# Transient Variability In Vapor Intrusion And The Factors That Influence It

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Abstract

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## 4 Introduction

- 5 Long term vapor intrusion (VI) studies in both residential and larger commercial struc-
- 6 tures have raised concerns regarding significant observed transient behavior in indoor air
- <sup>7</sup> contaminant concentrations. <sup>1–7</sup> VI involves the migration of volatilizing contaminants from
- 8 soil, groundwater or other subsurface sources into overlying structures. VI has been a rec-
- 9 ognized problem for some time, but many aspects remain poorly understood, particularly
- with respect to the causes of large temporal transients in indoor air concentrations. There
- is uncertainty within the VI community regarding how to best develop sampling strategies
- to address this problem. 1,3,8
- Results from a house operated by Arizona State University (ASU) near Hill AFB in
- Utah, an EPA experimental house in Indianapolis, IN and a large warehouse at the Naval
- Air Station (NAS) North Island, CA have all shown significant transient variations in indoor

air contaminant concentrations. All were outfitted with sampling and monitoring equipment that allowed tracking temporal variation in indoor air contaminant concentrations on time scales of hours. All have shown that these concentrations varied significantly with time orders of magnitude on the timescale of a day or days. <sup>5,9,10</sup>

In one instance the source of the variation was clearly established during the study; at
the ASU house a field drain pipe (or "land drain"), which connected to a sewer system, was
discovered beneath the house, and careful isolation of this source led to a clear conclusion that
this preferential pathway significantly contributed to observed indoor air contaminant levels
and their fluctuations. <sup>10,11</sup> While in this case the issue of a contribution from a preferential
pathway was clearly resolved, what it left open was a question of whether existence of such
a preferential pathway to an area beneath a structure would always be expected to lead to
large fluctuations in indoor air contaminant concentrations.

Similarly, a sewer pipe has recently been suggested to be a source of the contaminants 28 found in the EPA Indianapolis house. That site was also characterized by large indoor air 29 contaminant concentration fluctuations.  $^{7,12}$  Sewer lines have been generally implicated as VI sources at several sites. $^{12-15}$  A Danish study estimate that roughly 20% of all VI sites in 31 central Denmark involve significant sewer VI pathways. 16 Thus while the consideration of a role of possible sewer or other preferential pathways is now part of normal good practice in VI site investigation, 1 it is still not known whether the existence of such pathways automatically means that large temporal fluctuations are necessarily to be expected. In some of these cited 35 cases, <sup>13,15</sup> a sewer provided a pathway for direct entry of contaminant into the living space. 36 While potentially important in many cases, this scenario is not further considered here, 37 where the focus is on pathways that deliver contaminant via the soil beneath a structure. 38

It is, however, now known that even absent a preferential pathway, there may be significant transient variation in indoor air contaminant concentrations at VI sites. <sup>2,4,17</sup> One example is a site at NAS North Island at which no preferential pathways have been identified. Instead, a building at this site is characterized by significant temporal variations in indoor-outdoor pressure differential.<sup>5</sup> It is believed that this is the origin of the observed indoor air contaminant concentration fluctuations at that site.

This paper investigates the sources of the temporal variation in indoor air contaminant concentrations in both the presence and absence of preferential pathways. In this work, the latter scenarios are referred to as "normal" VI scenarios, in which there is typically a groundwater source of the contaminant. Specifically, we pose the question of just how much variation in indoor air contaminant concentration may be expected at such normal VI sites vs. those characterized by preferential pathways. The conditions required for preferential pathways to become significant contributors to temporal variations in indoor air contaminant concentrations are also explored, and the consequences for sampling strategies are also discussed.

## Methods

## 55 Statistical Analysis Of Field Data

To begin to characterize transient behavior in indoor air contaminant concentrations, actual datasets are analyzed to establish common levels of variability at VI sites. For this purpose, the datasets from the ASU house in Utah, the EPA Indianapolis site and North Island NAS were chosen for analysis. This paper relies on statistical analysis of published field data, and readers are referred to the original works for details regarding data acquisition. <sup>3,5,7,9,10</sup>

The ASU house data were obtained over a period of a few years. During part of this time, controlled pressure method (CPM) tests were being conducted, in which the house was underpressurized to an extent greater than that characterizing "normal" operation. This caused greater than normal advective flow from the subsurface into the house, thus increasing VI potential. <sup>6,9,18</sup> This period of CPM testing is considered separately from the otherwise "natural" VI conditions in the analysis. Likewise, the existence of a preferential pathway at the ASU house needs to be considered in examining the dataset, noting that

during some of the testing, this pathway was deliberately cut off, resulting in what we have termed "normal" VI conditions in which the main source of contaminant was believed to be groundwater.

The NAS North Island dataset has not (as far as is known) been influenced by a preferential pathway, but the structure there was subject to large internal pressure fluctuations, much more extensive than those typically recorded at the ASU house during normal operations. Additionally, the underlying soil at NAS North Island is sandy and more permeable than that at the ASU site, which, as will be shown, contributes to the indoor air contaminant concentrations being more sensitive to pressure fluctuations.<sup>5</sup>

Likewise, the Indianapolis site investigation spanned a number of years and periodically included the testing of a sub-slab depressurziation system (SSD). The goal of the SSD testing was to mitigate the VI risk by drastically depressurizing the sub-slab area underneath the house, preventing the contaminants from entering the structure above. Only the period before the installation of this system was considered in the present analysis. It is likely a sewer line beneath the structure acted as a preferential pathway, 12 however at no point was this preferential pathway removed, making it difficult to assess how significant the role of the preferential pathway was at this site. Regardless of this it is of interest to consider the data from this site due to how extensive and complete the data collection was.

The typical variation in indoor air contaminant concentrations with time will first be considered below in the case of the ASU house during "natural", (i.e. non-CPM conditions), in the case of the NAS North Island site over the entire available dataset, and for the Indianapolis case we consider the variations before the installation of the SSD system. The deviations in indoor air concentration from the mean TCE (and Chloroform and PCE at the Indianapolis site) values, as well as the indoor-outdoor pressure differentials associated with these concentrations were examined. Both univariate and bivariate kernel density estimations (KDE) were constructed. KDE is a technique that estimates the probability distribution of a random variable(s) by using multiple kernels, or weighting functions, and

in this case, Gaussian kernels are used to create the KDEs. This means that it is presumed that the variables of interest (i.e., indoor air contaminant concentrations and indoor-outdoor pressure differentials, as sampled) are normally distributed around mean values (and that there are statistical fluctuations associated with each sampling event). In this instance, the scipy statistical package was used to construct the KDEs, assuming a bandwidth parameter determined by Scott's rule. The distributions of the individual parameters and the relationship between them will be examined using the KDE method.

## $_{\scriptscriptstyle 2}$ Modeling Work

In addition to examining the actual field data, a previously described three -dimensional computational fluid dynamics model of a generic VI impacted house was used to elucidate certain aspects of the processes. This model was implemented in a finite element solver package, COMSOL Multiphysics. In the present work, there has been an addition of a preferential pathway to the "standard" model that has been described before in publications by this group. <sup>19–21</sup> As in the earlier studies, only the vadose zone soil domain is directly modeled.

The modeled structure is assumed to have a 10x10 m foundation footprint, with the 110 bottom of the foundation slab lying 1 m below ground surface (bgs), simulating a house 111 with a basement. The indoor air space is modeled as a continuously stirred tank (CST)<sup>1</sup> 112 and all of the contaminant entering the house is assumed to enter with soil gas through a 113 1 cm wide crack located between the foundation walls and the foundation slab around the 114 perimeter of the house. All of the contaminant leaving the indoor air space is assumed to 115 do so via air exchange with the ambient. The indoor control volume is assumed to consist 116 of only of the basement, assumed as having a total volume of 300 m<sup>3</sup>. Clearly different 117 assumptions could be made regarding the structural features and the size of the crack entry 118 route, but for present purposes, this is unimportant as the intent is only to show for "typical" values what the influence of certain other features can be. 120

The modeled surrounding soil domain extends 5 meters from the perimeter of the house, and is assumed to consist of sandy clay (except as noted). Directly beneath the foundation slab, there is assumed to be a 30 cm (one foot) thick gravel layer, except in certain cases where this sub-base material is assumed to be the same as the surrounding soil (termed a "uniform" soil scenario).

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Where relevant The preferential pathway is modeled as a 10 cm (4") pipe that exits into the gravel sub-base beneath the structure. The air in the pipe is assumed to be contaminated with TCE at a vapor concentration equal to the vapor in equilibrium with the groundwater contaminant concentration below the structure, modified by a scaling factor  $\chi$ , allowing the contaminant concentration in the pipe to be parameterized.

The groundwater beneath the structure is assumed to be homogeneously contaminated with trichloroethylene (TCE) as a prototypical contaminant. The groundwater itself is not modeled, as the bottom of the model domain is defined by the top of the water table. The ground surface and the pipe are both assumed to be sources of air to the soil domain. Both are assumed to be at reference atmospheric pressure.

Vapor transport in the soil is governed by Richard's equation, a modified version of
Darcy's Law, taking the variability of soil moisture in the vadose zone into account. <sup>22</sup> The
van Genuchten equations are used to predict the soil moisture content and thus the effective
permeability of the soil. <sup>23</sup> The effective diffusivity of contaminant in soil is calculated using
the Millington-Quirk model. <sup>24</sup> The transport of vapor contaminant in the soil is assumed
to be governed by the advection-diffusion equation, in which either advection or diffusion
may dominate depending upon position and particular circumstances. The equations and
the boundary conditions are given in Table 1.

Table 1: Governing equations, boundary conditions & model input parameters. (See below for table of nomenclature).

#### (a) Governing equations

Unsteady-CSTR	$V_{\frac{du}{dt}} = \int_{A_{ck}} j_{ck} dA - u A_e V$
Richard's equation	$\nabla \cdot \rho \left( -\frac{\kappa_s}{\mu} k_r \nabla p \right) = 0$
Millington-Quirk	$D_{ ext{eff}} = D_{ ext{air}} rac{ heta_g^{10/3}}{ heta_t^2} + rac{D_{ ext{water}}}{K_H} rac{ heta_w^{10/3}}{ heta_t^2}$
Advection-diffusion equation	$\frac{\partial}{\partial t} \Big( \theta_w c_w + \theta_g c \Big) = \nabla (D_{\text{eff}} \cdot \nabla c) - \vec{u} \cdot \nabla c$
van Genuchten equations	$Se = \frac{\theta_w - \theta_r}{\theta_t - \theta_r} = [1 +  \alpha z ^n]^{-m}$ $\theta_g = \theta_t - \theta_w$ $k_r = (1 - Se)^l [1 - (Se^{-m})^m]^2$ $m = 1 - 1/n$

## (b) Boundary conditions

Boundary	Richard's equation	Advection-diffusion equation
At foundation crack	$p = p_{\rm in/out} (Pa)$	$j_{\rm ck} = \frac{uc}{1 - \exp\left(uL_{\rm slab}/D_{\rm air}\right)}$
At groundwater source	N/A	$c = c_{\rm gw} K_H \left( \mu g / m^3 \right)$
At ground surface	p = 0  (Pa)	$c = 0 \; (\mu \mathrm{g/m^3})$
Exit of preferential pathway	p = 0  (Pa)	$c = c_{\rm gw} K_H \chi \; (\mu g/m^3)$

## (c) Soil & gravel properties $^{25-27}$

Soil	Permeability (m <sup>2</sup> )	Density (kg/m <sup>3</sup> )	$\theta_s$	$\theta_r$	$\alpha (1/m)$	n
Gravel	$1.3 \cdot 10^{-9}$	1680	0.42	0.005	100	3.1
Sand	$9.9 \cdot 10^{-12}$	1430	0.38	0.053	3.5	3.2
Sandy Clay	$1.7 \cdot 10^{-14}$	1470	0.39	0.12	3.3	1.2

## (d) Trichloroethylene (diluted in air) properties $^{26,27}$

$D_{\rm air}~({ m m}^2/{ m h})$	$D_{\rm water}~({\rm m}^2/{\rm h})$	Density $(kg/m^3)$	Viscosity (Pa·s)	$K_H$	M (g/mol)
$2.47 \cdot 10^{-2}$	$3.67 \cdot 10^{-6}$	1.614	$1.86 \cdot 10^{-5}$	0.403	131.39

#### (e) Building properties

$V_{\text{base}} (\text{m}^3)$	$L_{\rm slab}$ (cm)	$A_e (1/hr)$
300	15	0.5

# Results & Discussion

# 45 Statistical Analysis of Field Data

The pressure difference between the indoor and outdoor/ambient  $(p_{\rm in/out})$  is an important 7

driving force in VI, drawing in (or preventing) contaminants from entering a structure.

Figure 1: Distributions of indoor/outdoor pressure differences (top row), and its correlation with indoor air contaminant concentrations (middle rows), and air exchange rate (bottom row), at three different VI sites - ASU house, Indianapolis site, and North Island NAS.

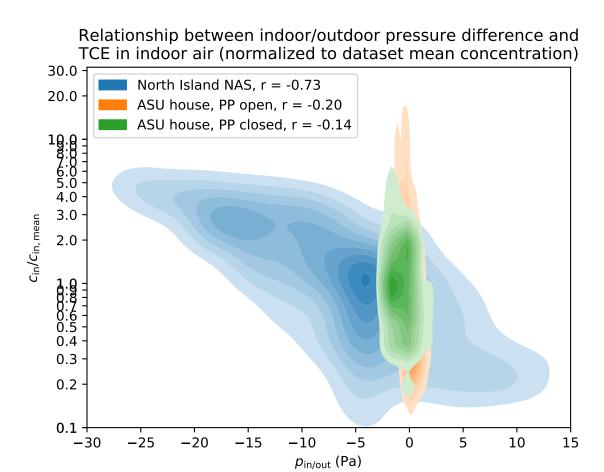


Table 2: 5th and 95th percentile values of  $p_{\rm in/out}$  and  $c_{\rm in}/c_{\rm in,mean}$  in Figure 1.

	North	Island NAS	ASU	house PP Open	ASU	House PP Closed
Percentile	$5 ext{th}$	$95 \mathrm{th}$	5th	$95 \mathrm{th}$	5th	$95  ext{th}$
$p_{\rm in/out}$ (Pa)	-19.9	7.4	-1.4	2.1	-2.1	2.27
$c_{\rm in}/c_{\rm in,mean}$	4.1	0.2	13.5	0.2	3.3	0.4

Changes in  $p_{\rm in/out}$  is also a dynamic and fast process, impacting VI more rapidly than e.g. fluctuations in groundwater depth or contaminant concentration does; these processes may take weeks or even months to impact the overlying structure. Therefore it is reasonable to assume that the temporal variability in indoor air contaminant concentration  $c_{\rm in}$  may be driven by changes in  $p_{\rm in/out}$ .

We examine the relationship between  $p_{\rm in/out}$  and  $c_{\rm in}$  by constructing the two-dimensional 153 kernel density estimation (KDE) plots seen in Figure 1. The KDE plot allows us to view 154 the distribution of  $p_{\text{in/out}}$  and  $c_{\text{in}}$ , and how well these correlate. For this analysis we consider 155 two VI sites, North Island NAS, and the ASU house, with the ASU house dataset divided 156 up into two periods, before and after the land drain (called preferential pathway (PP) from 157 here on) had been closed. The preferential pathway significantly impacted the ASU house 158 to such an extent that these two periods can essentially be considered as two different VI 159 sites, and allowing us to examine the effect it had. 160

In Figure 1  $c_{\rm in}$  is normalized to the mean  $c_{\rm in,mean}$  of each dataset, allowing us to compare the impact  $p_{\rm in/out}$  had on  $c_{\rm in}$ . A value of 10 on the y-axis indicate that  $c_{\rm in}$  is 10 times greater than the mean here, and 0.1 indicate that it is one tenth of the mean. The Pearson's r between  $p_{\rm in/out}$  and  $c_{\rm in}$  for each dataset is shown in the legend.

An interesting aspect of the North Island NAS site is that  $p_{\rm in/out}$  varies so significantly here, with the 5th and 9th percentile  $p_{\rm in/out}$  -19.9 and 7.4 Pa respectively. This may be contrasted with 5th and 95th percentile  $p_{\rm in/out}$  at the ASU house: -1.4 and 2.1 Pa (PP open), and -2.1 and 2.27 Pa (PP closed).

The large (roughly one order of magnitude) under- and overpressurization of the North Island NAS site leads to roughly a one order magnitude increase and decrease from the mean  $c_{\rm in}$ . Combined with r = -0.73, it is quite clear that at this site  $p_{\rm in/out}$  largely determines  $c_{\rm in}$  (there is still variability in  $c_{\rm in}$  for a given  $p_{\rm in/out}$ , which we address later). This is the same principle that governs the controlled pressure method (CPM) concept.

Turning to the ASU house datasets, we see a quite different situation. Here the variability of  $c_{\rm in}$  is just as large, or even larger than at North Island NAS, yet the  $p_{\rm in/out}$  varies far less. At first glance it may seem like the  $c_{\rm in}$  values for the periods when the PP is open and closed respectively are relatively comparable, but the 5th and 95th percentiles values of differ significantly  $c_{\rm in}/c_{\rm in,mean}$  as may be seen in Table 2.

It is clear that the PP dramatically increases the house's sensitivity to  $p_{\rm in/out}$ , and the

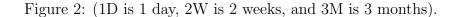
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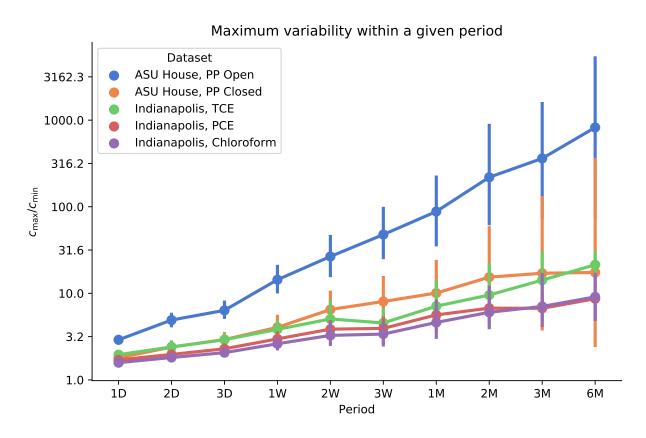
site is fundamentally different during the period when the PP was open. Yet, the magnitude of change in  $c_{\rm in}$  is far greater than the change in  $p_{\rm in/out}$ , leading to the conclusion that it is 181 not just increased flow rates into the structure that causes this, but there must also be an 182 increased amount of contaminant as well as a much larger spatial variability in the sub-base. 183 This is a topic that we will return to later. 184

The magnitude of change in  $c_{in}$  when the PP is closed is more in line with the magnitude 185 of change in  $p_{in/out}$ , but there is still more variability than one would expect. We hypothesize 186 that this is largely due to fluctuations in air exchange rate, which we will examine later.

## Change In Indoor Air Contaminant Concentration Over Time

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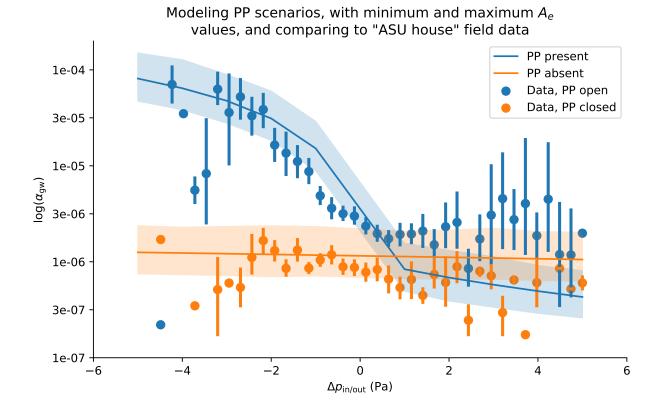


So far we have discussed how the relationship between  $c_{\rm in}$  and  $p_{\rm in/out}$ , and how signifi-189 cantly  $c_{\rm in}$  may vary at a site. But this tells us little about how much variability in  $c_{\rm in}$  may be expected over time. For this analysis we turn to the ASU house data (PP open/closed considered separately again), and another well-studied VI site, the Indianapolis site in Indiana. Both of these sites collected high frequency  $c_{\rm in}$  samples over a significant periods and are therefore suitable. Furthermore, at Indianapolis site the  $c_{\rm in}$  of three different contaminants, Chloroform, Trichloroethylene (TCE), and Tetrachloroethylene (PCE) were collected, allowing us to see if there is any difference in the variability of each. The North Island NAS dataset only spans a few days, and is therefore excluded from this analysis.

To demonstrate how significantly  $c_{\rm in}$  can vary across time, we calculate at the quotient of the maximum and minimum  $c_{\rm in}$  (denoted as  $c_{\rm max}/c_{\rm min}$ ) over a given time period. I.e. if  $c_{\rm in}$  samples were collected every four hours over a period of a year, and the data is resampled on a daily basis, then  $c_{\rm max}/c_{\rm min}$  is returned for within each day, giving 365 data points. Resampling periods of one, two, three, days, weeks, and months are chosen and Figure 2 shows the result of this analysis.

As one might expect, the longer the resampling period, the larger the maximum variability 204 is, spanning from less than a threefold difference, to two to three orders of magnitude. That 205 such a large variability is observed when the resampling time approaches the length of the 206 entire datasets is hardly surprising, nor is it surprising that the variability is much more 207 significant when the preferential pathway was open than closed. What may be surprising 208 is that absent a preferential pathway such as the one at the ASU house, it may take a few 209 weeks before an order of magnitude maximum variability is reached. The most significant 210 result of this analysis is that the maximum variability is quite small across a few days, 211 suggesting that e.g. 24-hour passive samplers will resolve the daily temporal variability in 212  $c_{\rm in}$  well  $(c_{\rm max}/c_{\rm min}\approx 1.5)$  and that a sampling frequency greater than a few days may yield 213 little extra information. There is also little difference between maximum variability of  $c_{\rm in}$ 214 the ASU house (when the PP is closed), and the three contaminants at the Indianapolis site. 215

Figure 3: Simulated cases



## 216 Influence Of Pressure On Indoor Air Variability

## 217 Attenuation To Subslab

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