

PREDICTING ADSORPTION ISOTHERMS IN NATURAL WATER USING
POLYPARAMETER LINEAR FREE ENERGY RELATIONSHIPS

BY

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DISSERTATION

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Abstract

Activated carbon is widely used in drinking water treatment to remove both natural organic matter (NOM) and trace organic contaminants. Adsorption isotherms for trace contaminants in the presence of NOM are needed to predict the performance of activated carbon for removing these compounds, but such data are not available for most of these contaminants and it would be expensive to carry out isotherm tests for all compounds of interest. This research was conducted to reduce the effort required to determine needed isotherm data by using molecular parameters to predict these data. Polyparameter linear free energy relationships (pp-LFER) for trace contaminants adsorbed on activated carbon from organic-free water have been well developed, and this research was carried out to extend the method to predict the adsorption capacity of activated carbon in natural water. Literature values of molecular descriptors of pharmaceuticals, endocrine disrupting compounds and some industrial solvents were used to predict natural water isotherms. Suwannee River NOM was used to represent natural water organic matter.

The first part of this study used oxidized graphite to represent a typical activated carbon surface. Oxidized graphite was used to eliminate the effect of pores on adsorption capacity so that surface chemical interactions can be better understood. pp-LFERs for trace contaminants adsorbed on oxidized graphite (OG) from organic-free water and water containing Suwannee River natural organic matter (SRNOM) were developed. The pp-LFERs were developed for measured partitioning coefficients K_d (i.e. the ratios of adsorption uptake q_e over the equilibrium concentration C_e). The best-fit pp-LFER correlation for predicting single solute $\log K_d$ was

determined to depend on solute activity a (i.e. the ratio of equilibrium concentration of a specific solute over its water solubility):

$$\log K_{d,i} = [(13.30 \pm 5.83) + (1.96 \pm 1.22)\log a_i]V + [(-9.97 \pm 5.37) - (1.16 \pm 1.11)\log a_i]B + \\ [(1.02 \pm 3.10) - (0.16 \pm 0.64)\log a_i]S + [(0.24 \pm 2.41) + (0.08 \pm 0.51)\log a_i]E + \\ [(-4.12 \pm 3.87) - (0.66 \pm 0.81)\log a_i]A + [(-11.87 \pm 5.20) - (2.34 \pm 1.08)\log a_i]$$

The best-fit pp-LFER correlation for adsorption from water containing SRNOM was determined to depend on solute equilibrium concentration:

$$\log K_{d,i} = [(1.57 \pm 0.63) + (0.36 \pm 0.26)\log C_{e,i}]V + [(-2.49 \pm 0.57) - (0.66 \pm 0.25)\log C_{e,i}]B + \\ [(2.11 \pm 0.25) + (0.30 \pm 0.15)\log C_{e,i}]S + [(0.03 \pm 0.23) + (0.36 \pm 0.10)\log C_{e,i}]E + \\ [(-0.51 \pm 0.25) + (0.04 \pm 0.14)\log C_{e,i}]A + [(-1.27 \pm 0.42) - (1.60 \pm 0.23)\log C_{e,i}]$$

The results showed that predictions by the pp-LFER model for adsorption by oxidized graphite are within a factor of two from the experimental values.

In the second part of this study, a pp-LFER was developed to predict activated carbon adsorption isotherms of trace organic contaminants in the presence of Suwannee River natural organic matter (SRNOM) using molecular descriptors of the trace organic contaminants. The trace organic contaminants used to develop the pp-LFER include pesticides, pharmaceuticals, endocrine disrupting compounds and some industrial solvents and the molecular descriptors (V , A , B , S , E) can be found in the literature.

The best-fit pp-LFER equation was determined to be:

$$\log K_{d,i} = [(-6.79 \pm 3.06) + (-0.71 \pm 0.77)\log C_{e,i}]V + [(7.84 \pm 2.27) + (1.63 \pm 0.64)\log C_{e,i}]B + \\ [(1.08 \pm 0.48) + (-0.14 \pm 0.15)\log C_{e,i}]S + [(-1.26 \pm 0.74) + (-0.83 \pm 0.19)\log C_{e,i}]E + \\ [(0.52 \pm 2.57) + (0.20 \pm 0.62)\log C_{e,i}]A + [(8.54 \pm 1.68) + (0.67 \pm 0.45)\log C_{e,i}].$$

This pp-LFER equation was able to predict activated carbon adsorption isotherms that are within a factor of 1.81 from the experimental values. The equation was also able to predict atrazine adsorption isotherms in two other natural waters that are within a factor of 3 from the experimental values.

To Father and Mother

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Chapter 1: Introduction

1.1 Applications of Activated Carbon

Activated carbon has been used for water treatment since the 1920s for taste and odor control (Sontheimer et al., 1988). Today, activated carbon is listed as the best available technology for removal of many target organic contaminants (Pontius, 2003). Both powdered activated carbon (PAC) and granular activated carbon (GAC) are widely used in the water treatment industry. PAC is added in powdered form at different points in the water treatment system. GAC is usually packed into a column and used as a filtration step in the water treatment system. The low capital cost and the ease of changing carbon dosage according to water quality changes make the addition of PAC very desirable for controlling taste and odor compounds. However, a GAC column would be more cost effective if the taste and odor problem exists throughout the year.

1.2 Characteristics of Activated Carbon

Both surface chemistry and pore size distribution (PSD) will affect the ability of activated carbon to adsorb. The type of surface chemical functional groups will determine the interactions of the adsorbate with the activated carbon surface. PSD, on the other hand, will determine how much adsorbate of certain size and shape can access the pore surface.

Extensive studies have been done to determine the effects of pore size distribution on adsorption. The pore with the highest adsorption energy is one that has multiple contact points with the

adsorbate, and this adsorbate will have a size that is close to the pore size (Sontheimer et al., 1988). It is important that the activated carbon used for water treatment has large amounts of suitable pores for target contaminants.

The International Union of Pure and Applied Chemistry (IUPAC) categorized the size of pores using four different pore widths. Primary micropores have width less than 8 Å, secondary micropores have width of 8-20 Å, mesopores have width of 20-500 Å and macropores have width larger than 500 Å (Lastoskie et al., 1993). Most of the surface area available for adsorption is in the micropore size (Sontheimer et al., 1988). Ebie et al. did isotherm experiments with single organic compounds and demonstrated that small halogenated organic compounds adsorbed primarily in pores smaller than 15 Å (Ebie et al., 1995). They also did experiments with one type of natural organic matter (NOM) and found that NOM adsorbed in the pore size range of 30-100 Å.

Many studies were also done to better understand the effects of surface chemistry of activated carbon on adsorption. The activated carbon surface is heterogeneous and has several chemical functional groups, most of which contain oxygen. The oxygen functional groups commonly present on the activated carbon surface are carboxylic acid groups, phenolic hydroxyl groups and quinone carboxyl groups (Boehm, 2002). The presence of these oxygen functional groups increases the surface acidity. The increased acidity causes the surface to become more hydrophilic and to decrease adsorption of hydrophobic organic compounds (Karanfil and Kilduff, 1999). This decrease in adsorption was accounted by Li et al. in their study by the

enhanced water adsorption due to the increase in hydrophilicity of the activated carbon surface (Li et al., 2002).

Besides surface acidity, other surface interactions were also studied. Electrostatic and dispersive interactions on the activated carbon surface were looked at in the work by Radovic et al. who showed that modification of the carbon surface will affect adsorption of aromatic compounds (Radovic et al., 1997). Franz et al. showed the major mechanisms through which oxygen functional groups affect adsorption are hydrogen bonding, dispersive/repulsive interactions and water adsorption (Franz et al., 2000). They also showed that the functional groups of aromatic compounds will affect how target molecules interact with the carbon surface. Details on the different types of surface interactions will be further discussed in Chapter 2.

1.3 Taste and Odor Compounds

The presence of taste and odor compounds has been a difficult aesthetic issue for the drinking water industry and GAC and PAC have been successful in removing them. Two common taste and odor compounds are 2-methylisoborneol (MIB) and geosmin (Gillogly et al., 1998); (Gillogly et al., 1999a); (Gillogly et al., 1999b). They are metabolites of algae and actinomycetes and consumers can detect the musty odor of both of these compounds at levels as low as 5-10 ng/L (Young et al., 1996); (Cook et al., 2001). The size of MIB and geosmin is 0.6-0.8 nm and it can be expected that they will adsorb on the smaller pores of activated carbon (Newcombe et al., 1997); (Newcombe, 2008). The presence of other constituents in the water will also affect the effectiveness of the activated carbon (Newcombe et al., 2002); (Newcombe et al., 2002);

(Hepplewhite et al., 2004). Thus, a careful selection of the right type of activated carbon for this application is important

1.4 Synthetic Organic Contaminants

Activated carbon is not only effective in removing taste and odor compounds, but also can remove synthetic organic contaminants. Synthetic organic chemicals (SOCs) are defined as man made organic compounds that are created by industrial synthesis. Some examples are industrial solvents, pesticides and polyaromatic hydrocarbons (PAHs). Trichloroethylene (TCE) is a common industrial solvent and it is widely used in activated carbon studies (Karanfil and Kilduff, 1999); (Kilduff and Wigton, 1999); (Kilduff and Karanfil, 2002); (Guo et al., 2007). Quinlivan et al. showed that the effective pores for TCE removal should be 1.5 times the kinetic diameter of TCE (Quinlivan et al., 2005).

Atrazine is a pesticide widely used in agriculture and commonly found in drinking water sources. There are a lot of studies in the literature which use atrazine as the target contaminant (Qi et al., 1994); (Pelekani and Snoeyink, 1999); (Pelekani and Snoeyink, 2001); (Pelekani and Snoeyink, 2000). It has been shown that both pore size distribution and surface chemistry affect how atrazine is removed by activated carbon.

1.5 Emerging Organic Contaminants

In recent years, the presence of pharmaceuticals and endocrine disrupting compounds in our drinking water has become of great concern to the public (Richardson, 2007). The presence of these compounds in our surface waters was evaluated by some researchers (Falconer et al., 2006); (Snyder et al., 2003). Since the compounds are present, it is important to evaluate current drinking water treatment systems to determine whether they can be removed. Many studies have been performed to evaluate our current drinking water treatment systems to determine if they are effective in reducing these emerging contaminants to an acceptable level (Westerhoff et al., 2005); (Snyder et al., 2007); (Chang et al., 2004).

Removal of some emerging organic contaminants by activated carbon was evaluated in several studies and it was shown that activated carbon is generally effective (Zhang and Zhou, 2005); (Yu et al., 2008); (Yoon et al., 2005); (Fukuhara et al., 2006). The removal of a mixture of endocrine disrupting compounds, pharmaceuticals, personal care products and flame retardants by several adsorbents was studied by Rossner et al. (Rossner et al., 2009) and they showed that activated carbon was the most effective adsorbent as compared to a carbonaceous resin and two high-silica zeolites. Fukuhara et al. showed that the amount of estrone and 17 β -estradiol removed by activated carbon depended on the pore diameter, specific surface area, the hydrophobicity of the target chemicals and the type and concentration of other water constituents (Fukuhara et al., 2006).

1.6 Natural Organic Matter

Natural organic matter (NOM) exists ubiquitously in the environment and in most of our water sources. NOM results from degrading plants and animals and consists of humic acids, carboxylic acids, proteins, lipids and amino acids. PAC and GAC are both effective in removing NOM. Under the Stage 2 Disinfection Byproduct Rule (DBPR), utilities are required to reduce the concentration of disinfection byproducts found in the distribution system. NOM is one of the major disinfection byproduct (DBP) precursors; by removing NOM, DBP formation can be reduced. The Stage 2 DBPR will drive utilities to consider using GAC for removing NOM.

The wide molecular weight distribution of NOM affects how it adsorbs on activated carbon. Studies showed that capacity of activated carbon for NOM increases as NOM size decreases (Kilduff et al., 1996); (McCreary and Snoeyink, 1980). The chemistry of NOM adsorption on activated carbon was studied by Karanfil et al. (Karanfil et al., 1996). They showed that NOM adsorption capacity increased as the amount of acidic functional groups increased, as long as the increase in functional groups did not increase NOM solubility.

Although activated carbon has been effective in removing trace organic contaminants, the presence of NOM in water sources has been a great challenge. The presence of NOM greatly reduces the capacity of activated carbon for trace organic contaminants. Many studies have been done to understand the competitive mechanism of NOM on the adsorption of trace organic contaminants. Qi et al. (Qi et al., 1994) found that NOM, present at a much higher concentration

in mg/L levels, will compete strongly with trace organic contaminants, which are present at $\mu\text{g/L}$ levels, for adsorption sites on activated carbon.

Many studies have been done to understand the mechanism of the competitive effects of NOM on the removal of trace organic contaminants. NOM competes with trace organic compounds for adsorption sites via two major mechanisms: direct site competition and pore blockage. Smaller molecular weight NOM that is of similar size and weight as the target compounds competes with trace compounds for the same sites. Larger molecular weight NOM adsorbs in larger pores and blocks the smaller molecular weight trace compounds from reaching the inner pores; thus, the rate of adsorption of trace compounds is reduced (Carter et al., 1992).

Li et al. used model compounds, p-dichlorobenzene and polystyrene sulfonates to mimic small molecular size and large molecular size NOM respectively (Li et al., 2003). They showed that NOM of small molecular size greatly reduced the capacity of activated carbon for trace organic compounds but did not significantly affect adsorption kinetics, whereas large molecules reduced adsorption kinetics without significantly affecting adsorption capacity.

Ding et al. (Ding et al., 2008) determined that the pore size distribution of activated carbon would determine the adsorption capacity for trace organic compounds. Several mechanisms of NOM competition were analyzed and it was determined that NOM that adsorbed in the 15-50 Å pore size range was the most effective at blocking pores. NOM that adsorbs in the pores reduces the rate of diffusion of organic compounds being adsorbed onto activated carbon. Thus, they

recommended that the volume of pores of this size range be increased to decrease pore blockage effects.

1.7 Motivation for Predicting Adsorption Isotherms for Activated Carbon

To predict the performance of activated carbon in a particular system, both the adsorption rate and capacity of the carbon have to be either experimentally determined or systematically estimated. The adsorption capacity of activated carbon generally shows which type of activated carbon is the most effective for the target purpose and can be used to estimate the performance of activated carbon in a treatment process. Adsorption capacity can be determined by conducting isotherm tests. However, since there are more than 100,000 chemicals and mixtures presently produced and used in the world and adsorption isotherm data are only readily available for a few of these, collecting isotherm data for all chemicals and mixtures will be expensive and unrealistic (Schwarzenbach et al., 2006).

As analytical techniques become more sophisticated and the detection limits of chemicals are lowered, it is expected that more stringent regulations will be set for water utilities. Being able to predict adsorption isotherms will be very valuable in determining the effectiveness of activated carbon for removal of target chemicals. The capacity of activated carbon in a natural water can be estimated with the equivalent background compound model using single solute isotherm data (Graham et al., 2000). However, if the adsorption isotherm data in natural water can be predicted without the need to determine isotherms, performance estimation can be done more easily and cheaply.

Several theories have been proposed to predict single solute adsorption isotherms of organic compounds on activated carbon. Crittenden et al. (Crittenden et al., 1999) integrated linear solvation energy relationship (LSER) parameters into Polanyi potential theory and correlated the adsorption isotherms of compounds with 4 LSER parameters. The Polanyi Potential equation was normalized by LSER parameters to achieve the correlation. The LSER parameters represent different types of interactions between organic compounds and activated carbon. These researchers used single solute isotherm data from the literature that included data for 4 different groups of adsorbents with a total of 8 adsorbents, and 4 different groups of adsorbates with a total of 56 organic compounds. Correlation curves were developed for compounds with specific functional groups for each adsorbent but an overall correlation curve for the adsorption isotherms of all organic compounds was not determined. The correlations of the adsorption isotherms for halogenated aliphatic organic compounds, halogenated aromatic organic compounds and aromatic organic compounds on F-400 carbon and LSER parameters were well developed, but correlations of adsorption isotherms for polyfunctional organic compounds and aromatic organic compounds and sulfonated aromatic organic compounds were not.

Li et al. (Li et al., 2005) developed another method to predict single solute isotherms using Polanyi-Dubinin-Manes (PDM) model. They developed a good correlation between the oxygen content of the activated carbons and the water affinity coefficients. This correlation integrated into the PDM model was successful in predicting single solute isotherms for three adsorbents and two adsorbates. The predictions agreed well with the experimental data.

A recent study by de Ridder et al. used quantitative structure activity relationships (QSARs) to develop a model to predict carbon loading (q_e) on F400 activated carbon in ultrapure water (de Ridder et al., 2010). Organic compounds were organized into four groups for good predictions of q_e . Polarizability of the solute, hydrophobic partitioning of the solute and the log of the equilibrium aqueous concentration, $\log C_e$, were found to be linearly correlated to $\log q_e$. The $\log q_e$ values of most solutes were predicted within 0.5 log units from the experimental values.

Another approach to predict the adsorption coefficients of several organic compounds on activated carbon was to develop a Linear Solvation Energy Relationship using the chemical activities of adsorbates. An adsorption coefficient is the ratio of equilibrium concentrations of an organic compound on activated carbon to that in the aqueous phase. The chemical activity of an adsorbate is the ratio of the concentration in the aqueous phase at equilibrium to that at saturation. Shih and Gschwend (Shih and Gschwend, 2009) directly fitted experimental adsorption coefficients and chemical activities of 14 organic compounds with the corresponding LSER parameters of the compounds. They were successful in correlating the experimental single solute adsorption coefficients with the LSER parameters.

Shih and Gschwend (Shih and Gschwend, 2009) demonstrated that polyparameter linear free energy relationships (LFERs) can be used to predict single solute isotherms on activated carbon. It is reasonable to hypothesize that polyparameter LFERs can be used to predict adsorption isotherms in the presence of one type of NOM.

1.8 Research Objectives

There are many studies that adsorption isotherms of organic compounds in pure water can be predicted. However in reality, there will always be NOM in drinking water sources and that will change how organic compounds interact with the surface of carbon. It will be very useful if adsorption isotherms in the presence of NOM can be predicted directly.

The objective of this research was to develop a correlative equation that can predict the adsorption capacity of activated carbon in natural water using molecular descriptors of trace organic contaminants. Surface chemistry and pore size distribution of activated carbon, molecular properties of the adsorbate, and the type of natural organic matter (NOM) can influence the capacity of activated carbon. In order to determine the role of NOM, it was decided to first determine the effect of one type of NOM on trace compound adsorption on a non-porous carbon, and then to determine the effect on porous activated carbon. Specific objectives of this research were:

1. Determine the influence of only surface chemistry on adsorption capacity for non-porous graphite. Without the presence of pores the chemistry of the surface adsorption can be specifically quantified. The highly hydrophobic surface of graphite can be oxidized before use to make its surface similar to that of a typical activated carbon;
2. Develop pp-LFER equations to predict the partitioning coefficient ($\log K_d$) of trace organic compounds on oxidized graphite in the absence and presence of one type of NOM;

3. Validate the above two equations. Adsorption isotherms for compounds not used to develop the correlations will be predicted and compared to experimental data;
4. Develop a pp-LFER equation to predict the partitioning coefficient ($\log K_d$) of trace compounds on activated carbon in the presence of one type of NOM;
5. Validate the activated carbon pp-LFER equation. Adsorption isotherms for compounds not used to develop the correlation will be predicted and compared to experimental data;
6. Apply the activated carbon pp-LFER equation to predict adsorption isotherms of atrazine in other natural waters with the same carbon;
7. Evaluate the critical number of trace compounds isotherms needed to develop a pp-LFER equation;
8. Compare the oxidized graphite equations and activated carbon equation to evaluate how the tendency of solute partitioning change with the presence of pore size distribution.

REFERENCES

- Boehm, H. P. (2002) Surface oxides on carbon and their analysis: a critical assessment. *Carbon*, **40**(2), 145-149.
- Carter, M. C., Weber, W. J., Olmstead, K. P. (1992) Effects of Background Dissolved Organic-Matter on Tce Adsorption by Gac. *Journal American Water Works Association* **84**(8), 81-91.
- Chang, S., Waite, T. D., Ong, P. E. A., Schafer, A. I., Fane, A. G. (2004) Assessment of trace estrogenic contaminants removal by coagulant addition, powdered activated carbon adsorption and powdered activated carbon/microfiltration processes. *J. Environ. Eng. -ASCE* **130**(7), 736-742.
- Cook, D., Newcombe, G., Sztajn bok, P. (2001) The application of powdered activated carbon for MIB and geosmin removal: Predicting PAC doses in four raw waters. *Water Res.* **35**(5), 1325-1333.
- Crittenden, J. C., Sanongraj, S., Bulloch, J. L., Hand, D. W., Rogers, T. N., Speth, T. F., Ulmer, M. (1999) Correlation of aqueous-phase adsorption isotherms. *Environ. Sci. Technol.* **33**(17), 2926-2933.
- de Ridder, D. J., Villacorte, L., Verliefde, A. R. D., Verberk, J. Q. J. C., Heijman, S. G. J., Amy, G. L., van Dijk, J. C. (2010) Modeling equilibrium adsorption of organic micropollutants onto activated carbon. *Water Res.* **44**(10), 3077-3086.

- Ding, L., Snoeyink, V. L., Marinas, B. J., Yue, Z., Economy, J. (2008) Effects of powdered activated carbon pore size distribution on the competitive adsorption of aqueous atrazine and natural organic matter. *Environ. Sci. Technol.* **42**(4), 1227-1231.
- Ebie, K., Li, F. S., Hagishita, T. (1995) Effect of pore size distribution of activated carbon on the adsorption of humic substances and trace organic compounds. *Water Supply* **13**(34), 65-70.
- Falconer, I. R., Chapman, H. F., Moore, M. R., Ranmuthugala, G. (2006) Endocrine-disrupting compounds: A review of their challenge to sustainable and safe water supply and water reuse. *Environ. Toxicol.* **21**(2), 181-191.
- Franz, M., Arafat, H. A., Pinto, N. G. (2000) Effect of chemical surface heterogeneity on the adsorption mechanism of dissolved aromatics on activated carbon. *Carbon* **38**(13), 1807-1819.
- Fukuhara, T., Iwasaki, S., Kawashima, M., Shinohara, O., Abe, I. (2006) Adsorbability of estrone and 17 beta-estradiol in water onto activated carbon. *Water Res.* **40**(2), 241-248.
- Gillogly, T. E. T., Snoeyink, V. L., Newcombe, G., Elarde, J. R. (1999a) A simplified method to determine the powdered activated carbon dose required to remove methylisoborneol. *Water Sci. Technol.* **40**(6), 59-64.
- Gillogly, T. E. T., Snoeyink, V. L., Vogel, J. C., Wilson, C. M., Royal, E. P. (1999b) Determining GAC bed life. *J. Am. Water Work Assoc.* **91**(8), 98-110.
- Gillogly, T. E. T., Snoeyink, V. L., Elarde, J. R., Wilson, C. M., Royal, E. P. (1998) C-14-MIB adsorption on PAC in natural water. *J. Am. Water Work Assoc.* **90**(1), 98-108.

Graham, M. R., Summers, R. S., Simpson, M. R., MacLeod, B. W. (2000) Modeling equilibrium adsorption of 2-methylisoborneol and geosmin in natural waters. *Water Research*, **34**(8), 2291-2300.

Guo, Y., Yadav, A., Karanfil, T. (2007) Approaches To Mitigate the Impact of Dissolved Organic Matter on the Adsorption of Synthetic Organic Contaminants by Porous Carbonaceous Sorbents. *Environ. Sci. Technol.*

Hepplewhite, C., Newcombe, G., Knappe, D. R. U. (2004) NOM and MIB, who wins in the competition for activated carbon adsorption sites? *Water Sci. Technol.* **49**(9), 257-265.

Karanfil, T. and Kilduff, J. E. (1999) Role of granular activated carbon surface chemistry on the adsorption of organic compounds. 1. Priority pollutants. *Environ. Sci. Technol.* **33**(18), 3217-3224.

Karanfil, T., Kilduff, J. E., Schlautman, M. A., Weber, W. J. (1996) Adsorption of organic macromolecules by granular activated carbon .1. Influence of molecular properties under anoxic solution conditions. *Environmental Science & Technology* **30**(7), 2187-2194.

Kilduff, J. E. and Karanfil, T. (2002) Trichloroethylene adsorption by activated carbon preloaded with humic substances: effects of solution chemistry. *Water Research* **36**(7), 1685-1698.

Kilduff, J. E. and Wigton, A. (1999) Sorption of TCE by humic-preloaded activated carbon: Incorporating size-exclusion and pore blockage phenomena in a competitive adsorption model. *Environmental Science & Technology* **33**(2), 250-256.

- Kilduff, J. E., Karanfil, T., Chin, Y. -, Weber, W. J. (1996) Adsorption of Natural Organic Polyelectrolytes by Activated Carbon: A Size-Exclusion Chromatography Study. *Environ. Sci. Technol.* **30**(4), 1336-1343.
- Lastoskie, C., Gubbins, K. E., Quirke, N. (1993) Pore-Size Distribution Analysis of Microporous Carbons - a Density-Functional Theory Approach. *J. Phys. Chem.* **97**(18), 4786-4796.
- Li, L., Quinlivan, P. A., Knappe, D. R. U. (2005) Predicting adsorption isotherms for aqueous organic micropollutants from activated carbon and pollutant properties. *Environ. Sci. Technol.* **39**(9), 3393-3400.
- Li, L., Quinlivan, P., Knappe, D. U. (2002) Effects of activated carbon surface chemistry and pore structure on the adsorption of organic contaminants from aqueous solution. *Carbon*, **40**(12), 2085-2100.
- Li, Q. L., Snoeyink, V. L., Mariaas, B. J., Campos, C. (2003) Elucidating competitive adsorption mechanisms of atrazine and NOM using model compounds. *Water Res.* **37**(4), 773-784.
- McCreary, J. J. and Snoeyink, V. L. (1980) Characterization and Activated Carbon Adsorption of several Humic Substances. *Water Res.* **14**(2), 151-160.
- Newcombe, G., Morrison, J., Hepplewhite, C. (2002) Simultaneous adsorption of MIB and NOM onto activated carbon. I. Characterisation of the system and NOM adsorption. *Carbon* **40**(12), 2135-2146.
- Newcombe, G., Morrison, J., Hepplewhite, C., Knappe, D. R. U. (2002) Simultaneous adsorption of MIB and NOM onto activated carbon - II. Competitive effects. *Carbon* **40**(12), 2147-2156.

Newcombe, G., Drikas, M., Hayes, R. (1997) Influence of characterised natural organic material on activated carbon adsorption .2. Effect on pore volume distribution and adsorption of 2-methylisoborneol. *Water Research* **31**(5), 1065-1073.

Newcombe, G. (2008) Adsorption from Aqueous Solutions: Water Purification. In: Eduardo J. Bottani, Juan M.D. Tascón (Eds.), *Adsorption by Carbons*, Elsevier, Amsterdam, 679-709.

Pelekani, C. and Snoeyink, V. L. (2001) A kinetic and equilibrium study of competitive adsorption between atrazine and Congo red dye on activated carbon: the importance of pore size distribution. *Carbon* **39**(1), 25-37.

Pelekani, C. and Snoeyink, V. L. (2000) Competitive adsorption between atrazine and methylene blue on activated carbon: the importance of pore size distribution. *Carbon* **38**(10), 1423-1436.

Pelekani, C. and Snoeyink, V. L. (1999) Competitive adsorption in natural water: Role of activated carbon pore size. *Water Res.* **33**(5), 1209-1219.

Pontius, F. W. (2003) USEPA's drinking water regulations. *J. Am. Water Work Assoc.* **95**(3), 57-.

Quinlivan, P. A., Li, L., Knappe, D. R. U. (2005) Effects of activated carbon characteristics on the simultaneous adsorption of aqueous organic micropollutants and natural organic matter. *Water Research* **39**(8), 1663-1673.

Radovic, L. R., Silva, I. F., Ume, J. I., Menendez, J. A., Leon, C. A. L. Y., Scaroni, A. W. (1997) An experimental and theoretical study of the adsorption of aromatics possessing electron-

withdrawing and electron-donating functional groups by chemically modified activated carbons. Carbon **35**(9), 1339-1348.

Richardson, S. D. (2007) Water Analysis: Emerging Contaminants and Current Issues. Anal. Chem. **79**(12), 4295-4324.

Rossner, A., Snyder, S. A., Knappe, D. R. U. (2009) Removal of emerging contaminants of concern by alternative adsorbents. Water Res. **43**(15), 3787-3796.

Schwarzenbach, R. P., Escher, B. I., Fenner, K., Hofstetter, T. B., Johnson, C. A., von Gunten, U., Wehrli, B. (2006) The Challenge of Micropollutants in Aquatic Systems. Science **313**(5790), 1072-1077.

Shaoying Qi, Samer S. Adham, Vernon L. Snoeyink, Ben W. Lykins, J. (1994) Prediction and Verification of Atrazine Adsorption by PAC. J. Envir. Engrg. **120**(1), 202-218.

Shih, Y. and Gschwend, P. M. (2009) Evaluating Activated Carbon-Water Sorption Coefficients of Organic Compounds Using a Linear Solvation Energy Relationship Approach and Sorbate Chemical Activities. Environmental Science & Technology **43**(3), 857.

Snyder, S. A., Westerhoff, P., Yoon, Y., Sedlak, D. L. (2003) Pharmaceuticals, personal care products, and endocrine disruptors in water: Implications for the water industry. Environ. Eng. Sci. **20**(5), 449-469.

Snyder, S. A., Adham, S., Redding, A. M., Cannon, F. S., DeCarolis, J., Oppenheimer, J., Wert, E. C., Yoon, Y. (2007) Role of membranes and activated carbon in the removal of endocrine disruptors and pharmaceuticals. Desalination, **202**(1-3), 156-181.

Sonthheimer, H., Crittenden, J., and Summers, S. (1988) Activated Carbon for Water Treatment.

Westerhoff, P., Yoon, Y., Snyder, S., Wert, E. (2005) Fate of Endocrine-Disruptor, Pharmaceutical, and Personal Care Product Chemicals during Simulated Drinking Water Treatment Processes. Environ. Sci. Technol. **39**(17), 6649-6663.

Yoon, Y., Westerhoff, P., Snyder, S. A. (2005) Adsorption of H-3-labeled 17-beta estradiol on powdered activated carbon. Water Air Soil Pollut. **166**(1-4), 343-351.

Young, W. F., Horth, H., Crane, R., Ogden, T., Arnott, M. (1996) Taste and odour threshold concentrations of potential potable water contaminants. Water Res. **30**(2), 331-340.

Yu, Z., Peldszus, S., Huck, P. M. (2008) Adsorption characteristics of selected pharmaceuticals and an endocrine disrupting compound - Naproxen, carbamazepine and nonylphenol - on activated carbon. Water Res. **42**(12), 2873-2882.

Zhang, Y. P. and Zhou, J. L. (2005) Removal of estrone and 17 beta-estradiol from water by adsorption. Water Res. **39**(16), 3991-4003.

Chapter 2: Theory and Applications of Polyparameter Linear Free Energy Relationship

2.1 Theory

Linear free energy relationships (LFERs) are extrathermodynamic correlations used to understand the characteristics of equilibrium and kinetic processes from the perspective of molecular structural contributions. LFERs are empirically determined and can be used to predict partition coefficients of organic compounds. Free energy change for the transfer of one mole of organic compound from the aqueous phase to activated carbon is related to the partition coefficient of the activated carbon as follows:

$$\Delta G_{overall} = -RT \ln(K_{eq}) \dots \dots \dots (2.1)$$

where $\Delta G_{overall}$ = Overall free energy change of the chemical equilibrium

R = Gas constant

T = Temperature

K_{eq} = Partitioning coefficient

LFERs have been commonly used in a form where the logarithm of reaction rate or equilibrium constants of one series of reactions is correlated linearly with the logarithm of reaction rate or equilibrium constants of another associated series of reactions. The Brønsted equation, developed in 1924 by Brønsted and Pederson, correlates the catalysis reaction rate with the ionization constant of acids was the first LFER that is recognized widely (Wells, 1968).

LFERs are used in the environmental engineering field to understand the equilibrium partitioning of organic compounds in several natural phases. The fate and transport of organic contaminants in the environment can then be better understood (Nguyen et al., 2005). LFERs have also been used in understanding the equilibrium partitioning of organic compounds on activated carbon. Free energy change for the transfer of one mole of organic compound from the aqueous phase to activated carbon is related to the adsorption capacity of activated carbon (Abe et al., 1983).

The type of LFER described above is known as a one-parameter LFER and it is mainly used to correlate a well-known parameter to unknown parameter. For example, in the following equation, the unknown natural organic carbon-water partition coefficient $K_{i,oc}$ is correlated with the well-known octanol-water partition coefficient $K_{i,ow}$. Coefficients a and b represent the differential interactions of the solute in the two phases (Schwarzenbach et al., 2003).

$$\log K_{i,oc} = a \log K_{i,ow} + b \dots \dots \dots (2.2)$$

A one-parameter LFER predicts well within one class of compounds. Although there are many one-parameter LFERs developed for different systems, the applicability of one-parameter linear free energy relationships have been very limited. It is often observed that the one-parameter LFER is unable to predict for compounds that are not in the same class. That is mainly because one particular parameter is picked to represent all the interactions between the solute and the natural phases of concern. However equilibrium partitioning coefficients are often governed by several different types of molecular interactions between a solute molecule and the aqueous and solid phase.

Unlike one-parameter LFERs, a polyparameter LFER is able to take into account the different molecular interactions between the solute and the natural phases. There are multiple kinds of molecular interactions between an organic compound and activated carbon. Thus, the polyparameter LFER can be useful in predicting the adsorption capacity of the activated carbon for different trace organic contaminants.

The types of interactions that affect adsorption capacity of activated carbon are dependent on the following:

1. Molecular properties of the organic compound
2. Type of NOM existing in the natural water
3. Surface chemistry of the activated carbon
4. Pore size distribution of the activated carbon

The overall free energy equation for polyparameter LFER can be represented as follows:

$$-\Delta G_{overall} = \Delta G_{solute-solvent} + \Delta G_{solvent-adsorbent} - \Delta G_{solute-adsorbent} \dots\dots\dots(2.3)$$

From the much-simplified equation, energy is needed to break the bond between the solvent and adsorbent and also the bond between the solute and solvent, in this case, the solvent being water.

$$\Delta G_{solute-solvent} = vW \dots\dots\dots(2.4)$$

The free energy of the interaction between the solute and water is proportional to molecule size, which can be represented by the McGowan's characteristic molecular volume (V) (McGowan and Mellors, 1986). The coefficient v would characterize the difference in the cohesive forces of the solute molecules in water and on carbon.

$$\Delta G_{\text{solvent-adsorbent}} - \Delta G_{\text{solute-adsorbent}} = eE + sS + aA + bB \dots\dots\dots(2.5)$$

The differential free energy change of the interaction between solute and adsorbent and the interaction between solvent and adsorbent is correlated to four parameters: E , S , A and B .

Parameter E being the excess molar refraction of the solute and it accounts for the induction of dipoles within the solute molecule and that correspond to the nonspecific interactions. Parameter S describes the polarity and polarizability of the solute. Parameter A is the overall hydrogen-bond acidity of the solute and that represent the solute's ability to accept electrons. Parameter B is the overall hydrogen-bond basicity and it represents the solute's ability to donate electrons. All parameters can be obtained from the literature. This nomenclature was introduced by Abraham et al (Abraham et al., 2004).

The values of the coefficients in front of the parameters, i.e. the small letters, will be determined by the chemical properties of both water and carbon. The coefficient e is determined by the differential non specific interactions of the solute and the two phases. The non specific interactions are caused by both induced dipole-induced dipole interactions and also dipole-induced dipole interactions. The coefficient s is determined by the differential dipole-dipole interactions of the solute in water and on carbon. It represents the electrostatic interactions determined by dipole-dipole interactions. The coefficient a is determined by the differential amount of H-bond acceptor sites in water and on carbon. The coefficient b is determined by the differential amount of H-bond donating sites in water and on carbon.

Combining all above the equations we obtain the pp-LFER for the partitioning coefficient $\log K_d$:

$$\log K_d = eE + sS + aA + bB + vV + c \dots\dots\dots(2.6)$$

The constant c is determined by the volume entropy effects and it is free energy contribution that is specific to the two phases.

Partitioning coefficients $\log K_d$ only represent the adsorption phenomena at infinite-dilution conditions. However, amount of organic compound adsorbed on carbon varies with the aqueous concentration of the organic compound. Experimental adsorption isotherms are non-linear because of this aqueous concentration dependence. Linear isotherms are only observed for a very narrow range of aqueous concentrations. Therefore, to be able to predict $\log K_d$ that is representative of environmental conditions, it will be important to include the concentration dependence into the predictive equation:

$$\log K_{d,i} = (e_1 + e_2 \log(a_{e,i}))E + (s_1 + s_2 \log(a_{e,i}))S + (a_1 + a_2 \log(a_{e,i}))A + (b_1 + b_2 \log(a_{e,i}))B + \dots\dots(2.7)$$

$$(v_1 + v_2 \log(a_{e,i}))V_x + c_1 + c_2 \log(a_{e,i})$$

where $C_{e,i}$ ($\mu\text{g/L}$) = Concentration of adsorbate i in solution at equilibrium

$q_{e,i}$ ($\mu\text{g/ mg carbon}$) = Concentration of adsorbate i on carbon at equilibrium

S_w ($\mu\text{g/L}$) = Solubility of adsorbate

$$a_{e,i} = C_{e,i} / S_w$$

$$K_{d,i} = q_{e,i} / C_{e,i}$$

The above equation includes the chemical activities of the adsorbates and represents the intermolecular interactions of compounds at varying concentration that will affect adsorption onto activated carbon.

2.2 Applications

Although the one-parameter LFER has very limited applications, it is useful for compounds that have similar functional groups. The one-parameter LFER can either account for van der Waals interactions or H-bond interactions because there is only one molecular parameter in the predictive equation. Better predictions will be obtained for a series of partitioning coefficients which vary only in one type of molecular interactions. For example, the application of the one-parameter LFER to predict reactions of a homologous series of chemicals usually is successful because H-bond interactions would either be unavailable or they would be a constant for each compound in the group (Goss and Schwarzenbach, 2001).

For predicting a wider range of chemical compounds, pp-LFERs are generally more successful, but the data set required for reliable predictions will be much larger as compared to one-parameter LFERs. The best results are obtained with a data set consisting of a huge array of neutral chemical compounds. For any given partitioning coefficient, the wider the range of functional groups included in the development of the polyparameter LFER, the better the predictions.

Polyparameter LFERs have been successful in predicting equilibrium partition coefficients of organic compounds between water and NOM in soils and sediments (Nguyen et al., 2005).

Polyparameter LFERs have also been successful in predicting the equilibrium partition coefficient of organic compounds in NOM-free water onto activated carbon (Shih and Gschwend, 2009).

REFERENCES

- Abe, I., Hayashi, K., Kitagawa, M., Hirashima, T. (1983) Prediction of Adsorbability of Organic-Compounds from Aqueous-Solution on Activated Carbon by Means of the Linear Free-Energy Relationship. *Bull. Chem. Soc. Jpn.* 56(4), 1002-1005.
- Abraham, M. H., Ibrahim, A., Zissimos, A. M. (2004) Determination of sets of solute descriptors from chromatographic measurements. *J. Chromatogr. A* 1037(1-2), 29-47.
- Goss, K. -. and Schwarzenbach, R. P. (2001) Linear Free Energy Relationships Used To Evaluate Equilibrium Partitioning of Organic Compounds. *Environ. Sci. Technol.* 35(1), 1-9.
- McGowan, J. C. and Mellors, A. (1986) *Molecular Volumes in Chemistry and Biology: Applications Including Partitioning and Toxicity.*
- Nguyen, T. H., Goss, K. -, Ball, W. P. (2005) Polyparameter Linear Free Energy Relationships for Estimating the Equilibrium Partition of Organic Compounds between Water and the Natural Organic Matter in Soils and Sediments. *Environ. Sci. Technol.* 39(4), 913-924.
- Schwarzenbach, R. P., Gschwend, P. M., Imboden, D. M. (Hoboken, N.J. : Wiley, c2003) *Environmental Organic Chemistry.* Wiley,
- Shih, Y. and Gschwend, P. M. (2009) Evaluating Activated Carbon-Water Sorption Coefficients of Organic Compounds Using a Linear Solvation Energy Relationship Approach and Sorbate Chemical Activities. *Environmental Science & Technology* 43(3), 857.
- Wells, P. R. (1968) *Linear free energy relationships.* Academic P., London, New York, 116 p.

Chapter 3: Linear Free Energy Relationship for Predicting Adsorption of Trace Organic Pollutants from Natural Water onto Oxidized Graphite

3.1 Introduction

There have been several approaches employed to predict single solute adsorption isotherms from NOM-free water. Crittenden et al. (Crittenden et al., 1999) integrated linear solvation energy relationship (LSER) parameters into the Polanyi potential theory and correlated the adsorption isotherms of compounds with 4 LSER parameters. A second method for NOM-free water using the Polanyi-Dubinin-Manes (PDM) model was developed by Li et al. (Li et al., 2005) to predict single solute activated carbon isotherms. They developed a good correlation between the oxygen content of the adsorbents and the water affinity coefficients. This correlation was then integrated into the PDM and they were successful in using it to predict single solute isotherms for three adsorbents and two adsorbates. A third approach was to develop a Linear Solvation Energy Relationship for $\log K_d$ using the chemical activities of adsorbates. (Shih and Gschwend, 2009) The adsorption coefficient, K_d , is the ratio of the adsorbed concentration of an organic compound to the aqueous phase concentration at equilibrium. Shih and Gschwend successfully fitted the experimental adsorption coefficients and chemical activities (the ratio of the equilibrium solution concentration to the saturation concentration) of 14 organic compounds with the corresponding LSER parameters of the compounds.

Natural organic matter (NOM) is ubiquitous in our water sources. It greatly reduces the capacity of activated carbon for trace organic contaminants via competition for adsorption sites because

NOM is present at mg/L levels while trace compounds are present at $\mu\text{g/L}$ levels and below.

Therefore, it will be useful to have an equation that is able to predict adsorption isotherms in the presence of NOM for utilities to use directly.

The objective of this chapter was to develop pp-LFER correlative equations that can predict the adsorption capacity of organic contaminants on oxidized graphite in pure water and natural water using molecular descriptors of the trace compounds. Oxidized graphite was used as a non-porous adsorbent representative of the activated carbon surface. The effect of activated carbon pore size distribution on adsorption capacity will be evaluated in chapter 5 and a pp-LFER correlative equation will be developed for activated carbon.

3.2 Materials and Methods

3.2.1 Water

Single solute isotherm experiments were performed in NOM-free water. Organic free water (OFW) was generated by passing de-ionized water through a NANOpure ultrapure water system (Barnstead, Dubuque, Iowa). The dissolved organic carbon (DOC) concentration of the organic-free water was lower than 0.3 mg/L.

Suwannee River NOM (International Humic Substances Society, St. Paul, Minnesota, USA), isolated by reverse osmosis, was received in dry form. The SRNOM was dissolved in OFW and filtered with a 0.45 μ m nylon membrane filter to remove suspended solids. The stock concentration of 200 mg/L was subsequently diluted to prepare the solution used to perform natural water isotherm experiments. The final solution TOC concentration was measured as 4 mg/L and the pH was 7.0.

3.2.2 Adsorbents

To understand the influence of surface chemistry on activated carbon adsorption, graphite, a non-porous adsorbent was selected. However, unlike activated carbon, the surface of graphite does not have oxygen functional groups. Thus graphite was oxidized to make an activated carbon-like surface before it was used for isotherm tests. Graphite powder (Sigma, St. Louis, Missouri, USA) was suspended in OFW, and ozone gas was bubbled through the suspension for 5 hrs to oxidize the graphite. (PCI Model GL-1, PCI Ozone & Control Systems, W. Caldwell, NJ) The oxidized graphite was filtered and then washed with 3 L of OFW. After this treatment, the oxidized

graphite was dried at 60°C for 24 hrs. The dried, oxidized graphite was then kept in a bottle in a desiccator for later use. The graphite was analyzed using X-ray-Photoelectron Spectroscopy (XPS) for its carbon and oxygen contents. XPS spectra were acquired from a Physical Electronics PHI 5400 (Perkin-Elmer, MN). The element peak areas were normalized using atomic sensitivity factors to obtain the carbon to oxygen ratios.

3.2.3 Adsorbates

Ten compounds were selected as the target trace organic contaminants, including organic solvents (trichloroethylene, 1-heptanol, nitrobenzene), endocrine disrupting compounds (estradiol, bisphenol A), an explosive (2, 4-dinitrotoluene), an insecticide (lindane), an herbicide (atrazine), a polycyclic aromatic hydrocarbon (naphthalene) and a pharmaceutical (diazepam). The 10 compounds have a range of molecular descriptors that covers the characteristics of most chemical contaminants. ¹⁴C-labeled forms of the 10 compounds were used (American Radiolabeled Chemicals, Inc). See Table 1 for structure and properties of the selected adsorbates.

3.2.4 Isotherm experiments with oxidized graphite

Single solute isotherms and competitive adsorption isotherms of 10 compounds were determined using traditional bottle-point isotherm tests. (Randtke and Snoeyink, 1983) The concentration of stock solution used for each compound was 10^{-3} of its solubility. A different amount of oxidized carbon was used for each compound to give a 10^{-4} to 10^{-6} of aqueous solubility range of final aqueous equilibrium concentration (C_e). Bottles were sealed with aluminum foil and mixed using a rotating tumbler. Seven days equilibrium time was used for all isotherms. All isotherm

experiments of volatile compounds were carried out in isotherm bottles with no headspace and sealed with aluminum foil and crimp top caps with a Teflon/silica septum.

After 7 days of equilibration, the aqueous phase concentration of the solute was sampled and the solid phase concentration was calculated using mass balance. Two blanks were included for each set of isotherm experiments to monitor the amount of solute lost besides adsorption. Less than 10% loss was observed for the 7 days equilibration time.

3.2.5 Analytical Methods

Samples were analyzed at the end of 7 days. All samples were taken using a 5 ml glass syringe. For non-volatile and low solubility compounds (estradiol, lindane, naphthalene, diazepam), the bottles were settled for 8 hrs before samples were taken. For non-volatile and high solubility compounds (atrazine, 2,4-dinitrotoluene, bisphenol A), aqueous samples were taken and filtered with a 0.45 μm nylon membrane filter. For volatile and high solubility compounds (trichloroethylene, 1-heptanol, nitrobenzene), the bottles were settled for 8 hrs before samples were taken.

The 2.5 mL of aqueous sample was then mixed with 18 mL of scintillation cocktail in a 20 mL scintillation vial and analyzed using a liquid scintillation analyzer (TriCarb Model 1600A, Packard Instrument Co. Downers Grove, IL). Each analysis session included 2 blanks.

3.2.6 Development of Predictive Polyparameter Linear Free Energy Relationship (pp-LFER)

K_d values were calculated using the following equation:

$$K_d = q_e / C_e \dots\dots\dots(3.1)$$

where q_e = adsorbed concentration at equilibrium and

C_e = aqueous concentration at equilibrium

These $\log K_d$ values of 9 compounds were then used to fit the adsorbate molecular properties using equation 2.7 described above. Multiple linear regressions were applied to the experimental $\log K_d$ data using Sigma Plot 2000 (SPSS, Chicago, IL) to obtain the coefficients. The partitioning of the aqueous solute to oxidized graphite will determine coefficients $e_1, e_2, s_1, s_2, a_1, a_2, b_1, b_2, v_1, v_2, c_1, c_2$.

3.2.7 Validation of the pp-LFER model

The fitted equations were used to predict $\log K_d$ values for the 10th compound that was not used for fitting. The predicted $\log K_d$ values were then compared with experimental $\log K_d$ values to determine if the fitted equations were applicable to other compounds.

3.3 Results and Discussion

3.3.1 Characterization of oxidized graphite

After ozonation, the oxygen concentration (mass percentage) of graphite was increased from 1.22% to 4.52%. This oxygen concentration was similar to the 7.21% of a commercial activated carbon SA-UF. Langley and Fairbrother (2007) showed that ozone increased the surface oxygen concentration of base char from 12% to 28%. The same study also showed that ozone increased the concentration of most surface oxygen functional groups (hydroxyl, carbonyl, carboxylic acid, lactone, pyrone and esters) thus increasing the surface chemical heterogeneity. (Langley and Fairbrother, 2007)

3.3.2 Oxidized graphite isotherms

To determine the effects of surface oxygen functional groups, adsorption isotherms were determined for atrazine on graphite and oxidized graphite. As expected the atrazine capacity on oxidized graphite was 3 times lower than on graphite because the surface of oxidized graphite was more hydrophilic. The Freundlich parameters $K_f = 0.0148 \text{ [(mg/g)/ (}\mu\text{g/L)}^{1/n}]$ and $1/n = 0.63$ for graphite decreased to $K_f = 0.0051 \text{ [(mg/g)/ (}\mu\text{g/L)}^{1/n}]$ and $1/n = 0.45$ for oxidized graphite.

Isotherm data were collected for 10 organic compounds (trichloroethylene, 1-heptanol, nitrobenzene, estradiol, 2, 4-dinitrotoluene, lindane, atrazine, naphthalene, diazepam, bisphenol A) in both OFW and dissolved SRNOM water. (OFW= Organic free water, SRNOM =

Suwannee river natural organic matter) The Freundlich parameters of 10 compounds in OFW and dissolved SRNOM water were tabulated in Table 2.

The capacity of oxidized graphite decreased in the presence of NOM for most compounds except nitrobenzene. Specifically, for atrazine the Freundlich parameters $K_f = 0.0085 \text{ [(mg/g)/ (}\mu\text{g/L)}^{1/n}]$ and $1/n = 0.74$ for adsorption from pure water decreased to $K_f = 0.0044 \text{ [(mg/g)/ (}\mu\text{g/L)}^{1/n}]$ and $1/n = 0.58$ for adsorption from SRNOM- containing water. The reduction in capacity was also observed by many studies. (Pelekani and Snoeyink, 1999); (Li, 2002); (Knappe et al., 1998); (Ding et al., 2006)

3.3.3 Single solute isotherm fitted pp-LFER

The best-fit pp-LFER equation for predicting single solute $\log K_d$ was determined to be:

$$\log K_{d,i} = [(13.30 \pm 5.83) + (1.96 \pm 1.22) \log a_i]V + [(-9.97 \pm 5.37) - (1.16 \pm 1.11) \log a_i]B + [(1.02 \pm 3.10) - (0.16 \pm 0.64) \log a_i]S + [(0.24 \pm 2.41) + (0.08 \pm 0.51) \log a_i]E + [(-4.12 \pm 3.87) - (0.66 \pm 0.81) \log a_i]A + [(-11.87 \pm 5.20) - (2.34 \pm 1.08) \log a_i] \dots \dots \dots (3.2)$$

Figure 3.1 showed that the predicted single solute $\log K_d$ values were well correlated with the experimental data ($R^2 = 0.97$). The best-fit pp-LFER equation showed that the coefficients of parameter S and parameter E were determined to be negligible. Thus, the amount of solute adsorbed onto oxidized graphite is indirectly proportional to the H-bond accepting nature of the solute (parameter B), directly proportional to the solute's dispersion interactions (parameter V) and slightly proportional to the solute's ability to donate an H-bond. The constant (coefficient C)

of the best-fit pp-LFER equation was significantly different from zero thus showing that the entropy interactions will affect how solutes adsorb on oxidized graphite.

Similar trends were observed by Shih and Gschwend in their work on developing LSERs for activated carbon. (Shih and Gschwend, 2009) They showed that both solute's dispersion interactions (parameter V) and the H-bond accepting nature of the solute (parameter B) were the major factors controlling adsorption of organic chemicals onto activated carbon. This comparison showed that with the presence of pores did not significantly affect the chemical interactions between solutes and the surface of carbon.

The best fit pp-LFER equation for predicting single solute $\log K_d$ was used to predict the single solute $\log K_d$ of the 10th compound that was not used in fitting procedure. Figure 3.1 shows the comparison of the predicted $\log K_d$ values of single solute Bisphenol A and experimental single solute $\log K_d$ values. The predicted single solute $\log K_d$ values for Bisphenol A agree well with the experimental single solute $\log K_d$ values. The best-fit pp-LFER equation predicted K_d values that are about 2 times higher than the experimental K_d .

3.3.4 Natural water isotherm fitted pp-LFER

The best-fit pp-LFER equation for predicting $\log K_d$ in the presence of NOM was determined to be:

$$\begin{aligned} \log K_d = & [(-3.72 \pm 5.86) - (1.39 \pm 1.35) \log a_i]V + [(4.16 \pm 5.21) + (1.63 \pm 1.17) \log a_i]B + \\ & [(-10.45 \pm 3.04) - (2.70 \pm 0.68) \log a_i]S + [(6.21 \pm 2.47) + (1.39 \pm 0.58) \log a_i]E + \\ & [(8.65 \pm 4.15) + (2.10 \pm 0.99) \log a_i]A + [(6.52 \pm 5.18) + (1.45 \pm 1.17) \log a_i] \dots \dots \dots (3.3) \end{aligned}$$

Figure 3.2 showed that the predicted $\log K_d$ values were well correlated with the experimental data with $R^2 = 0.92$.

The best-fit pp-LFER equation for predicting $\log K_d$ in the presence of NOM was used to predict the $\log K_d$ of Bisphenol A in the presence of NOM. The predicted $\log K_d$ values of Bisphenol A were plotted against the experimental $\log K_d$ values in Figure 3.2. Equation 3.3 predicted larger $\log K_d$ values as compared to the experimental $\log K_d$ values. The difference between the predicted $\log K_d$ values and experimental $\log K_d$ increased with increasing $\log a_i$ (activity). Unlike single solute isotherms, activity of trace compounds may not be a good parameter to relate to the amount of solute adsorbed when natural organic matter is present. NOM may increase in the solubility of trace organic compounds when it is present. Some studies demonstrated that in the presence of dissolved organic matter, solubility of hydrophobic organic compounds is enhanced, (Chiou et al., 1987); (Chiou et al., 1986); (Hassett and Anderson, 1979) and it is reasonable that enhanced solubility of some compounds caused the poor correlation.

Since $\log a_i$ was not a good representation of the surface saturation of adsorbates in the presence of NOM, $\log C_e$ values were used instead to get the following best-fit pp-LFER equation for predicting isotherms in the presence of NOM:

$$\log K_{d,i} = [(1.57 \pm 0.63) + (0.36 \pm 0.26) \log C_{e,i}]V + [(-2.49 \pm 0.57) - (0.66 \pm 0.25) \log C_{e,i}]B + [(2.11 \pm 0.25) + (0.30 \pm 0.15) \log C_{e,i}]S + [(0.03 \pm 0.23) + (0.36 \pm 0.10) \log C_{e,i}]E + [(-0.51 \pm 0.25) + (0.04 \pm 0.14) \log C_{e,i}]A + [(-1.27 \pm 0.42) - (1.60 \pm 0.23) \log C_{e,i}] \dots \dots \dots (3.4)$$

Figure 3.3 shows that the predicted $\log K_d$ values were well correlated with the experimental data with $R^2 = 0.99$. The best-fit pp-LFER equation showed that all coefficients were not negligible.

Nguyen et al (Nguyen et al., 2005) obtained pp-LFERs for the equilibrium partition of organic

compounds between water and natural organic matter in soils and sediments. Their pp-LFERs also showed that coefficients of the best fit equation were all statistically significant. Thus in the presence of natural organic matter, the chemical interactions of organic contaminants and carbon surface is affected by all molecular parameters.

The experimental isotherm in the presence of NOM of the 10th compound was used to validate the best-fit pp-LFER equation. The best-fit pp-LFER equation for predicting $\log K_d$ using $\log C_e$ was used to predict the $\log K_d$ of lindane, and these values are compared to the experimental values in Figure 3.3. Equation 3.4 predicted K_d values that are lower than the experimental K_d values by a factor of 1.17. $\log C_e$ values can better represent the solute concentration effects on adsorption because the equilibrium concentration already included the enhanced solubility effect of the trace compounds in the presence of NOM.

The validity of this approach was tested by withholding 1 out of 10 compounds for the pp-LFER fitting procedure for all 10 compounds. The predicted K_d values of the 10th compound were compared with experimental values and the difference was summarized in Table 3. Among the 10 compounds, prediction for lindane and trichloroethylene is the best with the ratios between predicted and experimental values being 1.17 and 1.16, respectively. The prediction for bisphenol A and estradiol were worse. For estradiol and bisphenol A the best-fit equation predicted NOM K_d values that are about 3.84 and 3.77 times lower than the experimental NOM K_d . Figure 3.4 shows the predicted NOM $\log K_d$ value for bisphenol A.

The correlation results showed that it is important to select a set of compounds that cover a wide range of molecular parameters to allow the predictive equation to be more robust. As shown in Table 1, the ranges represented are E from 0.21 to 2.08, S from 0.37 to 1.77, A from 0 to 0.99, B from 0.03 to 1.25, V from 0.71 to 2.19 and aqueous solubility from 0.26 to 1949 mg/L. But the fitted equation will not be able to make good predictions of compounds that have $\log K_d$ values that are beyond the fitting set. Examples were bisphenol A and estradiol where the experimental K_d values were beyond those used in the fitting set and the difference between predicted values and experimental values is about factor of 3.

3.3.5 Freundlich parameters from K_d values

Freundlich parameters are commonly used in models that predict performance of activated carbon in water treatment systems. Adsorption isotherms are usually fitted with the Freundlich equation as follows:

$$q_e = K_f C_e^{\frac{1}{n}} \dots\dots\dots(3.5)$$

The Freundlich equation can also be log-transformed to give the following equation:

$$\log q_{e,i} = \log K_{f,i} + \frac{1}{n} \log C_{e,i} \dots\dots\dots(3.6)$$

For single solute isotherms:

$$\log a_i = \frac{\log C_{e,i}}{\log S_{w,i}} \dots\dots\dots(3.7)$$

$$\log K_{d,i} = \frac{\log q_{e,i}}{\log C_{e,i}} \dots\dots\dots(3.8)$$

Using equations 3.2, 3.6, 3.7 and 3.8, we can attain single solute K_f and $1/n$ values using the molecular descriptors and solubility concentration of each compound:

$$\log K_{f,i} = (13.30V - 9.97B + 1.02S + 0.24E - 4.12A - 11.87) - (1.96V - 1.16B - 0.16S + 0.08E - 0.66A - 2.34) \log S_{w,i} \dots\dots\dots(3.9)$$

$$\frac{1}{n} = 1.96V - 1.16B - 0.16S + 0.08E - 0.66A - 1.34 \dots\dots\dots(3.10)$$

For NOM isotherms:

Using equations 3.4, 3.6 and 3.8, we can obtain the NOM K_f and $1/n$ using the molecular descriptors of each compound:

$$\text{Log} K_{f,i} = 1.57V - 2.49B + 2.11S + 0.03E - 0.51A - 1.27 \dots\dots\dots(3.11)$$

$$\frac{1}{n} = 0.36V - 0.66B + 0.30S + 0.36E + 0.04A - 0.60 \dots\dots\dots(3.12)$$

Thus, a correlation for K_d can be used to estimate Freundlich parameters (K_f and $1/n$).

3.4 Conclusion

- A pp-LFER equation has been developed for trace organic compounds adsorption on oxidized graphite in pure water and its able to predict K_d values that were about a factor of 2 different from the experimental K_d values.
- A pp-LFER equation has been developed for trace organic compounds adsorption on oxidized graphite in water containing NOM and its able to predict K_d values that were about a factor of 1.17 different from the experimental K_d values.

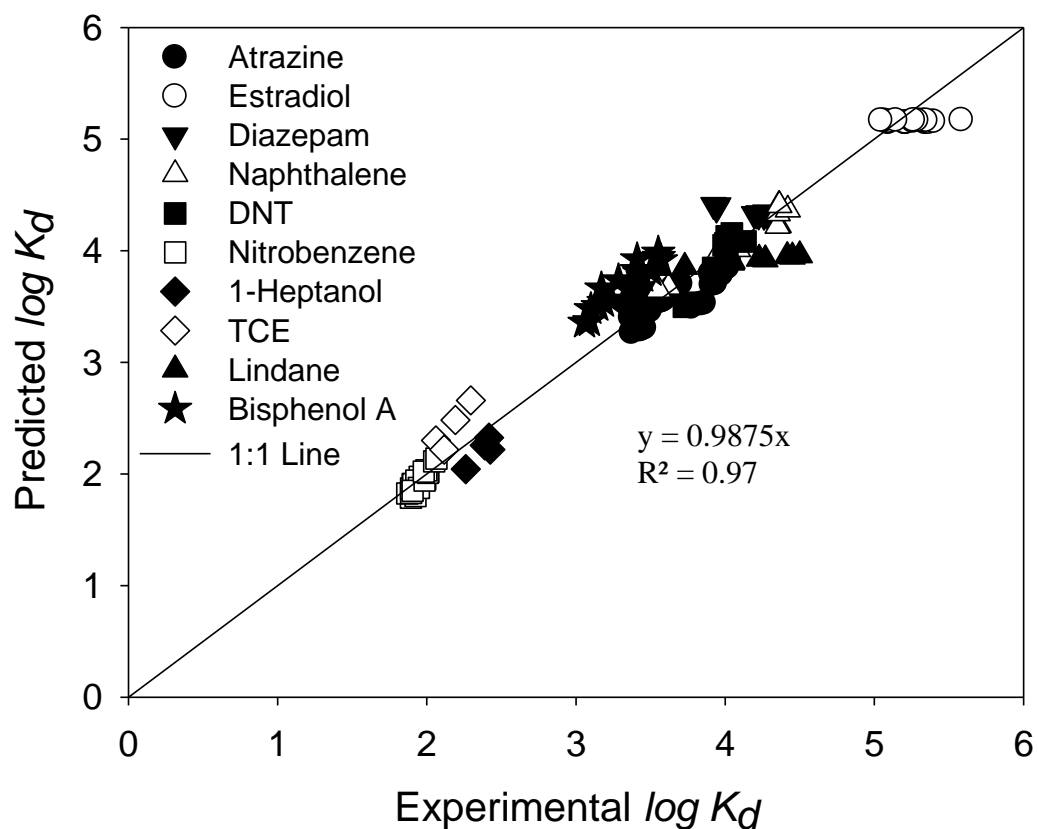


Figure 3.1: Plot of fitted $\log K_d$ from the correlation developed for single solute isotherm data against the experimental $\log K_d$ of 9 organic compounds. Single solute bisphenol A $\log K_d$ values were predicted and compared with the experimental $\log K_d$ on this plot marked as stars.

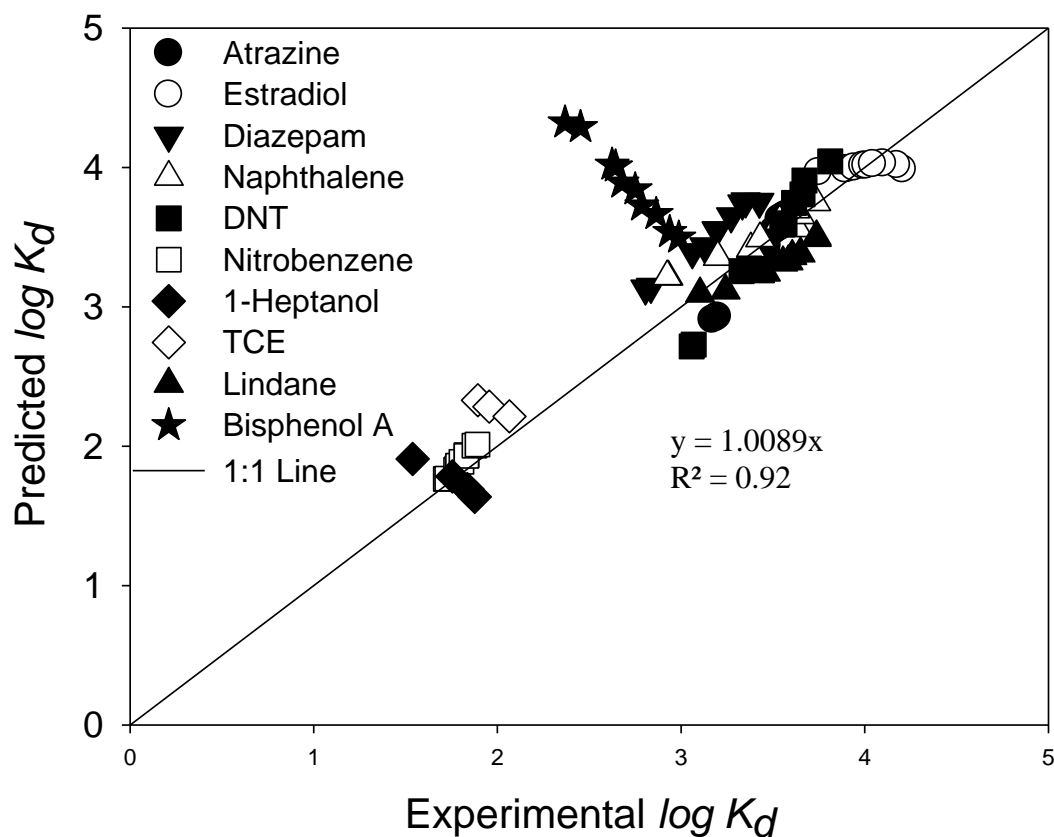


Figure 3.2: Plot of fitted $\log K_d$ from the correlation developed for NOM isotherm data against the experimental $\log K_d$ of 9 organic compounds using $\log a$ values. NOM bisphenol A $\log K_d$ values were predicted and compared with the experimental $\log K_d$ on this plot marked as stars.

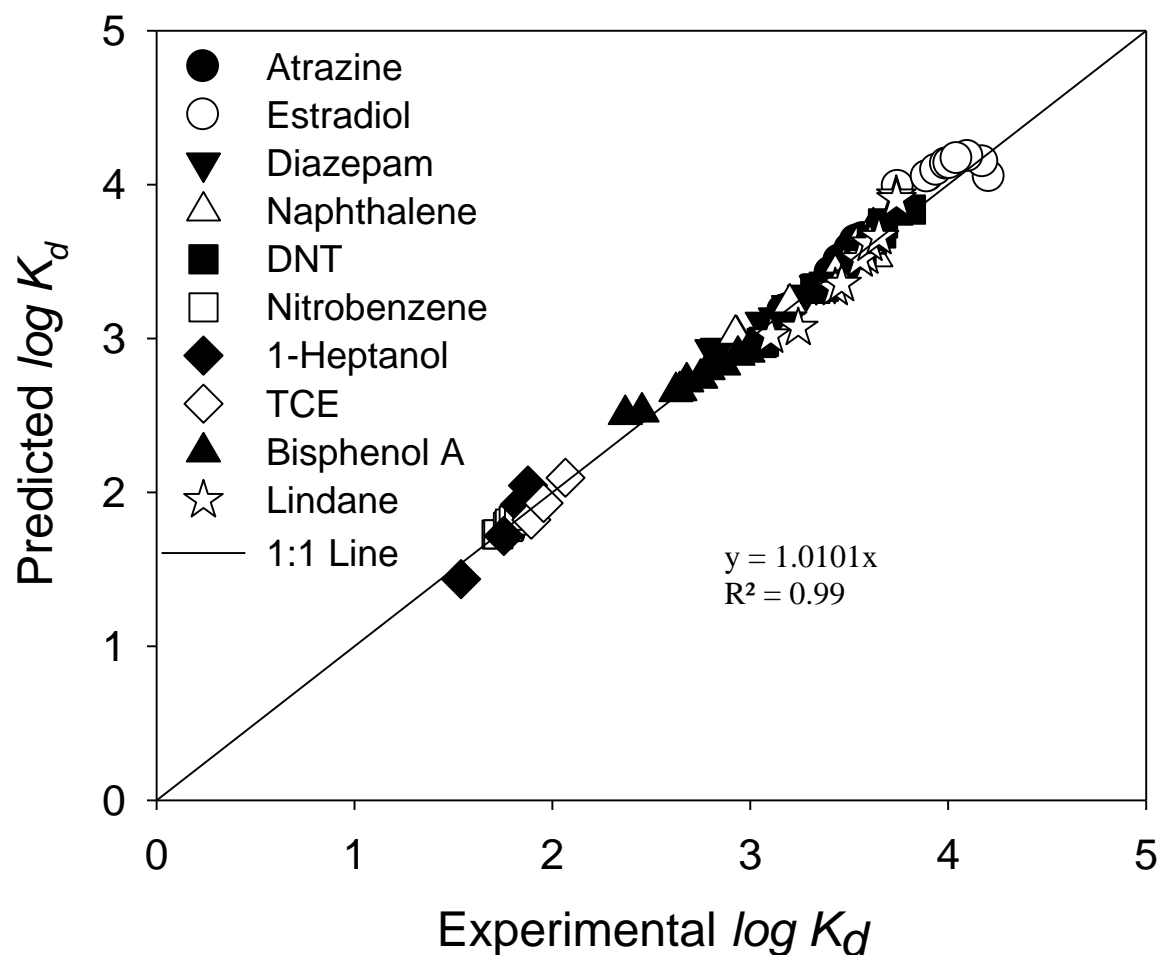


Figure 3.3: Plot of fitted $\log K_d$ from the correlation developed for NOM isotherm data against the experimental $\log K_d$ of 9 organic compounds using $\log C_e$ values. NOM lindane $\log K_d$ values were predicted and compared with the experimental $\log K_d$ on this plot marked as stars.

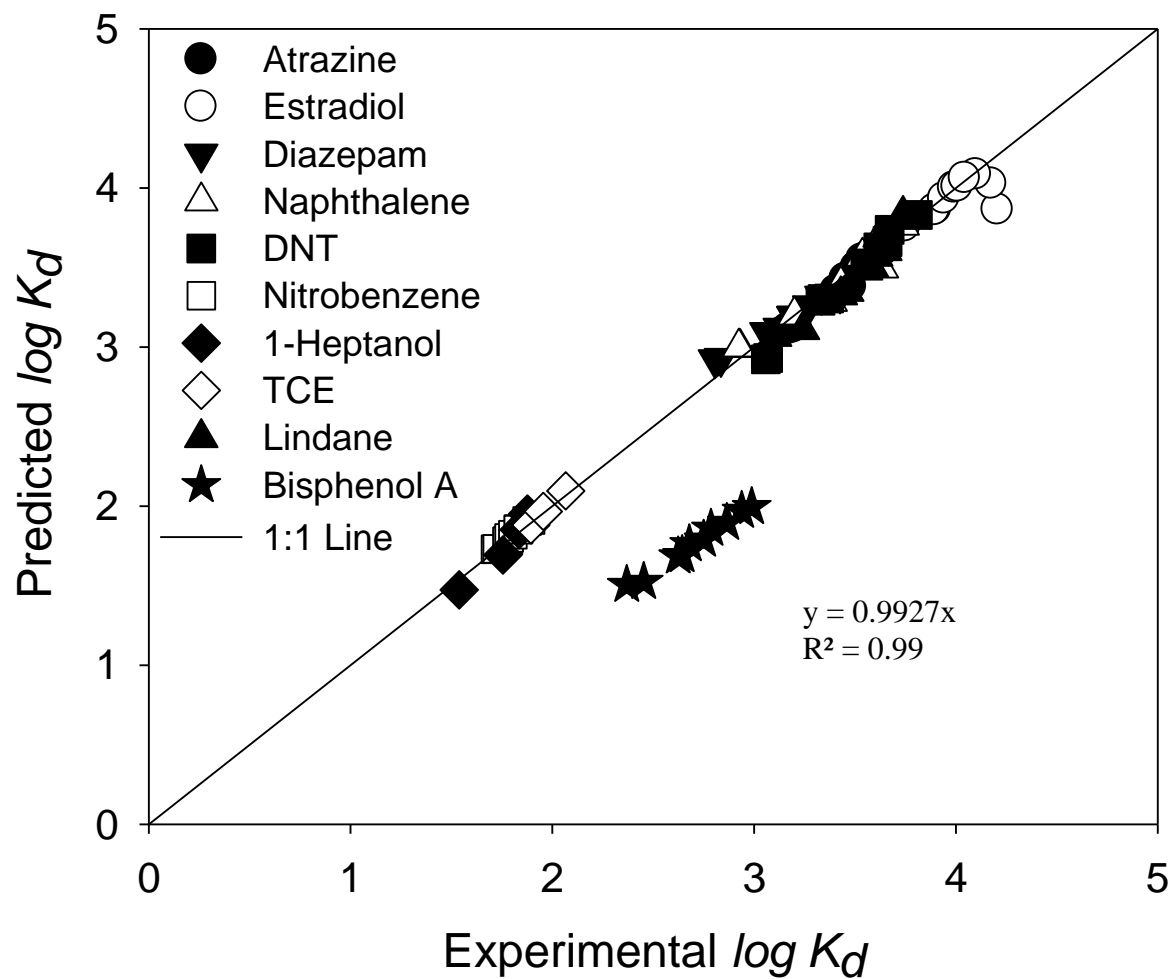


Figure 3.4: Plot of fitted $\log K_d$ from the correlation developed for NOM isotherm data against the experimental $\log K_d$ of 9 organic compounds using $\log C_e$ values. NOM bisphenol A $\log K_d$ values were predicted and compared with the experimental $\log K_d$ on this plot marked as stars.

Tables:

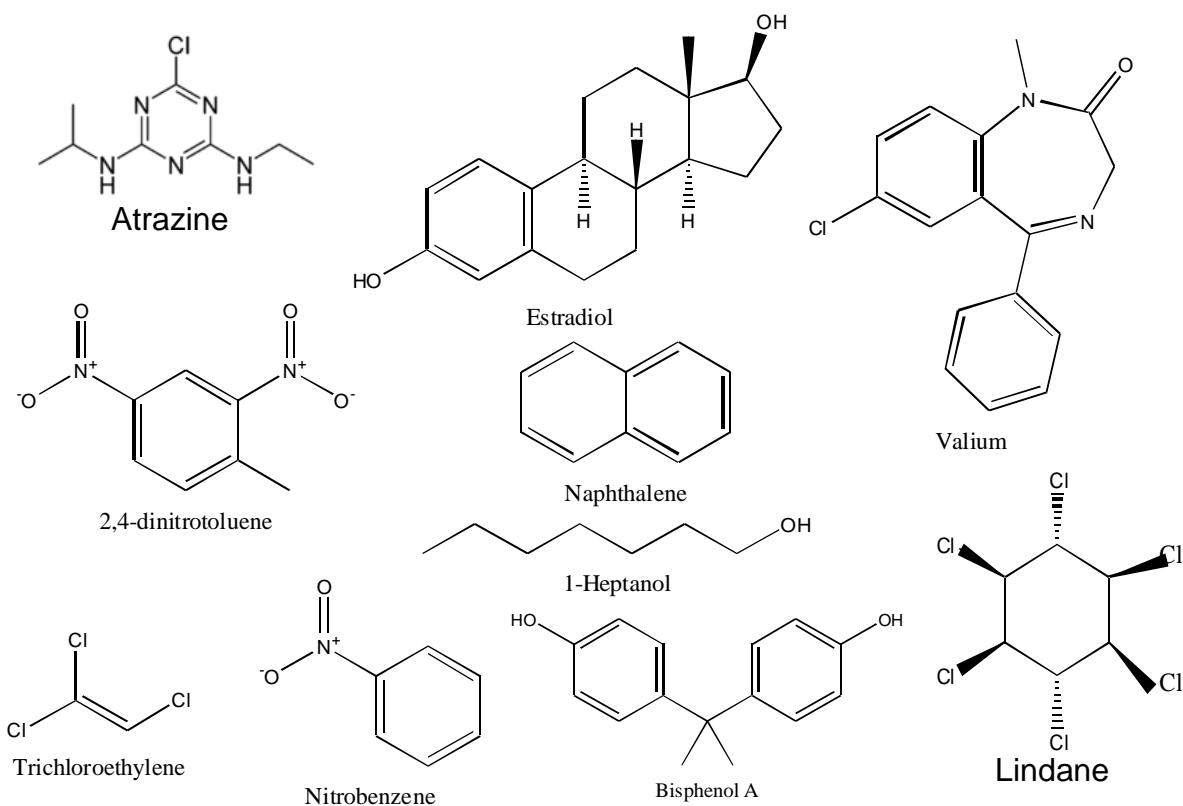


Table 3.1: Aqueous solubility (Yaw, 1999); (ADME/ Tox WEB) and molecular properties (Abraham et al., 1994a); (Abraham et al., 1994b) (ADME/ Tox WEB) of compounds used for fitting of both single solute and $\log K_d$ in the presence of NOM predictive equations

No	Compound	E	S	A	B	V_x	Aq. Solubility
							(mg/L)
1	Atrazine	1.95	1.64	0	1.04	1.61	70
2	Estradiol	1.80	1.77	0.86	1.10	2.19	0.26
3	Diazepam	2.08	1.55	0.09	1.25	2.07	50
4	Naphthalene	1.34	0.92	0	0.20	1.08	30
5	2,4-dinitrotoluene	1.17	1.27	0.07	0.51	1.20	270
6	Lindane	1.45	0.91	0	0.68	1.57	7.31
7	Nitrobenzene	0.87	1.11	0	0.28	0.89	1949
8	Trichloroethylene	0.52	0.37	0.08	0.03	0.71	1100
9	1-Heptanol	0.21	0.42	0.37	0.48	1.15	1740
10	Bisphenol A	1.61	1.56	0.99	0.94	1.86	300

Table 3.2: Experimental Freundlich isotherm parameters of 10 compounds

Compound	Single solute K_f	Single solute $1/n$	NOM K_f	NOM $1/n$
Atrazine	1416	0.74	250	0.58
Estradiol	101434	0.95	177	0.56
Diazepam	820	0.60	67	0.40
Naphthalene	245	0.45	70	0.40
2,4 dinitrotoluene	1170	0.63	185	0.41
Nitrobenzene	87	0.47	62	0.28
1-Heptanol	32	0.48	6	0.28
Trichloroethylene	25	0.52	12	0.32
Lindane	107	0.33	64	0.41
Bisphenol A	442	0.58	80	0.34

Table 3.3: Ratio of predicted K_d values and experimental K_d values of the 10th compound (not used in the fitting procedure)

Compounds	SS Predicted/ Experimental K_d	NOM Predicted K_d / Experimental K_d
Atrazine	8.42	1.33
Estradiol	4.33	3.84
Diazepam	12.60	1.50
Naphthalene	2.83	2.12
2,4-DNT	4.30	2.81
Lindane	5.02	1.17
Nitrobenzene	11.43	2.58
Trichloroethylene	1.84	1.16
1-Heptanol	4.25	1.54
Bisphenol A	2.50	3.77

REFERENCES

Abraham, M. H., Andonianhaftvan, J., Whiting, G. S., Leo, A., Taft, R. S. (1994a) Hydrogen-Bonding .34. the Factors that Influence the Solubility of Gases and Vapors in Water at 298-K, and a New Method for its Determination. J. Chem. Soc. -Perkin Trans. 2(8), 1777-1791.

Abraham, M. H., Chadha, H. S., Whiting, G. S., Mitchell, R. C. (1994b) Hydrogen-Bonding .32. an Analysis of Water-Octanol and Water-Alkane Partitioning and the Delta-Log-P Parameter of Seiler. J. Pharm. Sci. **83**(8), 1085-1100.

Chiou, C. T., Kile, D. E., Brinton, T. I., Malcolm, R. L., Leenheer, J. A., Maccarthy, P. (1987) A Comparison of Water Solubility Enhancements of Organic Solutes by Aquatic Humic Materials and Commercial Humic Acids. Environ. Sci. Technol. 21(12), 1231-1234.

Chiou, C. T., Malcolm, R. L., Brinton, T. I., Kile, D. E. (1986) Water Solubility Enhancement of some Organic Pollutants and Pesticides by Dissolved Humic and Fulvic-Acids. Environ. Sci. Technol. 20(5), 502-508.

Crittenden, J. C., Sanongraj, S., Bulloch, J. L., Hand, D. W., Rogers, T. N., Speth, T. F., Ulmer, M. (1999) Correlation of aqueous-phase adsorption isotherms. Environ. Sci. Technol. 33(17), 2926-2933.

Ding, L., Marinas, B. J., Schideman, L. C., Snoeyink, V. L. (2006) Competitive effects of natural organic matter: Parametrization and verification of the three-component adsorption model COMPSORB. Environ. Sci. Technol. 40(1), 350-356.

Hassett, J. P. and Anderson, M. A. (1979) Association of Hydrophobic Organic-Compounds with Dissolved Organic-Matter in Aquatic Systems. *Environ. Sci. Technol.* 13(12), 1526-1529.

Knappe, D. R. U., Matsui, Y., Snoeyink, V. L., Roche, P., Prados, M. J., Bourbigot, M. -. (1998) Predicting the Capacity of Powdered Activated Carbon for Trace Organic Compounds in Natural Waters. *Environ. Sci. Technol.* 32(11), 1694-1698.

Langley, L. A. and Fairbrother, D. H. (2007) Effect of wet chemical treatments on the distribution of surface oxides on carbonaceous materials. *Carbon* 45(1), 47-54.

Li, L., Quinlivan, P. A., Knappe, D. R. U. (2005) Predicting adsorption isotherms for aqueous organic micropollutants from activated carbon and pollutant properties. *Environ. Sci. Technol.* 39(9), 3393-3400.

Li, Q. (2002) Competitive adsorption of trace organic compounds by PAC in membrane filtration systems. Ph.D., University of Illinois at Urbana-Champaign, United States -- Illinois,

Pelekani, C. and Snoeyink, V. L. (1999) Competitive adsorption in natural water: Role of activated carbon pore size. *Water Res.* 33(5), 1209-1219.

Randtke, S. J. and Snoeyink, V. L. (1983) Evaluating Gac Adsorptive Capacity. *J. Am. Water Work Assoc.* 75(8), 406-413.

Shih, Y. and Gschwend, P. M. (2009) Evaluating Activated Carbon-Water Sorption Coefficients of Organic Compounds Using a Linear Solvation Energy Relationship Approach and Sorbate Chemical Activities. *Environmental Science & Technology* 43(3), 857.

Yaw, C.L. (1999) Chemical Properties Handbook: Physical, Thermodynamic, Environmental, Transport, Safety, and Health Related Properties for Organic and Inorganic Chemicals, McGraw-Hill: New York

<http://www.pharma-algorithms.com/webboxes/>

Chapter 4: Application of Oxidized Graphite Predictive Equations

4.1 pp-LFER Equation Development

In Chapter 3, nine compounds were chosen to represent several different intermolecular interactions with oxidized graphite. The LFER parameters and aqueous solubilities of these 9 compounds covered a wide range of values as shown in Table 1 of Chapter 3; the ranges represented for E are from 0.21 to 2.08, for S from 0.37 to 1.77, for A from 0 to 0.86, for B from 0.03 to 1.25 and for V from 0.71 to 2.19.

However, in order to efficiently develop pp-LFERs that are able to predict good isotherms, careful selection of organic compounds is essential. The linear free energy parameters values for atrazine, estradiol, diazepam, nitrobenzene, trichloroethylene and 1-heptanol cover the ranges of all the parameters for the 9 compounds in this study. Estradiol was the compound with the highest $V = 2.19$, $S = 1.77$ and $A = 0.86$. Diazepam was the compound with the highest $E = 2.08$ and $B = 1.25$. Trichloroethylene was the compound with the lowest $V = 0.71$, $B = 0.03$ and $S = 0.37$. 1-Heptanol was the compound with the lowest $E = 0.21$ and atrazine was the compound with the lowest $A = 0$. Estradiol had the lowest aqueous solubility = 0.26 mg/L and nitrobenzene had the highest solubility = 1949 mg/L. In Figure 4.1, it was observed that by choosing these 6 compounds, the predicted K_d values for single solute isotherms agree well with the experimental data with $R^2 = 0.99$. The equation for predicting single solute isotherms on oxidized graphite using only 6 compounds is:

$$\log K_{d,i} = [(-4.68 \pm 3.79) + (-1.00 \pm 0.88) \log a_i]V + [(2.88 \pm 5.00) + (0.66 \pm 1.10) \log a_i]B + \\ [(-0.49 \pm 2.51) + (-0.20 \pm 0.62) \log a_i]S + [(3.66 \pm 3.15) + (0.55 \pm 0.71) \log a_i]E + \\ [(6.41 \pm 2.63) + (1.00 \pm 0.64) \log a_i]A + [(0.92 \pm 2.91) + (-0.07 \pm 0.64) \log a_i] \dots \dots \dots (4.1)$$

However, when the single solute pp-LFER equation was used to predict adsorption isotherms of naphthalene, 2, 4-dinitrotoluene, bisphenol A and lindane, the K_d predictions were within a factor of 120, or within a 2 log factor, different from the experimental values. Although the parameters of these 4 compounds were within the ranges covered by the 6 compounds, the types of chemical interactions that determine single solute isotherms apparently are more complex than what these 6 compounds can represent. One factor contributing to the complexity of the system is oxidative coupling of organic compounds on the surface of activated carbon. Vidic et al. showed that in the presence of oxygen, adsorption of phenolic compounds on activated carbon surface was higher than what was observed when oxygen is absence in the system (Vidic et al., 1997); (Vidic and Suidan, 1991); (Tessmer et al., 1997). They determined that this phenomenon was caused by polymerization of trace organic compounds on the activated carbon surface. Thus in the case of single solute isotherm predictions, the chemical interactions were more complex than what the range of molecular parameters can represent.

For isotherms in the presence of NOM, using the same 6 compounds, Figure 4.2 showed that the predicted K_d values agree well with the experimental data with $R^2 = 0.99$. When the pp-LFER equation was used to predict K_d values of naphthalene, 2,4-dinitrotoluene and lindane, the predictions were within a factor of 2, or within 0.3 log factor, of the experimental values. The equation for predicting isotherms in the presence of NOM using only 6 compounds is:

$$\log K_{d,i} = [(-2.02 \pm 0.91) + (-1.24 \pm 0.78) \log C_{e,i}]V + [(1.05 \pm 0.90) + (0.96 \pm 0.82) \log C_{e,i}]B + \\ [(0.85 \pm 0.68) + (-0.25 \pm 0.73) \log C_{e,i}]S + [(0.76 \pm 0.81) + (0.56 \pm 0.74) \log C_{e,i}]E + \\ [(1.11 \pm 0.82) + (0.88 \pm 0.69) \log C_{e,i}]A + [(1.67 \pm 0.68) + (-0.08 \pm 0.55) \log C_{e,i}] \dots \dots \dots (4.2)$$

To test whether all the range of molecular parameters are needed to predict isotherms in the presence of NOM; one more compound was left out for pp-LFER equation fitting. In Figure 4.3 it was shown that when estradiol was left out, the K_d predictions for naphthalene, 2, 4-dinitrotoluene, bisphenol A, lindane and estradiol were factors of 1.05 to 49.96 different from the experimental values. Estradiol is the largest compound in our data set; it also has the lowest solubility and the highest K_d value. So without it in the data set, certain chemical interactions will not be well represented and thus the big discrepancies between the predicted values and the experimental values. Therefore we can conclude that to develop good pp-LFERs for isotherms in the presence of NOM, 6 compounds with values of LFER parameters and solubility that cover a range that includes the parameters of the compounds for which isotherms are to be predicted is required.

4.2 Prediction of Isotherm Data on Activated Carbon Using the Correlation Developed for Oxidized Graphite

In Chapter 4 we determined the effects of chemical interactions on adsorption of trace compounds on oxidized graphite in the presence and absence of natural organic matter.

However, to be able to predict activated carbon adsorption isotherms using the oxidized graphite equations, we will need to consider the higher surface area and pore size distribution of activated carbon.

Activated carbon being very porous has a higher surface area compared to oxidized graphite. It can be expected that activated carbon with a larger surface area will have a higher adsorption

capacity compared to oxidized graphite. Most activated carbons also have a range of pore sizes and this can affect the way trace organic contaminants adsorb. The effects of pore size distribution on the adsorption of trace compounds on activated carbon were well documented in the literature (Ebie et al., 2001); (Pelekani and Snoeyink, 1999); (Pelekani and Snoeyink, 2000); (Ding et al., 2008).

A preliminary evaluation of the application of the oxidized graphite pp-LFER equation for activated carbon predictions was carried out using activated carbon isotherms found in literature. These isotherms were compared to the predictions made using the oxidized graphite pp-LFER equation for the same compound. The difference in surface area of the activated carbon in comparison to that of the oxidized graphite was taken into account but the pore size distribution difference was not.

Atrazine in the absence and presence of NOM isotherm data can be found in work by To et al. and Ding et al. (To et al., 2008); (Ding et al., 2006). Both To et al. and Ding et al. used a commercial PAC, Norit SA-UF (NORIT France, S.a.r.l., Le Blanc Mesnil Cedex, France) with a nitrogen BET surface area of $1112 \text{ m}^2/\text{g}$ (Li et al., 2003) and an oxygen content that was measured to be 7.21 mass percent. To et al. also made use of SRNOM, which was obtained from the same source as the material used in this study and thus is thought to be similar to what was used in this study. Ding et al. used two different sources of natural water, one being the groundwater from Clinton Water Works (CWW), Clinton, IL and the other being the groundwater from a well located next to Newmark Civil Engineering Laboratory (NCEL) at the

University of Illinois. Relevant properties of the NOM in these waters can be found in the respective papers.

Trichloroethylene activated carbon isotherm data in the absence and presence of NOM can be found in work by Kilduff and Karanfil (Kilduff and Karanfil, 2002). Kilduff and Karanfil used a microporous coal base activated carbon, F400 from Calgon, Pittsburg, PA. The surface area of F400 was $948 \text{ m}^2/\text{g}$ and the oxygen content was 7.3% mass concentration (Karanfil and Kilduff, 1999); (Yue et al., 2006). Kilduff and Karanfil used a poly(maleic acid) (PMA) and the properties of this NOM component can be found elsewhere (Carter et al., 1992). The isotherm data for the two compounds were used to compare to the predicted values from the pp-LFER oxidized graphite equations.

2-Methylisoborneol (MIB) isotherm data in the absence and presence of NOM can be found in work by Newcombe et al. (Newcombe et al., 2002); (Newcombe et al., 2002). This study used several commercially available activated carbons, one of which was F400, which is the same as what was used in the TCE study. Newcombe et al. used Myponga reservoir water and the NOM was isolated into several different groups based on molecular weight. Extensive characterization of the raw water and groups of NOM isolates were carried out. MIB in the presence of NOM isotherm data can also be found in work by Cook et al. (Cook et al., 2001). Their study used a coconut based, steam activated carbon, Picatif 1100 PAC (PICA) with a surface area of $1241 \text{ m}^2/\text{g}$. Cook et al. used four different sources of natural water and the NOM was well characterized.

Figure 4.4 shows that by normalizing the K_d values of single solute activated carbon atrazine isotherm data with the surface area of the carbon, the experimental K_d values are within a factor of 3 compared to the predicted K_d values obtained from the oxidized graphite (OG) single solute predictive equation, which are also normalized with the surface area of oxidized graphite. Figure 4.4 also shows that the normalized experimental single solute activated carbon trichloroethylene K_d values are within a factor of 2 to 5 compared to the normalized K_d values predicted by the OG single solute predictive equation. It is also shown in Figure 4.4 that the normalized K_d values predicted by the OG single solute predictive equation of MIB are within an average factor of 2.5 compared with the normalized experimental K_d values.

These good predictions may be due to the similarities in the chemical interactions on oxidized graphite and activated carbon. Both SA-UF and F400 carbons have oxygen concentration of about 7% mass concentration and oxidized graphite has oxygen concentration of about 5%. This oxygen level is comparable to the oxygen concentration level on the oxidized graphite used in this study; thus this might be the reason why normalized predictions of adsorption isotherms agree well with the experimental isotherms. However, for carbons with higher oxygen concentration levels, the predictions might not agree as well because the chemical interactions of compounds with a highly oxidized surface may be very different compared to the oxidized graphite used in this study.

Pore size distribution data of SA-UF and F400 can be found in work of Ding et al. and Karanfil and Kilduff (Ding, 2010); (Karanfil and Kilduff, 1999). The surface areas and pore sizes of activated carbon were measured by nitrogen adsorption isotherms. The molecular size of

nitrogen was reported by Quirke et al. to be 3.5-3.7 Å (Quirke and Tennison, 1996). The molecular sizes of atrazine and trichloroethylene were reported by Guo et al. to be 8-20 Å and 5-8 Å (Guo et al., 2007). The smallest molecular size cutoff for pore size distribution measurements was 15-20 Å; the smallest pores that nitrogen, atrazine and trichloroethylene can adsorb on cannot be differentiated by current pore size measurements.

Since the pore size distribution cannot be further differentiated for the three compounds that was used in this study, future work will have to be done in order to determine if the good predictions are due to the similarities in chemical interactions between oxidized graphite and activated carbon or it is a compensating effects of chemical interactions and pore size distribution effects. One important issue to note is the predictions have worked well for compounds that are slightly larger than nitrogen. For compounds that are much larger than nitrogen, the pp-LFER equation of oxidized graphite might predict K_d values that are much greater than the experimental K_d . That is because some of the surface area accessible to nitrogen might not be accessible to the larger target organic contaminants.

Figure 4.5 shows that the predictive equation for isotherms in the presence of NOM determined for oxidized graphite could be used to predict atrazine, trichloroethylene and MIB isotherms of activated carbon in the presence of NOM by dividing K_d values with the surface area of the carbon. The normalized K_d values of the activated carbon atrazine isotherm in the presence of NOM are within a factor of 2 compared to the normalized K_d values predicted by the OG pp-LFER equation. For trichloroethylene, the adsorption isotherms predicted by the OG pp-LFER

equation were within a factor of 3 as compared to the experimental adsorption isotherms of activated carbon in the presence of NOM.

However for MIB, in the presence of NOM, the normalized K_d values predicted by the OG pp-LFER were about 1268 times higher than the normalized experimental K_d values. The type of NOM used in the MIB studies may explain these big discrepancies. The NOM used in the MIB studies consist of a substantial percentage of small molecular weight NOM that is similar in size as MIB. Thus, there might be greater competition from the NOM for the same pores and thus caused less adsorption of MIB as compared to what was predicted by the OG pp-LFER equation. These discrepancies demonstrate that the type of pores available for adsorption and the type of NOM present in the water are important variables. By considering just the surface area difference between oxidized graphite and activated carbon is insufficient to predict for a different carbon and a different natural water source.

From what was observed for atrazine, trichloroethylene and 2-methylisoborneol, it can be concluded that pore size distribution of activated carbon and type of NOM are important factors to consider for predicting good isotherm data with the pp-LFER approach. Therefore, a pp-LFER for one activated carbon and one type of NOM will be developed in the next chapter.

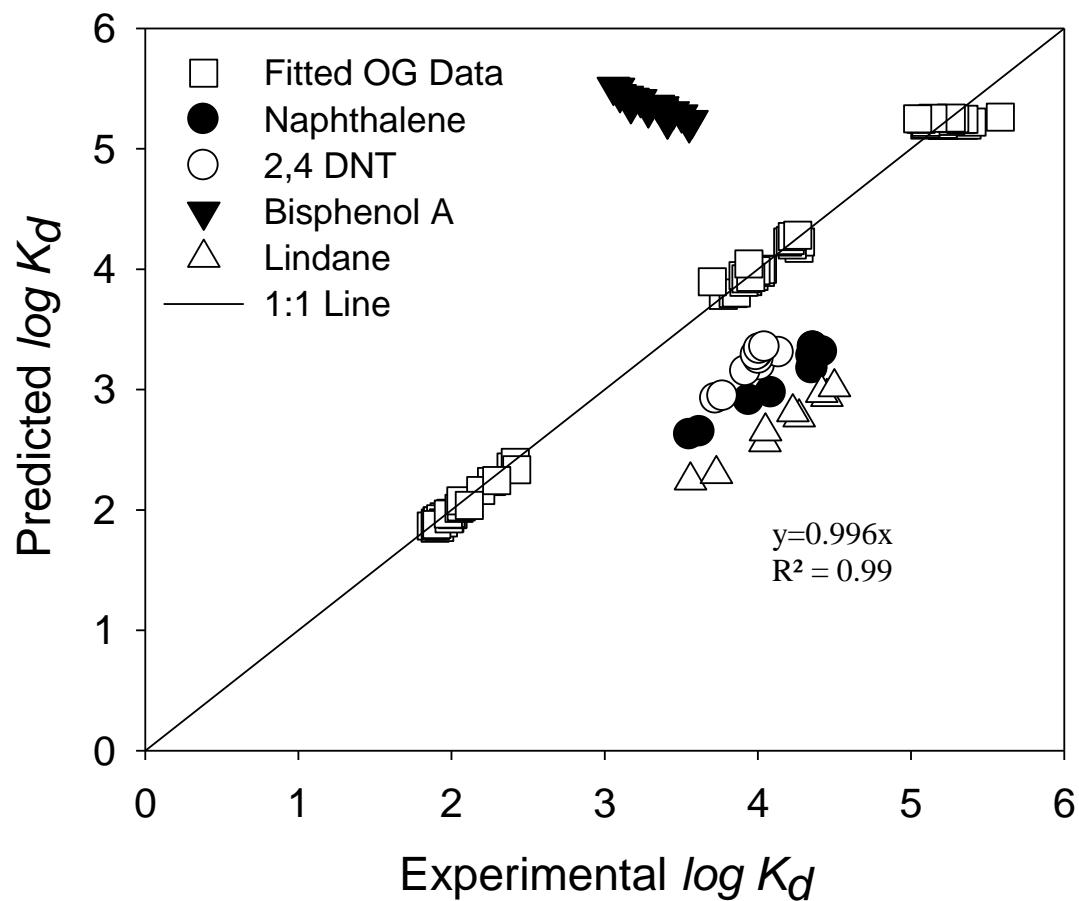


Figure 4.1: Plot of fitted $\log K_d$ from the correlation developed for single solute isotherm data against the experimental $\log K_d$ of 6 organic compounds. Single solute Naphthalene, 2, 4 DNT, Bisphenol A and Lindane $\log K_d$ values were predicted and compared with the experimental $\log K_d$ on this plot.

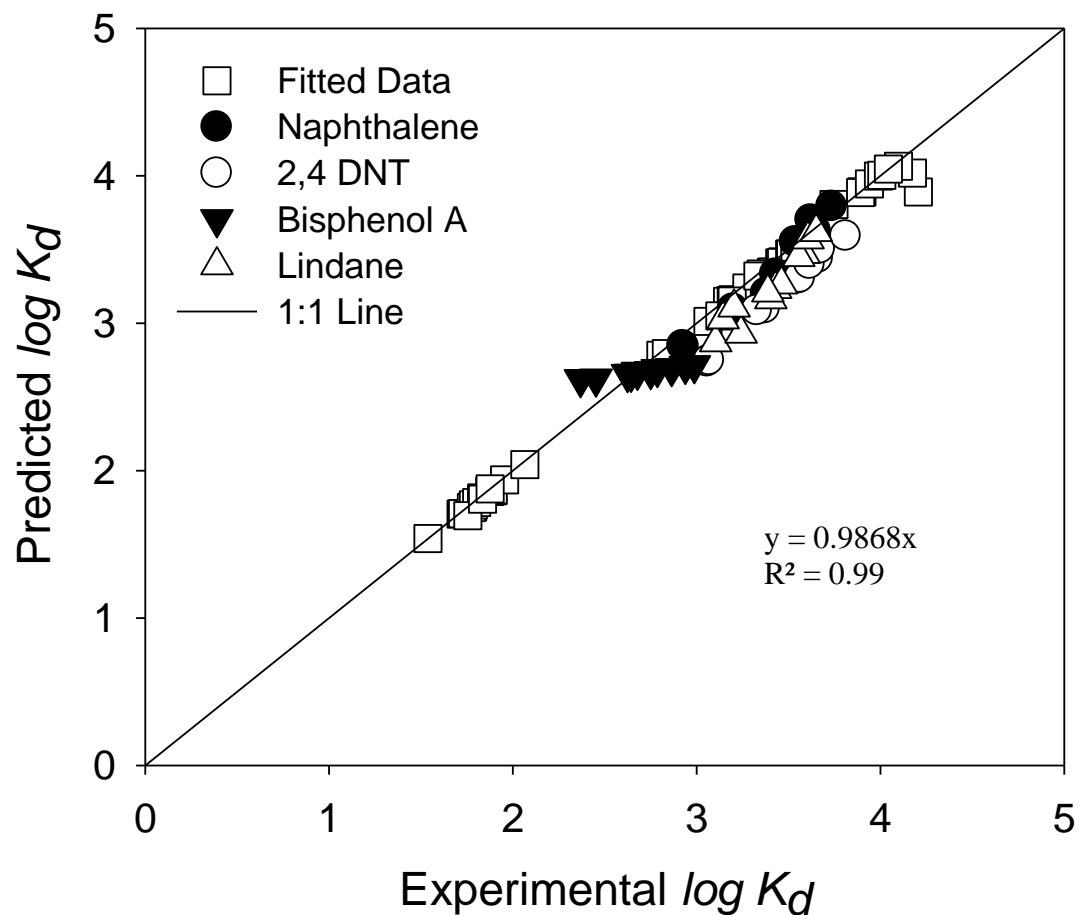


Figure 4.2: Plot of fitted $\log K_d$ from the correlation developed for isotherms in the presence of NOM against the experimental $\log K_d$ of 6 organic compounds. Naphthalene, 2, 4 DNT, Bisphenol A and Lindane $\log K_d$ values in the presence of NOM were predicted and compared with the experimental $\log K_d$ on this plot.

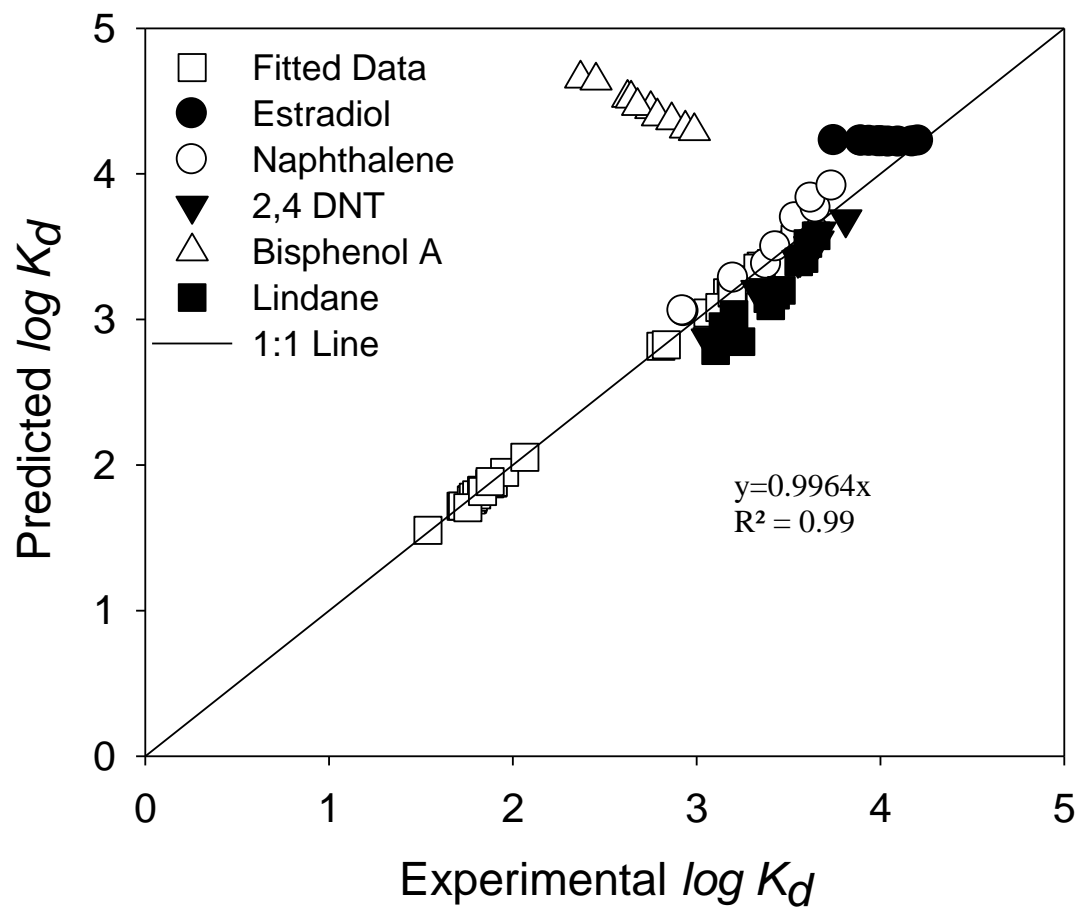


Figure 4.3: Plot of fitted $\log K_d$ from the correlation developed for isotherms in the presence of NOM against the experimental $\log K_d$ of 5 organic compounds. Estradiol, naphthalene, 2, 4 DNT, bisphenol A and lindane $\log K_d$ values in the presence of NOM were predicted and compared with the experimental $\log K_d$ on this plot.

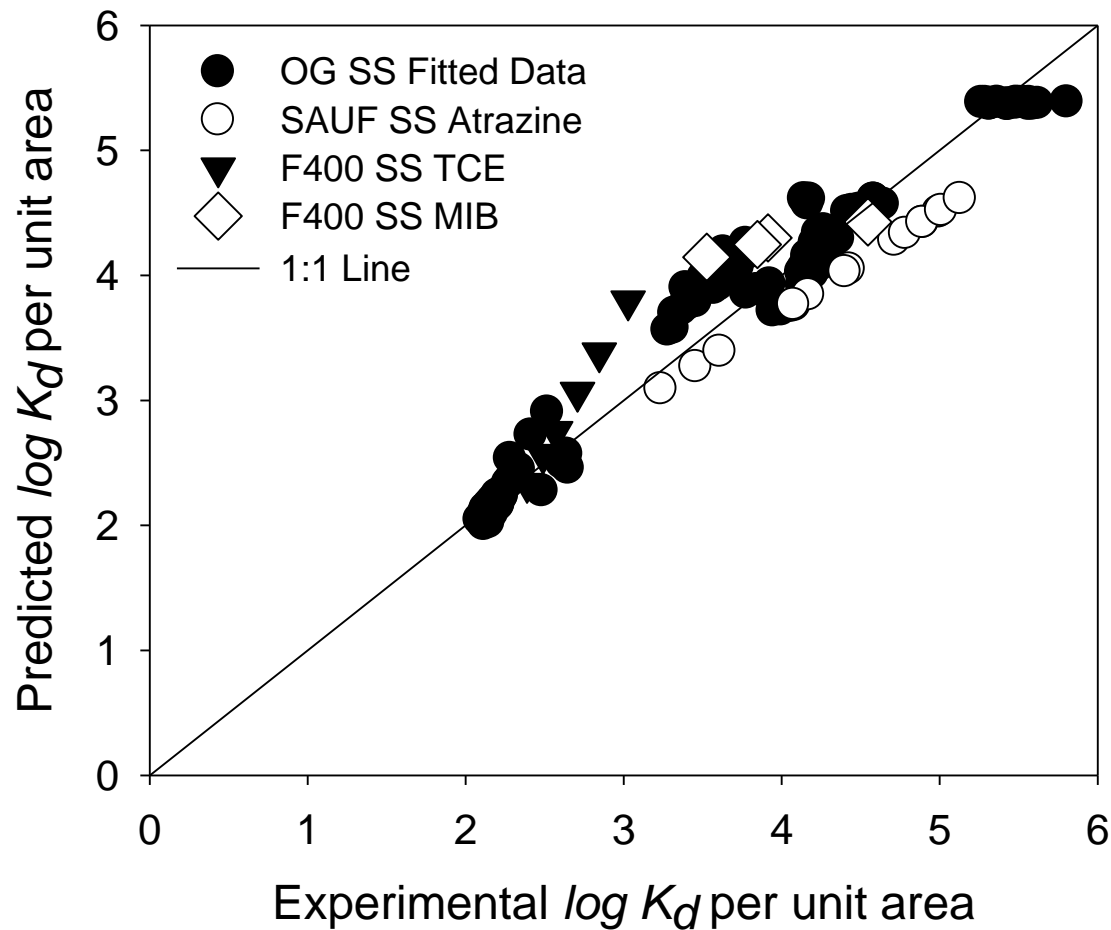


Figure 4.4: Plot of predicted single solute (SS) $\log K_d$ values (per unit area) using the determined equation for oxidized graphite (OG) vs experimental SS $\log K_d$ values (per unit area) of atrazine on SA-UF activated carbon (SAUF), trichloroethylene on F400 activated carbon and MIB on F400 activated carbon. The predicted SS $\log K_d$ values were normalized with the surface area of OG ($0.6\text{m}^2/\text{g}$) while the experimental SS $\log K_d$ values were normalized with the surface area of SAUF carbon ($1112\text{ m}^2/\text{g}$) and F400 carbon ($948\text{ m}^2/\text{g}$) (To et al., 2008); (Ding et al., 2006); (Newcombe et al., 2002); (Newcombe et al., 2002).

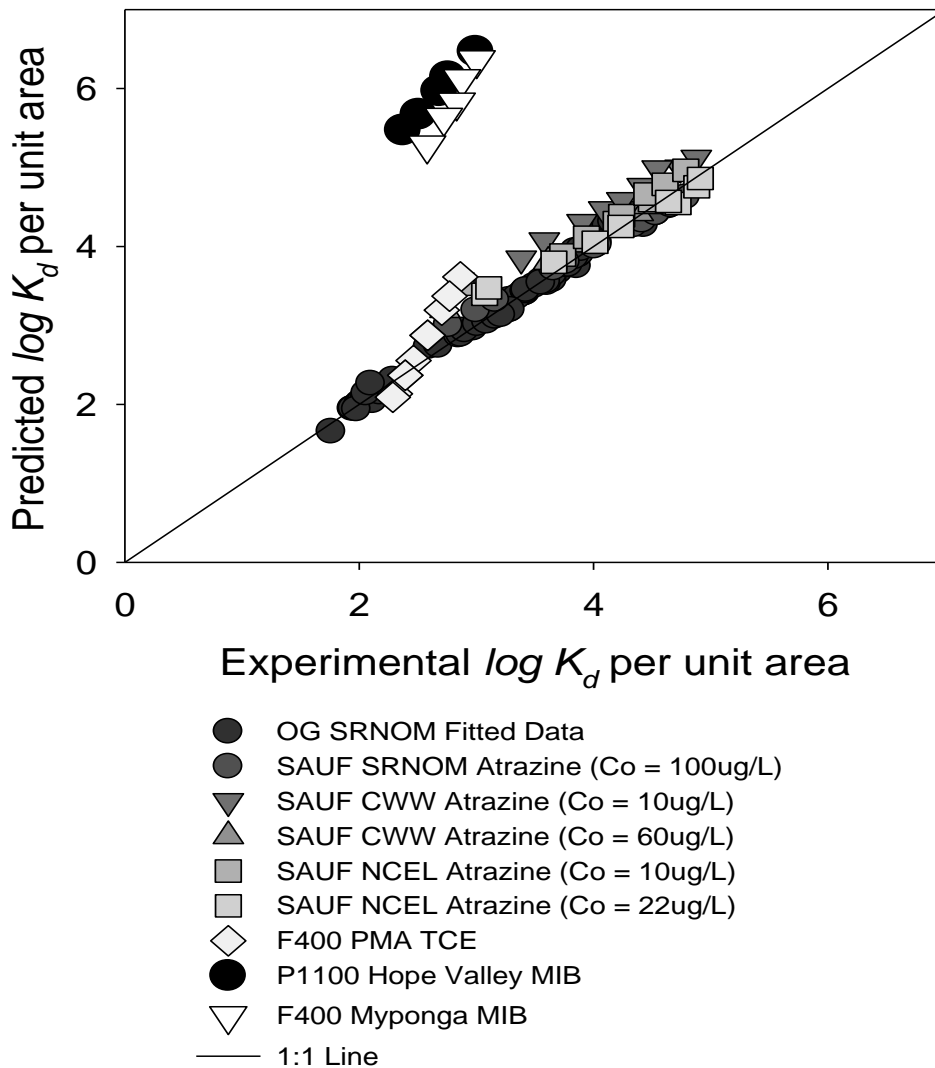


Figure 4.5: Plot of predicted Suwannee River NOM (SRNOM), Clinton Water Works NOM (CWW) and Newmark Groundwater NOM (NCEL) $\log K_d$ values (per unit area) using the determined equation for OG vs experimental SRNOM, CWW and NCEL $\log K_d$ values (per unit area) of atrazine on SA-UF. The plot also include predicted poly(maleic acid) (PMA) $\log K_d$ values (per unit area) using the OG equation vs experimental PMA $\log K_d$ values (per unit area) of TCE on F400. The plot also include predicted Hope Valley NOM (HV) and Myponga reservoir NOM (MRNOM) $\log K_d$ values (per unit area) using the OG equation vs experimental HV and MRNOM $\log K_d$ values (per unit area) of MIB on PICA and F400. The predicted $\log K_d$ values were normalized with the surface area of OG ($0.6\text{m}^2/\text{g}$) while the experimental SRNOM, CWW and NCEL $\log K_d$ values were normalized with the surface area of SAUF AC ($1112\text{m}^2/\text{g}$), the experimental PMA $\log K_d$ values and HV $\log K_d$ values were normalized with the surface area of F400 ($948\text{m}^2/\text{g}$) and the experimental MRNOM $\log K_d$ values were normalized with the surface area of PICA ($1241\text{m}^2/\text{g}$) (To et al., 2008); (Ding et al., 2006); (Cook et al., 2001) (Newcombe et al., 2002); (Newcombe et al., 2002).

REFERENCES

- Carter, M. C., Weber, W. J., Olmstead, K. P. (1992) Effects of Background Dissolved Organic-Matter on Tce Adsorption by Gac. *Journal American Water Works Association* **84**(8), 81-91.
- Cook, D., Newcombe, G., Sztajn bok, P. (2001) The application of powdered activated carbon for MIB and geosmin removal: Predicting PAC doses in four raw waters. *Water Res.* **35**(5), 1325-1333.
- Ding, L., Marinas, B. J., Schideman, L. C., Snoeyink, V. L. (2006) Competitive effects of natural organic matter: Parametrization and verification of the three-component adsorption model COMPSORB. *Environ. Sci. Technol.* **40**(1), 350-356.
- Ding, L. (2010) Mechanisms of competitive adsorption between trace organic contaminants and natural organic matter on activated carbon. Ph.D., University of Illinois at Urbana-Champaign,
- Ding, L., Snoeyink, V. L., Marinas, B. J., Yue, Z., Economy, J. (2008) Effects of powdered activated carbon pore size distribution on the competitive adsorption of aqueous atrazine and natural organic matter. *Environ. Sci. Technol.* **42**(4), 1227-1231.
- Ebie, K., Li, F. S., Azuma, Y., Yuasa, A., Hagishita, T. (2001) Pore distribution effect of activated carbon in adsorbing organic micropollutants from natural water. *Water Res.* **35**(1), 167-179.
- Guo, Y., Yadav, A., Karanfil, T. (2007) Approaches To Mitigate the Impact of Dissolved Organic Matter on the Adsorption of Synthetic Organic Contaminants by Porous Carbonaceous Sorbents. *Environ. Sci. Technol.*

Karanfil, T. and Kilduff, J. E. (1999) Role of granular activated carbon surface chemistry on the adsorption of organic compounds. 1. Priority pollutants. *Environ. Sci. Technol.* **33**(18), 3217-3224.

Kilduff, J. E. and Karanfil, T. (2002) Trichloroethylene adsorption by activated carbon preloaded with humic substances: effects of solution chemistry. *Water Research* **36**(7), 1685-1698.

Li, Q. L., Snoeyink, V. L., Marinas, B. J., Campos, C. (2003) Elucidating competitive adsorption mechanisms of atrazine and NOM using model compounds. *Water Res.* **37**(4), 773-784.

Newcombe, G., Morrison, J., Hepplewhite, C. (2002) Simultaneous adsorption of MIB and NOM onto activated carbon. I. Characterisation of the system and NOM adsorption. *Carbon* **40**(12), 2135-2146.

Newcombe, G., Morrison, J., Hepplewhite, C., Knappe, D. R. U. (2002) Simultaneous adsorption of MIB and NOM onto activated carbon - II. Competitive effects. *Carbon* **40**(12), 2147-2156.

Pelekani, C. and Snoeyink, V. L. (2000) Competitive adsorption between atrazine and methylene blue on activated carbon: the importance of pore size distribution. *Carbon* **38**(10), 1423-1436.

Pelekani, C. and Snoeyink, V. L. (1999) Competitive adsorption in natural water: Role of activated carbon pore size. *Water Res.* **33**(5), 1209-1219.

Quirke, N. and Tennison, S. R. R. (1996) The interpretation of pore size distributions of microporous carbons. *Carbon* **34**(10), 1281-1286.

Tessmer, C. H., Vidic, R. D., Uranowski, L. J. (1997) Impact of Oxygen-Containing Surface Functional Groups on Activated Carbon Adsorption of Phenols. *Environ. Sci. Technol.* **31**(7), 1872-1878.

To, P. C., Marinas, B. J., Snoeyink, V. L., Ng, W. J. (2008) Effect of pore-blocking background compounds on the kinetics of trace organic contaminant desorption from activated carbon. *Environ. Sci. Technol.* **42**(13), 4825-4830.

Vidic, R. D., Tessmer, C. H., Uranowski, L. J. (1997) Impact of surface properties of activated carbons on oxidative coupling of phenolic compounds. *Carbon*, **35**(9), 1349-1359.

Vidic, R. D. and Suidan, M. T. (1991) Role of Dissolved-Oxygen on the Adsorptive Capacity of Activated Carbon for Synthetic and Natural Organic-Matter. *Environ. Sci. Technol.* **25**(9), 1612-1618.

Yue, Z. R., Economy, J., Bordson, G. (2006) Preparation and characterization of NaOH-activated carbons from phenolic resin. *J. Mater. Chem.* **16**(15), 1456-1461.

Chapter 5: Linear Free Energy Relationship for Predicting Adsorption of Trace Organic Pollutants from Natural Water onto Activated Carbon

5.1 Introduction

Both pore size distribution (PSD) and surface chemistry of activated carbon will affect how organic compounds adsorb. Adsorbates of a particular shape and size can only access some portions of the pore surface, which are determined by the pore size distribution of the activated carbon. The types of chemical functional groups on the activated carbon surface determine interactions between adsorbates and the activated carbon surface.

In chapter 3, the effects of surface chemistry on adsorption isotherm prediction in the absence and presence of natural organic matter (NOM) were determined on non-porous oxidized graphite. It was shown that a polyparameter linear free energy relationship (pp-LFER) is able to predict oxidized graphite adsorption isotherms for both cases successfully. The pp-LFER approach correlated 5 molecular descriptors of organic compounds and the aqueous equilibrium concentration with the adsorption partition coefficient on oxidized graphite. The pp-LFER theory was introduced and developed by Abraham et al. (Abraham et al., 2004). It has been used to predict several partition coefficients in different environmental systems (Nguyen et al., 2005); (Goss, 1997)). A detailed explanation of pp-LFERs can be found in chapter 2.

The equation that was developed to predict isotherms in natural water is as follows:

$$\log K_{d,i} = (e_1 + e_2 \log(C_{e,i}))E + (s_1 + s_2 \log(C_{e,i}))S + (a_1 + a_2 \log(C_{e,i}))A + (b_1 + b_2 \log(C_{e,i}))B + (v_1 + v_2 \log(C_{e,i}))V_x + c_1 + c_2 \log(C_{e,i}) \quad \dots\dots(5.1)$$

where $K_{d,i} = q_{e,i} / C_{e,i}$; $C_{e,i}$ ($\mu\text{g/L}$) is the concentration of adsorbate i in solution at equilibrium and $q_{e,i}$ ($\mu\text{g/ mg carbon}$) is the concentration of adsorbate i on carbon at equilibrium

Besides the surface chemistry of activated carbon, the pore size distribution of activated carbon will also affect trace organic compounds and NOM adsorption. Based on the knowledge from the oxidized graphite study, the objective of this chapter was to develop pp-LFER for activated carbon adsorption capacity of trace organic compounds in the presence of NOM using molecular descriptors of the trace compounds.

5.2 Materials and Methods

5.2.1 Water

De-ionized water was passed through a NANOpure ultrapure water system (Barnstead, Dubuque, Iowa) to produce distilled deionized (DDI) water. The simulated natural water was prepared by dissolving dry form Suwannee River NOM (International Humic Substances Society, St. Paul, Minnesota, USA), isolated by reverse osmosis, in DDI water. The total organic concentration (TOC) of the stock concentration was measured to be 200 mg/L. The stock was then diluted to prepare the solution used to perform isotherm experiments and the final TOC concentration was measured as 4 mg/L and the pH was 7.0.

5.2.2 Adsorbent

A commercially available activated carbon, Norit SA-UF powered activated carbon (Norit France, S.a.r.l., Le Blanc Mesnil Cedex, France) was chosen. The SA-UF PAC was dried at 105°C overnight, and cooled in a desiccator before weighing. The surface area of SA-UF was determined to be 1112 m²/g and the majority of the pores are in the micropores and primary mesopores size range (Li et al., 2003). The surface oxygen content of SA-UF carbon was found to be 7.21 mass percent (Chapter 3).

5.2.3 Adsorbates

¹⁴C-labeled target trace organic contaminants were used (American Radiolabeled Chemicals, Inc): four organic solvents (trichloroethylene, 1-heptanol, nitrobenzene, benzene), two endocrine

disrupting compounds (estradiol, bisphenol A), one explosive (2, 4-dinitrotoluene), one herbicide (atrazine), one polycyclic aromatic hydrocarbon (naphthalene) and a pharmaceutical (diazepam). The molecular descriptors of the 10 target organic contaminants are listed in Table 1. The molecular descriptors have a range of values that are broad enough to cover a wide range of organic contaminants.

5.2.4 Isotherm Tests

Bottle-point isotherm tests were determined for each compound for 7 days (Randtke and Snoeyink, 1983). The initial concentration used for each compound was 10^{-3} to 10^{-5} of its solubility. For each compound, a different amount of activated carbon was used to achieve an aqueous equilibrium concentration (C_e) of 10^{-4} to 10^{-8} of its solubility. The isotherm bottles were prepared with no headspace, sealed with aluminum foil and mixed for seven days. Isotherms of non-volatile compounds were done using screw top caps with Teflon septa while isotherms of volatile compounds were done with crimp top caps with Teflon/silica septa.

After 7 days, aqueous samples were taken from the isotherm bottles and analyzed. The solid phase concentration was computed using a mass balance. Each set of isotherm experiments had two blanks to determine loss due to factors other than adsorption. Less than 10% loss was found after 7 days.

5.2.5 Analytical Methods

The isotherms samples were taken with a 5 ml glass syringe. Samples of estradiol, naphthalene and bisphenol A solution were filtered with 0.45 μm glass fiber membrane filters. Samples of atrazine, 2, 4-dinitrotoluene and diazepam were filtered with 0.45 μm nylon membrane filters. Samples of the solutions with volatile and high solubility compounds (trichloroethylene, 1-heptanol, nitrobenzene, benzene) were taken after the carbon in the bottles was allowed to settle for 8 hours.

Eighteen mL of scintillation cocktail was mixed with 2.5 mL of aqueous sample in a 20 mL scintillation vial. The mixture was then analyzed using a liquid scintillation analyzer (TriCarb Model 1600A, Packard Instrument Co. Downers Grove, IL).

5.2.6 Development of Predictive Polyparameter Linear Free Energy Relationship (pp-LFER)

K_d values of isotherms in natural water were calculated for 10 compounds as follow:

$$K_d = q_e / C_e \dots\dots\dots(5.2)$$

where q_e = adsorbed concentration at equilibrium and

C_e = aqueous concentration at equilibrium

The experimental $\log K_d$ values, $\log C_e$ values and 5 molecular descriptors of 9 compounds were tabulated and fitted with multiple linear regression using equation 5.1 (Sigma Plot 2000, SPSS, Chicago, IL) to obtain the coefficients. The coefficients e_1 , e_2 , s_1 , s_2 , a_1 , a_2 , b_1 , b_2 , v_1 , v_2 , c_1 , c_2 determine the tendency of how trace organic compounds partition onto activated carbon in the presence of NOM.

5.2.7 Validation of the pp-LFER model

After the pp-LFER equation was determined, it was used to predict $\log K_d$ values for the 10th compound, the compound that was not used for fitting. The predicted $\log K_d$ values were compared to experimental $\log K_d$ values to determine whether the fitted pp-LFER equation could be used to predict isotherms.

5.3 Results and Discussion

5.3.1 Activated carbon isotherms

Freundlich isotherm parameters for the 10 target organic contaminants in water containing Suwannee river natural organic matter (SRNOM) on activated carbon are reported in Table 2. The $1/n$ of heptanol and bisphenol A were observed to be greater than 1. That can be due to the hydroxyl functional group of one molecule interacting with another hydroxyl functional group of another molecule or the hydroxyl functional groups of NOM thus causing more molecules to adsorb beyond the activated carbon surface.

Freundlich isotherm parameters for the 10 target organic contaminants in water containing SRNOM on oxidized graphite are reported in Table 3. In the presence of pores, the order of K_d describing adsorption on activated carbon is different from what was observed on oxidized graphite. Thus, it was shown that the pore size distribution of activated carbon strongly affects the interactions of compounds and NOM with the activated carbon surface.

5.3.2 Natural water isotherm fitted pp-LFER

Adsorption data for 9 compounds (atrazine, estradiol, diazepam, 1-heptanol, trichloroethylene, nitrobenzene, benzene, 2, 4- dinitrotoluene, naphthalene) were used to fit the pp-LFER equation and thus determine the coefficients of the parameters in the equation. In Figure 5.1 it is shown that the predicted $\log K_d$ values were well correlated with experimental $\log K_d$ ($R^2 = 0.98$).

The best fit pp-LFER equation was established to be:

$$\log K_d = [(-6.79 \pm 3.06) + (-0.71 \pm 0.77) \log C_{e,i}]V + [(7.84 \pm 2.27) + (1.63 \pm 0.64) \log C_{e,i}]B + [(1.08 \pm 0.48) + (-0.14 \pm 0.15) \log C_{e,i}]S + [(-1.26 \pm 0.74) + (-0.83 \pm 0.19) \log C_{e,i}]E [(0.52 \pm 2.57) + (0.20 \pm 0.62) \log C_{e,i}]A + [(8.54 \pm 1.68) + (0.67 \pm 0.45) \log C_{e,i}] \dots (5.3)$$

The best-fit equation showed that coefficients v_2 , s_2 , a_1 and a_2 were statistically negligible. The two most important parameters that affect the adsorption were V and B . This was similar to what Shih and Gschwend observed in their work with single solute isotherm predictions (Shih and Gschwend, 2009).

5.3.3 Validation of pp-LFER equation

To determine if the pp-LFER equation, equation 5.3, is able to predict isotherms for other compounds not used in the fitting, the isotherm of bisphenol A was predicted and compared with its experimental isotherm. The data in Figure 5.1 show that the predicted $\log K_d$ values of bisphenol A, were well correlated with its experimental $\log K_d$ values. The predicted K_d values are within a factor of 1.81 from the experimental K_d values.

Further validation of the pp-LFER approach was done by withholding 1 out of 10 compounds for the fitting procedure, for all 10 compounds. Table 4 is a tabulation of the average difference between the predicted K_d value of the 10th compound and its experimental K_d values. Predictions of K_d values for most compounds were within a factor of 2 with the exceptions of heptanol, atrazine and nitrobenzene. Heptanol has the lowest K_d values and atrazine has the highest K_d values among all 10 compounds. Thus, when either heptanol or atrazine was not used in the fitting process, the predictions of the K_d values were not as good because the predictions will be

an extrapolation of the K_d range used in the fitting. But the extrapolation still yields predicted values that were within a factor of 10 of the experimental values. Nitrobenzene is the compound with the highest aqueous solubility and when it is not used in the fitting, the predictions were an average factor of 6.23 from the experimental values. This indicates that aqueous solubility is an important factor to consider during the selection of compounds to use for fitting.

5.3.4 Comparison of the pp-LFER equations for oxidized graphite and activated carbon

The pp-LFER equation for prediction of isotherms in natural water on oxidized graphite was:

$$\log K_{d,i} = [(1.57 \pm 0.63) + (0.36 \pm 0.26) \log C_{e,i}]V + [(-2.49 \pm 0.57) - (0.66 \pm 0.25) \log C_{e,i}]B + [(2.11 \pm 0.25) + (0.30 \pm 0.15) \log C_{e,i}]S + [(0.03 \pm 0.23) + (0.36 \pm 0.10) \log C_{e,i}]E + [(-0.51 \pm 0.25) + (0.04 \pm 0.14) \log C_{e,i}]A + [(-1.27 \pm 0.42) - (1.60 \pm 0.23) \log C_{e,i}] \dots \dots \dots (3.4)$$

In the presence of pore size distribution effects, the signs of v_1 and v_2 of the activated carbon pp-LFER equation changed from positive values to negative values. The signs of b_1 and b_2 also changed from negative values to positive values. In terms of chemical interactions, these changes imply that the chemical interactions of target organic compound with the activated carbon surface in the presence of NOM changed in the presence of pores.

The negative sign of coefficients v_1 and v_2 indicates that the smaller the compound, the higher the K_d and thus more adsorption on activated carbon in the presence of NOM. One good comparison is between estradiol and atrazine. Atrazine is smaller than estradiol with all other parameters to be close with the exception of parameter A but coefficient a_1 and a_2 have been shown to be negligible. The K_d of atrazine is higher than that of estradiol; and thus more atrazine is removed by activated carbon in the presence of NOM than estradiol. The signs of coefficients

v_1 and v_2 in this study was also opposite to what was observed in work by Shih and Gschwend on activated carbon adsorption in pure water (Shih and Gschwend, 2009). The signs of coefficients v_1 and v_2 were positive, meaning that larger target organic compounds will adsorb more onto activated carbon. This difference indicates that the accessibility of larger target organic contaminants to the pores of activated carbon will be hindered more as compared to smaller target organic contaminants in the presence of NOM.

The positive signs of coefficients b_1 and b_2 indicate that the higher the tendency of the target organic compound towards accepting a hydrogen bond, the more it will adsorb onto activated carbon in the presence of NOM. This trend is again opposite to what was observed for oxidized graphite and for what was observed for activated carbon adsorption in pure water. But it can be observed that the coefficient b_1 gradually increased from -9.97 (oxidized graphite in pure water) to -4.47 (activated carbon in pure water) to -2.49 (oxidized graphite in water with NOM) to 7.84 (activated carbon in water with NOM). In the presence of NOM and pores in the system, the differential interactions of target organic compounds with the activated carbon surface and aqueous NOM phase evolved and caused activated carbon to have a higher removal of target organic contaminants that have more hydrogen bond accepting functional groups.

5.3.5 Predicting isotherms for another natural water

The pp-LFER equation was developed for one type of activated carbon and one kind of natural water. The type of NOM in different natural water varies tremendously in terms of molecular weights and chemical functional groups. The effect of a difference in NOM composition was

tested by using the pp-LFER equation determined above using SRNOM to predict atrazine isotherms in other natural waters.

Ding et al. measured adsorption isotherms of atrazine in presence of two different natural waters on SA-UF activated carbon. (Ding et al., 2006). The two different natural waters were groundwater from Clinton Water Works (CWW), Clinton, IL and groundwater from a well located next to Newmark Civil Engineering Laboratory (NCEL) at the University of Illinois. The properties of these two natural waters were determined by Ding et al..

In Figure 5.2, it was observed that for higher atrazine initial concentrations, the predicted K_d values were within a factor of 1.5 from the experimental K_d values. For higher NOM initial concentration of 7.5 mg/L, the predicted K_d values were on average, 3 times higher than the experimental K_d values. For a lower NOM initial concentration of 2.5 mg/L, the predicted K_d values were an average of 1.9 higher than the experimental K_d values. The initial concentration of the SRNOM used in this study was 4 mg/L; thus we can speculate that for NOM with an initial concentration that is close to that used in the fitting, the prediction will be better. The initial concentration of the target organic compound is also important. For higher initial concentrations, the predictions were better. However, this exercise was only done for one compound, atrazine, which was used in the fitting process of this study. More adsorption data for other compounds in different natural waters on SA-UF will have to be collected in order to show that one pp-LFER equation will be sufficient to predict adsorption isotherms in different natural waters, on one carbon, with reasonable accuracy.

5.3.6 Selection of compounds to develop pp-LFER

The pp-LFER approach has been shown above to be an effective method to predict adsorption isotherms on activated carbon in natural water. However, for each combination of natural water and activated carbon, we will need to develop a pp-LFER equation to predict adsorption isotherms. Thus, it will be valuable to evaluate the selection of compounds required to develop a good pp-LFER.

Six compounds (atrazine, heptanol, nitrobenzene, trichloroethylene, estradiol and diazepam) were selected based on the range of their molecular descriptors. This selected group of compounds covers the wide range of molecular descriptors and aqueous solubility shown in Table 1 ($V = 0.71$ to 2.19 , $B = 0.03$ to 1.25 , $S = 0.37$ to 1.77 , $E = 0.21$ to 2.08 , $A = 0$ to 0.86 and aqueous solubility = 0.26 mg/L to 1949 mg/L). Using the adsorption isotherms of these six compounds in natural water, the pp-LFER equation was fitted. In Figure 5.3, it was shown that for fitting with 6 compounds, the K_d predictions for bisphenol A, benzene, naphthalene and 2,4-dinitrotoluene were within a factor of 5 from the experimental values.

5.3.7 Application to predict elution order of compounds

Granular activated carbon (GAC) columns have been used to remove organic compounds from drinking water. Estimation of an activated carbon adsorber bed's life has been a challenge for the water industries, especially when a new target organic contaminant that is of concern to the public emerges. Snoeyink et al. used isotherm data to predict the remaining lifetime of an

adsorber bed and they showed that reasonable predictions could be made with isotherm data (Snoeyink et al., 1989).

For good estimation of the lifetime of a GAC column, both equilibrium and kinetics data are required (Schideman et al., 2007). But to determine if new compounds will breakthrough a GAC column quickly, or if it will break through before another compound the predictive tool developed in this study will be of great use. By assuming an ideal breakthrough curve, the sequence of the elution of organic compounds will be determined by their K_d values. A higher K_d value will have a longer retention time in the GAC column while a lower K_d will indicate a faster breakthrough. For initial design of GAC column, the pp-LFER equation can help water utilities plan ahead for a variety of organic compounds. When a new organic compound emerges, water utilities can obtain the molecular descriptors and the aqueous concentration of the compound and apply the pp-LFER equation developed to get the K_d value of the new compound. This will help water utilities predict if the new compound will be of great concern and if the GAC column does not have a high capacity for it.

One point to note is that the pp-LFER was determined for simultaneous adsorption of organic compound and NOM on activated carbon. Frequently a GAC column will be preloaded with NOM before the organic contaminant is introduced (Gilligly et al., 1999); (Knappe et al., 1997); (Speth, 1991). Further study of NOM preloading effects on GAC column will have to be done in order to determine the limits of this approach to predict elution order.

5.4 Conclusion

- K_d values of adsorption of organic compounds in SRNOM on activated carbon can be predicted by a pp-LFER within a factor of 1.81 from the experimental adsorption isotherms.
- The pp-LFER equation can be extrapolated and used to predict K_d of atrazine in two different natural waters on SAUF carbon within a factor of 3.
- To predict adsorption isotherms effectively, at least 6 organic compounds adsorption isotherms will have to be determined, in order to predict K_d values that are within a factor of 5 from the experimental value.
- The pp-LFER equation can be applied to predict the elution order of organic compounds from a GAC column.

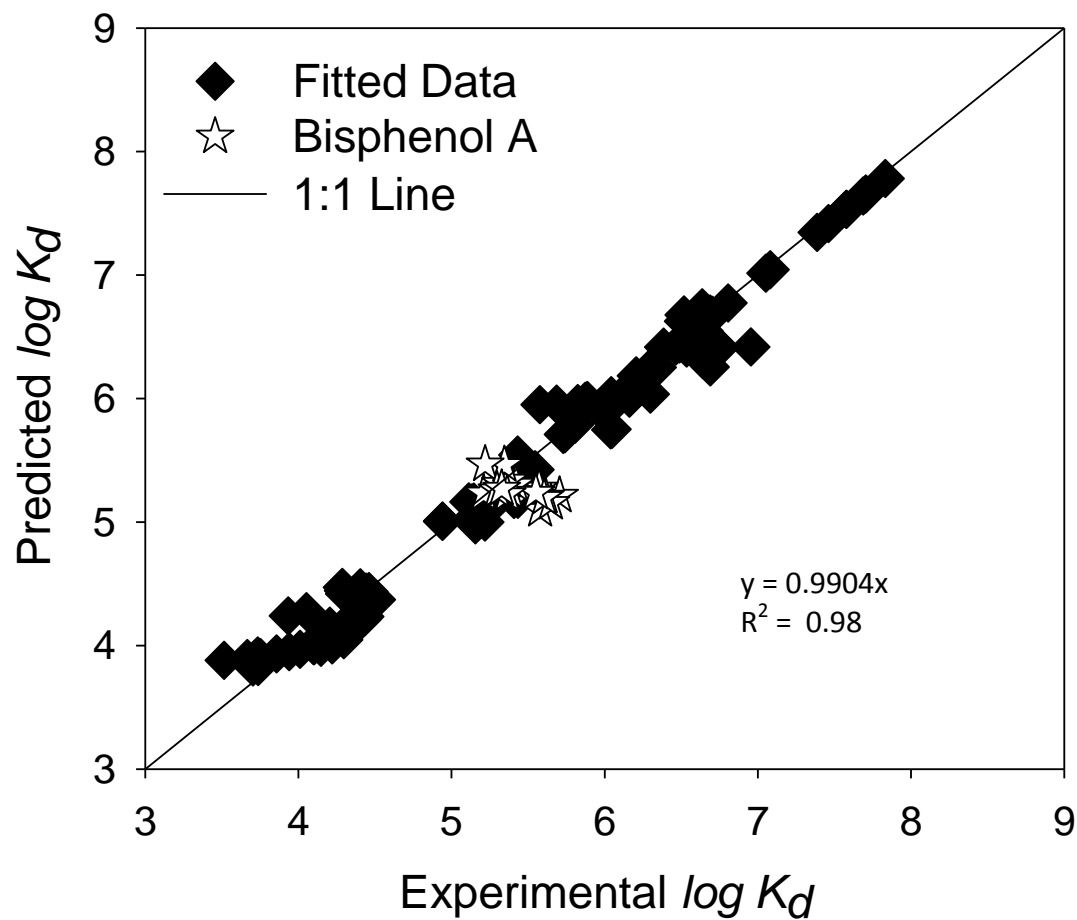


Figure 5.1: Plot of $\log K_d$ values predicted by pp-LFER equation fitted using isotherms of 9 organic compounds against the experimental $\log K_d$ values of the 9 organic compounds. $\log K_d$ values of Bisphenol A (not used in fitting) were predicted and compared with the experimental values on this plot marked as stars.

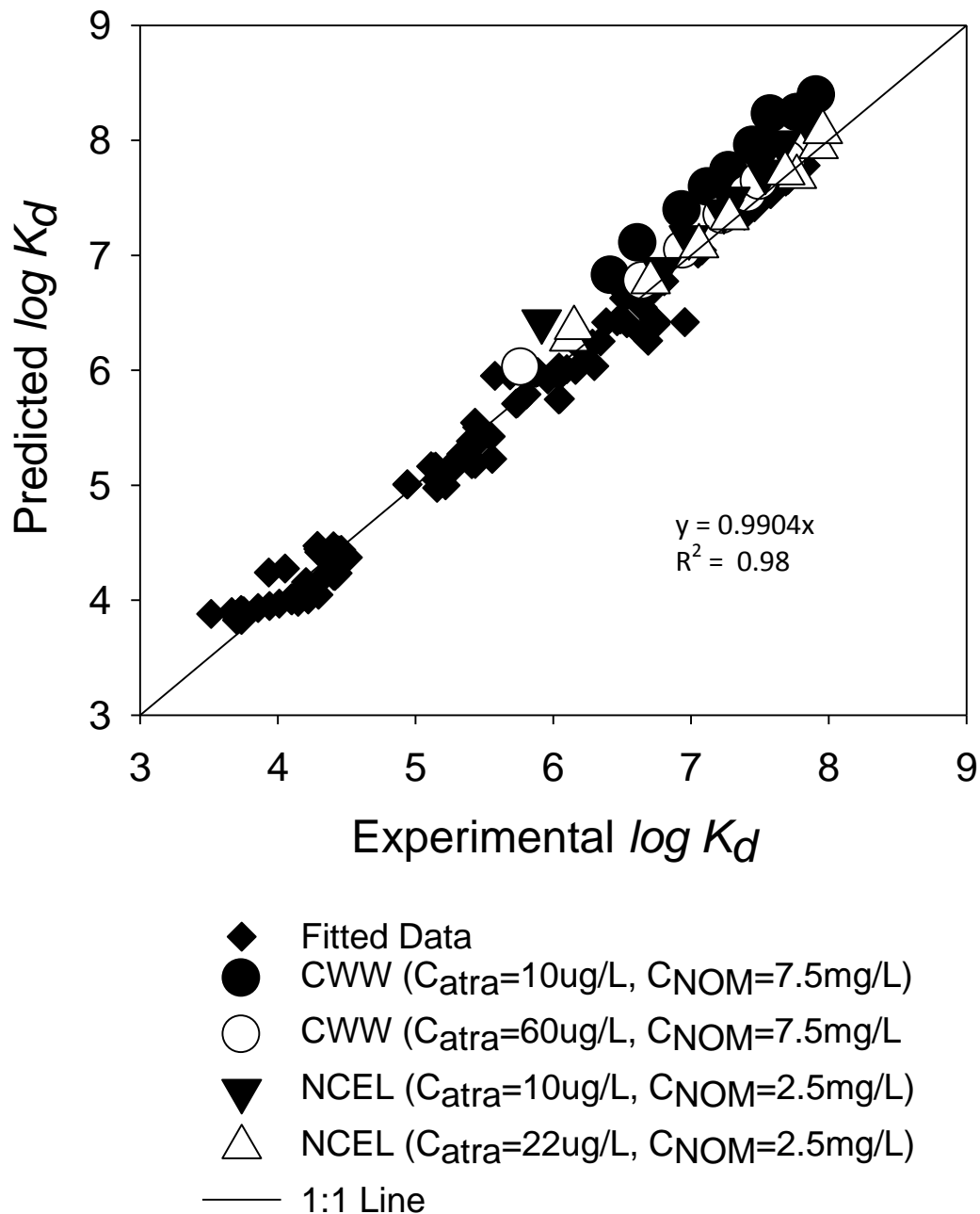


Figure 5.2: Comparison of predicted $\log K_d$ values for SA-UF isotherms of atrazine in Clinton Water Works (CWW) and Newmark Groundwater (NCEL) with the experimental $\log K_d$ values. The total organic carbon (TOC) concentration of CWW was 7.5mg/L and NCEL is 2.5mg/L. Two initial concentrations of atrazine was used for both natural waters (Ding et al., 2006).

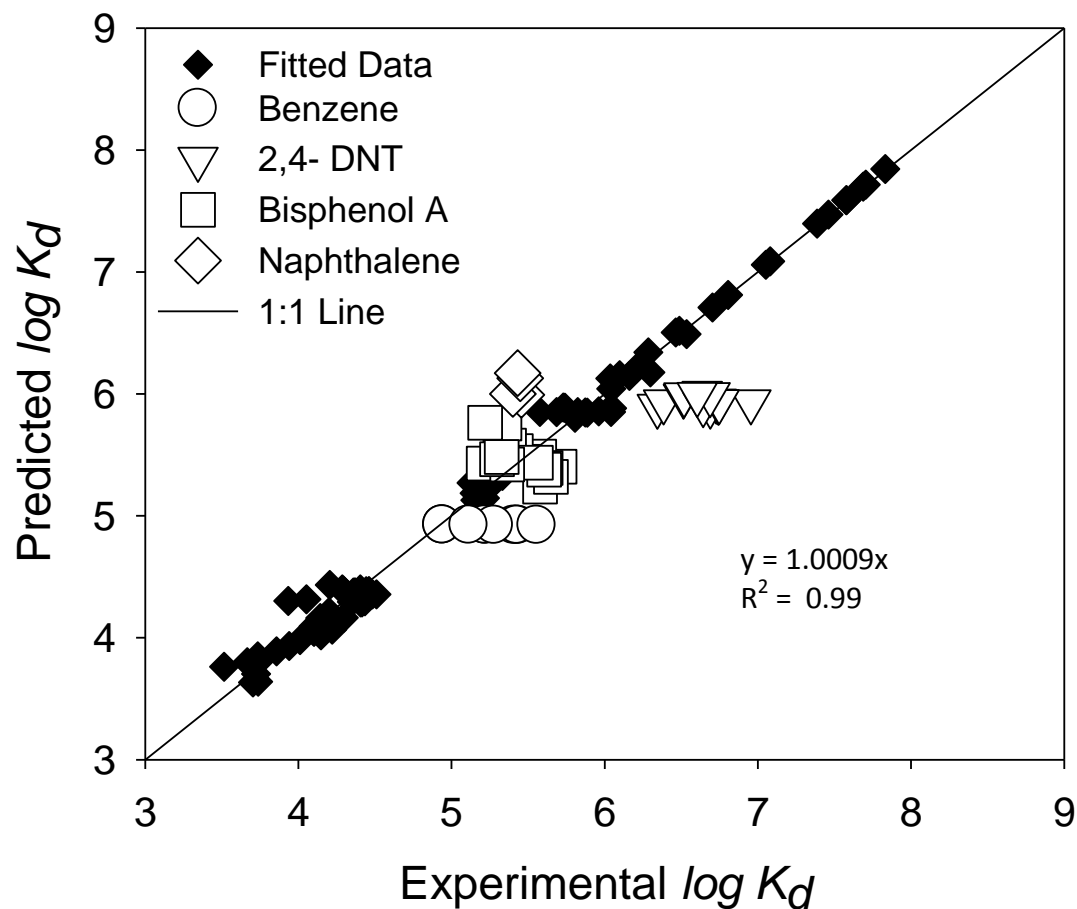


Figure 5.3: Plot of $\log K_d$ values predicted by pp-LFER equation fitted using isotherms of 6 compounds (atrazine, estradiol, diazepam, trichloroethylene, 1-heptanol, nitrobenzene) against the experimental $\log K_d$ values of the 6 organic compounds. Naphthalene, 2, 4 DNT, Bisphenol A and Benzene $\log K_d$ values in the presence of SRNOM were predicted and compared with the experimental $\log K_d$ on this plot.

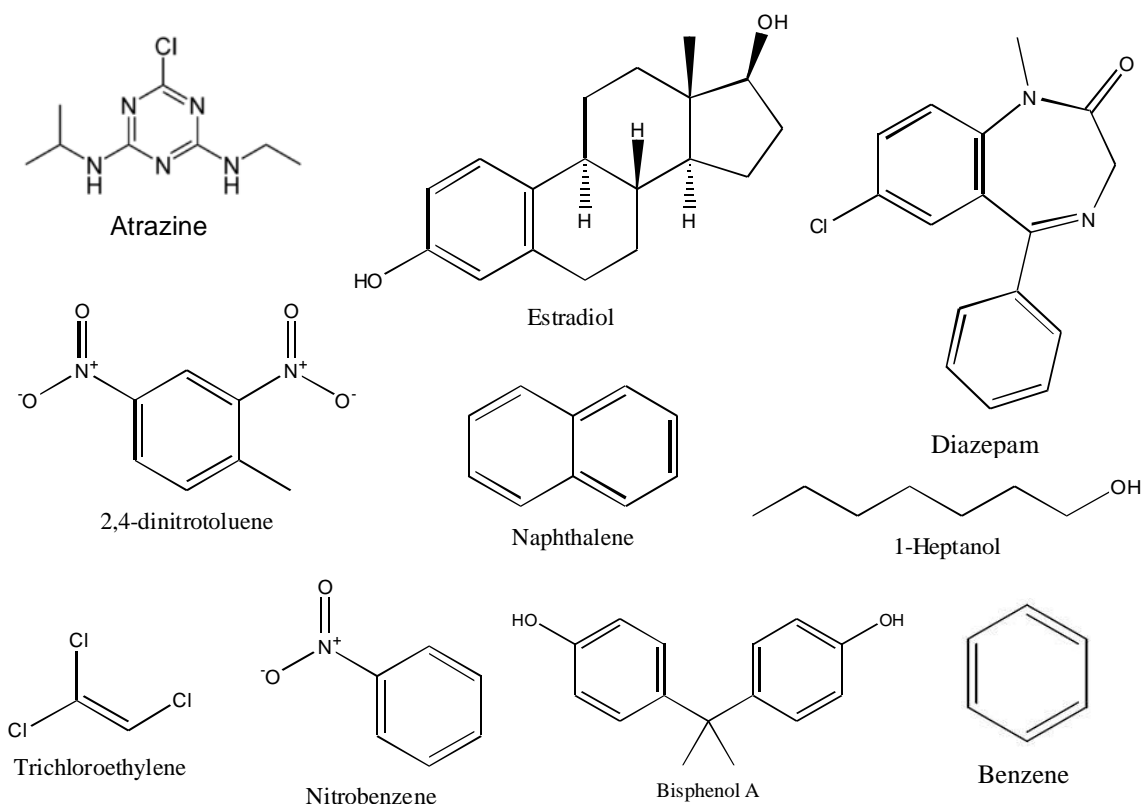


Table 5.1: Aqueous solubility (Yaw, 1999); (ADME/ Tox WEB) and molecular properties (Abraham et al., 1994a); (Abraham et al., 1994b); (ADME/ Tox WEB) of compounds used for fitting of $\log K_d$ predictive equation.

No							Aq. Solubility
	Compound	<i>E</i>	<i>S</i>	<i>A</i>	<i>B</i>	<i>V_x</i>	(mg/L)
1	Atrazine	1.95	1.64	0	1.04	1.61	70
2	Estradiol	1.80	1.77	0.86	1.10	2.19	0.26
3	Diazepam	2.08	1.55	0.09	1.25	2.07	50
4	Naphthalene	1.34	0.92	0	0.20	1.08	30
5	2,4-Dinitrotoluene	1.17	1.27	0.07	0.51	1.20	270
6	Bisphenol A	1.60	1.56	0.99	0.94	1.86	300
7	Nitrobenzene	0.87	1.11	0	0.28	0.89	1949
8	Trichloroethylene	0.52	0.37	0.08	0.03	0.71	1280
9	Benzene	0.61	0.52	0	0.14	0.71	1760
10	1-Heptanol	0.21	0.42	0.37	0.48	1.15	1740

Table 5.2: Experimental Freundlich isotherm parameters of 10 compounds on Activated Carbon in the presence of NOM

No	Compound	$K_f \times 10^4$	$1/n$
1	Atrazine (To et al., 2008)	12	0.36
2	Estradiol	0.114	0.48
3	Diazepam	0.895	0.43
4	Nitrobenzene	100	1.06
5	Benzene	0.751	0.54
6	2,4 Dinitrotoluene	500	1.00
7	Bisphenol A	2000	1.49
8	Trichloroethylene	1.25	0.89
9	Naphthalene	18	0.94
10	Heptanol	100	2.15

Table 5.3: Experimental Freundlich isotherm parameters of 10 compounds on Oxidized Graphite in the presence of NOM

Compound	K_f	$1/n$
Atrazine	250	0.58
Estradiol	177	0.56
Diazepam	67	0.40
Naphthalene	70	0.40
2,4 Dinitrotoluene	185	0.41
Nitrobenzene	62	0.28
1-Heptanol	6	0.28
Trichloroethylene	12	0.32
Lindane	64	0.41
Bisphenol A	80	0.34

Table 5.4: Ratio of predicted K_d values and experimental K_d values of the 10th compound (not used in the fitting procedure)

No.	Compounds	<i>Average</i>(K_d Predicted/ K_d Experimental)	<i>Range</i>(K_d Predicted/ K_d Experimental)
1	Naphthalene	5.13	3.66-6.89
2	Bisphenol A	1.81	1.01-3.11
3	2, 4- Dinitrotoluene	1.80	1.03-3.81
4	Benzene	2.22	1.08-3.88
5	Nitrobenzene	6.23	3.77-10.99
6	Trichloroethylene	1.41	1.06-2.54
7	Heptanol	10.18	6.47-19.00
8	Diazepam	1.66	1.14-2.36
9	Estradiol	2.02	1.48-3.68
10	Atrazine	8.17	1.09-39.94

REFERENCES

- Abraham, M. H., Ibrahim, A., Zissimos, A. M. (2004) Determination of sets of solute descriptors from chromatographic measurements. *J. Chromatogr. A* **1037**(1-2), 29-47.
- Abraham, M. H., Andonianhaftvan, J., Whiting, G. S., Leo, A., Taft, R. S. (1994a) Hydrogen-Bonding .34. the Factors that Influence the Solubility of Gases and Vapors in Water at 298-K, and a New Method for its Determination. *J. Chem. Soc. -Perkin Trans. 2*(8), 1777-1791.
- Abraham, M. H., Chadha, H. S., Whiting, G. S., Mitchell, R. C. (1994b) Hydrogen-Bonding .32. an Analysis of Water-Octanol and Water-Alkane Partitioning and the Delta-Log-P Parameter of Seiler. *J. Pharm. Sci.* **83**(8), 1085-1100.
- Ding, L., Marinas, B. J., Schideman, L. C., Snoeyink, V. L. (2006) Competitive effects of natural organic matter: Parametrization and verification of the three-component adsorption model COMPSORB. *Environ. Sci. Technol.* **40**(1), 350-356.
- Gillogly, T. E. T., Snoeyink, V. L., Vogel, J. C., Wilson, C. M., Royal, E. P. (1999) Determining GAC bed life. *J. Am. Water Work Assoc.* **91**(8), 98-110.
- Goss, K. U. (1997) Conceptual model for the adsorption of organic compounds from the gas phase to liquid and solid surfaces. *Environ. Sci. Technol.* **31**(12), 3600-3605.
- Knappe, D. R. U., Snoeyink, V. L., Roche, P., Prados, M. J., Bourbigot, M. M. (1997) The effect of preloading on rapid small-scale column test predictions of atrazine removal by GAC adsorbers. *Water Res.* **31**(11), 2899-2909.

- Li, Q. L., Snoeyink, V. L., Mariaas, B. J., Campos, C. (2003) Elucidating competitive adsorption mechanisms of atrazine and NOM using model compounds. *Water Res.* **37**(4), 773-784.
- Nguyen, T. H., Goss, K. -, Ball, W. P. (2005) Polyparameter Linear Free Energy Relationships for Estimating the Equilibrium Partition of Organic Compounds between Water and the Natural Organic Matter in Soils and Sediments. *Environ. Sci. Technol.* **39**(4), 913-924.
- Randtke, S. J. and Snoeyink, V. L. (1983) Evaluating Gac Adsorptive Capacity. *J. Am. Water Work Assoc.* **75**(8), 406-413.
- Schideman, L. C., Snoeyink, V. L., Marinas, B. J., Ding, L., Campos, C. (2007) Application of a three-component competitive adsorption model to evaluate and optimize granular activated carbon systems. *Water Res.* **41**(15), 3289-3298.
- Shih, Y. and Gschwend, P. M. (2009) Evaluating Activated Carbon-Water Sorption Coefficients of Organic Compounds Using a Linear Solvation Energy Relationship Approach and Sorbate Chemical Activities. *Environmental Science & Technology* **43**(3), 857.
- Snoeyink, V. L., Qi, S., Halloran, A., Beck, E., Koffskey, W., Lykins, B. (1989) Application of isotherms for predictions of GAC performance: A preliminary analysis. *Design and Use of Granular Activated Carbon: Practical Aspects Conference Proceedings*, 43-66.
- Speth, T. F. (1991) Evaluating Capacities of Gac Preloaded with Natural-Water. *J. Environ. Eng. -ASCE* **117**(1), 66-79.

To, P. C., Marinas, B. J., Snoeyink, V. L., Ng, W. J. (2008) Effect of pore-blocking background compounds on the kinetics of trace organic contaminant desorption from activated carbon. Environ. Sci. Technol. **42**(13), 4825-4830.

Yaw, C.L. (1999) Chemical Properties Handbook: Physical, Thermodynamic, Environmental, Transport, Safety, and Health Related Properties for Organic and Inorganic Chemicals, McGraw-Hill: New York

<http://www.pharma-algorithms.com/webboxes/>

Chapter 6: Summary and Future Research

6.1 Summary

Activated carbon has long been used to treat trace organic chemicals and natural organic matter (NOM) in drinking water. Industries have produced many new organic compounds that will eventually make it to our drinking water sources. To be able to assure the public that the current water treatment system is capable of dealing with these new organic compounds, good predictive tools will be useful for a quick initial evaluation of expected activated carbon performance.

Polyparameter linear free energy relationships (pp-LFERs) have been successful in predicting partition coefficients of many environmental systems. There are also many studies that have been done with pp-LFERs and quantitative structure-activity relationships (QSARs) to predict isotherms in organic-free water. But there has not been a good way to predict the expected performance of activated carbon for removal of trace organic compounds in the natural water. In this work, it has been demonstrated that pp-LFERs are applicable to carbon systems and that they can be used to make good predictions of adsorption capacity. Adsorption capacity values can then be used to estimate performance.

The first part of this research demonstrated that pp-LFERs are effective in predicting non-porous oxidized graphite isotherms in waters in absence and presence of NOM. In absence of pores, the surface interactions were greatly influenced by the molecular volume (V) and overall hydrogen-

bond basicity (B) of the organic compound. Larger compounds with lower hydrogen-bond basicity will adsorb more strongly onto oxidized graphite. In the presence of NOM, similar trends were observed but the magnitude of the dependency on parameters V and B decreased indicating the influence of surface interactions by NOM adsorption. Both oxidized graphite pp-LFER equations developed for absence and presence of NOM predicted adsorption isotherms that are within a factor of 2 from the experimental isotherms.

The second part of this research demonstrated that pp-LFERs are also effective in predicting activated carbon isotherms in the presence of NOM. In the presence of pores, the pp-LFER equation changed significantly emphasizing the importance of pore size distribution. The signs of parameters V and B changed signs from what was determined for oxidized graphite.

Adsorption of organic compounds is not only determined by surface interactions but also by the amount and size of pores accessible for removal. The pp-LFER predicted adsorption isotherms that are within a factor 2 from the experimental isotherms. The predicted adsorption isotherm data can be used to determine the elution order of organic compounds from a granular activated carbon (GAC) column. It was also shown that atrazine in other natural waters can also be predicted by the same pp-LFER equation with reasonable accuracy.

6.2 Future Research

This study has shown that pp-LFERs are effective in predicting adsorption isotherms on one type of activated carbon in the presence of one type of NOM. Future research should focus on extending this study to undertake the following specific tasks:

- To determine if the pp-LFER equation developed using one type of NOM is applicable to other natural waters. The NOM in different natural waters varies widely in terms of chemistry and molecular weight distribution. It was shown in this work that the equation developed using Suwanee River NOM could be applied to atrazine in two other natural waters. Isotherms of more compounds in other natural waters on the same activated carbon must be determined to demonstrate that the effects of NOM variability are not great enough to require the development of a new pp-LFER.
- To determine the limit of the pp-LFER equation to predict isotherms of NOM preloaded activated carbon. Preloading of NOM has been a great problem in GAC columns. Thus, it will be important to determine the amount of NOM preloading that will cause a change in the elution order determined by the pp-LFER equation.
- To determine a parameter that can represent the effects of pore size distribution of activated carbon on adsorption capacity. Different activated carbons have different pore size distribution and that will greatly influence the adsorption process. Thus if a parameter can be determined to represent this in the pp-LFER equation then it will make the pp-LFER equation more useful for predicting adsorption capacity of activated carbon.

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Appendix:

Table A1: Oxidized graphite single solute isotherm data of 10 trace organic compounds

Atrazine

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
3.466922986	0.003467	0.020636	20.6355	5952.109	3.774671
3.026535356	0.003027	0.020825	20.82523	6880.88	3.837644
2.911329818	0.002911	0.02107	21.07008	7237.27	3.859575
1.253571306	0.001254	0.010782	10.78244	8601.374	3.934568
1.255061912	0.001255	0.010735	10.73464	8553.074	3.932122
1.325711082	0.001326	0.010739	10.73879	8100.402	3.908507
0.895775059	0.000896	0.008082	8.082179	9022.554	3.955329
0.894913701	0.000895	0.008133	8.133221	9088.274	3.958481
0.859639822	0.00086	0.008122	8.122245	9448.428	3.97536
0.813325544	0.000813	0.006514	6.514043	8009.146	3.903586
0.631607672	0.000632	0.006562	6.562229	10389.72	4.016604
1.278942527	0.001279	0.00645	6.449823	5043.091	3.702697
0.507316535	0.000507	0.005445	5.44512	10733.18	4.030728
0.519581426	0.00052	0.005456	5.455725	10500.23	4.021199
0.560635284	0.000561	0.005457	5.457097	9733.774	3.988281

Table A1 (cont.)**Estradiol**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
0.080438053	8.04E-05	0.017671	17.6706	219679.6	5.34179
0.070860888	7.09E-05	0.016455	16.45536	232220.7	5.365901
0.082718143	8.27E-05	0.013245	13.24516	160124	5.204456
0.07978083	7.98E-05	0.012973	12.97348	162614	5.211158
0.053950824	5.4E-05	0.008719	8.71858	161602.4	5.208448
0.070920954	7.09E-05	0.008726	8.726462	123044.9	5.090064
0.053473957	5.35E-05	0.006687	6.68698	125051.2	5.097088
0.021837215	2.18E-05	0.005458	5.45757	249920.6	5.397802
0.016816417	1.68E-05	0.00305	3.050279	181386.9	5.258606
0.022908315	2.29E-05	0.002867	2.867484	125172.2	5.097508
0.007469367	7.47E-06	0.001579	1.579331	211441	5.325189
0.007205778	7.21E-06	0.001598	1.597848	221745.5	5.345855
0.002798712	2.8E-06	0.001072	1.072409	383179.4	5.583402
0.005547319	5.55E-06	0.00107	1.070226	192926.7	5.285392
0.006856247	6.86E-06	0.000798	0.797571	116327.6	5.065683
0.004361668	4.36E-06	0.000799	0.798554	183084.5	5.262652
0.004655032	4.66E-06	0.000647	0.647048	138999.6	5.143014
0.005811775	5.81E-06	0.000642	0.641969	110460	5.043205

Diazepam

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
2.868910794	0.002869	0.023957	23.95675	8350.468	3.921711
2.66449695	0.002664	0.023949	23.94924	8988.278	3.953676
1.392070876	0.001392	0.012158	12.15795	8733.717	3.941199
0.651401441	0.000651	0.012095	12.09503	18567.71	4.268758
0.483199393	0.000483	0.009171	9.171038	18979.82	4.278292
0.557489987	0.000557	0.009357	9.35653	16783.31	4.224878
0.480484649	0.00048	0.007514	7.514141	15638.67	4.1942
0.448344503	0.000448	0.007299	7.298523	16278.83	4.211623
0.367872845	0.000368	0.006166	6.165786	16760.64	4.224291
0.345165512	0.000345	0.006286	6.285671	18210.6	4.260324

Table A1 (cont.)**Naphthalene**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
4.924223976	0.004924	0.020513	20.51308	4165.749	3.619693
5.585295767	0.005585	0.019978	19.97805	3576.901	3.553507
1.00892852	0.001009	0.01228	12.27977	12171.1	4.08533
1.375492831	0.001375	0.01202	12.01979	8738.531	3.941438
0.369984941	0.00037	0.008443	8.443229	22820.47	4.358325
0.379441351	0.000379	0.008519	8.519113	22451.73	4.35125
0.225320781	0.000225	0.005062	5.062379	22467.43	4.351553
0.192308278	0.000192	0.005056	5.055718	26289.65	4.419785
0.157552578	0.000158	0.003617	3.617212	22958.76	4.360948
0.1593956	0.000159	0.003676	3.675724	23060.39	4.362867

2,4-dinitrotoluene

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
13.13200742	0.013132	0.06951	69.50979	5293.158	3.723715
11.78587499	0.011786	0.069516	69.51565	5898.218	3.770721
3.493969635	0.003494	0.036068	36.0675	10322.79	4.013797
4.403600155	0.004404	0.036595	36.59541	8310.339	3.919619
2.082156508	0.002082	0.028528	28.52802	13701.19	4.136758
2.687073912	0.002687	0.027185	27.1845	10116.77	4.005042
2.198808405	0.002199	0.02203	22.03006	10019.09	4.000828
2.328085384	0.002328	0.022726	22.72603	9761.681	3.989525
1.804318846	0.001804	0.018464	18.464	10233.23	4.010013
1.665794786	0.001666	0.018485	18.48476	11096.66	4.045192

Table A1 (cont.)**Nitrobenzene**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
1308.479069	1.308479	0.102268	102.2679	78.15781	1.892972
1268.457913	1.268458	0.106328	106.3282	83.82479	1.923372
1202.938061	1.202938	0.088792	88.79196	73.81258	1.86813
1189.703357	1.189703	0.093308	93.30774	78.42942	1.894479
1084.541751	1.084542	0.084936	84.93573	78.31486	1.893844
1063.019228	1.063019	0.088168	88.16833	82.94143	1.918772
993.5573082	0.993557	0.079884	79.88407	80.40207	1.905267
996.6836764	0.996684	0.080692	80.69202	80.96052	1.908273
859.0478924	0.859048	0.076982	76.98185	89.61299	1.952371
927.4367696	0.927437	0.078768	78.76754	84.93036	1.929063
1075.455654	1.075456	0.095119	95.11863	88.44496	1.946673
1141.329819	1.14133	0.09153	91.5297	80.19566	1.904151
904.920007	0.90492	0.088131	88.1306	97.39048	1.988517
928.3101554	0.92831	0.089528	89.52798	96.44189	1.984266
780.2540235	0.780254	0.079804	79.80403	102.2796	2.009789
793.1572421	0.793157	0.080575	80.57544	101.5882	2.006843
755.4387778	0.755439	0.074952	74.95184	99.2163	1.996583
775.7419891	0.775742	0.07395	73.94979	95.32782	1.97922
629.4976928	0.629498	0.070381	70.38099	111.805	2.048461
596.2008488	0.596201	0.069032	69.03186	115.7862	2.063657

Lindane

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
4.192221025	0.004192	0.015221	15.22141	3630.871	3.560011
3.401835315	0.003402	0.018244	18.24352	5362.847	3.729395
1.203364672	0.001203	0.013358	13.35766	11100.26	4.045333
0.843460525	0.000843	0.009454	9.453932	11208.51	4.049548
0.535121797	0.000535	0.009938	9.937994	18571.46	4.268846
0.450631379	0.000451	0.007612	7.612062	16891.99	4.227681
0.273805241	0.000274	0.007729	7.72897	28227.99	4.45068
0.240479488	0.00024	0.006277	6.277316	26103.33	4.416696
0.199136255	0.000199	0.006296	6.296108	31617.09	4.499922

Table A1 (cont.)**Bisphenol A**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
43.09845546	0.043098	0.074043	74.04266	1717.989	3.23502
47.24465434	0.047245	0.072218	72.21817	1528.6	3.184294
13.19264106	0.013193	0.041959	41.95865	3180.458	3.50249
27.10196628	0.027102	0.040147	40.14658	1481.316	3.170648
12.67292339	0.012673	0.032455	32.45463	2560.942	3.4084
8.491654278	0.008492	0.031488	31.48788	3708.097	3.569151
7.359752116	0.00736	0.026421	26.42055	3589.869	3.555079
7.636076622	0.007636	0.026791	26.7914	3508.529	3.545125
8.150727128	0.008151	0.020955	20.95546	2570.993	3.410101
6.157147118	0.006157	0.021928	21.92763	3561.329	3.551612
110.9396013	0.11094	0.126443	126.4428	1139.744	3.056807
104.2603102	0.10426	0.128164	128.1643	1229.272	3.089648
55.80411467	0.055804	0.078176	78.17643	1400.908	3.14641
63.35856468	0.063359	0.079735	79.73477	1258.469	3.099842
27.64406114	0.027644	0.061791	61.79135	2235.249	3.349326
24.7022699	0.024702	0.059061	59.06068	2390.901	3.378562
16.80133303	0.016801	0.047062	47.06205	2801.09	3.447327
19.83549627	0.019835	0.046985	46.98546	2368.756	3.37452
14.24770025	0.014248	0.03785	37.85041	2656.598	3.424326
19.34377516	0.019344	0.03726	37.26006	1926.204	3.284702

1-Heptanol

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
31.46977771	0.03147	0.005738	5.73832	182.3438	2.260891
18.16841559	0.018168	0.00444	4.439922	244.3758	2.388058
15.23106174	0.015231	0.003976	3.976252	261.062	2.416744
20.09884959	0.020099	0.005354	5.353613	266.3641	2.425476

TCE

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
32.17446037	0.032174	0.003698	3.698468	114.9504	2.060511
21.25480943	0.021255	0.0033	3.300018	155.2598	2.191059
14.19457195	0.014195	0.0028	2.799979	197.257	2.295032
38.95538104	0.038955	0.005087	5.087394	130.5954	2.115928

Table A2: Oxidized graphite competitive isotherm data of 10 trace organic compounds**Atrazine**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
12.14056476	0.012140565	0.017912382	17.9123817	1475.415852	3.168914445
11.40409832	0.011404098	0.018018728	18.01872824	1580.022176	3.198663183
11.76287693	0.011762877	0.018136806	18.13680574	1541.868189	3.188047248
3.963278933	0.003963279	0.010131219	10.13121946	2556.272126	3.407607084
3.975100216	0.0039751	0.010325133	10.3251325	2597.452125	3.414547552
3.45810308	0.003458103	0.010414399	10.41439864	3011.592886	3.478796263
2.788410765	0.002788411	0.007895152	7.895151556	2831.416251	3.45200372
2.675983502	0.002675984	0.007894937	7.8949368	2950.293526	3.469865226
2.726253356	0.002726253	0.007913602	7.913601605	2902.738878	3.46280797
1.988019792	0.00198802	0.006385821	6.385820808	3212.151526	3.506796024
1.976706184	0.001976706	0.006411565	6.411565252	3243.560071	3.511021946
1.863699936	0.0018637	0.006398518	6.398517933	3433.233971	3.535703401
1.496826906	0.001496827	0.005371764	5.371763541	3588.767359	3.554945306
1.442913477	0.001442913	0.005372955	5.372954501	3723.684466	3.570972873
1.560906557	0.001560907	0.005357285	5.357284585	3432.162267	3.535567812

Estradiol

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
0.18900723	0.000189007	0.001055752	1.055752462	5585.778187	3.747083686
0.120159477	0.000120159	0.001931092	1.93109176	16071.07322	4.20604488
0.119362654	0.000119363	0.000948001	0.948001293	7942.193464	3.899940462
0.12258624	0.000122586	0.000958133	0.95813267	7815.988751	3.892983926
0.089521739	8.95217E-05	0.000780677	0.780676546	8720.524836	3.940542623
0.060182744	6.01827E-05	0.000899143	0.899142908	14940.21115	4.174356735
0.066884598	6.68846E-05	0.000653404	0.653404214	9769.128244	3.989855811
0.065276677	6.52767E-05	0.000663518	0.663518043	10164.70306	4.007094696
0.04698694	4.69869E-05	0.000589183	0.589182736	12539.28713	4.098272847
0.051719563	5.17196E-05	0.000571535	0.571534869	11050.65151	4.043387883

Table A2 (cont.)**Diazepam**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
21.62891409	0.021628914	0.013830377	13.83037654	639.439247	2.805799289
20.71272496	0.020712725	0.014177934	14.17793436	684.5035784	2.835375723
8.894670798	0.008894671	0.010238875	10.23887498	1151.124669	3.061122361
7.695446825	0.007695447	0.010307425	10.30742477	1339.41862	3.126916332
5.206882218	0.005206882	0.008144991	8.144990963	1564.274094	3.194312853
2.592998595	0.002592999	0.006921024	6.921024113	2669.119886	3.426368081
3.698941119	0.003698941	0.006935969	6.93596928	1875.122922	3.273029743
5.217304124	0.005217304	0.008155204	8.155203804	1563.106848	3.193988666
2.56598179	0.002565982	0.005762408	5.762408342	2245.693389	3.35135046
2.705187457	0.002705187	0.005832738	5.832738105	2156.130841	3.333675112

Naphthalene

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
12.6292347	0.012629235	0.010773761	10.77376094	853.0810615	2.930990301
12.43500505	0.012435005	0.010458297	10.45829723	841.0368303	2.924815015
3.961395555	0.003961396	0.009495531	9.495531186	2397.016671	3.379671054
5.586216111	0.005586216	0.008852398	8.85239762	1584.685849	3.19994318
1.713379641	0.00171338	0.007371164	7.371164268	4302.119678	3.633682488
2.594352285	0.002594352	0.006977526	6.977525688	2689.505866	3.429672496
1.27633871	0.001276339	0.004393813	4.393812554	3442.512964	3.536875584
1.002582059	0.001002582	0.004466338	4.466338245	4454.835597	3.648831681
0.790777116	0.000790777	0.003305257	3.305257431	4179.758574	3.621151197
0.587345267	0.000587345	0.003195768	3.195767662	5441.037571	3.735681725

2,4-dinitrotoluene

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
45.193074	0.045193074	0.051671048	51.67104827	1143.339979	3.05817539
43.99119107	0.043991191	0.051538595	51.538595	1171.566256	3.068766854
13.52667944	0.013526679	0.031833777	31.83377705	2353.406627	3.371696972
14.10402588	0.014104026	0.030081251	30.08125059	2132.813059	3.328952791
6.85402737	0.006854027	0.024385422	24.38542235	3557.823895	3.551184448
6.73624288	0.006736243	0.024613664	24.61366397	3653.915753	3.56275853
4.303668959	0.004303669	0.019715734	19.71573374	4581.145514	3.660974087
4.916312047	0.004916312	0.020268442	20.26844209	4122.692354	3.615180928
3.47112573	0.003471126	0.016288713	16.28871287	4692.631191	3.671416423
2.586332894	0.002586333	0.016764188	16.76418795	6481.836884	3.811698098

Table A2 (cont.)**Nitrobenzene**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
1232.105438	1.232105438	0.06440096	64.40096029	52.26903341	1.718244469
1226.109706	1.226109706	0.065287006	65.28700597	53.2472793	1.726297422
1053.056773	1.053056773	0.063657506	63.65750596	60.45021275	1.781397834
1072.570758	1.072570758	0.063860237	63.86023714	59.53941654	1.774804574
1008.287546	1.008287546	0.061480296	61.48029624	60.97496343	1.785151549
954.0930394	0.954093039	0.060741611	60.74161067	63.66424254	1.803895576
862.2532322	0.862253232	0.058493753	58.49375257	67.8382526	1.831474653
873.2352291	0.873235229	0.059158978	59.15897791	67.7468979	1.830889414
747.9168923	0.747916892	0.056015545	56.0155447	74.89541321	1.874455221
735.9998565	0.735999857	0.057288306	57.28830594	77.83738737	1.89118825

Lindane

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
5.662747816	0.005662748	0.007183971	7.183971158	1268.636957	3.103337359
4.788971578	0.004788972	0.008359093	8.359093434	1745.488211	3.24191692
2.007158626	0.002007159	0.005393636	5.393635894	2687.199619	3.429299929
1.833239764	0.00183324	0.005304292	5.304291993	2893.39785	3.461408154
1.010302145	0.001010302	0.003883826	3.883826362	3844.222624	3.58480853
1.076013173	0.001076013	0.003863173	3.863172934	3590.265464	3.555126561
0.76290487	0.000762905	0.003075685	3.075685096	4031.544712	3.605471481
0.660896178	0.000660896	0.002954029	2.954029166	4469.732563	3.650281539
3.615569159	0.003615569	0.005031891	5.031890763	1391.728533	3.143554531
2.448100856	0.002448101	0.006193363	6.193363191	2529.864395	3.403097243
2.875472689	0.002875473	0.004592715	4.592715048	1597.203502	3.203360254
2.141092382	0.002141092	0.00522027	5.220269836	2438.133861	3.387057546

Bisphenol A

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
170.9122581	0.170912258	0.03992445	39.92445027	233.5961779	2.368465733
157.6959174	0.157695917	0.044718699	44.7186995	283.5755056	2.452668715
81.75083136	0.081750831	0.034408897	34.40889707	420.8996593	2.624178574
79.23382989	0.07923383	0.034886462	34.88646228	440.2975638	2.643746283
53.09147997	0.05309148	0.029860811	29.86081129	562.4407402	2.750076771
59.76186894	0.059761869	0.0285266	28.52659957	477.3378087	2.678825835
40.07028924	0.040070289	0.024486448	24.48644773	611.0873715	2.786103309
34.53901462	0.034539015	0.025304478	25.30447825	732.6346315	2.864887444
25.26629203	0.025266292	0.021912439	21.91243864	867.2597708	2.938149201
22.90715995	0.02290716	0.022273444	22.27344415	972.3354705	2.987816129

Table A2 (cont.)**TCE**

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
60.85759093	0.060857591	0.004762022	4.762022372	78.24861779	1.893476675
48.02925466	0.048029255	0.004332903	4.332903173	90.2138333	1.955273137
33.59486403	0.033594864	0.0039086	3.908600413	116.3451773	2.065748386

1-Heptanol

C_e (ug/L)	C_e (mg/L)	q (ug/mg)	q (mg/kg)	K_d	$\log K_d$
83.31605035	0.08331605	0.002878115	2.878114525	34.54453869	1.538379397
50.39603142	0.050396031	0.002867798	2.867797852	56.90523184	1.755152197
35.04907678	0.035049077	0.002401408	2.40140766	68.51557531	1.835789309
28.03250693	0.028032507	0.002111353	2.111353043	75.31802447	1.87689892

Table A3: Activated carbon competitive isotherm data of 10 trace organic compounds**Atrazine (To et al., 2008)**

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
30.661507	33.778321	0.0306615	33778.321	1101652.3	6.0420446
1.9401091	12.401822	0.0019401	12401.822	6392332.4	6.8056594
0.7190289	8.6497496	0.000719	8649.7496	12029766	7.0802572
0.2372795	5.7839442	0.0002373	5783.9442	24376077	7.3869638
0.1188512	4.5001913	0.0001189	4500.1913	37864065	7.5782272
0.0799391	3.8967829	7.994E-05	3896.7829	48746899	7.687947
71.986873	46.04554	0.0719869	46045.54	639638	5.8059343
16.97323	27.252675	0.0169732	27252.675	1605626.9	6.2056446
2.8041184	14.176032	0.0028041	14176.032	5055432.8	6.7037583
0.7976834	8.9819184	0.0007977	8981.9184	11260004	7.0515386
0.1817246	5.2501508	0.0001817	5250.1508	28890703	7.4607581
0.1191763	4.5046554	0.0001192	4504.6554	37798242	7.5774716
0.0754183	3.8152994	7.542E-05	3815.2994	50588526	7.704052
0.047619	3.2287841	4.762E-05	3228.7841	67804467	7.8312583

Estradiol

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.1208699	0.0173033	0.0001209	17.303284	143156.24	5.1558103
0.1119093	0.0185092	0.0001119	18.509193	165394.64	5.2185214
0.0938763	0.0133073	9.388E-05	13.307339	141754.05	5.1515355
0.0564383	0.0110608	5.644E-05	11.06082	195980.81	5.2922136
0.05073	0.0108894	5.073E-05	10.889384	214653.68	5.3317383
0.0255638	0.0089811	2.556E-05	8.9810597	351319.98	5.5457028
0.0432971	0.0093108	4.33E-05	9.3107916	215044.42	5.3325282
0.070825	0.0128337	7.083E-05	12.833731	181203.3	5.2581661
0.0828102	0.0132937	8.281E-05	13.293697	160532.09	5.2055619
0.0650861	0.0087898	6.509E-05	8.7898162	135049.01	5.1304914
0.0635909	0.008887	6.359E-05	8.8870341	139753.15	5.1453616

Table A3 (cont.)**Benzene**

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
4.8174609	0.4199928	0.0048175	419.99275	87181.351	4.9404236
4.8057834	0.4198564	0.0048058	419.85643	87364.825	4.9413366
1.3251332	0.2208847	0.0013251	220.88471	166688.68	5.2219061
1.3941148	0.229786	0.0013941	229.78602	164825.75	5.2170251
0.5615831	0.1523199	0.0005616	152.31994	271233.12	5.4333427
0.5716649	0.1463927	0.0005717	146.39265	256081.21	5.4083777
0.4379333	0.1169681	0.0004379	116.96809	267091.14	5.4266595
0.323325	0.1165976	0.0003233	116.59762	360620.47	5.5570504
0.4962514	0.0942079	0.0004963	94.207935	189839.14	5.2783858
0.7179693	0.0931395	0.000718	93.139542	129726.35	5.1130282

2,4 dinitrotoluene

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.1069377	0.2362693	0.0001069	236.2693	2209409.9	6.3442763
0.0440742	0.2363456	4.407E-05	236.3456	5362447.5	6.7293631
0.0265646	0.1186469	2.656E-05	118.64689	4466347.6	6.6499525
0.0489826	0.1187201	4.898E-05	118.72006	2423721.1	6.3844826
0.0142054	0.0589403	1.421E-05	58.940313	4149149.1	6.617959
0.0181266	0.0592168	1.813E-05	59.216791	3266843.5	6.5141283
0.0108695	0.0470715	1.087E-05	47.071451	4330605.1	6.6365486
0.0143177	0.0471518	1.432E-05	47.151801	3293249.4	6.5176246
0.1049717	0.5132185	0.000105	513.2185	4889114.7	6.6892302
0.048716	0.4393422	4.872E-05	439.34219	9018445.6	6.9551317
0.0800734	0.3518047	8.007E-05	351.80472	4393527.3	6.6428133
0.052731	0.2912157	5.273E-05	291.21569	5522665.5	6.7421487
0.0480022	0.268013	4.8E-05	268.01301	5583347.9	6.7468947
0.0137541	0.0671745	1.375E-05	67.174494	4883972	6.6887732
0.013361	0.0537412	1.336E-05	53.741167	4022235.2	6.6044675

Table A3 (cont.)**TCE**

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
1.8865851	0.0523161	0.0018866	52.316122	27730.592	4.4429591
1.8011818	0.0521186	0.0018012	52.118603	28935.781	4.4614352
1.6924186	0.033311	0.0016924	33.310972	19682.467	4.2940795
1.3611965	0.0345918	0.0013612	34.591828	25412.811	4.4050527
1.3294591	0.02576	0.0013295	25.759981	19376.287	4.2872706
3.2726493	0.0997438	0.0032726	99.743845	30478.013	4.4839866
3.2684645	0.1057376	0.0032685	105.73765	32350.863	4.5098859
2.6177638	0.0710348	0.0026178	71.034769	27135.668	4.4335405
2.6090649	0.0681288	0.0026091	68.128754	26112.326	4.4168456
2.3305454	0.0533241	0.0023305	53.32405	22880.503	4.3594656
2.3017061	0.0539032	0.0023017	53.903211	23418.807	4.3695648
2.133466	0.0426493	0.0021335	42.649294	19990.613	4.3008261
13.693115	0.3484174	0.0136931	348.41743	25444.716	4.4055976
13.777033	0.3587083	0.013777	358.70827	26036.686	4.4155857
12.292463	0.2747108	0.0122925	274.71075	22347.902	4.3492368
11.149187	0.3036141	0.0111492	303.61411	27231.95	4.4350787
7.7624738	0.0876543	0.0077625	87.654296	11292.057	4.0527731
10.583787	0.0909686	0.0105838	90.968634	8595.0929	3.9342506

Table A3 (cont.)**1-Heptanol**

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
31.123394	0.4991881	0.0311234	499.18805	16038.998	4.2051772
18.212559	0.3602818	0.0182126	360.28178	19782.052	4.2962713
19.922562	0.3162354	0.0199226	316.23541	15873.231	4.2006653
17.12611	0.2490425	0.0171261	249.04249	14541.685	4.1626147
18.091466	0.2514519	0.0180915	251.45188	13898.922	4.1429811
15.773158	0.2089068	0.0157732	208.90684	13244.452	4.122034
15.812079	0.1990822	0.0158121	199.08222	12590.515	4.1000435
14.711451	0.1843357	0.0147115	184.3357	12530.083	4.0979539
14.420731	0.1823476	0.0144207	182.34755	12644.821	4.1019127
12.559268	0.1287462	0.0125593	128.74624	10251.094	4.0107702
11.418809	0.0993787	0.0114188	99.378691	8703.0698	3.9396725
10.465228	0.0753504	0.0104652	75.350392	7200.0714	3.8573368
9.0967948	0.0514606	0.0090968	51.460566	5656.9998	3.7525862
8.5534286	0.0445701	0.0085534	44.570092	5210.7867	3.7169033
7.2218712	0.0382585	0.0072219	38.258474	5297.5846	3.7240779
6.3532685	0.0346633	0.0063533	34.663306	5455.9801	3.7368728
6.2695648	0.0317229	0.0062696	31.722931	5059.8299	3.7041359
14.95411	0.2484618	0.0149541	248.46175	16614.948	4.220499
13.595034	0.1909229	0.013595	190.92286	14043.574	4.1474776
9.652292	0.0523984	0.0096523	52.398418	5428.5985	3.7346877
8.7580464	0.0406129	0.008758	40.612926	4637.2129	3.666257
8.1183039	0.0266203	0.0081183	26.620282	3279.0448	3.5157473

Diazepam

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.1395722	0.2773198	0.0001396	277.3198	1986927.1	6.2981819
0.155966	0.2260489	0.000156	226.04894	1449347.3	6.1611725
0.1696491	0.1843786	0.0001696	184.37858	1086823.4	6.036159
0.1524556	0.1910329	0.0001525	191.03291	1253039.4	6.0979647
0.0398783	0.1364937	3.988E-05	136.49373	3422754.7	6.5343758
0.0723668	0.1395	7.237E-05	139.49997	1927678.2	6.2850345
0.0377721	0.1097509	3.777E-05	109.7509	2905609.4	6.4632372
0.0357111	0.1097711	3.571E-05	109.77111	3073869.1	6.4876854

Table A3 (cont.)**Nitrobenzene**

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.97759	1.0751476	0.0009776	1075.1476	1099794	6.0413113
0.9152562	0.6112626	0.0009153	611.26261	667859.54	5.8246851
0.9373935	0.4556226	0.0009374	455.62256	486052.64	5.6866833
0.9662831	0.3643389	0.0009663	364.33888	377051.92	5.5764012
3.195254	3.5780046	0.0031953	3578.0046	1119787.2	6.0491355
3.2883414	3.6003193	0.0032883	3600.3193	1094873.9	6.0393641
4.102382	2.2000825	0.0041024	2200.0825	536293.91	5.7294029
3.9688862	2.1835498	0.0039689	2183.5498	550166.9	5.7404945
1.1671864	1.0667134	0.0011672	1066.7134	913918.64	5.9609075
0.7884644	0.6075339	0.0007885	607.53387	770527.97	5.8867884
0.820682	0.6135376	0.0008207	613.53762	747594.79	5.8736663

Naphthalene

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.5502966	0.1572073	0.0005503	157.20732	285677.43	5.4558759
0.5429718	0.1367315	0.000543	136.73152	251820.67	5.4010914
0.4130808	0.1159145	0.0004131	115.91449	280609.75	5.4481028
0.3776519	0.1019884	0.0003777	101.98836	270059.18	5.4314589

Bisphenol A

C_e (ug/L)	q (mg/g)	C_e (mg/L)	q (mg/kg)	K_d	$\log K_d$
0.1736978	0.0882108	1.74E-04	88.210803	507840.59	5.7057274
0.1314095	0.0490923	1.31E-04	49.092316	373582.71	5.5723868
0.1514128	0.0449854	1.51E-04	44.985433	297104.62	5.4729094
0.1491466	0.0363317	1.49E-04	36.331745	243597.59	5.386673
0.1127538	0.0296084	1.13E-04	29.608409	262593.43	5.4192839
0.1597079	0.025861	1.60E-04	25.860979	161926.74	5.2093186
0.1087793	0.0226485	1.09E-04	22.648483	208205.74	5.3184927
0.0966691	0.0229362	9.67E-05	22.936239	237265.51	5.3752346
0.0521536	0.0115507	5.22E-05	11.550738	221475.42	5.3453255
0.0512199	0.0084793	5.12E-05	8.479319	165547.31	5.2189221
0.3031492	0.1148277	3.03E-04	114.82766	378782.7	5.5783901
0.2306125	0.1028516	2.31E-04	102.85155	445992.93	5.649328
0.1802571	0.0740867	1.80E-04	74.086676	411005.59	5.6138477
0.189101	0.0760293	1.89E-04	76.029262	402056.35	5.6042869
0.1555274	0.0553231	1.56E-04	55.323061	355712.61	5.5510993
0.1636879	0.0378448	1.64E-04	37.844778	231200.84	5.3639894
0.1410478	0.0279712	1.41E-04	27.971248	198310.4	5.2973455
0.1325186	0.0282018	1.33E-04	28.201778	212813.66	5.3279995

A4: XPS Data:

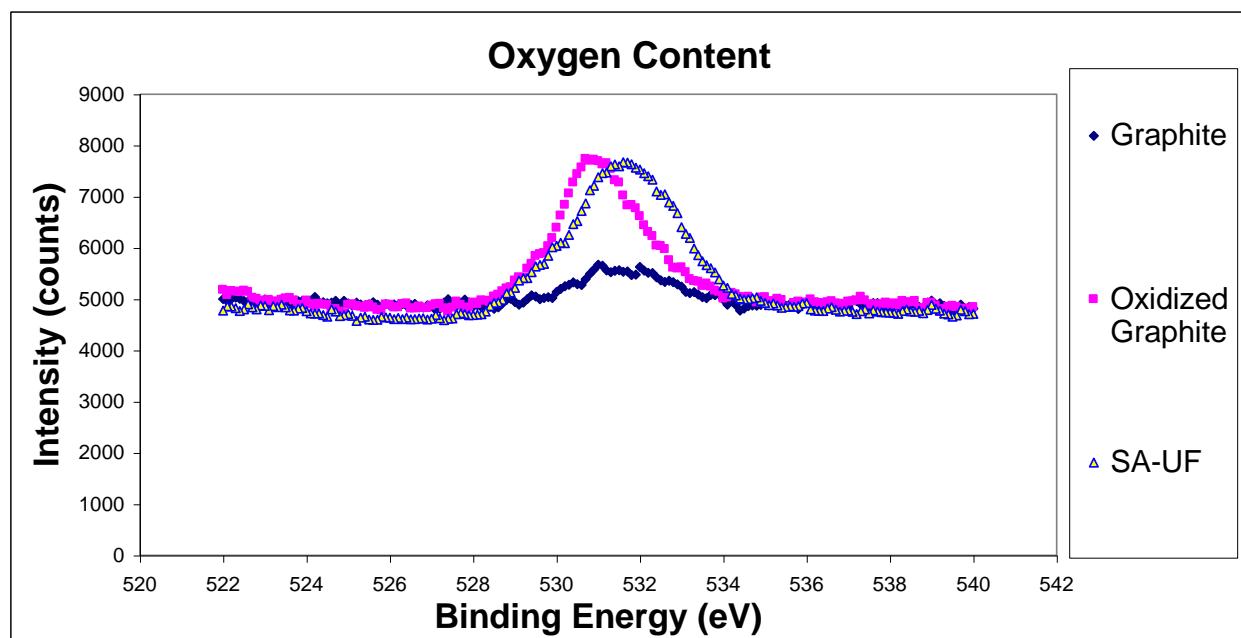


Figure A1: Oxygen content of carbons measured by XPS

Table A4: Mass Concentration of oxygen and carbon on three different carbons

Carbon	Graphite	Oxidized Graphite	SA-UF
Mass Conc (%)			
Oxygen	1.22	4.52	7.21
Carbon	98.78	95.48	92.79