fracture water profiles of these species (Spence *et al.* 2005*b*). It indicates that there is rapid chemical equilibrium between the fracture water and matrix pore water and that the original reservoir of O_2 and NO_3 in the matrix pore water has diffused into the fractures and been consumed by biodegradation. The exception is SO_4, which remains as a residual (incompletely used) fraction, with lower concentrations in the plume than in the uncontaminated aquifer.

In estimating $M_{\text{diffusion EA}}$ it was assumed that the aquifer matrix now occupied by the plume originally had dissolved O_2, NO_3 and SO_4 concentrations ($C_{\text{background EA}}$) equivalent to those in uncontaminated groundwater at MW14. Also, the relationships noted above between contaminant and aqueous electron acceptor distribution in the matrix pore water, fracture water and plume contour plots, suggest that aqueous electron acceptors in the matrix will be depleted over a volume of aquifer corresponding to the BTEX plume. Values of $M_{\text{diffusion EA}}$ were therefore estimated for O_2, NO_3 and SO_4 using the following expression:

$$M_{\text{diffusion EA}} = \varepsilon_m \left(1 - \varepsilon_f \right) \int C_{\text{background EA}} \, \partial V_{\text{benzene plume}} \tag{13}$$

where $V_{\text{benzene plume}}$ is the volume of aquifer (L^3) occupied by the benzene plume in 2002 (representing the maximum spatial distribution of dissolved BTEX contaminants).

Probabilistic analysis

Probabilistic analyses were performed using Oracle Crystal Ball®. Different scenarios were evaluated to assess the variation in mass balance predictions due to variability in parameter values. This was achieved using Monte Carlo methods and probability distribution functions (pdfs) assigned to each data range in Table 1. The significance of parameter uncertainty on the mass balance was also evaluated using a sensitivity analysis for each input. Uncertainty is assessed in the sensitivity analysis using the pdfs assigned to parameters and is propagated through the various embedded forecasts in the

simulation. In this way the probabilistic analysis assessed uncertainty in the conceptual model and input data used in the mass balance.

Monte Carlo simulations were run for 10,000 realisations and the significance of forecasts was evaluated at the 95% confidence interval (Environment Agency, 2001b; Anderson et al., 2007). Specific pdfs describing parameter inputs for the Monte Carlo simulations were obtained by statistical analysis of the site data using the Crystal Ball software (Supplementary Table 1), to account for the site-specific variability in parameter values (McNab and Dooher, 1998; Soutter and Musy, 1998; Wang *et al.* 2004; Anderson *et al.* 2007; Liang *et al.* 2011). However, in some cases an assumed pdf was assigned to a parameter input where data was limited, according to guidance and previous studies (Environment Agency, 2001; Liang, 2009; Liang *et al.* 2011).

Plume source term

Estimates of $M_{SZ\,dissolved\,released}$ were evaluated for three scenarios, considering time-variant contaminant concentrations for MW7 *only*, MW13 *only* and MW7 *combined* with MW13. This evaluated uncertainty in the conceptual model for the plume source term. Four values (0.0021, 0.0022, 0.0024 and 0.0031) of the hydraulic gradient, I_x, obtained from separate groundwater elevation surveys were used in this analysis. A uniform pdf was assigned to this range (Environment Agency, 2001), which was also used to calculate other input terms requiring this parameter.

Plume residuals

The mass of plume residuals (dissolved organic contaminants and electron acceptors/biodegradation products) was estimated by assigning a triangular pdf to each set of concentration ($C_{residual\,ED}$ or $C_{residual\,EA}$) contours, using the lower, arithmetic mean and upper solute concentration as respective minimum, most likely and maximum values in the distribution (Thornton *et al.* 2001). An extreme value probability distribution was obtained by fitting Crystal Ball to the data range for matrix porosity, ε_m. This distribution was truncated at the minimum and maximum values of ε_m as these defined the range used.

Sorbed contaminant mass

Three parameter values, f_{oc}, ε_m and ρ_b, were evaluated in the calculation of $M_{SZ\,sorbed}$. Crystal Ball was used to fit a Weibull pdf, truncated at the minimum and maximum values defining the range, to 29 values of f_{oc}, obtained by analysis of rock core from MW20. The values and pdfs for ε_m used in the calculation of plume residuals (see above) were included to estimate this input. Values of ρ_b were obtained using an expression which includes ε_m and so the same probability distribution (Extreme value) was used for this parameter in the Monte Carlo simulations.

Biodegraded contaminant mass

In the calculation of $M_{advected\,EA}$ all data from two monthly and three annual groundwater quality surveys of MW14 between 2000 and 2003 were used to determine the electron acceptor concentration, $C_{background\,EA}$, used in Eq. 12. This accounts for temporal variability in the background electron acceptor mass flux entering the plume via advection of uncontaminated groundwater. In this case, $C_{background\,EA}$ was obtained by fitting a pdf to the time-series data for each monitoring point on MW14 (seven in total) using Crystal Ball. This provides different values of $C_{background\,EA}$ as a function of depth and time over a significant part of the plume history. Estimates of $M_{diffusion\,EA}$ also included the pdfs fitted by Crystal Ball to $C_{background\,EA}$.

Results

Contaminant mass balance

The mass balance for the saturated zone of the aquifer is shown in Table 2. This includes estimates of total contaminant mass and mass of BTEX *only*, where relevant. Mean values of predictions and the output range between the 95% confidence level are provided for each forecast. All masses are in kilograms.

Values of $M_{SZ\,dissolved\,released}$ predicted using the MW7 data are higher than those using the MW13 data. These estimates vary by orders of magnitude at the 95% confidence interval. However, mean values of $M_{SZ\,dissolved\,released}$ predicted using a combination of MW7 and MW13 data vary much less. The residual contaminants in the saturated zone include dissolved mass in the matrix ($M_{SZ\,dissolved}$), fractures ($M_{SZ\,fract\,dissolved}$) and sorbed to the matrix ($M_{SZ\,sorbed}$). Residual dissolved contaminants in the matrix

represent the largest fraction, with approximately 50% (110kg) comprising BTEX alone. The dissolved contaminant mass in fractures is very small, due to the low fracture porosity. Most (94% based on mean values) of the contaminant fraction sorbed to the aquifer matrix is accounted for by BTEX. This group also accounts for ca. 62% of the total residual contaminants (205 kg) in the saturated zone (Table 2).

Plume balance

The consumption of electron acceptors in the saturated zone ($M_{SZ \ biodegraded}$) is equivalent to biodegradation of 1,595 kg of BTEX, with a relatively restricted range (Table 2). This is based on the reaction stoichiometry for individual biodegradation processes (Wiedemeier *et al.* 1999) and assumes that oxidant consumption results primarily from BTEX biodegradation, which is supported by stable isotope studies of contaminant biodegradation (Spence *et al.* 2005*a*).

The sum of the total residual contaminant and biodegraded hydrocarbon fractions can be compared with the predicted total dissolved contaminant mass released in the saturated zone, according to Eq 1. The latter implies that the dissolved contaminant mass released into the groundwater from LNAPL dissolution should be balanced by the residual dissolved contaminant mass distributed between different aquifer fractions and the contaminant mass biodegraded within the plume. The respective input terms in Eq. 14 are summarised in the "plume balance" (Table 2). There is good agreement in the results when using the combined MW7 and MW13 data as the plume source term, particularly for the "all contaminant" fraction. The residual and biodegraded contaminant mass is 1,927 kg (range of 1,647-2,169 kg) and this brackets the mean value of $M_{SZ \ dissolved \ released}$ (1,796 kg) predicted using the combined MW7 and MW13 data (Table 2).

Oxidant mass balance

The consumption of electron acceptors contributed by advection, matrix diffusion and plume residuals during biodegradation in the saturated zone ($M_{SZ\,biodegraded}$) is presented in Table 3. An estimated 7,681 kg of electron acceptors have been consumed, with aqueous oxidants dominating the budget.

The consumption of dissolved NO_3 and SO_4 accounts for 96% of the total electron acceptor budget in the plume, whereas Mn/Fe mineral oxides contribute very little (<1% in total). The contribution of methanogenesis in biodegradation is negligible. The mass balance suggests that the supply of dissolved O_2 and NO_3 by advection into the plume is more important for biodegradation than diffusion from the matrix, over the timescale of the analysis. The true contribution of advection and diffusion in the SO_4 budget cannot be estimated because the residual SO_4 in the plume may include contributions from both sources. Instead, the estimates of SO_4 from advection and diffusion are maximum values over the timescale of the analysis and the SO_4 consumption is obtained by subtracting the residual plume SO_4 estimate from the sum of these two components (Table 3).

Sensitivity analyses

The value of $M_{SZ\,dissolved\,released}$ was primarily sensitive to contaminant concentrations rather than the aquifer properties (hydraulic gradient). The aquifer matrix porosity was most important for estimates of $M_{SZ\,dissolved}$, whereas estimates of $M_{SZ\,sorbed}$ are primarily affected by uncertainty in the f_{oc} of the aquifer matrix, which accounts for 93% of the variability. This is consistent with similar studies assessing organic contaminant fate (e.g. Soutter and Musy, 1998; Dubus *et al*. 2003) and is expected, since hydrocarbon sorption is strongly controlled by the organic matter content of the aquifer matrix.

Electron acceptor inputs in the mass balance are controlled primarily by the aquifer properties. Estimates of $M_{advected\,EA}$ and $M_{diffusion\,EA}$ are influenced by, respectively, the hydraulic gradient and matrix porosity, more than variability in the background groundwater electron acceptor concentration. Similarly, $M_{SZ\,biodegraded}$ also depended primarily on the variation in aquifer hydraulic gradient, through the influence on $M_{advected\,EA}$.

Regarding the plume balance, variability in the aquifer f_{oc} significantly (90%) controls the predicted total residual contaminants in the saturated zone (Inputs 9 and 10 in Table 2). However, the sum of residual contaminants and equivalent BTEX biodegraded in the saturated zone (Inputs 12 and 13 in Table 2) were most sensitive to variation in aquifer hydraulic conductivity (79-81% of total variance). The variation in other aquifer and hydrochemical properties (e.g. matrix porosity, f_{oc} and background SO_4 concentration) contributed much less (<15% in total) to the uncertainty in this prediction.

Discussion

Sensitivity of mass balance to uncertainty of input parameters

By identifying critical inputs which introduce most uncertainty in the mass balance, the sensitivity analysis indicates system parameters or aquifer properties which need greater characterisation during a site investigation for performance assessment of NA using this methodology (Ho *et al.* 2007; Zhang *et al.* 2010; Beck and Mann, 2012). Additional data which accounts for a greater range in magnitude of a sensitive parameter or property may in turn significantly reduce the uncertainty in the mass balance input terms, providing greater confidence in predictions (Hosseini, 2009).

Most pdfs assigned to parameters were obtained by fitting the Crystal Ball library to site-specific data sets. This means that the sensitivity analyses probably reflect the true effect of data variability on predictions rather than the assigned pdf (Anderson *et al.* 2007). A uniform distribution was assumed for the hydraulic gradient (Environment Agency, 2001), but a previous study by the authors shows this property is relatively insensitive to the pdf assigned to it in a plume-scale mass balance (Thornton *et al.* 2001).

The mass balance is most sensitive to the aquifer hydraulic gradient, and less so to other aquifer properties such as f_{oc}, matrix porosity or the concentration of electron acceptors in background groundwater. This probably occurs because the hydraulic gradient typically exhibits significant

variability in the Chalk aquifer (MacDonald and Allen, 2001) and controls the groundwater flux through the plume. In turn, this controls the advective mass flux of aqueous electron acceptors into the plume and quantity of contaminants biodegraded.

Estimated plume source term

The plume source term was evaluated using contaminant concentrations in MW7 and MW13. These are the highest in the saturated zone but with considerable spatial-temporal variation. Both monitoring wells are transverse to the plume flow path (Figure 1) and could therefore contribute dissolved contaminants to the plume. For this reason, the mass balance considered MW7 and MW13 in isolation and in combination.

Contaminant concentrations in MW13 are generally much lower than those in MW7 (cf Figures 2 and 3) and $M_{SZ\,dissolved\,released}$ predicted for the MW13 data is less than the sum of residual and biodegraded contaminants in the plume (Input 12 and 13 in Table 2). This suggests that the plume is not generated by contaminant concentrations found exclusively in MW13. The large range in $M_{SZ\,dissolved\,released}$ predicted for the MW7 data results from the Monte Carlo sampling of the data used in this forecast. However, it is unrealistic that the highest value will define most of the mass released from this location, because contaminant concentrations in MW7 peak in October 2000 and are much lower over most of the simulation period. This also suggests that the plume source is not described by the MW7 data exclusively.

Intuitively the plume should result from dissolved contaminants contributed by both MW7 and MW13. This is because there are contaminated MLS boreholes downgradient of MW7 (e.g. MW15 and MW19) and MW13 (e.g. MW17 and MW16) along the plume flow path (Figure 1). The range in forecast values of $M_{SZ\,dissolved\,released}$ produced from a combination of MW7 and MW13 data is less than that for the MW7 data (Table 2). Again, the range predicted by the probabilistic analysis is based on the Monte Carlo sampling of the combined data set and these extreme values may not be realistic for

the plume. This is because the possibility of all parameters taking their extreme values simultaneously is less likely than the possibility of any one of them taking its extreme value individually (Environment Agency, 2001). It is more appropriate to compare the mean values of predicted estimates, although the simulation will include statistical analysis of the distribution. A statistical distribution can be fitted to $M_{SZ\,dissolved\,released}$ forecast for MW7 and MW13 using Crystal Ball (Figure 5). The forecast has a lognormal distribution, which allows the plume source term in the saturated zone to be predicted (with respect to dissolved contaminant load) from the time-series contaminant concentration data for these two monitoring wells. This can also be undertaken to forecast future concentrations or mass flux estimates of individual or grouped contaminants. The uncertainty in this forecast can be further reduced by including more time-series data from these monitoring wells in the simulation.

Residual plume contaminant mass

The residual contaminant mass in the saturated zone, represented by dissolved compounds in the matrix ($M_{SZ\,dissolved}$), fractures ($M_{SZ\,fract\,dissolved}$) and sorbed to the matrix ($M_{SZ\,sorbed}$) comprises 332 kg of total contaminants (Input 9, Table 2). The respective estimates for the BTEX group are much lower (Input 10, Table 2). This difference occurs because there is a much higher mass of dissolved MTBE and TAME in the matrix, rather than sorbed to the aquifer compared with BTEX. This is shown in Table 2, where estimates of $M_{SZ\,dissolved}$ for BTEX are approximately half those of the total contaminants, whereas estimates of $M_{SZ\,sorbed}$ for BTEX are only slightly less than those for the total contaminants. Due to higher solubility, MTBE and TAME will sorb less to the aquifer matrix compared with the BTEX compounds. This is illustrated by the respective distribution coefficients (K_d) and corresponding retardation factors (R_f) estimated for these compounds (Supplementary Table 2), which show that significant attenuation of the hydrocarbon plumes by sorption can be expected. However, there will be negligible retardation of the MTBE and TAME plumes. This is confirmed by comparing groundwater velocities from tracer tests with the size of the MTBE plume, which indicate minimal attenuation by sorption (Bottrell *et al.* 2010).

Electron acceptor consumption

Advection is predicted to supply most of the dissolved O_2, NO_3 and SO_4 consumed by biodegradation in the plume (Table 3). However, consumption of these electron acceptors originally present in the matrix now contained within the plume is also important. These respective masses ($M_{advected\ EA}$ and $M_{diffusion\ EA}$) must be considered maximum possible inputs in each case, because the residual SO_4 present in the plume cannot be separated into components contributed by advection or matrix diffusion. However, if all the residual SO_4 in the plume was contributed by diffusion, the estimated (adjusted) value of $M_{diffusion\ EA}$ (2,482 kg) would still not exceed the value of $M_{advected\ EA}$ (3,570 kg).

The contribution of electron acceptors in the matrix is finite and limited to the volume of aquifer occupied by the plume. It is reasonable to assume that aqueous O_2, NO_3 and SO_4 originally present in the uncontaminated aquifer matrix has diffused into the fractures to support contaminant biodegradation in the plume (Spence *et al.* 2005*a*). The absence or reduced concentration of O_2, NO_3 and SO_4 in the plume indicates the matrix has been depleted in these oxidants. When this reservoir is exhausted, biodegradation will be primarily sustained by the supply of these electron acceptors from advection. However, as the plume expands downgradient into the aquifer, these oxidants will be supplied by diffusion from the uncontaminated matrix to support contaminant biodegradation in the fractures (Spence *et al.* 2005*a*). The rate of plume expansion will then be a function of the type and quantity of electron acceptors in the matrix, the composition of the contaminant plume, matrix-fracture mass transfer rate (controlled by diffusion) and biodegradation kinetics. More contaminant mass can be biodegraded by NO_3 and SO_4 than O_2, and the relatively high concentrations of NO_3 and SO_4 in the matrix has supported significant attenuation of the BTEX plume by biodegradation.

The contribution of Mn(IV)-reduction, Fe(III)-reduction and methanogenesis in contaminant biodegradation is minor compared with aerobic respiration, denitrification and sulphate reduction. The consumption of mineral oxidants in the plume is <1% of the total electron acceptor budget. The

estimate for Fe is a minimum value because dissolved Fe can precipitate with sulphide to form iron sulphide phases (FeS) under the SO_4-reducing conditions in the plume (Spence *et al.* 2005*a*). However, metal oxide coatings were observed on some core samples, suggesting the contribution of mineral oxidants to biodegradation in the saturated zone is small. Methanogenesis is not an important pathway for contaminant degradation in the plume (Table 2), most likely due to the greater availability of NO_3 and SO_4 in the groundwater.

Plume balance

The sum of all dissolved contaminants released into the groundwater (Input 1, Table 2) is in good agreement with the residual and biodegraded mass of all contaminants in the plume (Input 12, Table 2). The difference in the mass balance is only 7%.

As the estimate of $M_{SZ\,dissolved\,released}$ (Input 1, Table 2) and plume balance (Input 12 and 13, Table 2) for the combined MW7 and MW13 scenario have been independently determined, but agree well in the mass balance, it is reasonable to assume that this scenario adequately represents the conceptual model for the plume source term. This is further evaluated in Figure 5, which shows overlays of $M_{SZ\,dissolved\,released}$ and the plume balance forecast for different scenarios. The distributions of each forecast are mutually exclusive when using the MW13 data to estimate $M_{SZ\,dissolved\,released}$ (Figure 5a), but overlap for both the total and BTEX-only plume balance when the combined MW7 and MW13 data is used (Figure 5b and 5c). This suggests that the plume balance reflects the release of dissolved contaminants from MW7 and MW13 in combination and that the amount of BTEX biodegradation attributable to electron acceptor consumption in Table 2, $M_{SZ\,biodegraded}$, is a reasonable estimate of BTEX mass turnover for the period of study.

A plume-scale biodegradation rate for BTEX can be estimated using the equivalent mass of BTEX biodegraded in the plume, $M_{SZ\,biodegraded}$, and the plume balance (Inputs 11 and 12 in Table 2). The plume balance gives a total contaminant load of 1,927 kg in the saturated zone, of which 1,595 kg has been biodegraded between December 1999 and July 2003. Assuming a first-order kinetic model, this

mass loss corresponds to a first-order biodegradation rate, λ, of 0.0013 day^{-1} or 0.49 yr^{-1} for BTEX. This corresponds to a half-life ($T_{0.5}$) of 1.41 years for the BTEX plume. It should be noted that the estimated BTEX biodegradation rate is a spatially and temporally averaged value, scaled over the volume and history of the BTEX plume, for the time period assessed. However, the biodegradation rate compares well with estimates obtained for BTEX plumes in other studies using different methods (Suarez and Rifai, 1999; Wiedemeier *et al.* 1999).

Conclusion

The mass balance methodology developed in this study supports field-scale performance assessment of NA for dissolved petroleum fuel components in fractured, dual porosity aquifers. It provides a robust quantitative framework to estimate site-specific mass balances for groundwater plumes using site investigation data, to evaluate the distribution of dissolved contaminants in the saturated zone, estimate plume source terms, contaminant biodegradation and plume status (shrinking, stable, expanding), and identify potential limits on NA performance (e.g. oxidant supply). It is a useful screening tool to evaluate sites where MNA is being considered with other management options and also prior to undertaking more sophisticated numerical fate and transport modelling for site assessment. The methodology is transferable to, and can be further developed for, other dual porosity aquifer settings, when supported by appropriate data collection and site characterization. The approach can be also used to assess the NA of other organic contaminants in such aquifers and when combined with probabilistic analysis can address the inherent uncertainty in site investigation data.

For the case study, probabilistic analysis was particularly helpful in deducing the likely plume source term from different scenarios and predicting the long-term plume source composition as a probability distribution. This is advantageous in providing a site-specific estimate of the contaminant mass flux for forward modelling of plume development and NA performance. It also showed that errors or uncertainty in the conceptual model for the plume source term had a significant influence on the overall plume mass balance and interpretation of NA. However, the sensitivity analysis enabled the

parameters which have greatest control on predictions to be identified, in order to prioritise the site investigation to reduce the uncertainty in data inputs.

The biodegradation rate estimated for the BTEX plume using the probabilistic analysis is consistent with that reported for other BTEX plumes in the literature. Biodegradation using NO_3 and SO_4 supplied by advection and matrix diffusion appears to be a significant process limiting BTEX migration in the aquifer. Other biodegradation processes are limited by oxidant availability. Sorption to the aquifer matrix is also important for the attenuation of BTEX but less so for MTBE and TAME, which are more mobile in the groundwater.

Acknowledgements

References

Abbaspour, K.C., Schulin, R., van Genuchten, M.Th. and Schlappi, E. (1998). Procedures for uncertainty analyis applied to a landfill leachate plume. Ground Water, 36, 874-883.

Allen, D.J., Brewerton, L.J., Coleby, L.M., Gibbs, B.R., Lewis, M.A., MacDonald, A.M., Wagstaff, S.J. and Williams, A.T. (1997). The physical properties of major aquifers in England and Wales. BGS Technical Report, WD/97/34, Environment Agency R&D Publication, 8.

Anderson, B.A., Maslia, M.L., Caparoso, J.L. and Ausdemore, D. (2007). Probabilistic analysis of pesticide transport in shallow groundwater at the Oatland Island Education Center, Oatland Island, Georgia. Atlanta, GA: Agency for Toxic Substances and Disease Registry; 2007., pp.64.

Arey, J.S. and Gschwend, P.M. (2005). A physical-chemical screening model for anticipating widespread contamination of community water supply wells by gasoline constituents. Journal of Contaminant Hydrology, 76, 109-138.

ASTM (2016). Standard guide for remediation of ground water by natural attenuation at petroleum release sites. ASTM E1943-98(2015), pp.43.

Baalousha, H. (2006). Groundwater pollution risk using a modified Latin hypercube sampling. Journal of Hydroinformatics, 8, 223-234.

Bartlett, R., Bottrell, S.H., Sinclair, K., Thornton, S., Fielding, I.D. and Hatfield, D. (2010). Lithological controls on biological activity and groundwater chemistry in Quaternary sediments. Hydrological Processes, 24, 726-735.

Beck, P.H. and Mann, B. (2012). Fate and transport modelling for monitored natural attenuation projects: what should be considered to maximize the value in decision making? In *Models –*

Repositories of Knowledge, Proceedings ModelCARE2011 held at Leipzig, Germany, in September 2011), IAHS Publ. 355, 149-156.

Benekos, I.D., Shoemaker, C.A. and Stedinger, J.R. (2007). Probabilistic risk and uncertainty analysis for bioremediation of four chlorinated ethenes in groundwater. Stochastic Environmental Research and Risk Assessment, 21, 375-390.

Bloomfield, J. (1996). Characterisation of hydrogeologically significant fracture distributions in the Chalk: an example from the Upper Chalk of southern England. Journal of Hydrology, 184, 355-379.

Bottrell, S.H., Moncaster, S.J., Tellam, J.H., Lloyd, J.W., Fisher, Q.J. and Newton, R.J. (2000). Controls on bacterial sulphate reduction in a dual porosity aquifer system: the Lincolnshire Limestone aquifer, England. Chemical Geology, 169, 461-470.

Bottrell, S.H., Thornton, S.F., Spence, M.J., Allshorn, S. and Spence, K.H. (2010). Assessment of the use of fluorescent tracers in a contaminated Chalk aquifer. Quarterly. J. of Engineering Geology & Hydrology, 43, 1-14.

Bolster, D., Barahona, M., Dentz, M., Fernandez-Garcia, D., Sanchez-Vila, X., Trinchero, P., Valhondo, C. and Tartakovsky, D.M. (2009). Probabilistic risk analysis of groundwater remediation strategies. Water Resources Research, 45, W06413, https://doi.org/10.1029/2008WR007551.

Bredehoeft, J.D. (2003). From models to performance assessment: The conceptualisation problem. Ground Water, 41, 571-577.

Bredehoeft, J. D. (2005). The conceptualisation model problem-surprise. Hydrogeology Journal, 13, 37-46.

Burgess, W.G., Dottridge, J. and Symington, R.M. (1998). Methyl tertiary butyl ether (MTBE): a groundwater contaminant of growing concern. In: Mather, J., Banks, D., Dumpleton, S & Fermor, M. (eds), Groundwater Contaminants and their Migration. Geological Society, London, Special Publications, 128, 29-34.

Chen, Y-C. and Ma, H-W. (2007). Combining the cost of reducing uncertainty with the selection of risk assessment models for remediation decision of site contamination. Journal of Hazardous Materials, 141, 17-26.

Chen, X., Goeke, J. and Summerside, S. (1999). Hydraulic properties and uncertainty analysis for an unconfined aquifer. Ground Water, 37, 845-854.

Chisala, B.N., Tait, N.G. and Lerner, D.N. (2007). Evaluating the risks of methyl tertiary butyl ether (MTBE) pollution of urban groundwater. Journal of Contaminant Hydrology, 91, 128-145.

CL:AIRE (2014). An illustrated handbook of LNAPL transport and fate in the subsurface. CL:AIRE, London. ISBN 978-1-905046-24-9., pp.105. Download at www.claire.co.uk/LNAPL.

Davis, L. C. and Erickson, L. E. (2004). A review of bioremediation and natural attenuation of MTBE. Environmental Progress, 23, 243-252.

Dewsbury, P., Thornton, S.F. and Lerner, D.N. (2003). An improved analysis of MTBE, TAME and TBA in petroleum fuel-contaminated ground water by SPME using deuterated internal standards with GC-MS. Environmental Science and Technology, 37, 1392-1397.

Dubas, I.G., Brown, C.D. and Beulke, S. (2003). Sources of uncertainty in pesticide fate modelling. The Science of the Total Environment, 317, 53-72.

Einarson, M.D. and Cherry, J.A. (2002). A new multilevel ground water monitoring system using multichannel tubing. Ground Water Monitoring and Remediation, 22, 52-65.

Environment Agency (2000). Guidance on the assessment and monitoring of natural attenuation of contaminants in groundwater. R&D Publication 95, pp.143.

Environment Agency (2001a). Guide to good practice for the development of conceptual models and the selection and application of mathematical models of contaminant transport processes in the subsurface. National Groundwater and Contaminated Land Centre report NC/99/38/2, pp.121.

Environment Agency (2001b). Guidance on assigning values to uncertain parameters in subsurface contaminant fate and transport modelling. National Groundwater and Contaminated Land Centre report NC/99/38/3. pp.79.

Environment Agency (2002). The effects of contaminant concentration on the potential for natural attenuation. Environment Agency R&D Technical Report P2-228/TR. pp.166.

Essaid, H.I. and Hess, K.M. (1993). Monte Carlo simulations of multiphase flow incorporating spatial variability of hydraulic properties. Ground Water, 31, 123-134.

Fan, M., Thongsri, T., Axc, L. and Tyson, T.A. (2005). Using a probabilistic approach in an ecological risk assessment simulation tool: test case for depleted uranium (DU). Chemosphere, 60, 111-125.

Fiorenza, S. and Rifai, H.S. (2003). Review of MTBE biodegradation and bioremediation. Bioremediation Journal, 7, 1-35.

Fredrickson, J.K., McKinley, J.P., Bjornstad, B.N., Long, P.E., Ringelberg, D.B., White, D.C., Krumholz, L.R., Suflita, J.M., Colwell, F.S., Lehman, R.M., Phelps, T.J. and Onstott, T.C. (1997). Pore size constraints on the activity and survival of subsurface bacteria in a late cretaceous shale-sandstone sequence, northwestern New Mexico. Geomicrobiology Journal, 14, 183–202.

Gonzalez, D., Janardhanan, S., Pagendam, D.R. and Gladish, D.W. (2020). Probabilistic groundwater flow, particle tracking and uncertainty analysis for environmental receptor vulnerability assessment of a coal seam gas project. Water 2020, 12(11), 3177; https://doi.org/10.3390/w12113177.

Gwo, J-P., Jardine, P.M. and Sanford, W.E. (2005). Modelling field-scale multiple tracer injection at a low-level waste disposal site in fractured rocks: Effect of multiscale heterogeneity and source term uncertainty on conceptual understanding of mass transfer processes. Journal of Contaminant Hydrology, 77, 91-118.

Hardisty, P.E., Wheater, H.S., Johnston, P.M. and Bracken, R.A. (1998). Behaviour of light immiscible liquid contaminants in fractured aquifers. Géotechnique, 48, 747-760.

Hardisty, P.E., Wheater, H.S., Birks, D. and Dottridge, J. (2003). Characterisation of LNAPL in fractured rock. Quarterly J. of Engineering Geology and Hydrogeology, 36, 343-354.

Ho, C.K., Goering, T.J., Peace, J.L. and Miller, M.L. (2007). Probabilistic performance-assessment modeling of the mixed waste landfill at Sandia National Laboratories (2nd Edition). Sandia Report SAND2007-0170, pp.87.

Hosseini, A.M. (2009). Probabilistic modeling of natural attenuation of petroleum hydrocarbons. PhD thesis, Dept of Civil and Environmental Engineering, University of Alberta, Canada, pp. 359.

James, A.L. and Oldenburg, C.M (1997). Linear and Monte Carlo uncertainty analysis for subsurface contaminant transport simulations. Water Resources Research, 33, 2495-2508.

Jefferis, M., Hall, D., Hinchliff, J. and Aitken, M. (1993). Risk assessment: where are we, and where are we going ? In: Bentley, S.P (ed) Engineering Geology of Waste Disposal, Geol Soc. Eng. Geol. Special Publ., pp.241-359.

Jones, H.K. and Robins, N.S. (eds) (1999). The Chalk aquifer of the South Downs. Hydrogeological Report Series of the British Geological Survey.

King, M.W.G., Barker, J.F., Devlin, J.F. and Butler, B.J. (1999). Migration and natural fate of a coal tar creosote plume, 2. Mass balance and biodegradation indicators. J. Contaminant Hydrology, 39, 281-307.

Knopman, D.S. and Hollyday, E.F. (1993). Variation in specific capacity in fractured rocks, Pennsylvania. Ground Water. 31, 135-145.

La Cecilia, D., Porta, G.M., Tang, F.H.M., Riva, M. and Maggi, F. (2020). Probabilistic indicators for soil and groundwater contamination risk assessment. Ecological Indicators, 115, 1-13. 10.1016/j.ecolind.2020.106424.

Landmeyer, J.E., Bradley, P.M., Trego, D.A., Hale, K.G. and Haas, J.E. (2010). MTBE, TBA and TAME attenuation in diverse hyporheic zones. Ground Water, 48, 30-41.

Liang, H. (2009). Probabilistic remediation evaluation model for chlorinated solvents considering uncertainty. PhD thesis, Clemson University, pp.198.

Liang, H., Falta, R.W., Hendersen, J.K. and Shoemaker, S. (2011). Probabilistic simulation of remediation at a site contaminated by trichloroethylene. Groundwater Monitoring and Remediation. 32, 131-141. https://doi.org/10.1111/j.1745-6592.2011.01378.x

Liang, S.H., Kao, C.M., Kuo, Y.C., Chen, K.F. and Yang, B.M. (2011). In situ oxidation of petroleum-hydrocarbon contaminated groundwater using passive ISCO system, Water Research, 45, 2496-2506.

MacDonald, A.M. and Allen, D.J. (2001). Aquifer properties of the Chalk of England. Quarterly Journal of Engineering Geology & Hydrogeology, 34, 371-384.

McNab, W.W. and Dooher, B.P. (1998). Uncertainty analyses of fuel hydrocarbon biodegradation signatures in ground water by probabilistic modeling. Ground Water, 36, 691-698.

Mercer, J.W. and Cohen, R.M. (1990) A review of immiscible fluids in the subsurface: properties, models, characterization and remediation. Journal of Contaminant Hydrology, 6, 107-163. https://doi.org/10.1016/0169-7722(90)90043-G.

Middlemis, H. and Peeters, L. (2018). Uncertainty analysis—Guidance for groundwater modelling within a risk management framework. A report prepared for the Independent Expert Scientific Committee on Coal Seam Gas and Large Coal Mining Development through the Department of the Environment and Energy, Commonwealth of Australia. Canberra: Independent Expert Science Committee on Coal Seam Gas and Large Coal Mines; 2018. csiro:EP181142. http://hdl.handle.net/102.100.100/86118?index=1. pp 60.

Newell, C.J., Acree, S.D., Ross, R.R. and Huling, S.G. (1995). Groundwater Issue: Light Nonaqueous Phase Liquids. United States Environmental Protection Agency Office of Solid Waste and Emergency Response. EPA/540/S-95/500, pp.28.

O'Malley, D. and Vesselinov, V.V. (2014). A combined probabilistic/nonprobabilistic decision analysis for contaminant remediation. SIAM/ASA Journal on Uncertainty Quantification, 2, 10.1137/140965132.

Peckenham, J.M. (2010). The occurrence and persistence of MTBE in groundwater in Windham, Maine, USA. Groundwater Monitoring and Remediation, 30, 53-62.

Pena-Haro, S., Pulido-Velazquez, M. and Llopis-Albert, C. (2011). Stochastic hydro-economic modeling for optimal management of agricultural groundwater nitrate pollution under hydraulic conductivity uncertainty. Environmental Modelling and Software, 26, 999-1008.

Price, M. (1996). Introducing groundwater 2nd edition. Chapman & Hall, London.

Puls, R.W. and Paul, C.J. (1997). Multi-layer sampling in conventional monitoring wells for improved estimation of vertical contaminant distributions and mass. J. Contaminant Hydrology, 25, 85-111.

Ravi, V., Chen, J.S and Gierke, (1997). Evaluating the natural attenuation of transient-source compounds in groundwater. Proc. 4th International Symposium on In-situ and On-site Bioremediation, San Diego, USA, 205-211. Batelle Press.

Rebata-Landa, V. and Santamarina, J.C. (2006). Mechanical limits to microbial activity in deep sediments. Geochemistry, Geophysics and Geosystems, 7, Q11006. DOI: 10.1029/2006GC001355.

Reimann, C. and Filzmoser, P. (1999). Normal and lognormal data distribution in geochemistry: death of a myth. Consequences for the statistical treatment of geochemical and environmental data. Environmental Geology, 39, 1001-1014.

Rivett, M.O. and Thornton, S.F. (2008). Monitored natural attenuation of organic contaminants in groundwater: Principles and application. Water Management Journal, 161, 381-392. https://doi.org/10.1680/wama.2008.161.6.381.

Rivera-Velasquez, M.F., Fallico, C., Guerra, C. and Straface, S. (2013). A Comparison of deterministic and probabilistic approaches for assessing risks from contaminated aquifers: An Italian case study, Waste Management and Research, 31, https://doi.org/10.1177/0734242X135073.

Rodak, C. and Silliman, S. (2012). Probabilistic risk analysis and fault trees: Initial discussion of application to identification of risk at a wellhead. Advances in Water Resources, 36, 133-145.

Ronen, D. Berkowitz, B. and Magaritz, M. (1993). Vertical heterogeneity in horizontal components of specific discharge: a case study analysis. Ground Water, 31, 33-40.

Rosell, M, Lacorte, S. and Barceló D. (2006). Analysis, occurrence and fate of MTBE in the aquatic environment over the past decade. Trends in Analytical Chemistry, 25, 1016-1029.

Schmidt, T.C., Schirmer, M., Weiß, H. and Haderlein, S. B. (2004). Microbial degradation of methyl *tert*-butyl ether and *tert*-butyl alcohol in the subsurface. Journal of Contaminant Hydrology, 70, 173-203.

Schreiber, M.E. and Bahr, J.M. (1999) Spatial electron acceptor variability: Implications for assessing bioremediation potential. Bioremediation Journal, 3, 363-378.

Shah N.W., Thornton S.F., Bottrell, S.H. and Spence, M.J. (2009). Biodegradation potential of MTBE in a fractured chalk aquifer under aerobic conditions in long-term uncontaminated and contaminated aquifer microcosms. Journal of Contaminant Hydrology, 103, 119-133.

Snelling, J., Barnett, M.O., Zha, D. and Arey, J.S. (2007). Methyl-tert-hexyl ether and methyl-tert-octyl ether as gasoline oxygenates: Anticipating widespread risks to community water supply wells. Environmental Toxicology and Chemistry, 26, 2253-2259.

Soutter, M. and Musy, A. (1998). Coupling 1D Monte Carlo simulations and geostatistics to assess groundwater vulnerability to pesticide contamination on a regional scale. Journal of Contaminant Hydrology, 32, 25-39.

Spence, M.J., Bottrell, S.H., Thornton, S.F., Richnow, H. and Spence, K.H. (2005a). Hydrochemical and isotopic effects associated with fuel biodegradation pathways in a chalk aquifer. Journal of Contaminant Hydrology, 79, 67-88.

Spence, M.J., Thornton, S.F., Bottrell, S.H. and Spence, K.H. (2005b). Determination of interstitial water chemistry and porosity in consolidated aquifer materials by diffusion equilibrium-exchange as an alternative to centrifugation. Environmental Science and Technology, 39, 1158-1166.

Srivastava, A. and Singh, R.N. (2005). Use of multimedia mass balance model to predict concentrations of benzene in microenvironments in air. Environmental Modelling and Software, 20, 1-5.

Suarez, M.P. and Rifai, H.S. (1999). Biodegradation rates for fuel hydrocarbons and chlorinated solvents in groundwater. Bioremediation Journal, 3, 337-362.

Thornton, S.F., Lerner, D.N. and Banwart, S.A. (2001). Assessing the natural attenuation of organic contaminants in aquifers using plume-scale electron and carbon balances: model development with analysis of uncertainty and parameter sensitivity. Journal of Contaminant Hydrology, 53, 199-232.

Thornton, S.F., Bottrell, S.H. Pickup, R.P. Spence, M.J. and Spence, K.H. (2006). Processes controlling the natural attenuation of fuel hydrocarbons and MTBE in the UK Chalk aquifer, RP3, pp.96. Published by CL:AIRE (www.claire.co.uk).

Thornton, S.F. and Wealthall, G.P. (2008). Site characterization for improved assessment of contaminant fate in fractured aquifers. Water Management Journal, 161, 343-356.

Thornton, S.F., Bottrell, S.H., Spence, K.H., Pickup, R., Spence, M.J., Shah, N., Mallinson, H.E.H. and Richnow, H.H. (2011). Assessment of MTBE biodegradation in contaminated groundwater using ^{13}C and ^{14}C analysis: Field and laboratory microcosm studies. Applied Geochemistry, 26, 828-837.

Thornton, S.F., Tobin, K. and Smith, J.W.H (2013). Comparison of constant and transient-source zones on simulated contaminant plume evolution in groundwater: Implications for hydrogeological risk assessment. Groundwater Monitoring and Remediation, 33, 78-91.

Troldborg, M., Gagkas, Z., Vinten, A., Lilly, A. and Miriam Glendel, M. (2022). Probabilistic modelling of the inherent field-level pesticide pollution risk in a small drinking water catchment using spatial Bayesian belief networks. Hydrol. Earth Syst. Sci., 26, 1261–1293, https://doi.org/10.5194/hess-26-1261-2022.

Uhlenbrook, S. and Sieber, A. (2005). On the value of experimental data to reduce the prediction uncertainty of a process-oriented catchment model. Environmental Modelling and Software, 20, 19-32.

USEPA (1999). Reliability-based uncertainty analysis of groundwater contamination and remediation. Office of Research and Development, Washington DC 20460, EPA/6000/R-99/028., pp.82.

van Afferden, M., Rahman, K.Z., Mosig, P., De Biase, C., Thullner, M., Oswald, S. E. and Mueller, R. A. (2011). Remediation of groundwater contaminated with MTBE and benzene: The potential of vertical-flow soil filter systems. Water Research, 45, 5063-5074.

van Reeuwijk, L.P. (1987). Procedures for soil analysis. Technical Paper no. 9. International Soil Reference and Information Centre, Wageningen, The Netherlands.

Wang, F. and Bright, J. (2004). Scale effect and calibration of contaminant transport models. Ground Water, 42, 760-766.

Wang, X.L., Tao, S., Dawson, R.W. and Wang, X.J. (2004). Uncertainty analysis of parameters for modelling the transfer and fate of benzo(a)pyrene in Tianjin wastewater-irrigated areas. Chemosphere, 55, 525-531.

Wealthall, G.P., Thornton, S.F. and Lerner, D.N. (2002). Assessing the transport and fate of MTBE-amended petroleum hydrocarbons in the UK Chalk aquifer. In GQ2001: Natural and Enhanced Restoration of Groundwater Pollution, Sheffield, U.K., 16-21 June 2001. (eds, Thornton, S. F and Oswald, S. O.), IAHS Publ. No. 275, 205-212.

Whelan, G., McDonald, J.P., Taira, R.Y., Gnanapragasam, E.K., Yu, C., Lew, C.S. and Mills, W.B. (2000). Source-term development for a contaminant plume for use by multimedia risk assessment models. Journal of Contaminant Hydrology, 41, 205-223.

Wiedemeier, T.H., Downey, D.C., Wilson. J.T., Kampbell. D.H., Miller. R.N. and Hansen, J.E. (1995). Technical protocol for implementing the intrinsic remediation (natural attenuation) with long term monitoring option for dissolved-phase fuel contamination in groundwater. Air Force Center for Excellence, Brooks Air Force Base, USA.

Wiedemeier, T.H., Rifai, H.S., Newell, C.J. and Wilson, J.T. (1999). Natural attenuation of fuels and chlorinated solvents in the subsurface. John Wiley & Sons, pp. 617.

Woodbury, A., Render, F. and Ulrich, T. (1995). Practical probabilistic ground-water modeling. Groundwater, 33, 532-538. https://doi.org/10.1111/j.1745-6584.1995.tb00307.x.

Winter, C.L. and Tartakovsky, D.M. (2008). A reduced complexity model for probabilistic risk assessment of groundwater contamination. Water Resources Research, 44, W06501, doi:10.1029/2007WR006599.

Wilson, R.D., Thornton, S.F. and Mackay, D.M. (2004). Challenges in monitoring the natural attenuation of spatially variable plumes. Biodegradation, 15, 359-369.

Xie, Y., Li, X., Liu, X., Du, X., Li, F. and Cao, Y. (2016). Performance evaluation of remediation scenarios for DNAPL contaminated groundwater using analytical models and probabilistic methods. Procedia Environmental Sciences, 31, 264-273.

Yang, A., Yang, Q., Fan, Y., Suo, M. Fu, H., Liu, J. and Lin, X. (2018). An integrated simulation, inference and optimization approach for groundwater remediation with two-stage health-risk assessment. Water, 10, 694; doi:10.3390/w10060694.

Yenigul, N.B., Elfeki, M.M.A., Gehrels, J.C., van den Akker, C., Hensbergen, A.T. and Dekking, F.M. (2005). Reliability assessment of groundwater monitoring networks at landfill sites. Journal of Hydrology, 308, 1-17.

Younger, P.L. and Elliot, T. (1995). Chalk fracture system characteristics: Implications for flow and solute transport. Quarterly Journal of Engineering Geology, 28 (Special Issue): S39-S50. 1995. ISSN: 0481-2085, http://dx.doi.org/10.1144/GSL.QJEGH.1995.028.S1.04.

Zhang, D., Shi, L., Chang, H. and Yang, J. (2010). A comparative study of numerical approaches to risk assessment of contaminant transport. Stochastic Environmental Research and Risk Assessment, 24, 971-984.

Zhang, P. (2010). Probabilistic methods used in environmental risk evaluation for groundwater protection. PhD thesis, Dept of Geosciences, University of Oslo, pp.86.

Zhang, P., Aagaard, P. and Gottschalk, L. (2010). Probability method used in predicting contaminant risk in groundwater adjacent to airport. Water, Air and Soil Pollution, 211, 323-339.
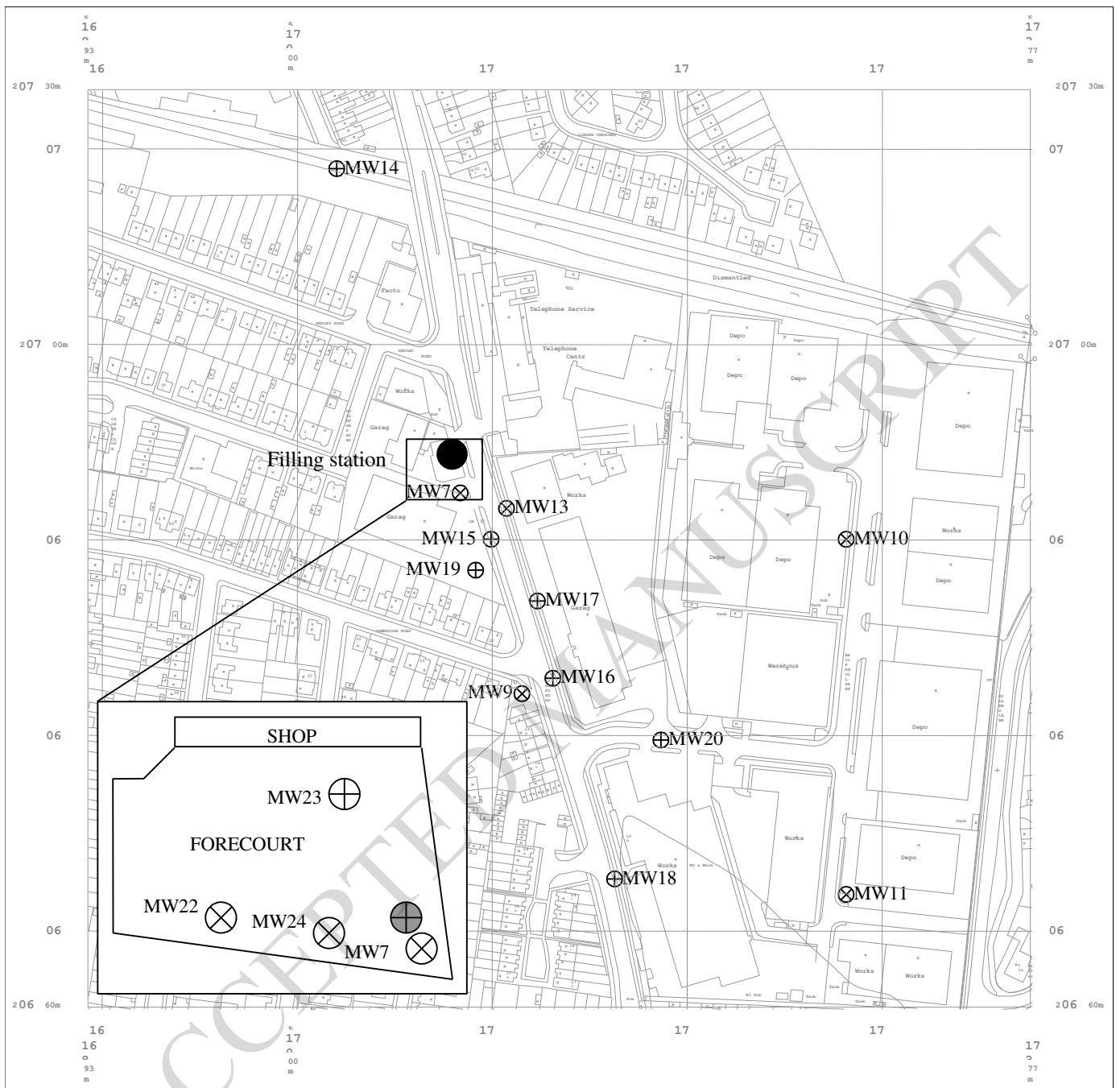
Figure 1. Location map for monitoring wells installed at the study site (see main text for description of network and function of each well type).
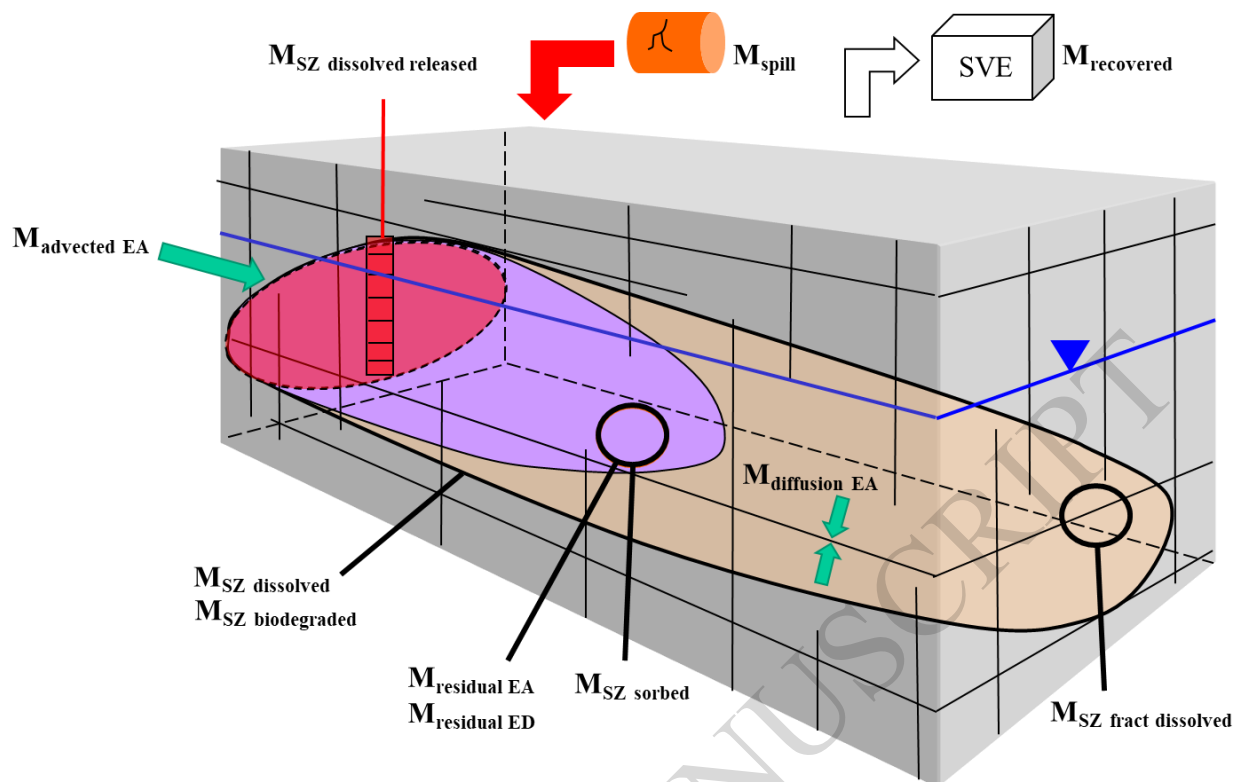
Figure 2. Graphical conceptual model of contaminant fate in a dual-porosity aquifer system that underpins the calculations in this study.
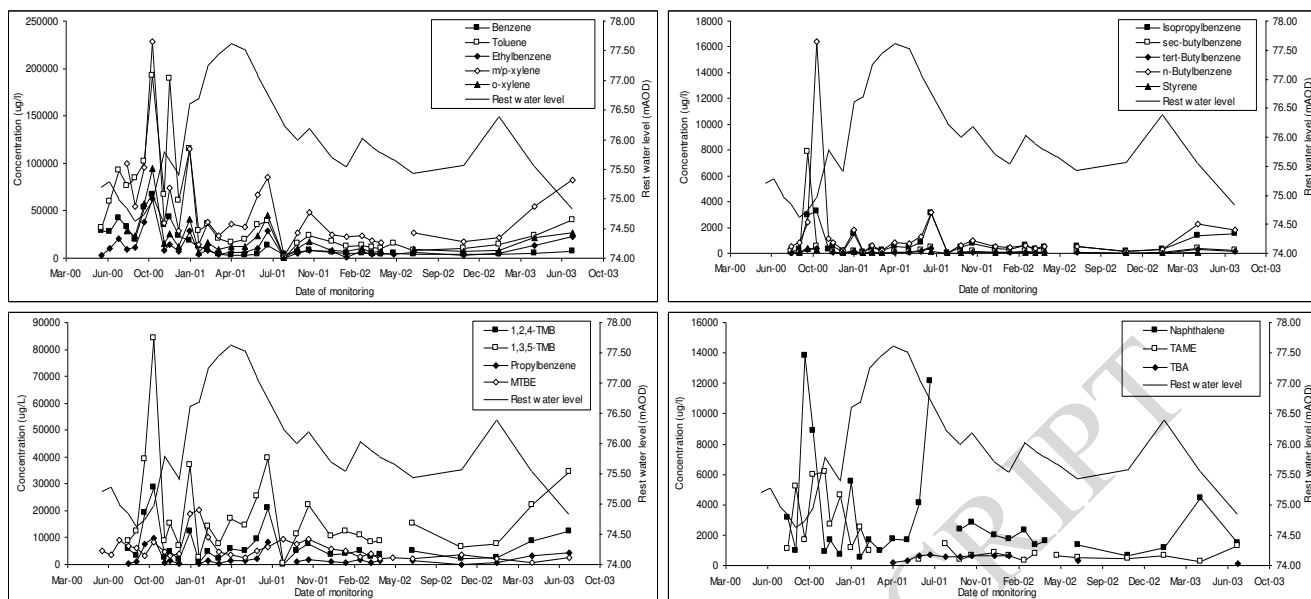
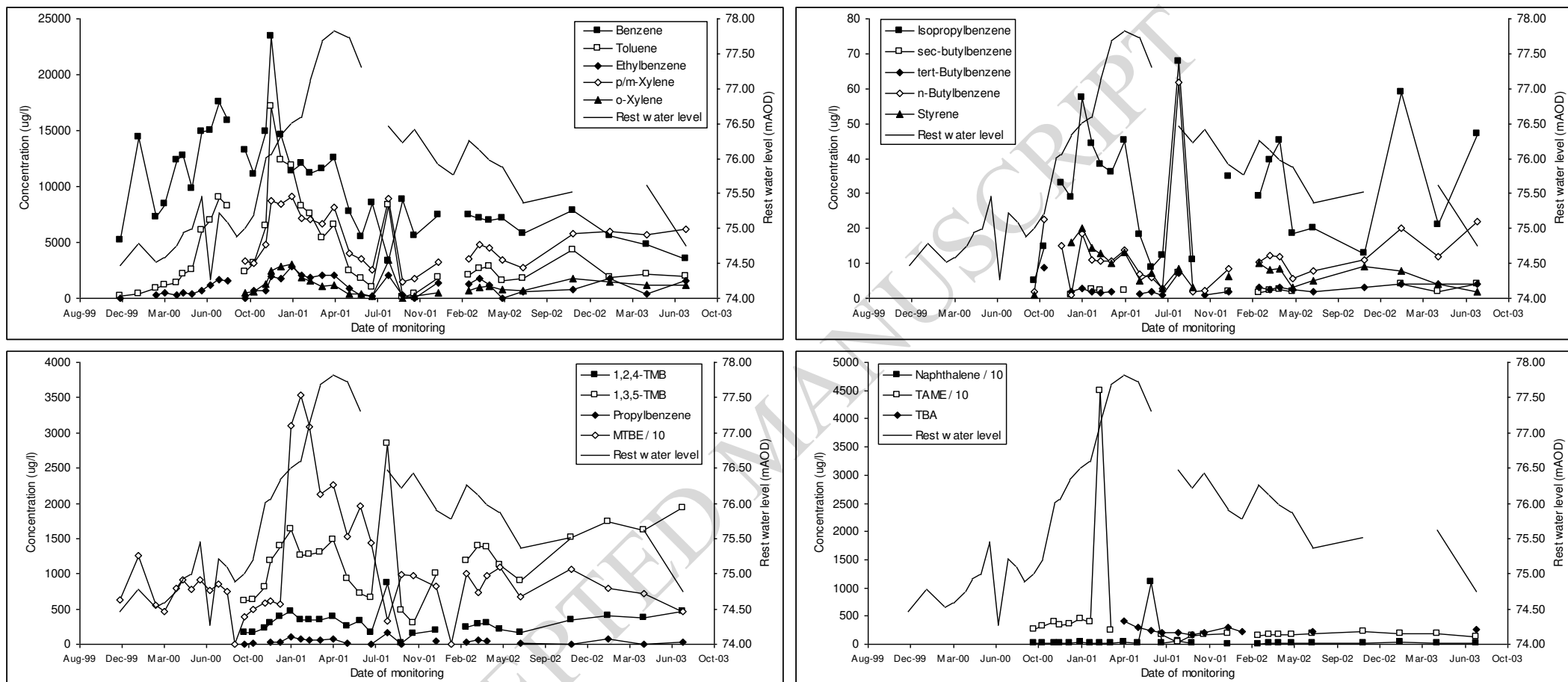Figure 3. Variation in concentration of dissolved hydrocarbon compounds sampled in MW7

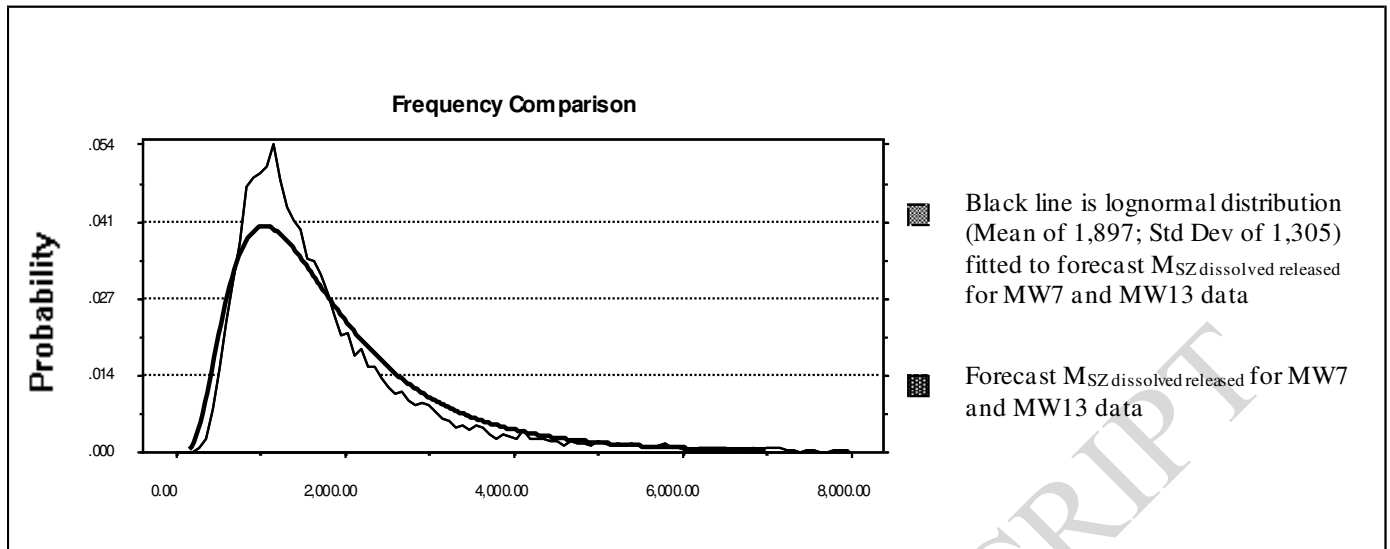Figure 4. Variation in concentration of dissolved hydrocarbon compounds sampled in MW13

Figure 5. Overlay of statistical distribution fitted by Crystal Ball to forecast of total dissolved contaminant mass released ($M_{SZ\ dissolved\ released}$) and the same forecast predicted by the Monte Carlo simulation. Horizontal scale has units of kg.

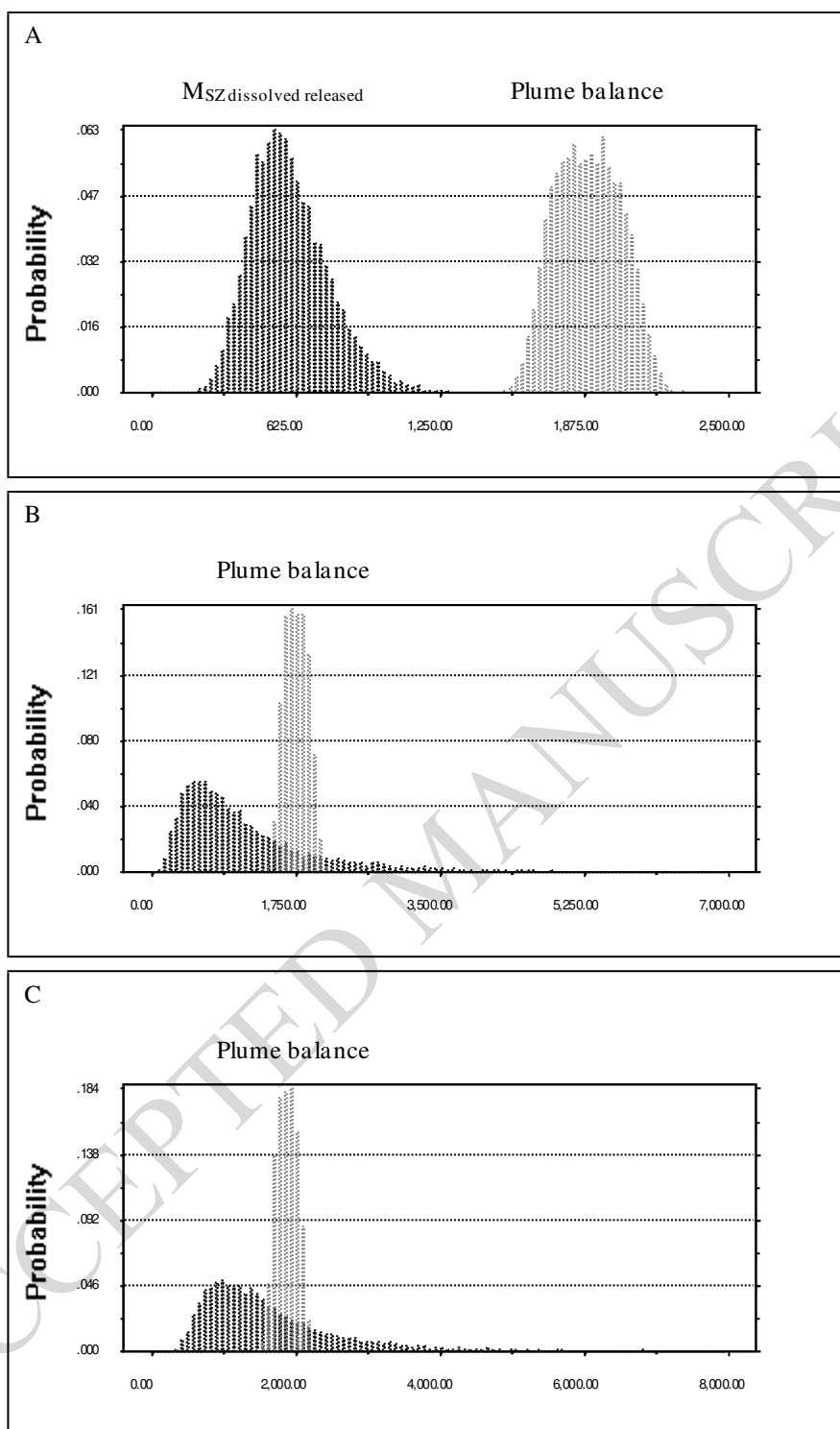Figure 6. Overlay of Monte Carlo forecast of total dissolved contaminant mass released into saturated zone ($M_{SZ\,dissolved\,released}$) and plume balance for input data from a) scenario using MW13 data only, b) scenario using combined MW7 and MW13 data with plume balance for BTEX-only hydrocarbons, c) scenario using combined MW7 and MW13 data with plume balance for all contaminants. Horizontal scale has units of kg.

Table 1. Summary of input parameters for mass balance

Parameter	Data range	Data source
Plume dimensions		
Plume width (m)	33.75	Maximum distance between MW13 and MW19, normal to the plume flow path
Plume thickness (m)	15.35	Vertical distance of saturated zone contamination identified from MLS profile in MW15
Plume source concentrations		
MW7-MW13		
Benzene, C_{source} (μg/l)	3,198 - 67,133	Monthly data from December 1999-July 2003 for both monitoring wells
Toluene, C_{source} (μg/l)	97 - 192,586	"
Ethylbenzene, C_{source} (μg/l)	3 - 62,261	"
m/p-Xylene, C_{source} (μg/l)	1,487 - 228,293	"
o-Xylene, C_{source} (μg/l)	23 - 94,929	"
Styrene, C_{source} (μg/l)	1-407	"
Isopropylbenzene, C_{source} (μg/l)	4 – 3,248	"
Propylbenzene, C_{source} (μg/l$^)$	1-9,801	"
1,2,4-TMB, C_{source} (μg/l)	20 – 28,771	"
1,3,5-TMB, C_{source} (μg/l)	301 – 84,027	"
sec-Butylbenzene, C_{source} (μg/l)	1 – 7,837	"
tert-Butylbenzene, C_{source} (μg/l)	1 – 382	"
n-Butylbenzene, C_{source} (μg/l)	1 – 16,445	"
Naphthalene, C_{source} (μg/l)	16 – 13,789	"
MTBE, C_{source} (μg/l)	829 – 35,365	"
TAME, C_{source} (μg/l)	270 – 44,935	"
TBA, C_{source} (μg/l)	156 – 736	"
Background groundwater		
O_2, $C_{background\ EA}$ (mg/l)	0 - 7.3	Vertical profile from 2 monthly surveys and 3 annual surveys between 2000 – 2003 for MW14
NO_3, $C_{background\ EA}$ (mg/l)	58.7 – 109.3	"
SO_4, $C_{background\ EA}$ (mg/l)	38.2 - 153	"
Aquifer hydrogeology		
Hydraulic conductivity, K_x (m/d)	0.04 – 75.69	Vertical profile obtained from straddle packer pumping tests of MW15
Hydraulic gradient, I_x	0.0021 – 0.0031	Groundwater elevation for all MW on 4 surveys between Nov 2000 – Feb 2001
Matrix porosity, ε_m	0.37 – 0.51	Vertical profile from analysis of rock core from MW23 over plume interval
Fracture porosity, ε_f	0.011	Taken from Price (1996)
Aquifer properties		
Organic carbon content (wt %)	0.002 – 0.126	Vertical profile from analysis of uncontaminated rock core from MW20
Bulk density, ρ_b (g/ml)	1.3 - 1.67	Estimated using formula and assumed particle density for chalk

Table 2. Summary of mass balance

Forecast[1]	Input No.	Scenarios[2,3]		
		MW7 only	MW13 only	MW7 & MW13
Plume source				
$M_{SZ\ dissolved\ released}$				
All contaminants	1	4,142 (1,351-575,963)	602 (362-1,099)	1,796 (643-7,993)
BTEX only	2	3,073 (813-498,982)	339 (180-648)	1,280 (322-6,979)
Plume contaminant residuals				
$M_{SZ\ dissolved}$				
All contaminants	3	228 (180-252)	228 (180-252)	228 (180-252)
BTEX only	4	110 (89-123)	110 (89-123)	110 (89-123)
$M_{SZ\ fract\ dissolved}$				
All contaminants	5	5.55 (5.25-5.87)	5.55 (5.25-5.87)	5.55 (5.25-5.87)
BTEX only	6	2.68 (2.5-2.9)	2.68 (2.5-2.9)	2.68 (2.5-2.9)
$M_{SZ\ sorbed}$				
All contaminants	7	98 (2.5-138)	98 (2.5-138)	98 (2.5-138)
BTEX only	8	91.8 (20-138)	91.8 (20-138)	91.8 (20-138)
Total all contaminants (3+5+7)	9	332 (225-375)	332 (225-375)	332 (225-375)
Total BTEX only (4+6+8)	10	205 (125-245)	205 (125-245)	205 (125-245)
Electron acceptor consumption[4]				
$M_{SZ\ biodegraded}{}^{5}$	11	1,595 (1,338-1,827)	1,595 (1,338-1,827)	1,595 (1,338-1,827)
Plume balance				
All contaminant residual and biodegraded (9+11)	12	1,927 (1,647-2,169)	1,927 (1,647-2,169)	1,927 (1,647-2,169)
BTEX residual and biodegraded (10+11)	13	1,800 (1,525-2,036)	1,800 (1,525-2,036)	1,800 (1,525-2,036)

Notes

1. All values in kg

2. Monte Carlo simulations include **three scenarios** for prediction of $M_{SZ\ dissolved\ released}$, using contaminant concentration data from MW7 only, MW13 only, or MW7 and MW13 in combination (see text for discussion). However, forecast predictions for **all** other inputs are identical for each scenario.

3. Forecasts for Monte Carlo simulations are mean values with range at 95% confidence level in brackets

4. Total electron acceptor consumption determined from sum of $M_{advected\ EA}$, $M_{residual\ EA}$ and $M_{diffusion\ EA}$ for O_2, NO_3, SO_4, Mn(II), Fe(II) and CH_4

5. Equivalent BTEX mass biodegraded for estimated electron acceptor consumption, obtained using conversion factors for reaction stoichiometry (Wiedemeier et al, 1999).

Table 3. Electron acceptor (EA) consumption estimated from biodegradation ($M_{SZ\ biodegraded}$) for different mass inputs.

Process descriptor[1]	Advection[2] ($M_{advected\ EA}$)	Matrix diffusion[2] ($M_{diffusion\ EA}$)	Plume residual[2] ($M_{residual\ EA}$)	Total EA consumption[2,3]	Equivalent BTEX biodegraded ($M_{SZ\ biodegraded}$)[2,4]
O_2	146	27		173	55.4
NO_3	1,716	1,219		2,935	587
SO_4	3,570	1,715	767	4518	949
Mn			10	10	0.92
Fe			44	44	1.98
CH_4			0.9	0.9	1.14
Total	**5,432**	**2,961**	**822**	**7,681**	**1,595**

Notes

1. O_2: aerobic respiration; NO_3: denitrification; SO_4: SO_4-reduction; Mn: Mn(IV) oxide-reduction; Fe: Fe(III) oxide-reduction; CH_4: methanogenesis
2. All values in kg
3. Total electron acceptor consumption obtained by summation of $M_{advected\ EA}$, $M_{diffusion\ EA}$ and $M_{residual\ EA}$, except for SO_4-reduction which is given by ($M_{advected\ EA} + M_{diffusion\ EA} - M_{residual\ EA}$)
4. Equivalent BTEX hydrocarbon mass biodegraded for estimated electron acceptor consumption, obtained using values of $M_{SZ\ biodegraded}$ and conversion factors for reaction stoichiometry (Wiedemeier et al, 1999).