Theoretical Aspects of Multicomponent Adsorption Equilibria
Part I: Development of a Computationally Fast Adsorption Isothem Model that can be Applied to Systems Which Demonstrate Energetic Heterogeneity

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Theoretical Aspects of Multicomponent Adsorption Equilibria
Part I: Development of a Computationally Fast Adsorption Isotherm Model that can be
Applied to Systems Which Demonstrate Energetic Heterogeneity

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Theoretical aspects of multicomponent adsorption equilibria are investigated with a focus on the thermodynamically based Adsorbed Solution Theory (AST). This report details the development of a computationally fast, heterogeneous, binary adsorption relation using the Langmuir isotherm with the Ideal AST (IAT) as the local, single component and mixture adsorption models. Equations are developed for binary systems which display both equal and unequal saturation/monolayer coverages. Additionally, equations for random and perfectly correlated heterogeneous site matching are developed. Theoretical results plotted as deviations from exact solutions indicate that the developed equations generate consistently better results than analogous homogeneous relations.
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NOMENCLATURE

\( A_s \) = Area of the Adsorbent
\( b \) = Langmuir isotherm fitting parameter (1/MPa)
\( F(\epsilon) \) = Cumulative (integral) energy distribution
\( f(\epsilon) \) = Single probability density function
\( g(\epsilon_1, \epsilon_2) \) = Joint probability density function
\( H, L \) = Exponential curve fitting parameters
\( K, u \) = Freundlich parameters
\( m \) = \( \sigma_2^2/\sigma_1^2 \)
\( N \) = Global amount adsorbed (moles/kg)
\( n \) = Amount adsorbed in local homogeneous patch (moles/kg)
\( P \) = Partial pressure (MPa)
\( Q \) = Saturation adsorption value (moles/kg)
\( R \) = Universal gas constant (kJ/mole*K or MPa*L/mole*K)
\( R_i \) = Constant separation factor isotherm parameter
\( T \) = Temperature (K)
\( X \) = Adsorbed phase mole fraction
\( Y \) = Vapor phase mole fraction
\( Z \) = Total number of components in the mixture

GREEK LETTERS

\( \gamma_i \) = Activity coefficient in the adsorbed phase
\( \epsilon \) = Adsorption site energy (kJ/mole)
\( \mu_p \) = \( p^{th} \) central moment of a probability density function
\( \pi \) = Spreading pressure
\( \rho \) = Site matching parameter or covariance of the probability density function
\( \sigma \) = Square root of the variance of a probability density function
\( \sigma_e \) = Excess area of mixing (kg/mole)
\( \phi \) = Fugacity coefficient
\( \psi \) = Spreading pressure function \( \pi A_s/RT \) (moles/kg)
THEORETICAL ASPECTS OF MULTICOMPONENT ADSORPTION EQUILIBRIA

INTRODUCTION

The need exists for a method to accurately model mixture adsorption equilibria. This need arises not only from an academic desire to understand the fundamentals of the adsorption process, but also from the more pragmatic requirement to provide the nucleus of mathematical models which describe dynamic adsorption systems. In the course of generating results for a dynamic adsorption system using accepted mathematical models, adsorption isotherm relationships are regenerated hundreds of thousands of times. This points to the importance of generating equilibrium adsorption data with a minimum of computational burden and a high degree of accuracy.

Several models have been developed to describe adsorption systems. The simpler models often are limited in their applicability and accuracy, while the more complicated systems become too cumbersome or unsolvable under some conditions. Several of the models are ultimately correlative in nature. While these models provide important information in understanding the mixture adsorption process, their utility is limited due to the paucity of, and difficulty in obtaining, mixture adsorption data. As a result, models that can predict mixture adsorption from single component adsorption data are becoming more popular.

This study will investigate one of the more popular mixture adsorption theories, the Adsorbed Solution Theory (AST) (Myers and Prausnitz, 1965), and its ability to predict mixture adsorption from single component data. In an effort to isolate pure theory from the effects and inherent errors associated with experimental data, this work will focus mostly on theoretical manipulations and evaluations of the theory. This report details the development of a computationally fast adsorption isotherm model that can be applied to systems which demonstrate energetic heterogeneity. Since this work focuses primarily on the AST, a background section providing the details of this theory is included below.

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I: ADSORBED SOLUTION THEORY

BACKGROUND

The Adsorbed Solution Theory as proposed by Myers and Prausnitz (1965) is based on the assumption that an adsorbed mixture can be represented as a homogeneous, well-mixed, two-dimensional phase. This assumption leads to a series of thermodynamic relations confined to a two-dimensional plane. Standard three-dimensional properties (Pressure and Volume) are reduced to areal properties (Spreading Pressure and Area). Functions which describe nonidealities, such as activity coefficients and excess volume of mixing, also translate to two-dimensional analogs. A consequence of this approach is that adsorbate/adsorbent interactions in the AST are completely separate from adsorbate/adsorbate interactions which are modeled as mixture nonidealties. The theory uses the spreading pressure of the mixture as the reference state. The resulting equations are as follows:

\[
\frac{1}{n_{\text{TOT}}} = \sum_{i=1}^{Z} \frac{X_i}{n_i^o} + \sigma^o \quad (1.1)
\]

\[
1 = \sum_{i=1}^{Z} X_i \quad (1.2)
\]

\[
P_i \phi_i = X_i P_i^o \gamma_i \phi_i^o \quad (1.3)
\]

\[
P_i^o = \mathcal{F}_1(n_i^o) \quad (1.4)
\]

\[
\psi = \mathcal{F}_2(n_i^o) \quad (1.5)
\]

\[
n_i = X_i n_{\text{TOT}} \quad (1.6)
\]

The function \( \mathcal{F}_1(n_i^o) \) is determined directly from the single component isotherm for each specie. The function \( \mathcal{F}_2(n_i^o) \) is found using the following equation:
\[ \mathcal{F}_2(n^*) = \int_0^n \frac{d\ln P}{d\ln n} \ dn \]  

(1.7)

Or when \( \psi \) is a function of \( P_i^* \), the \( \mathcal{F}_2 \) integral can take the form:

\[ \mathcal{F}_2(P_i^*) = \int_0^{P_i^*} \frac{n}{P} \ dP \]  

(1.8)

The integrand of Equation 1.7 or 1.8 is obtained using the single component isotherm for each specie using the mixture spreading pressure as the standard state. A more rigorous derivation of the thermodynamics associated with the AST can be found in Van Ness (1969).

If it is assumed that the vapor and adsorbed phases are ideal (Ideal Adsorbed Solution Theory - IAST) \( \sigma^\circ \) becomes 0 and \( \gamma_i, \phi_i, \) and \( \phi_i^\circ \) all become 1. The resulting 5Z+2 unknowns in 4Z+2 equations can be determined if Z conditions are known (i.e. partial pressures for each component). Thus, the method can predict multi-component adsorption values, requiring only single component isotherm information for each specie in the mixture.

In the IAST, the resultant equations are greatly simplified. However, this simplification restricts the application of the theory to adsorbates that behave ideally as a mixture. In adsorption systems that deviate significantly from ideality, a model which describes the adsorbed phase activity coefficients and excess area of mixing must be employed. Costa et al. (1981) developed the Real Adsorbed Solution Theory (RAST) to model non-ideal adsorption systems using liquid phase activity coefficient models. Talu and Zwiebel (1986) followed with a proof that the adsorbed phase activity coefficients must depend on the system spreading pressure in order to be thermodynamically consistent. Gamba et al. (1989 and 1990) then modified the RAST to minimize the errors associated with low pressure regions and lack of thermodynamic consistency. However, use of this modified RAST is limited to applications requiring calculations in a small concentration range where binary adsorption data already exist.

II: THE DEVELOPMENT OF A FAST HETEROGENEOUS BINARY ADSORPTION ISOTHERM

BACKGROUND

An important practical purpose for an adsorption equilibrium equation is to aid in the modeling of adsorption-based separation systems. The models that describe such systems
accurately are often complex and typically involve several layers of nested iterations. At the root of all the calculations is the equilibrium relation which may get called hundreds of thousands of times in any given simulation. Therefore, a simple, fast equilibrium relation is desired to reduce the computational burden and create a manageable overall model.

In an effort to address the need for a computationally fast equilibrium relation, O'Brien and Myers (1985) developed a fast adsorption equilibrium algorithm (FastIAS) based upon the Ideal Adsorbed Solution Theory (Myers and Prausnitz, 1965). Although this model is many times faster than the analogous iterative solution, it is restricted to homogeneous adsorption.

To complicate matters further, the single component isotherm relation used in the FastIAS, the TALAN (Taylor series expansion of the Langmuir isotherm) (O'Brien and Myers, 1984), is a heterogeneous isotherm model. This application of a heterogeneous single component isotherm relation in a homogeneous mixture adsorption model seems somewhat inappropriate. Additionally, since the majority of adsorption systems of interest can be categorized as heterogeneous, a new, fast algorithm is needed. This section details the development of a fast heterogenous binary adsorption equilibrium relation. Because of the need for speed, only those solutions that produce a single analytical equation are sought.

Adsorbent heterogeneity has been historically modelled by describing the surface as either patchwise or randomly heterogeneous (Jaroniec and Madey, 1988, p.8). Patchwise heterogeneity assumes that like sites are grouped together to form independent homogeneous patches, while random heterogeneity assumes that the individual homogeneous sites are ungrouped and randomly distributed across the surface. The patchwise model has been favored by researchers in the past because the location or relative placement of the sites is not needed. In the random model, site layout is needed to determine lateral interactions between nearest neighbors. In the patchwise model, since the patches are homogenous and independent, all nearest neighbors are known to be other homogeneous sites. If, however, lateral interactions are ignored, the two models become identical.

For a patchwise heterogenous adsorbent, Jaroniec and Rudzinski (1975) found that the general equation for heterogeneous adsorption takes on a multiple integral form. In a binary system the equation becomes:

\[
N_1 = \int_{e_2}^{e_1} \int_{e_1}^{e_2} n_1(T,P_1,P_2,e_1,e_2) g(e_1,e_2) \, de_1 \, de_2 \quad (2.1)
\]

Where \( N_1 \) is the global amount of i adsorbed, \( e_1 \) is the energy of adsorption, \( n_1 \) is the local amount of i adsorbed for a patch characterized by the adsorption site energies \( e_1 \) and \( e_2 \), and \( g(e_1, e_2) \) is the joint probability density function for the distribution of adsorption energies. A similar equation can be used to describe the adsorption of component 2. Since transposing the component subscripts in the equations produces the component 2 analog, the remainder of this
work will consider only the component 1 equation.

The choice of the local isotherm function is determined by the type of adsorption (monolayer or multilayer) and lateral interactions that are expected on the homogeneous patch. The majority of past research has focused on monolayer adsorption without lateral interactions (Jaroniec and Madey, 1988, p. 220). While Jaroniec and Madey (1988, pp. 213-215, 220) have investigated heterogenous adsorption using local isotherms with lateral interactions and multi-layer adsorption, the equations are too cumbersome to produce an analytical solution. In addition, as Jaroniec and Madey (1988, p. 220) state, there is no satisfactory theory for multilayer adsorption that can be used as a homogeneous local isotherm. For these reasons, this work will focus on monolayer adsorption with no lateral interactions.

The most common adsorption equation for monolayer adsorption with no lateral interactions is the Langmuir adsorption isotherm. If it is assumed that the saturated molar surface coverage is the same for both components (this can later be modified by adding the $Q_1 \ast Q_2$ correction detailed below), the local isotherm then takes the form:

$$n_1 = \frac{Qb_1P_1e^{E_1/kT}}{1 + b_1P_1e^{E_1/kT} + b_2P_2e^{E_2/kT}}$$  \hspace{1cm} (2.2)

The joint probability density function is typically developed by combining the individual probability density functions with a site matching parameter. The common statistical representation of this site matching parameter is the covariance of the two distributions $\rho$. $\rho$ can take on any value from -1 to +1 and is a measure of how the individual distribution functions are related: $\rho = -1$ indicates perfect negative correlation, $\rho = 0$ indicates random site matching, $\rho = +1$ indicates perfect positive correlation. Perfect positive correlation means that the site with the highest adsorption energy for component 1 will have the highest adsorption energy for component 2 with each subsequent site behaving in the same manner down to the lowest energy site for each component. Perfect negative correlation means that the site with the highest adsorption energy for component 1 will have the lowest adsorption energy for component 2 with each subsequent site behaving in the same contrary manner. Random site matching means that for each adsorption site, the adsorption energy for component 1 has no bearing on, and is completely uncorrelated with, the adsorption energy for component 2. All other values of $\rho$ indicate varying degrees of positive or negative site matching.

The majority of past research into the predetermined selection of covariance has come from the two research groups associated with Jaroniec and Myers. Valenzuela et al. (1988) have indicated that, since most adsorption phenomena are controlled by dispersion forces, adsorbates that are similar in size and shape should behave similarly. Therefore, most mixture adsorption systems, in the absence of chemical interactions, are perfectly correlated ($\rho = 1$).
Indeed, Hoory and Prausnitz (1967) found that, for mixtures of ethane and ethylene on charcoal, a $\rho = 1$ value gave an excellent fit of the data. Jaroniec (Jaroniec and Madey, 1988, pp. 199-202, 203-208) has also investigated perfectly correlated adsorption, but limited their efforts to adsorbates which demonstrate identical single component energy distributions, offset only by the differences in the means of the individual distributions. The remainder of Jaroniec’s work in this area focuses on the $\rho = 0$ case.

What follows is an evaluation of the possible solutions to Equation 1. In an effort to be comprehensive, solutions are sought for each type of site matching: $\rho = 0$, $\rho = 1$ ($\rho = -1$ producing mathematically similar results to $\rho = +1$), and also for $\rho$ as a binary fit parameter. In addition, two methods of solution (direct integration and expansion) are investigated. The evaluation is separated by the solution approach and is subdivided by the method of site matching.

THEORY

Direct Integration Solution

A direct, analytical solution from integration is the preferred method of solution for Equation 2.1. Other methods such as numerical integration, mathematical approximations, and expansions prior to integration may introduce unwanted errors into the calculation. Furthermore, a numerical integration or inclusion of an extended series defeats the purpose of developing a fast, analytical solution to the binary heterogeneous adsorption equation.

A direct integration requires the selection of a joint probability density function. The simplest function, the bivariate normal distribution used by Hoory and Prausnitz (1967), takes the form:

$$g(e_1, e_2) = \frac{1}{2\pi\sigma_1\sigma_2(1 - \rho^2)} e^{\frac{\Psi_1^2 - 2\rho\Psi_1\Psi_2 + \Psi_2^2}{2(1 - \rho^2)}}$$  \hspace{1cm} (2.3)

with:

$$\Psi_1 = \frac{e_1 - \bar{e}_1}{\sigma_1} \quad \Psi_2 = \frac{e_2 - \bar{e}_2}{\sigma_2}$$  \hspace{1cm} (2.4)

$\rho$ as a Fit Parameter. Allowing $\rho$ to be a fit parameter in the binary heterogeneous adsorption equation is appealing because it removes any presuppositions as to the relative interaction of each component with the surface. It does, however, add the complexity that
binary data are required to determine $\rho$. In the strictest sense, the model becomes a correlative rather than a predictive solution. Substituting Equations 2.2 - 2.4 into Equation 2.1 produces a double integral equation which is horribly complex. However, with a little manipulation and substitution, the inner (component 1) integral can be reduced to the form:

$$\int e^{(Ax^2 + Bx)} \frac{dx}{F + Ge^{dx}}$$

(2.5)

with $x$ representing $\epsilon_1$ and $A$, $B$, $F$, $G$, and $D$ being complex functions of $\epsilon_2$, $\rho$, and assorted constants from the above equations. Unfortunately, no analytical solution to this integral has been found.

$\rho=0$ — Random Site Matching. If random site matching is assumed ($\rho=0$), Equation 2.3 takes the form:

$$g(\epsilon_1, \epsilon_2) = \frac{1}{2\pi\sigma_1\sigma_2} e^{-\frac{\epsilon_1^2}{2\sigma_1^2}} e^{-\frac{\epsilon_2^2}{2\sigma_2^2}} = f_1(\epsilon_1) f_2(\epsilon_2)$$

(2.6)

Unfortunately, using Equations 2.2 and 2.6 to solve Equation 2.1 gives an equation of the same form as Equation 2.5. As with the case where $\rho$ is a variable, no solution has been found.

$\rho=1$ — Perfect Positive Correlation. If perfect positive correlation is assumed ($\rho=+1$), then the joint probability density function becomes the individual probability density function of one of the components. Valenzuela et al. (1988) used this information to simplify Equation 1 to a single integral form with respect to the reference component 1:

$$N_1 = \int_{\epsilon_1} n_1(T, P_{1p}, P_{2p}, \epsilon_1, \epsilon_2) f(\epsilon_1) d\epsilon_1$$

(2.7)

Since the energy distributions of the two components are positively correlated, $\epsilon_2$ can be represented as a function of the reference energy $\epsilon_1$. According to Valenzuela et al. (1988), this function ($\epsilon_2^*$) takes the form:

$$F_2(\epsilon_2^*) = F_1(\epsilon_1)$$

(2.8)

The function $F$ represents the cumulative (integral) energy distributions for each component. Equation 2.8 states that the value of $\epsilon_2^*$ can be found by establishing the $\epsilon_1$ value for a given
site, finding the corresponding area under the component 1 probability density function defined by \( \epsilon_1 \), and choosing the component 2 energy value which gives an equivalent area under the component 2 probability density function. If the energy distributions are balanced around the means (i.e. not skewed), then Equation 2.8 simplifies to the linear form:

\[
\epsilon_2^* = \frac{\sigma_2^2}{\sigma_1^2} (\epsilon_1 - \overline{\epsilon}_1) + \overline{\epsilon}_2
\] (2.9)

Jaroniec and Madey (1988, pp. 201-202, 203-208) performed a similar simplification, but restricted the application to identically shaped energy distributions by fixing the slope of the linear relationship to one. Unfortunately, as with the other cases noted above, substituting Equations 2.2 and 2.9 into Equation 2.7 produces the familiar integral form found in Equation 2.5.

**O'Brien and Myers Expansion Procedure**

A second approach to solving Equation 2.1 follows the Taylor series expansion procedure O'Brien and Myers (1984) used to develop the TALAN single component heterogeneous isotherm. The authors expanded the energy dependence of the local Langmuir isotherm about the mean of the energy distribution. This approach is appealing because it does not require an *a priori* specification of a distribution function. In addition, even though this method uses an expansion to approximate a portion of the equation, the penalty for truncation at each level is well defined. Each additional term represents an additional central moment of the probability density function. Therefore, truncating the series at any given point results in the specification of the general form of the probability density function. For example, truncation of the O'Brien and Myers expansion after the second term (as in the TALAN) fixes the shape of the probability function to the symmetric, normal distribution form. Truncation after the third term allows for skewness. Truncation after the fourth term allows for kurtosis. It should be noted, however, that each additional term in the series may provide a superior fit to the data, but adds an additional fit parameter. And, with each additional fit parameter, it becomes more difficult to determine whether a superior fit is provided because the energy distribution is better described by the additional moments, or because more parameters can make a fundamentally poor model seem better.

To apply the approach of O'Brien and Myers to Equation 2.1, their theory must be extended to account for more than one component. In the case of a binary isotherm, the local (binary) isotherm needs to be expanded about the means of the energy distributions of both components yielding:
\[ n_1(T,p_1,p_2,e_1,e_2) = \sum_{p=0}^{\infty} \sum_{q=0}^{\infty} \frac{n_1^{(p,q)}(\bar{e}_1,\bar{e}_2)(e_1 - \bar{e}_1)^p(e_2 - \bar{e}_2)^q}{(p+q)!} \]  

(2.10)

where \( n_1^{(p,q)} \) is the multiple derivative of the local isotherm with respect to \( e_1 \) (p\textsuperscript{th} derivative) and \( e_2 \) (q\textsuperscript{th} derivative). If we follow O'Brien and Myers, substitute Equation 2.10 into Equation 2.1 and then interchange the order of integration:

\[ N_1 = \sum_{p=0}^{\infty} \sum_{q=0}^{\infty} \frac{n_1^{(p,q)}(\bar{e}_1,\bar{e}_2)}{(p+q)!} \int_{e_2} \int_{e_1} (e_1 - \bar{e}_1)^p(e_2 - \bar{e}_2)^q g(e_1,e_2) \, de_1 \, de_2 \]  

(2.11)

And, as with direct integration, the possible solutions can be separated by the method used to match the sites.

\( \rho \) as a Fit Parameter. Using the distribution defined in Equation 2.3 and a variable \( \rho \) value, \( g(e_1, e_2) \) cannot be broken into two independent functions of \( e_1 \) and \( e_2 \). However, the distribution can be divided such that the inner (component 1) integral becomes:

\[ (e_2 - \bar{e}_2)^q \int_{e_1} (e_1 - \bar{e}_1)^p g_1(e_1,e_2) \, de_1 \]  

(2.12)

Since the integral is with respect to \( e_1 \), \( g_1(e_1,e_2) \) becomes essentially \( f_1(e_1) \). Thus, the integral in Equation 2.12 is the p\textsuperscript{th} central moment of \( f_1(e_1) \) or \( \mu_p \). Equation 2.11 then becomes:

\[ N_1 = \sum_{p=0}^{\infty} \sum_{q=0}^{\infty} \frac{n_1^{(p,q)}(\bar{e}_1,\bar{e}_2)}{(p+q)!} \int_{e_2} \mu_p(e_2 - \bar{e}_2)^q f_2(e_2) \, de_2 \]  

(2.13)

Unfortunately, \( \mu_p \) is a function of \( e_2 \) and therefore cannot be removed from the integral. Thus, no solution to this equation has been found.

\( \rho = 0 \) -- Random Site Matching. If random site matching is assumed, the joint probability density function can be separated into two independent functions as shown in Equation 2.6. These separate functions \( f_1(e_1) \) and \( f_2(e_2) \) can be directly substituted for \( g(e_1,e_2) \) in Equation 2.11. Following the work-up of Equations 2.12 and 2.13, and recognizing the p\textsuperscript{th} and q\textsuperscript{th} central moments leads to the equation:
\[ N_1 = \sum_{p=0}^{\infty} \sum_{q=0}^{\infty} n_1^{(p,q)}(\bar{\varepsilon}_1, \bar{\varepsilon}_2) \mu_p \mu_q (p+q)! \]  

(2.14)

\( \mu_p \) and \( \mu_q \) are respectively the \( p^{\text{th}} \) central moment of the component 1 distribution and the \( q^{\text{th}} \) central moment of the component 2 distribution. Following O'Brien and Myers (1984), \( \mu_0 = 1 \) and \( \mu_1 = 0 \), so that Equation 2.14 can be written:

\[ N_1 = n_1(\bar{\varepsilon}_1, \bar{\varepsilon}_2) + \sum_{p=2}^{\infty} \frac{n_1^{(p)}(\bar{\varepsilon}_1)\mu_2}{p!} + \sum_{q=2}^{\infty} \frac{n_1^{(q)}(\bar{\varepsilon}_2)\mu_q}{q!} + \sum_{p=2}^{\infty} \sum_{q=2}^{\infty} \frac{n_1^{(p,q)}(\bar{\varepsilon}_1, \bar{\varepsilon}_2)\mu_p \mu_q}{(p+q)!} \]  

(2.15)

The sums can be extended knowing that \( \mu_2/\sigma^2 \) is the variance of the distribution, while \( \mu_3/\sigma^3 \) represents the skewness and \( \mu_4/\sigma^4 \) represents the kurtosis.

Using Equation 2.2 for the local isotherm and following O'Brien and Myers (1984) in their development of the TALAN (truncating the sums after the second terms) gives:

\[ N_1 = Q \left( \frac{x}{\xi} + \frac{x(1+y)(1-x+y)\sigma_1^2 + xy(y-x-1)\sigma_2^2}{2R^2T^2\xi^3} \right) \]  

(2.16)

where:

\[ x = P_1 b_1 e^{\frac{\varepsilon_1}{RT}} \quad y = P_2 b_2 e^{\frac{\varepsilon_2}{RT}} \]  

(2.17)

\[ \xi = 1 + x + y \]

It can be seen that Equation 2.16 reduces to the TALAN for single component \( (P_2 = y = 0) \):

\[ N_1 = Q \left( \frac{x}{1+x} + \frac{x(1-x)\sigma_1^2}{2R^2T^2(1+x)^3} \right) \]  

(2.18)

Indeed, fitting each component to a TALAN isotherm gives the parameters (\( b, Q, \bar{\varepsilon}, \) and \( \sigma \)) for the binary equation. In addition, assuming a homogeneous surface (\( \sigma_1 = \sigma_2 = 0, \bar{\varepsilon}_1 = \varepsilon_1, \) and \( \bar{\varepsilon}_2 = \bar{\varepsilon}_2 \)) yields the familiar Langmuir (Equation 2.2) form for both binary and single component systems.
\[ \rho = 1 \quad \text{Perfect Positive Correlation.} \quad \text{As shown above, in the case of perfect positive correlation, Equation 2.1 simplifies to Equation 2.7. If Equation 2.9 is used to replace the } \epsilon_2^* \text{ dependence, Equation 2.7 can be expanded a single time about the mean of the component 1 energy distribution giving:} \]

\[
N_1 = \sum_{p=0}^{\infty} \frac{n_1(\epsilon_1, \epsilon_2)}{p!} \int_{\epsilon_1} (\epsilon_1 - \overline{\epsilon_1})^{p} f(\epsilon_1) d\epsilon_1 \tag{2.19}
\]

The integral in Equation 2.19 is the \( p \)-th central moment of the component 1 energy distribution. To evaluate Equation 2.19, Equation 2.2 is used as the local isotherm. Prior to differentiation in Equation 2.19, the \( \epsilon_2 \) dependence is replaced in Equation 2.2 using the \( \epsilon_2^* \) function in Equation 2.9. The sum is then evaluated at \( \overline{\epsilon}_1 \) and \( \overline{\epsilon}_2 \). Truncating the sum after the second term gives:

\[
N_1 = Q \left( \frac{x}{\xi} \right) + \frac{x \sigma_1^2}{2 RT^2 \xi} (1 - \frac{3x + m(y + 2)}{\xi} + \frac{2(x + m y)^2}{\xi^2}) \tag{2.20}
\]

where:

\[
x = P_1 b_1 e^{\frac{\overline{\epsilon}_1}{RT}} \quad y = P_2 b_2 e^{\frac{\overline{\epsilon}_2}{RT}}
\]

\[
m = \frac{\sigma_2^2}{\sigma_1^2} \quad \xi = 1 + x + y \tag{2.21}
\]

As with the \( \rho = 0 \) case, Equation 2.20 reduces to the TALAN (Equation 2.18) for single component (\( P_2 = y = 0 \)), and to the basic Langmuir for a homogeneous surface (\( \sigma_1 = \sigma_2 = 0 \), \( \overline{\epsilon}_1 = \epsilon_1 \), and \( \overline{\epsilon}_2 = \epsilon_2 \)).

Binary Systems with Different Molar Monolayer Coverages (Q1 * Q2)

Equations 2.16 and 2.20 represent analytical solutions for a binary heterogeneous system which adsorbs locally as a monolayer without lateral interactions. These solutions cannot, however, be applied when the molar monolayer coverages for the two components are not equal. LeVan and Vermeulen (1981) addressed this problem with binary homogeneous adsorption using a Taylor series expansion about the mean of the two different saturation (monolayer) coverages. Frey and Rodrigues (1994) extended this solution to allow for more than two components. Since this research is focusing on binary adsorption, only the LeVan and Vermeulen, binary solution will be considered here. The equations developed from the LeVan and Vermeulen effort include both a two-term and three-term Taylor expansion. The
primary equation for the three-term expansion takes the form:

\[ n_1 = \frac{\overline{Q} x}{\xi} + \Delta_{L2} (1 + \Delta_{L3}) \]  (2.22)

with:

\[ \Delta_{L2} = (Q_1 - Q_2) \frac{xy}{(x + y)^2} \ln(\xi) \]

\[ \Delta_{L3} = \frac{(Q_1 - Q_2)}{(Q_1 + Q_2) x + y} \left( \frac{y^2 + 2y - 4x - x^2}{x + y} \ln(\xi) \right. \\
+ \left. \frac{3x^2 + 4x + xy - 2y - 2y^2}{\xi} \right) \]  (2.23)

\[ \overline{Q} = \frac{Q_1 x + Q_2 y}{x + y} + \frac{2xy(Q_1 - Q_2)^2}{(Q_1 + Q_2)(x + y)^2} \left( \frac{1}{x + y} + \frac{1}{2} \ln(\xi) - 1 \right) \]  (2.24)

For the two term expansion, the following simplifications to the above equations apply:

\[ \overline{Q} = \frac{Q_1 x + Q_2 y}{x + y} \]  (2.25)

\[ \Delta_{L3} = 0 \]

This analytical solution for different monolayer coverages assumes ideal adsorbed solution theory thermodynamic interactions. As a result, this solution may be used as the local isotherm in the proposed heterogeneous model. The O'Brien and Myers (1984) expansion, detailed in equation 2.11, can be applied using equation 2.22 as the local isotherm, and the ultimate solution obtained. Due to the complexity of the resultant equations, only the two-term Taylor expansion form of Equation 2.22 will be considered.

\( \rho = 0 \) -- Random Site Matching. When random site matching is assumed, Equation 2.22 is used for the local isotherm in Equation 2.15. Truncating the expansion after the second term and using the definitions of Equations 2.17, 2.23, 2.24, and 2.25 gives:
\[ N_1 = \frac{\bar{Q}x}{\xi} + \Delta_{L2} + \frac{\sigma_1^2}{2R^2T^2}(\alpha + \beta) + \gamma \frac{(Q_1 - Q_2)xy}{2R^2T^2(x + y)^2} \]  

(2.26)

where:

\[ \alpha = \frac{3x^2(Q_1 - \bar{Q})}{(x + y)\xi} + \frac{2x^3(\bar{Q} - Q_1)}{(x + y)^2\xi} + \frac{2x^3(\bar{Q} - Q_1)}{(x + y)\xi^2} + \frac{\bar{Q}x}{\xi} - \frac{3\bar{Q}x^2}{\xi^2} + \frac{2\bar{Q}x^3}{\xi^3} \]  

(2.27)

\[ \beta = \frac{xy(Q_2 - \bar{Q})}{(x + y)\xi} + \frac{2xy^2(\bar{Q} - Q_2)}{(x + y)^2\xi} + \frac{2xy^2(\bar{Q} - Q_2)}{(x + y)\xi^2} - \frac{\bar{Q}xy}{\xi^2} + \frac{2\bar{Q}xy^2}{\xi^3} \]  

(2.28)

\[ \gamma = (\sigma_1^2 + \sigma_2^2)\ln\xi + (x\sigma_1^2 + y\sigma_2^2)\left(\frac{3}{\xi} - \frac{6\ln\xi}{(x + y)}\right) + \] 

\[ + (x^2\sigma_1^2 + y^2\sigma_2^2)\left(\frac{6\ln\xi}{(x + y)^2} - \frac{4}{(x + y)\xi\xi'} - \frac{1}{\xi^2}\right) \]  

(2.29)

\[ \rho = +1 \text{ – Perfect Positive Correlation.} \]  

When perfect positive correlation is assumed, Equation 2.22 is used for the local isotherm in Equation 2.19. Truncating the expansion after the second term and using the definitions of Equations 2.21, 2.23, 2.24, and 2.25 gives:

\[ N_1 = \frac{\bar{Q}x}{\xi} + \Delta_{L2} + \frac{\sigma_1^2}{2R^2T^2}\left[\frac{x}{\xi} + \frac{xy(Q_1 - Q_2)\ln\xi}{(x + y)^2}(\beta - \gamma)\right] \]  

(2.30)

where:

\[ \alpha = (\bar{Q} + \bar{Q}')\left(1 - \frac{2(x + my)}{\xi}\right) + (\bar{Q}' + \bar{Q}'') - \frac{\bar{Q}(x + m^2y)}{\xi} + \frac{2\bar{Q}(x + my)^2}{\xi^2} \]  

(2.31)

\[ \beta = m(m + 1)^2\ln(\xi) + \frac{(x + m^2y) + 2(x + my)(m + 1)}{\xi} - \frac{(x + my)^2}{\xi^2} \]  

(2.32)
\[
\gamma = \frac{4(x+my)^2}{(x+y)\xi} + \frac{2m(2x+3my-y)\ln(\xi)}{(x+y)} \tag{2.33}
\]

\[
\overline{Q'} = \frac{Q_1x + Q_2my - \overline{Q}(x+my)}{(x+y)} \tag{2.34}
\]

\[
\overline{Q''} = \frac{Q_1x + Q_2m^2y - \overline{Q}(x+m^2y)}{(x+y)} - \frac{2(x+my)}{(x+y)} \overline{Q'} \tag{2.35}
\]

It can be seen that, with a little mathematical manipulation, Equations 2.26 and 2.30 reduce to the TALAN (Equation 2.18) in the limit of one component \((P_2 = y = 0)\), to the Langmuir (Equation 2.2) when the surface is homogeneous \((\alpha_1 = \alpha_2 = 0, \bar{\varepsilon}_1 = \varepsilon_1, \text{ and } \bar{\varepsilon}_2 = \varepsilon_2)\), and to Equations 2.16 and 2.20 when the monolayer coverages are equal \((Q_1 = Q_2 = Q)\). These equations may be extended to a higher number of components or to a greater level of complexity in any of the Taylor expansions.

**COMPARISONS AND DISCUSSION**

Equations 2.16, 2.20, 2.26 and 2.30 represent a step forward in the development of a computationally fast heterogeneous adsorption equation. However, what ultimately determines the utility of the equations and underlying assumptions is the quality of the fit produced when compared with an exact solution. Since a universal exact solution to Equation 1 has not been found, the "exact" solution presented for each case will be one found when assuming specific constraints on the system. What follows is a discussion of how the developed equations compare to the "exact" solutions calculated for each specific case.

**Equal Monolayer Coverages: \(Q_1 = Q_2\)**

\(\rho = 0 \text{ -- Random Site Matching.}\) Exact solutions were generated and compared to the results obtained from Equation 2.16 for a variety of conditions. Results are plotted in Figures 2.1 - 2.3. Results from the Langmuir isotherm (Equation 2.2 with \(\bar{\varepsilon}_1 = \varepsilon_1 \text{ and } \bar{\varepsilon}_2 = \varepsilon_2\)) were also compared with the exact solutions. The exact solutions were generated by substituting Equations 2.2 and 2.6 into Equation 2.1 and integrating the results numerically using the Euler method. It should be noted that the "exact" solution is constrained by the assumptions inherent in Equations 2.2 and 2.6 (i.e. local Langmuir isotherm and Gaussian shaped energy distribution curves). The Euler integration was centered on \(\bar{\varepsilon}\) for each component using 200 integration steps. The selected model isotherm parameters were based on typical values as
indicated by O'Brien and Myers (1984) and are listed in Table 2.1. A copy of the BASIC code written to perform the integration and compare the results to the calculated Equation 2.16 and Langmuir results can be found in Appendix A.

**TABLE 2.1: Variable values used in numerical calculations**

<table>
<thead>
<tr>
<th>Component 1</th>
<th>Component 2</th>
</tr>
</thead>
<tbody>
<tr>
<td>T</td>
<td>298 K</td>
</tr>
<tr>
<td>Q</td>
<td>1.0</td>
</tr>
<tr>
<td>$\bar{e}$</td>
<td>10*RT</td>
</tr>
<tr>
<td>$b$</td>
<td>1.0*10^-10</td>
</tr>
</tbody>
</table>

Figure 2.1 compares the percent error for Equation 2.16 and the Langmuir isotherm as a function of total system pressure and breadth of the heterogeneous energy distributions ($\sigma$/RT). In this plot, $\sigma$/RT and the partial pressure for each component are held equal ($Y_1 = 0.5$). As the plot shows, the deviations from the exact solution for Equation 2.16 are consistently smaller than those of the Langmuir equation. While the Langmuir errors get consistently worse with increasing pressure, the Equation 2.16 errors appear oscillatory. It cannot be determined from the plot, however, whether the Equation 2.16 series will ultimately converge at higher pressures.

Figures 2.2 and 2.3 show the dramatic difference between the Equation 2.16 solution and the Langmuir solution. Figure 2.2 gives the Equation 2.16 and Langmuir errors when the energy distributions for both components are held constant and equal ($\sigma$/RT = 0.75) while the relative vapor phase concentrations are varied. Figure 2.3 compares Equation 2.16 and Langmuir errors when $Y_1$ and $\sigma_1$/RT are held constant at 0.5 and 0.75 respectively, and $\sigma_2$/RT is varied. In both plots, the Langmuir errors increase consistently with increasing pressure while the Equation 2.16 errors oscillate around 0. As the plots show, for any given point, the errors obtained for the Equation 2.16 solution are approximately an order of magnitude better than those obtained from the Langmuir solution.

$\rho$=1 -- **Perfect Positive Correlation.** Exact solutions were generated and compared to the results obtained from Equation 2.20 for a variety of conditions and plotted in Figures 2.4 - 2.6. The exact solutions were generated by substituting Equations 2.2 and 2.9 into Equation 2.7 and integrating the results numerically using the Euler method. As with the $\rho$=0 case above, the "exact" solution is constrained by the assumptions inherent to Equations 2.2 and 2.9 (i.e. local Langmuir isotherm and Gaussian or non-skewed energy distribution curves). The Euler integration was centered on $\bar{e}$, using 200 integration steps. The selected model isotherm parameters are the same as those used in the $\rho$=0 calculations above, and can be found in Table 2.1. A copy of the BASIC code written to perform the integration and compare the results to the calculated Equation 2.20 and Langmuir results can be found in Appendix B.
FIGURE 2.1: Percent Error for the Langmuir Isotherm and Equation 2.16

\[ Y_1 = 0.5 \]
\[ \sigma_1/RT = \sigma_2/RT = \sigma/RT \]
FIGURE 2.2: Percent Error for the Langmuir Isotherm and Equation 2.16
FIGURE 2.3: Percent Error for the Langmuir Isotherm and Equation 2.16

\[ \frac{Y_1}{\sigma_1/RT} = 0.75 \]

\[ \frac{Y_1}{\sigma_1/RT} = 0.50 \]
\[ Y_1 = 0.5 \]
\[ \sigma_1/RT = \sigma_2/RT = \sigma/RT \]

**Equation 2.20**

---

**FIGURE 2.4**: Percent Error for the Langmuir Isotherm and Equation 2.20
\[
\sigma_1/RT = \sigma_2/RT = 0.75
\]

**FIGURE 2.5:** Percent Error for the Langmuir Isotherm and Equation 2.20
FIGURE 2.6: Percent Error for the Langmuir Isotherm and Equation 2.20
Figure 2.4 compares the percent error for Equation 2.20 and the Langmuir isotherm as a function of total system pressure and breadth of the heterogeneous energy distributions (σ/RT). In this plot, σ/RT and the partial pressure for each component are held equal (Y_i = 0.5). As the plot shows, the deviations from the exact solution for Equation 2.20 are consistently smaller than those of the Langmuir equation. As would be expected for an expansion solution, the deviations from the exact values are oscillatory with decreasing absolute magnitudes as the breadth of the distribution gets narrower. Although the error gets as high as 8% for the σ/RT = 1.25 solution, the results still converge with increasing pressure values.

Figure 2.5 shows how Equation 2.20 and the Langmuir isotherm perform when the energy distributions for both components are held constant and equal (σ/RT = 0.75) while the relative vapor phase concentrations are varied. Once again, Equation 2.20 gives consistently better results than the Langmuir isotherm.

The final plot in this group, Figure 2.6, provides the most insight into the quality of the Equation 2.20 results as compared to those of the Langmuir equation. In this plot, Y_1 and σ_1/RT are held constant at 0.5 and 0.75 respectively, while σ_2/RT is varied. As σ_2/RT drops from 0.75 to 0.25, the Equation 2.20 error shifts upward in the lower pressure region, but converges more rapidly as the pressure is increased. The Langmuir solution, however, is so poor that the σ_2/RT = 0.25 curve does not fit on the plot.

Unequal Monolayer Coverages: Q_1 ≠ Q_2

p=0 — Random Site Matching. Figures 2.7-2.12 are plots comparing Equation 2.26 results to those obtained from the homogeneous Langmuir solutions of LeVan and Vermeulen (Equation 2.22). The results from each of these equations are plotted as a percentage of error calculated from an "exact" solution. The exact solutions were generated by substituting Equation 2.6 and an ideal adsorbed solution theory, iterative mixture solution based on Equation 2.2, into Equation 2.1 and integrating the results numerically using the Euler method. As with the Q_1 = Q_2 case detailed previously, the "exact" solutions are subject to the constraints inherent in Equations 2.2 and 2.6 (i.e. local Langmuir isotherm and Gaussian shaped energy distribution curves). The Euler integration was centered on ε for each component using 200 integration steps. The selected model isotherm parameters are listed in Table 2.1. A copy of the BASIC code written to perform the integration, and compare the results to the calculated Equation 2.22 and Equation 2.26 results can be found in Appendix C.

The plots can be divided into two series: Figures 2.7 - 2.9 using Q_1 = 0.9 and Q_2 = 1.1, and Figures 2.10 - 2.12 using Q_1 = 0.8 and Q_2 = 1.2. The latter of the two cases should represent a fairly extreme spread in saturation values, with Q_2 being 50% larger than Q_1. As the plots indicate, as values for site energy distribution breadth, vapor phase mole fraction, and site energy distribution ratios are changed,
FIGURE 2.7: Percent Error Equations 2.22 and 2.26

$Y_1 = 0.5$
$\sigma_1/RT = \sigma_2/RT = \sigma/RT$
$Q_1 = 0.9 \quad Q_2 = 1.1$

$\sigma/RT$

- ****** Eq 2.26 0.25
- ***** Eq 2.22 0.25
- ++++ Eq 2.26 0.50
- +++++ Eq 2.22 0.50
- ****** Eq 2.26 0.75
- +++++ Eq 2.22 0.75
$Q_1 = 0.9 \quad Q_2 = 1.1$

$\sigma_1/RT = \sigma_2/RT = 0.25$

FIGURE 2.8: Percent Error Equations 2.22 and 2.26
FIGURE 2.9: Percent Error Equations 2.22 and 2.26

\[ Q_1 = 0.9 \quad Q_2 = 1.1 \]
\[ Y_1 = 0.5 \]

- - - - - Eq 2.26 0.75 0.25
- - - - - Eq 2.22 0.75 0.25
- - - - - Eq 2.26 0.75 0.50
- - - - - Eq 2.22 0.75 0.50
- - - - - Eq 2.26 0.75 0.75
- - - - - Eq 2.22 0.75 0.75
\[ Y_1 = 0.5 \]
\[ \sigma_1/RT = \sigma_2/RT = \sigma/RT \]
\[ Q_1 = 0.8 \quad Q_2 = 1.2 \]

FIGURE 2.10: Percent Error Equations 2.22 and 2.26
Error (Percent)

$Q_1 = 0.8 \quad Q_2 = 1.2$

$Y_1 = 0.5$

FIGURE 2.12: Percent Error Equations 2.22 and 2.26
definite error patterns develop which are consistent for both the Equation 2.22 and Equation 2.26 solutions. The major difference between the two solutions is that the Equation 2.26 solution consistently provides a corrective offset. This offset makes the Equation 2.26 solution significantly superior (as much as an order of magnitude) to the Equation 2.22 solution. In addition, as would be expected for a Taylor series expansion solution, as the spread in Q values widens, the solutions for both equations become less accurate and more divergent. It is expected that this trend would continue if mixtures demonstrating larger differences in Q values were investigated.

ρ=1 — Perfect Positive Correlation. Figures 2.13-2.18 are plots comparing Equation 2.30 results to those obtained from the homogeneous Langmuir solutions of LeVan and Vermeulen (Equation 2.22). The results from each of these equations are plotted as a percentage of error calculated from an "exact" solution. The exact solutions were generated by substituting Equations 2.9 and 2.22 into Equation 2.7 and integrating the results numerically using the Euler method. As detailed previously, the "exact" solutions are subject to the constraints inherent in Equations 2.7 and 2.22 (i.e. local Langmuir isotherm and Gaussian shaped energy distribution curves). The Euler integration was centered on ε for each component using 200 integration steps. The selected model isotherm parameters are listed in Table 2.1. A copy of the BASIC code written to perform the integration, and compare the results to the calculated Equation 2.22 and Equation 2.30 results can be found in Appendix D.

The plots can be divided into two series: Figures 2.13 - 2.15 using Q₁ = 0.9 and Q₂ = 1.1, and Figures 2.16 - 2.18 using Q₁ = 0.8 and Q₂ = 1.2. As with the ρ = 0 case outlined above, as values for site energy distribution breadth, vapor phase mole fraction, and site energy distribution ratios are changed, definite error patterns develop which are consistent for both the Equation 2.22 and Equation 2.30 solutions. However, unlike the ρ = 0 case, the corrective offset provided by the heterogeneity terms inherent in the Equation 2.30 solution does not always provide a superior solution. This offset consistently shifts the homogeneous (or Equation 2.22) solutions to the more positive errors. Since the error function was defined as: moles adsorbed actual - moles adsorbed theory, positive error readings indicate that the theory is under-predicting the amount adsorbed. As the plots show, Equation 2.30 gives superior results at the lower total pressure (P_TO) values. As P_TO increases, both solutions drift toward more positive error readings. Thus, as P_TO is increased past a threshold value, the Equation 2.30 solution becomes inferior to that of Equation 2.22.

CONCLUSIONS

As the various plots indicate, a non-iterative, heterogeneous solution has been developed that provides predictive results superior to those of analogous homogeneous solutions. Variations in the model's constituent parameters indicate that the results hold for a substantial range of conditions. The one (general) case where the new heterogeneous solution did not outperform the homogeneous solution is rendered insignificant when the following is
\[ \text{Error (Percent)} \]

\[ Y_1 = 0.5 \]
\[ \sigma_1/RT = \sigma_2/RT = \sigma/RT \]
\[ Q_1 = 0.9 \quad Q_2 = 1.1 \]

**FIGURE 2.13:** Percent Error Equations 2.22 and 2.30
FIGURE 2.15: Percent Error Equations 2.22 and 2.30

$Q_1 = 0.9 \quad Q_2 = 1.1$

$Y_1 = 0.5$

$Q_1 = 0.9 \quad Q_2 = 1.1$

$Y_1 = 0.5$

$σ_1/RT \quad σ_2/RT$

- ✔✔✔✔ Eq 2.30 0.75 0.25
- ✔✔✔✔ Eq 2.22 0.75 0.25
- ✔✔✔ Eq 2.30 0.75 0.50
- ✔✔✔ Eq 2.22 0.75 0.50
- ✔✔✔✔ Eq 2.30 0.75 0.75
- ✔✔✔ Eq 2.22 0.75 0.75
considered: 1) the error for the fast heterogeneous model, within the range of parameters investigated, was not appreciably higher than that obtained from the homogeneous equation; 2) the parametric conditions where this occurred were limited to higher total pressures and divergent Q values; and 3) the errors for these specific cases were very low relative to the homogeneous solution errors at lower total pressures. As a result, when all conditions are considered (i.e. the full range of pressure values), the new heterogeneous model provides the most accurate solution.

An important caveat to this analysis is that the "exact" solutions which are used as error standards have inherent assumptions that may have had impact on the results. In addition, the "exact" solution is a theoretical one that is based fundamentally on a heterogeneous adsorption system. Ultimately, the utility of a developed solution is dependent on its ability to fit and predict actual mixture data.

An indication of the importance of developing fast solutions can be found in the time it took to generate the "exact" solutions in the analysis. On a relatively fast PC (80486 running at 50 MHz), the compiled code to generate 100 data points for each plot required between 10 and 200 minutes to run. This computational burden can be compared to the immeasurably short time (milliseconds or less) it takes to execute the single calculation associated with the fast heterogeneous solution. Considering that, as stated above, any isotherm subroutine in a dynamic model will be called hundreds of thousands of times, it becomes apparent that a fast model of the type developed here is desirable.

REFERENCES


APPENDIX A

BASIC code for $Q_1 = Q_2$ and $\rho = 0$
10  DEFDBL A-H, I-Z
20  DEFINT I-K
30  PI = 3.1415926#: R = 8.3124400000000001D-03: ' KJ/MOLE K
40  RT = R * 298
50  DIM FX(502), GX(502)
70  INC = 200
80  GTOT1 = 0
90  INPUT "ENTER OUTPUT FILE NAME ---> ", FIELDS
95  OPEN FIELDS FOR OUTPUT AS #1
100  EB1 = 10 * RT: EB2 = 12 * RT
110  B1 = EXP(-10): B2 = 1.2 * EXP(-10)
115  INPUT "ENTER SIGMA1/RT --> ", SRT1
117  INPUT "ENTER SIGMA2/RT --> ", SRT2
118  S1 = SRT1 * RT: S2 = SRT2 * RT
120  M = (S2 / S1) ^ 2
130  INPUT "ENTER GAS PHASE MOLE FRACTION Y1 ---> ", Y1
200  FOR PTOT = 0 TO 1! STEP .01
220  Q = 1
230  P1 = Y1 * PTOT
240  P2 = PTOT - P1
260  FX(0) = 0: GX(0) = 0
290  GCN = 0
295  H1 = 2 * EB1 / INC: H2 = 2 * EB2 / INC
300  FOR I = 1 TO INC
310     E1 = H1 * I
320     XS = P1 * B1 * EXP(E1 / RT)
325     PS11 = (E1 - EB1) / S1
340     G1 = EXP(-(PS11 ^ 2) / 2) / (S1 * (2 * PI) ^ .5)
345     FCN = 0
350  FOR J = 1 TO INC
360     E2 = H2 * J
370     XS = P2 * B2 * EXP(E2 / RT)
390     PS12 = (E2 - EB2) / S2
500     N1 = Q * XS / (1 + XS + XS)
520     G2 = EXP(-(PS12 ^ 2) / 2) / (S2 * (2 * PI) ^ .5)
530  IF (PTOT=0) AND (I=1) THEN GTOT2 = GTOT2 + G2 * H2: GOTO 600
550  IF (J = INC) THEN FCN = FCN + H2 * FX(J) / 3: GOTO 600
580     FCN = FCN + H2 * ((2 + 2 * (J MOD 2)) * FX(J)) / 3
600  NEXT J
630  IF (PTOT = 0) THEN GTOT1 = GTOT1 + G1 * H1: GOTO 790
650  IF (I = INC) THEN GCN = GCN + H1 * GX(I) / 3: GOTO 790
680  GCN = GCN + H1 * ((2 + 2 * (I MOD 2)) * GX(I)) / 3
790  NEXT I
800  X = P1 * B1 * EXP(EB1 / RT)
810  Y = P2 * B2 * EXP(EB2 / RT)
820  XI = 1 + X + Y
850  HH = X * (1+Y) * (1-X-Y) * S1 ^ 2 + X * Y * (Y-X-1) * S2 ^ .2
880  HX1 = Q * (X / XI + HH / (2 * XI ^ 3 * RT ^ 2))
890  LANG = Q * X / XI
895  IF (FCN = 0) THEN GOTO 920
900  DEL = (GCN - HX1) * 100 / FCN
910  DEL2 = (GCN - LANG) * 100 / FCN
920  PRINT USING "##.##### ": PTOT; GCN; HX1; DEL; LANG; DEL2; GTOT1; GTOT2
930  PRINT #1, USING "##.######## ": PTOT; DEL; DEL2
940  NEXT PTOT
945  CLOSE #1
950  END

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APPENDIX B

BASIC code for $Q_1 = Q_2$ and $\rho = 1$
10 DEFDBL A-H, L-Z
20 DEFINT I-K
30 PI = 3.1415926#: R = 8.3124400000000001D-03: ' KJ/MOLE K
40 RT = R * 298
50 DIM FX(5002)
70 INC = 1000
80 GTOT = 0
90 INPUT "ENTER OUTPUT FILE NAME ---> ", FILS
95 OPEN FILS FOR OUTPUT AS #1
100 EB1 = 10 * RT: EB2 = 12 * RT
110 B1 = EXP(-10): B2 = 1.2 * EXP(-10)
115 INPUT "ENTER SIGMA1/RT ---> ", SRT1
117 INPUT "ENTER SIGMA2/RT ---> ", SRT2
118 S1 = SRT1 * RT: S2 = SRT2 * RT
120 M = (S2 / S1)^2
130 INPUT "ENTER GAS PHASE MOLE FRACTION Y1 ---> ", Y1
200 FOR PTOT = 0 TO 1! STEP .01
220 Q = 1
230 P1 = Y1 * PTOT
240 P2 = PTOT - P1
260 FX(0) = 0
290 FCN = 0
295 H = 2 * EB1 / INC
300 FOR I = 1 TO INC
320 E1 = 2 * EB1 * I / INC
340 E2 = M * (E1 - EB1) + EB2
345 IF (E2 < 0) THEN PRINT "E2 < 0 ERROR": END
350 XS = P1 * B1 * EXP(E1 / RT)
360 YS = P2 * B2 * EXP(E2 / RT)
380 PSI1 = (E1 - EB1) / S1
390 PSI2 = (E2 - EB2) / S2
500 N1 = Q * XS / (1 + XS + YS)
510 G1 = EXP(-(PSI1^2) / 2) / (S1 * (2 * PI)^.5)
520 G2 = EXP(-(PSI2^2) / 2) / (S2 * (2 * PI)^.5)
530 IF (PTOT = 0) THEN GTOT = GTOT + G1 * H: GOTO 700
550 FX(I) = N1 * G1 / GTOT
560 IF (I = INC) THEN FCN = FCN + H * FX(I) / 3: GOTO 700
580 FCN = FCN + H * ((2 + 2 * (I MOD 2)) * FX(I)) / 3
700 NEXT I
800 X = P1 * B1 * EXP(EB1 / RT)
810 Y = P2 * B2 * EXP(EB2 / RT)
820 XI = 1 + X + Y
850 HH = 1 - (3*X + M*Y*(M+2)) / XI + (2*(X+M*Y)^2) / XI^2
880 HX1 = Q * (X / XI + (X * HH * S1^2) / (2 * XI * RT^2))
890 LANG = Q * X / XI
895 IF (FCN = 0) THEN GOTO 920
900 DEL = (FCN - HX1) * 100 / FCN
910 DEL2 = (FCN - LANG) * 100 / FCN
920 PRINT USING "###.###": PTOT; FCN; HX1; DEL; LANG; DEL2; GTOT
930 PRINT #1, USING "###.###": PTOT; DEL; DEL2
940 NEXT PTOT
945 CLOSE #1
950 END
APPENDIX C

BASIC code for $Q_1 \ast Q_2$ and $\rho = 0$
10 DEFDBL A-H, K-Z
20 DEFINT I-J
25 DX = .0000000001#  
30 PI = 3.1415926#: R = 8.31244000000001D-03: ' KJ/MOLE K
40 RT = R * 298
50 DIM FX(202), GX(202)
70 INC = 200
80 GTOT1 = 0: GTOT2 = 0
90 INPUT "ENTER OUTPUT FILE NAME -----> ", FILS$
100 EB1 = 10 * RT: EB2 = 12 * RT
110 B1 = EXP(-10): B2 = 1.2 * EXP(-10)
112 INPUT "ENTER SIGMA1/RT ----> ", SRT1
113 INPUT "ENTER SIGMA2/RT ----> ", SRT2
115 INPUT "ENTER Q1 -----> ", Q1
117 INPUT "ENTER Q2 -----> ", Q2
118 S1 = SRT1 * RT: S2 = SRT2 * RT
120 M = (S2 / S1) ^ 2
130 INPUT "ENTER GAS PHASE MOLE FRACTION Y1 -----> ", Y1
200 FOR PTOT = 0 TO 1: STEP 0.01
210 OPEN FILS$ FOR APPEND AS #1
230 P1 = Y1 * PTOT
240 P2 = PTOT - P1
260 FX(0) = 0: GX(0) = 0
290 GCN = 0
295 H1 = 2 * EB1 / INC: H2 = 2 * EB2 / INC
300 FOR I = 1 TO INC
310 E1 = H1 * I
320 XS = P1 * B1 * EXP(E1 / RT)
325 PSI1 = (E1 - EB1) / S1
340 G1 = EXP(-(PSI1 ^ 2) / 2) / (S1 * (2 * PI) ^ .5)
345 FCN = 0
350 FOR J = 1 TO INC
360 E2 = H2 * J
370 YS = P2 * B2 * EXP(E2 / RT)
390 PSI2 = (E2 - EB2) / S2
500 GOSUB 2000
520 G2 = EXP(-(PSI2 ^ 2) / 2) / (S2 * (2 * PI) ^ .5)
530 IF (PTOT = 0) AND (I = 1) THEN GTOT2 = GTOT2 + G2 * H2: GOTO 600
550 FX(J) = N1 * G2 / GTOT2
560 IF (J = INC) THEN FCN = FCN + H2 * FX(J) / 3: GOTO 600
580 FCN = FCN + H2 * ((2 + 2 * (J MOD 2)) * FX(J)) / 3
600 NEXT J
630 IF (PTOT = 0) THEN GTOT1 = GTOT1 + G1 * H1: GOTO 790
650 GX(I) = FCN * G1 / GTOT1
660 IF (I = INC) THEN GCN = GCN + H1 * GX(I) / 3: GOTO 790
680 GCN = GCN + H1 * ((2 + 2 * (I MOD 2)) * GX(I)) / 3
750 NEXT I
775 IF (PTOT = 0) THEN HX1 = 0: LANG = 0: PRINT GTOT1, GTOT2: GOTO 940
800 X = P1 * B1 * EXP(EB1 / RT)
810 Y = P2 * B2 * EXP(EB2 / RT)
820 XI = 1 + X + Y
830 QB = (Q1 * X + Q2 * Y) / (X + Y)
835 DL2 = (Q1 - Q2) * X * Y * LOG(XI) / (X + Y) ^ 2
855 A = 3 * X ^ 2 * (Q1 - QB) / ((X + Y) * XI) + 2 * X ^ 3 * (QB - Q1)
856 + QB * X / XI - 3 * QB * X ^ 2 / XI ^ 2 + 2 *QB * X * XI ^ 2 / XI ^ 3
860 B = X * Y * (Q2 - QB) / ((X + Y) * XI) + 2 * X * Y ^ 2 * (QB - Q2)
861 / (X + Y) ^ 2 * XI + 2 * X * Y ^ 2 * (QB - Q2) / ((X + Y) * XI
865 G = (S1 ^ 2 + S2 ^ 2) * LOG(XI) + (X * S1 ^ 2 + Y * S2 ^ 2) * (3 /
XI = 6 * LOG(XI) / (X + Y) + (X ^ 2 * S1 ^ 2 + Y ^ 2 * S2 ^ 2) * 
   (6 * LOG(XI) / (X + Y) ^ 2 - 4 / ((X + Y) * XI) - 1 / XI ^ 2)
870
875 HH = (Q1 - Q2) * X * Y / (2 * RT ^ 2 * (X + Y) ^ 2)
880 HX1 = QB * X / XI + DL2 + S1 ^ 2 * (A + B) / (2 * RT ^ 2) + G * HH
890 LANGL = QB * X / XI + DL2
895 IF (GCN = 0) THEN GOTO 920
900 DEL = (GCN - HX1) * 100 / GCN
910 DEL2 = (GCN - LANGL) * 100 / GCN
915 DEL3 = (GCN - (Q1 + Q2) / 2) * X / XI) * 100 / GCN
920 PRINT USING "#####.##### "; PTOT; GCN; HX1; DEL; LANGL; DEL2; DEL3
930 PRINT #1, USING "#####.##### "; PTOT; DEL; DEL2; DEL3
940 CLOSE #1
945 NEXT PTOT
950 END
960 '
970 '
2000 ' LANGMUIR AST ALGORITHM
2100 '
2110 ' IF (PTOT = 0) THEN N1 = 0: RETURN
2120 K1 = B1 * EXP(E1 / RT)
2130 K2 = B2 * EXP(E2 / RT)
2200 IC = 0
2210 PSI = .0000001
2220 FOR IJ = 0 TO 1 STEP 1
2230 PSIG = PSI * (1 + DX * IJ)
2240 P10 = (EXP(PSIG / Q1) - 1) / K1
2250 P20 = (EXP(PSIG / Q2) - 1) / K2
2260 X1 = P1 / P10
2270 X2 = P2 / P20
2280 IF (IJ = 0) THEN ER1 = X1 + X2 - 1 ELSE DER1 = X1 + X2 - 1
2290 NEXT IJ
2310 IF (ABS(ER1) < DX * 1000) THEN GOTO 2500
2320 IC = IC + 1
2330 IF (IC > 50) THEN PRINT "NON CONVERGE": END
2340 DDER1 = (DER1 - ER1) / (PSI * DX)
2350 PSI = PSI - ER1 / DDER1
2400 GOTO 2220
2420 '
2450 '
2470 '
2500 N10 = Q1 * K1 * P10 / (1 + K1 * P10)
2510 N20 = Q2 * K2 * P20 / (1 + K2 * P20)
2520 NTOT = 1 / (X1 / N10 + X2 / N20)
2530 N1 = NTOT * X1
2550 '
2600 RETURN
APPENDIX D

BASIC code for $Q_1 \ast Q_2$ and $\rho = 1$
10 DEFDBL A-H, K-Z
20 DEFINT I-J
25 DX = .000000001#
30 PI = 3.1415926#: R = 8.312440000000001D-03: ' KJ/MOLE K
40 RT = R * 298
50 DIM FX(500)
70 INC = 200
80 GTOT = 0
90 INPUT "$ENTER OUTPUT FILE NAME --- > ", FILS$
95 OPEN FIL$ FOR OUTPUT AS #1
100 EB1 = 10 * RT: EB2 = 12 * RT
110 B1 = EXP(-10): B2 = 1.2 * EXP(-10)
112 INPUT "$ENTER SIGMA1/RT ---- > ", SRT1
113 INPUT "$ENTER SIGMA2/RT ---- > ", SRT2
115 INPUT "$ENTER Q1 ---- > ", Q1
117 INPUT "$ENTER Q2 ---- > ", Q2
118 S1 = SRT1 * RT: S2 = SRT2 * RT
120 M = (S2 / S1)^2
130 INPUT "$ENTER GAS PHASE MOLE FRACTION Y1 ---- > ", Y1
200 FOR PTOT = 0 TO 1: STEP .01
230 P1 = Y1 * PTOT
240 P2 = PTOT - P1
260 FX(0) = 0
290 FCN = 0
295 H = 2 * EB1 / INC
300 FOR I = 1 TO INC
320 E1 = H * I
340 E2 = M * (E1 - EB1) + EB2
345 IF (E2 < 0) THEN PRINT "E2 < 0 ERROR": END
350 XS = P1 * B1 * EXP(E1 / RT)
360 YS = P2 * B2 * EXP(E2 / RT)
380 PS11 = (E1 - EB1) / S1
390 PS12 = (E2 - EB2) / S2
500 GOSUB 2000
510 G1 = EXP(-(PS11 ^ 2) / 2) / (S1 * (2 * PI) ^ .5)
520 G2 = EXP(-(PS12 ^ 2) / 2) / (S2 * (2 * PI) ^ .5)
530 IF (PTOT = 0) THEN GTOT = GTOT + G1 * H: GOTO 700
550 FX(I) = N1 * G1 / GTOT
560 IF (I = INC) THEN FCN = FCN + H * FX(I) / 3: GOTO 700
580 FCN = FCN + H * ((2 + 2 * (I MOD 2)) * FX(I)) / 3
700 NEXT I
750 IF (PTOT = 0) THEN HX1 = 0: LANG = 0: GOTO 920
800 X = P1 * B1 * EXP(EB1 / RT)
810 Y = P2 * B2 * EXP(EB2 / RT)
820 XI = 1 + X + Y
830 QB = (Q1 * X + Q2 * Y) / (X + Y)
835 DL2 = (Q1 - Q2) * X * Y * LOG(XI) / (X + Y) ^ 2
840 QBP = -((Q1 * X + Q2 * M * Y) - QB * (X + M * Y)) / (X + Y)
850 QBP = ((Q1 * X + Q2 * Y * M) * 2) - QB * (X + M * Y) * 2 * (X + M * Y) * QBP) / (X + Y)
850 A = (QBP + QBP) * (1 - 2 * (X + M * Y)) / XI + ((2 + QBP * (X + M * Y) ^ 2) / (XI ^ 2)
860 B = M * ((M + 1) ^ 2) * LOG(XI) + ((X + Y * M ^ 2) + 2 * (X + M * Y) + 2 * (X + M * Y) * Y) * (M + 1) / XI
865 G = ((X + M * Y) ^ 2) / (XI ^ 2) + (4 * (X + M * Y) ^ 2) / ((X + Y) * XI) + 2 * M * (2 * X + 3 * M * Y - Y) * LOG(XI) / (X + Y)
870 ' HH = X * A / XI + (Q1 - Q2) * X * Y * (B - G) / ((X + Y) ^ 2)
880 HX1 = QB * X / XI + DL2 + S1 ^ 2 * HH / (2 * RT ^ 2)
890 LANGL = QB * X / XI + DL2
895 IF (FCN = 0) THEN GOTO 920
900  DEL = (FCN - HX1) * 100 / FCN
910  DEL2 = (FCN - LANGL) * 100 / FCN
915  DEL3 = (FCN - ((Q1 + Q2) / 2) * X / XI) * 100 / FCN
920  PRINT USING "####.####### "; PTOT; FCN; HX1; DEL; LANGL; DEL2; GTOT; DEL3
930  PRINT #1, USING "####.####### " ; PTOT; DEL; DEL2; DEL3
940  NEXT PTOT
945  CLOSE #1
950  END
960  
970  
2000  ' LANGMUIR AST ALGORITHM
2100  
2110  
2115  IF (PTOT = 0) THEN N1 = 0: RETURN
2120  K1 = B1 * EXP(E1 / RT)
2130  K2 = B2 * EXP(E2 / RT)
2200  IC = 0
2210  PSI = .0000001
2220  FOR IJ = 0 TO 1 STEP 1
2230   PSIG = PSI * (1 + DX * IJ)
2240   P10 = (EXP(PSIG / Q1) - 1) / K1
2250   P20 = (EXP(PSIG / Q2) - 1) / K2
2260   X1 = P1 / P10
2270   X2 = P2 / P20
2280   IF (IJ = 0) THEN ER1 = X1 + X2 - 1 ELSE DER1 = X1 + X2 - 1
2300  NEXT IJ
2310  IF (ABS(ER1) < DX * 1000) THEN GOTO 2500
2320  IC = IC + 1
2330  IF (IC > 50) THEN PRINT "NON CONVERGE"; END
2340  DDER1 = (DER1 - ER1) / (PSI * DX)
2350  PSI = PSI - ER1 / DDER1
2400  GOTO 2220
2420  
2450  
2470  
2500  N10 = Q1 * K1 * P10 / (1 + K1 * P10)
2510  N20 = Q2 * K2 * P20 / (1 + K2 * P20)
2520  NTOT = 1 / (X1 / N10 + X2 / N20)
2530  N1 = NTOT * X1
2550  
2600  RETURN