"Receptor-Carrier Based Models for Antigen-Antibody Interactions"
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Therefore the system needs to have many restrictions upon it to make it manageable. Most models assume the area for interaction is an infinite sheet (i.e. an endless two dimensional cell membrane). These models are limited to only discussing the average number and types of interactions. This model is receptor-carrier based which counts interaction on a cell/vesicle surface to allow questions about actions and conditions per cell to be addressed. The models which have been addressed are the "sandwich" model which takes a two-step interaction with a linking ligand and then the antibody, and the agglutination model which addresses the inter-cellular binding between two adjacent cells through a bridging ligand. Results show that the sandwich model and the agglutination model are less sensitive immunoassay techniques than the direct binding assay, although these models are very useful for describing more complex interactions.
"Receptor-Carrier Based Models
for Antigen-Antibody Interactions"

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ABSTRACT

A theoretical model was developed for antigen-antibody interaction in a receptor-carrier based system. Many models have been developed, both kinetic and equilibrium, for receptor-to-ligand interaction in the immune response. The immune response is extremely complex, with very large numbers of receptor sites on many distinct cell types in the system. Therefore the system needs to have many restrictions upon it to make it manageable. Most models assume the area for interaction is an infinite sheet (i.e. an endless two dimensional cell membrane). These models are limited to only discussing the average number and types of interactions. This model is receptor-carrier based which counts interactions on a cell/vesicle surface to allow questions about actions and conditions per cell to be addressed. The models which have been addressed are the "sandwich" model which takes a two-step interaction with a linking ligand and then the antibody, and the agglutination model which addresses the inter-cellular binding between two adjacent cells through a bridging ligand. Results show that the sandwich model and the agglutination model are less sensitive immunoassay techniques than the direct binding assay, although these models are very useful for describing more complex interactions.
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I. INTRODUCTION

The immune response in humans is a highly organized and very specific defense system [1]. The immune system defends the body against foreign molecules, cells, or viruses, which are called antigens. The human system not only recognizes non-human cells, but also those from other individuals, which explains the difficulty in transplanting organs from one person to another [2].

There are two main types of cells, lymphocytes, which are involved in the immune response: T-cells (which mature in the thymus) and B-cells (which mature in bone marrow). The humoral immune response is produced by the B-cells and is most effective against pathogenic (disease causing) cells, viruses and free toxins. The cell-mediated immune response uses the T-cells and is effective against larger cells, such as parasites, cancer, and infected cells.

The cell-mediated response (Figure 1) is more complex than the humoral immune response (Figure 2). This is the system which is attacked by the AIDS virus. In many responses the antigen is picked up by a macrophage and is presented to a "virgin" T-cell which initiates a response. The simplest T-cell response would then produce the cytotoxic T-cell which kills the antigen bearing cell and then looks for other antigens. The
Cell-Mediated Immune Response

Figure 1
Humoral Immune Response

B-lymphocyte plasma cell

Antibody secretion

Antibody binding to antigen

Phagocytosis by macrophages

Destruction by K lymphocytes

Lysis mediated by complement process

Figure 2
other response initiated by the T-cell involves helper T-cells and suppressor T-cells. The helper cells secrete lymphokines and interleukins which activate macrophages and mast cells in the area to ingest the antigen. The helper cells can also induce local plasma cells to produce antibodies. The helper cells also tell the antigen bound T-cells to reproduce. At the same time suppressor T-cells are doing the opposite of the helper cells in a dynamic balance to induce an appropriate level of response. An appropriate response is necessary since too little response will not rid the body of the antigen and over response is what causes side effects like sneezing, coughing and hives.

The B-cell response is more direct and simpler to understand. The B-cells move freely through the blood and lymph system of the body. When a B-cell encounters a specific antigen, it will begin to grow, and divide. This division produces two types of cells, the memory cells and the plasma cells. The memory cells allow a more rapid response to be initiated if the same antigen were to attack again. The plasma cells actually produce antibodies which are used to attack the antigen. Figure 3 shows the layout of a typical IgG antibody. There are more kinds of antibodies than just IgG, but IgG is the most common, with serum levels up to 12.0 mg/ml. The antibodies are made up of two light and two heavy protein
Antibody Structure

Figure 3
chains. Antibodies are very specific in their binding sites (Fab portion), and hence differ in the amino acid sequencing in that region, but all contain an Fc tail which is constant in amino acid sequencing from one antibody molecule to another. This Fc tail is the portion of the antibody which helps initiate the complement system. This system is shown in Figure 4. The chain of proteins forms a pathway to destroy the antigen, and acts as a catalyst to promote larger response further along in the process. The system has two methods of ridding the body of the antigen: macrophages in the area can be activated to engulf the antigen or to form a hole in the cell membrane of the antigen to lyse the cell. The humoral response can also produce K lymphocytes which mark the antigen and kill it through an unknown mechanism [2].

The purpose of this paper is to describe in detail the derivation of a theoretical model for cellular interaction. This interaction occurs in many places in the immune response; for example: antibody to antigen, macrophage to antigen, and helper T-cell to plasma cell.

Actual modeling of the B-cell response, considering the huge number of receptor sites on the B-cell [3], on an antigenic bacteria, or on a virus is unreasonable. The system needs to be easier to control and have fewer parameters. A more tractable system would be to model an
Complement System

CELL MEMBRANE

recognition → activation → attack

IgG IgG

completion of complement system causes cell lysis

incomplete protein chain activates local macrophage

complement proteins

"cascade" of complement proteins

Figure 4
artificial system. The purpose of this paper is to describe in detail the derivation of a theoretical model for interactions within the immune system, its usefulness, and the results of two separate applications of the theoretical model.

The system which is being modeled uses vesicles in immunolysis assays, shown in Figure 5 [4]. These assays use covalently bound antigens on the surface of phospholipid vesicles [5]. The vesicles can be made with specific types and numbers of receptor sites. The vesicles can encapsulate markers like spin-labelled compounds [6] or fluorescent compounds like carboxyfluorescein [7]. Carboxyfluorescein does not fluoresce in high concentrations like that which is encapsulated in the vesicles, but when the vesicle is lysed through the complement process or by another means, the material is diluted and fluoresces. This system can be modeled and a series of questions can be answered such as the number of antigenic sites required to activate a response, and a theoretical limit to detection of antigen or antibody with this system.

Another more modern and exciting application of such modeling is for the system recently developed involving immunoadhesins for AIDS therapy (Figure 6) [8]. It has been shown that the AIDS virus kills by depleting the body's T4 lymphocytes (a type of T-cell) and the victim
Immunolysis Assays

monovalent or bivalent ligands or receptors

Figure 5
AIDS Virus Immunoadhesins

virus tied up by CD4 (20-30 min)

coat protein (gp120)

AIDS virus

immunoadhesin

virus with immunoadhesins (? 21 days)

T-cell

CD4 receptor

Figure 6
becomes susceptible to infections and neoplasms (new growth, tumors). The T4 lymphocytes have numerous CD4 protein sites as receptors. The Human Immunodeficiency Virus (HIV) envelope glycoprotein (gp120) has been shown to be strongly attracted to the CD4 protein. Most therapies are ineffective against the AIDS virus because its variable receptors keep changing and modifying. The one common denominator is its attraction to the CD4 site on the T-cells. After the virus has attached itself to the T-cell membrane, it is internalized, uncoated, and alters the cellular DNA. It would seem that an increase in free CD4 protein in the blood would tie up the coat protein on the virus (gp120) and not allow it to attach to the T-cells. The problem with this approach is that CD4 only has an effective lifespan of twenty to thirty minutes in the blood. Researchers have found a way to develop a special antibody, called an immunoadhesin, which has CD4 receptor sites and an Fc tail. Experiments predict a lifespan approaching 21 days, the same as for a natural IgG. The goal of this program is to tie up the virus so it cannot attack the T4 cells and allow the normal immune response, involving the Fc portion of the immunoadhesin and the complement system, to kill the virus. This system can be represented by monovalent receptors (the gp120 coat protein sites) on the viral surface and bivalent ligands, the immunoadhesins. This
model can describe the specific actions on the viral surface.

Many questions still remain about the specific methods and limiting characteristics of the immune system. A theoretical model would be useful to answer some of these questions. Since the antibody to antigen binding constant is ambiguous [9] and a crucial factor in explaining cell to cell interactions, it is a central goal of any theoretical model to find a realistic value for this constant. Varying this parameter while holding the other conditions to those of an actual experiment should allow this constant to be approximated.

Attachment to the cell by first one protein followed by another is a possibility; modeling this two step process could show interesting effects. It is also believed that agglutination of the antigens into clumps helps the killer cells, macrophages, and the complement system to work more efficiently [1]. A model of these situations would help find the system’s limits, characteristic constants, ideal conditions and number of bound antibodies needed to produce cell lysis.

The models in this paper are receptor-carrier based [10]. This form of modeling is significant since it allows questions to be asked concerning the number of antibodies bound or cross-links per cell/vesicle. Other models have been proposed [11,12] which are kinetic or
equilibrium studies of monovalent or multivalent receptors or ligands. These models assume the surface of interaction to be infinite so only questions of types or average number of structures can be asked. The counting done by this model keeps track of structures on the vesicle surface. It is assumed that the initial antibody attachment to the vesicles is caused by a receptor-ligand interaction. This initial attachment rate is considered to be independent of number of sites occupied if there are sites remaining (steric considerations are not taken into account). The subsequent attachment of multivalent antibodies (in the agglutination model, the bivalent receptor model, and the bivalent ligand sandwich model) do take steric effects into account. Finally it is assumed that the receptor sites are able to migrate freely [13] to allow adequate mixing for the equilibrium equations to be relevant.
II. GENERIC MODEL DERIVATION

The models are defined by a system of conservation equations. The derivations are not intuitive, so an example will be provided. The simplest case is one of monovalent receptors on the cell and monovalent ligands [14]. This system is represented in Figure 7.

This system has a vesicle, or cell, covered by \( n \) monovalent antigenic receptors. The binding ligands are monovalent, like Fab fragments. The system has a degeneracy, (i.e., number of arrangements) which is defined as \( d_i \). Degeneracy is the number of equivalent arrangements of bound ligands.

\[
d_i = \frac{n!}{i!(n-i)!} = \binom{n}{i} \tag{1}
\]

This is derived by taking the number of possible arrangements of receptors (\( n! \)) and dividing by the number of arrangements of the different sites which are on the cell, including \( i! \) the number of identical sites which are bound and \( (n-i)! \) unbound sites. The ligands need to be specific for the receptor sites and are governed by the binding and dissociative rate constants, \( k_1 \) and \( k_{-1} \). These constants are related to \( K \), the equilibrium binding constant, by

\[
K = \frac{k_1}{k_{-1}} \tag{2}
\]
Monovalent Receptor-Monovalent Ligand Model

Figure 7
The Law of Mass Action representing the $i$th antibody binding is represented by

$$V_{i-1} + a = V_i,$$  \hspace{2cm} (3)

which can be represented by the following arrangement where the ' represents one of the specific arrangements among the total $d_i$, possible arrangements of $V_i$, and "a" represents the free antibody concentration:

$$V_i' = (aK)V_{i-1}'$$  \hspace{2cm} (4)

The total concentration of $V_i$ can be defined as

$$V_i' = V_i/d_i$$  \hspace{2cm} (5)

Repeating equation (4) $i$ times gives the general equation

$$V_i = d_i(aK)^iV_0$$  \hspace{2cm} (6)

where $V_0$ is the concentration of cells without any antibodies bound.

The total number of vesicles and antibodies in the system cannot change. This allows conservation equations to be found. The conservation equations for the
system can be derived as

\[ V_0 = \sum_{i=0}^{n} d_i (aK)^i V_0 \]  
(7)

\[ Ab^o = \sum_{i=0}^{n} id_i (aK)^i V_0 \]  
(8)

where \( V_0 \) is the initial concentration of vesicles and \( Ab^0 \) is the initial concentration of antibodies.

The mathematical identity of

\[ \sum_{i=0}^{n} \binom{n}{i} x^i = (1+x)^n \]  
(9)

allows equation 7 to be reduced to

\[ V_0 = V_0 (1+aK)^n \]  
(10)

and the derivative of equation 9 allows equation 8 to be reduced to

\[ Ab^0 = a + nV_0 aK/(1+aK) \]  
(11)
A. General Derivation Summary

This is the format which will be used to derive all future models:

1. Identify possible structures and parameters governing receptor to ligand binding.
2. Calculate degeneracy through possible arrangements, including unbound sites.
3. Strip (stepwise) possible arrangements to $V_0$ (entirely empty) arrangement.
4. Derive conservation equations for vesicles and antibodies.
5. Address specific questions.

III. SANDWICH MODEL

The first model which will be discussed in depth is the "sandwich" model [4]. This model is shown in Figure 8. This model is different from the monovalent receptor/monovalent ligand model by the ligand being bivalent, with one end specific to the receptors on the vesicle, and the other specific to the monovalent antibody. Now the vesicle receptors can have three possible arrangements: empty, with the bivalent "link" attached ($f$), and having the link attached to an antibody and the receptor site ($s$).
Sandwich Model
Monovalent receptors - specific links
Monovalent ligands
The degeneracy for this model is derived as in the monovalent model and it results in

$$d_{f,s} = \frac{n!}{[f!s!(n-f-s)!]}$$  \hspace{1cm} (12)

The Law of Mass Action for the model is represented by

$$K \quad \text{Equation} \quad V_{f,s} = V_{f+1,s-1}$$  \hspace{1cm} (13)

where $K$ is represented by Equation 2 for the $s$(second) step. By repeating this equation $s$ times the general equation for the concentration with $f$ links and 0 sandwiches can be shown to be

$$V_{f,s}/d_{f,s} = (Ksa)^{s}V_{f+s,0}/d_{f+s,0}$$  \hspace{1cm} (14)

where $a$ is the free antibody concentration and $K_s$ is the equilibrium constant for that reaction. The Law of Mass Action equation

$$V_{f+s,0}/d_{f+s,0} = (Kfl)(Ksa)^{s}V_{f+s-1,0}/d_{f+s-1,0}$$  \hspace{1cm} (15)

can be repeated $f+s$ times to yield the general equation

$$V_{f,s} = d_{f,s}(Kfl)^{f+s}(Ksa)^{s}V_{0,0}$$  \hspace{1cm} (16)
where \( l \) is the free link concentration, and \( V_{o,0} \) is the concentration of vesicles without any links or sandwiches attached.

The conservation equations can be derived to yield

\[
V^0 = \sum_{f=0}^{n} \sum_{s=0}^{n-f} d_{f,s}(K_{f}^{l})^{f+s}(K_{sa})^{s}V_{o,0} \quad (17)
\]

\[
A^0 = a + K_{sa}l + \sum_{f=0}^{n} \sum_{s=0}^{n-f} (s)d_{f,s}(K_{f}^{l})^{f+s}(K_{sa})^{s}V_{o,0} \quad (18)
\]

\[
L^0 = l + K_{sa}l + \sum_{f=0}^{n} \sum_{s=0}^{n-f} (f+s)d_{f,s}(K_{f}^{l})^{f+s}(K_{sa})^{s}V_{o,0} \quad (19)
\]

which can be simplified with Equation 9 and its derivative to the formulas below

\[
V^0 = (1+xy+y)^{n}V_{o,0} \quad (20)
\]

\[
A^0 = a + K_{sa}l + nxy(1+xy)^{n-1}(1+y+xy)^{n-1} \quad (1+y+xy-y/(1+xy))V_{o,0} \quad (21)
\]

\[
L^0 = l + K_{sa}l + ny(1+xy)^{n-1}(1+y+xy)^{n-1} [1+x(1+2y+xy-y/(1+xy))]V_{o,0} \quad (22)
\]

where \( x = K_{sa} \)

\( y = K_{f}^{l} \)
A. Sandwich Model Results

The sandwich model is best used to represent possible applications in immunolysis assays [4]. A vesicle can be made with antibody receptors, antigenic links, and antibody ligands. This would provide more flexibility in the types of systems to be modelled. In Figures 9 and 10 the fraction of vesicles with at least one receptor/link/ligand sandwich is shown with typical values for the binding constants for both binding steps, $K_f$ and $K_s$. The value of $8.3 \times 10^{-9}$ M for the constants $L^0$ and $Ab^0$ was chosen to allow for comparison with previous models. In Waite and Chang's paper on the monovalent receptor and monovalent ligand [14], results were shown for varying values of $n$ (number of receptor sites) versus fraction of vesicles with one or more antibodies bound to the antigenic receptors (Figure 11). At values of $n = 100$ and $Ab^0 = 8.3 \times 10^{-9}$ M, approximately seven percent of the vesicles had one or more ligands bound. In the sandwich model, with the link concentration held at the value of $8.3 \times 10^{-9}$ M it takes relatively large concentrations of $Ab^0$, about $8.8 \times 10^{-6}$ M, to reach the same seven percent. The sandwich model also allows the antibody concentration to be held constant and the link concentration to be varied. In this context the link concentration would need to be about $7.9 \times 10^{-5}$ M, with the $[Ab^0] = 8.3 \times 10^{-9}$ M, to
Sandwich model

\begin{align*}
L^0 &= 8.3 \times 10^{-9} \\
V^0 &= 4.2 \times 10^{-13} \\
K_f &= 10^5 \\
K_s &= 10^5
\end{align*}

Figure 9
Sandwich model

$V_0 = 4.2 \times 10^{-13}$
$A_0 = 8.3 \times 10^{-9}$
$K_f = 10.5$
$K_s = 10.5$

Figure 10

Fraction with 1 or more sandwiches

Log $L$
Figure 11 - Monovalent receptor-Monovalent ligand model. $\text{Ab}_0=8.3\times10^{-9}$ M. Number of receptors ($n$) vs. fraction with antibody ligands attached.

Plot from Waite and Chang's mono-mono paper (used by permission)
reach seven percent. This shows that the sandwich model is less sensitive than the direct binding model. This result is to be expected since this model requires a two step process to be completed versus a one step process. The plots also show a smooth increase in the number of sites "sandwiched" up to a maximum. On the plot which varies $Ab^0$, the maximum is one. This plot will not drop back off, as does the monovalent receptor, bivalent ligand model, shown in Figure 15 [15], since there isn’t any competition between the links and the ligands to get to the receptor cites. There is no competition since the links are specific to the receptor on one end and the ligand on the other. The plot of varying $L^0$ concentrations reaches a maximum at approximately eight percent. This is logical since it was shown that the sandwiches form at higher concentrations of $Ab^0$ on the first plot, then the low concentration of the $Ab^0$ in this plot lowers the possible number of sandwiches. The number of links attached does not affect the number of sandwiches after a point if there are not enough ligands present to form the second step of the process.

The values of $10^5$ for the binding constants are typical [16] but can vary from system to system. This system is a simple two step process. Changing one binding constant by some orders of magnitude, while holding the second constant, then changing the second constant
Fractions sandwiched $K_f = 10^{10}$, $K_S = 10^5$ (square)
$K_f = 10^5$, $K_S = 10^{10}$ (triangle)

Figure 12
while holding the first to its original value keeps the results constant, as is shown in Figure 12. It is not pictured but is logical that an increase in a single binding constant or increasing both would increase the relative response.

IV. AGGLUTINATION MODEL

Another model which can be used to represent the B-cell response is the monovalent receptor, multivalent antibody, with possibility of inter-linking between vesicles, called agglutination [4]. This is diagrammed in Figure 13. The antibody can be either IgG or IgM. The IgG antibody has two binding sites, while the IgM has ten binding sites, \((\lambda = 2 \text{ for IgG, } \lambda = 10 \text{ for IgM, where } \lambda \text{ is the number of binding sites on the antibody})\) but for simplicity in this model it will be assumed that only two sites will be available for binding at one time. Due to steric considerations of the planar shape of the IgM, this approximation is not too severe.

From Waite and Chang's paper on multivalency effects in the direct binding model the final equilibrium equation for a single vesicle with monovalent receptors and multivalent ligands is
Agglutination Model
Monovalent receptor–bivalent ligand

Figure 13
\[ V_{i,j} = d_{i,j}(K_1/\alpha)^j(\lambda K_1 a)^i j V_0,0 \]  \hspace{1cm} (23)

where
\[ d_{i,j} = n!/[i!j!2^j(n-i-2j)!] \hspace{1.5cm} (24) \]

and \( V_0,0 \) is the concentration of vesicles without any antibodies bound, either as danglers (i or i') or two-footers (j or j'), \( K_1 \) is the intrinsic association constant, \( K_2 \) is the intra-cross-linking constant, and \( \alpha \) is the geometric factor which takes into account the local concentration enhancement of forming a two-footer from a dangler.

Now the concentration of the agglutinated vesicles can be derived from this information, with some modifications. For mathematical simplicity, the number of vesicles which can form inter-cellular links will be limited to two cells. Therefore the degeneracy of the agglutinated vesicles can be shown to be the degeneracy of a single vesicle squared with the number of bridges (k) taken into account.

\[ d_{\text{agglut}} = n!n!/[i!j!2^j(n-i-2j)!k!i'!j'!2^j'(n-i'-2j')!] \hspace{1.5cm} (25) \]

\( K_1 \) is the binding constant for the initial bridge, \( K_2 \) is the binding constant for the bridges following the initial bridge, and \( \beta \) is the geometric factor which takes the steric effects and local concentration effects.
for bridging into account. Using the Law of Mass Action as has been shown before, the equilibrium equation for the concentration of two agglutinated vesicles \((C_2)\) can be given by

\[
C_2 = d_{\text{agglut}}(K_2/\alpha)^j(\lambda K_1 a)_{i+j} V_{o, o}(K_2/\alpha)^{j'}(\lambda K_1 a)_{i'+j'} V_{o, o} K_i (K_2/\beta)^{k-1}
\]  

(26)

which can be simplified to

\[
C_2 = d_{\text{agglut}}(K_2/\alpha)^{j+j'}(\lambda K_1 a)_{i+i'+j+j'} V_{o, o} K_i (K_2/\beta)^{k-1}
\]  

(27)

Equations 23 and 27 can be combined to give the conservation equation, which has been abbreviated.

\[
V^o = \sum \sum (n) V_{i,j} + 2 \sum \sum \sum \sum \sum \sum \sum C_2
\]  

(28)

Combining similar terms and substituting allow the simplified equations to be found.

\[
V^o = G(n) V_{o, o} + H(n) V_{o, o}^2
\]  

(29)

where

\[
G(n) = \sum_{i=0}^{n/2^4} S(i)(x^i + x^{n-i})/(1+\delta n/2, i)
\]  

(30)
The \( n/2^i \) means the term will be truncated, and the Kronecker delta function \( \delta_{i,j} = 1 \) for \( i = j \) and \( \delta_{i,j} = 0 \) for \( i \neq j \).

\[
S(i) = \sum_{j=0}^{i} \frac{n!y!}{[(i-j)!j!2^j(n-i-j)!]} \quad (31)
\]

and

\[
H(n) = \sum_{k=1}^{n} \frac{(n!n!/k!(n-k)!(n-k)!)x_kz^{k-1}G(n-k)}{2G(n-k)} \quad (32)
\]

\[
x = aK_1
\]

\[
y = K_2/\alpha
\]

\[
z = K_2/\beta
\]

The antibody conservation equation can be described in similar terms

\[
Ab^0 = a + J(n)V_{0,o} + W(n)V_{2,o,o} \quad (33)
\]

\( J(n) \) and \( W(n) \) are defined as

\[
J(n) = \sum_{i=0}^{n/2^i} (ix^i + (n-i)x^{n-i})S(i)/(1 + \delta_{n/2,i}) \quad (34)
\]

\[
W(n) = \sum_{k=1}^{n} \frac{(n!n!/k!(n-k)!(n-k)!x_kz^{k-1}G(n-k)}{2G(n-k)}[kG^2(n-k) + 2G(n-k)J(n-k)] \quad (35)
\]

and \( x, y, \) and \( z \) are defined as before.
AGGLUTINATION MODEL RESULTS

The agglutination model is useful for modeling many systems. It can be used to model the immunolysis assays with monovalent (Fab fragment) receptors and IgG antibody or bivalent antigen ligands or bridging between the AIDS viruses in immunoadhesin experiments [8]. Figure 14 shows the fraction agglutinated of the system with realistic binding constants for the formation of two-footers (j) and danglers (i) while the binding constant for the formation of the initial bridge is varied. This is done since no realistic value for this binding constant is known. The values of \( K_i \) on this plot vary from values of \( 10^{-11} \) to \( 10^3 \). A desired value would be to find the antibody concentration where the agglutination would be maximized. It is not intuitive that the maximum be consistent for all values of \( K_i \), but as the plot shows, no matter what the probability of forming the first bridge is, the maximum occurs at the same concentration, about \( 5.0 \times 10^{-6} \) M.

From Waite and Chang's paper on multivalency effects [15] on the direct binding model it was shown that the maximum value for two-footers is about \( 5.0 \times 10^{-6} \) M (shown in Figure 15). This maximum is the same for fraction agglutinated, which implies that the two are related. In low antibody concentration very few danglers are formed since the local concentration of receptors is so high that
Figure 19 - Plot from Waite and Chang's Multivalency effects paper, used by permission.
Plot of antibody concentration vs. fraction with antibody, two-footed attached.
most move directly to two-footers. A bridge cannot be formed if both sites on the antibody are occupied, therefor it requires danglers to be present. After the maximum concentration of two footers is reached, the relative number of two-footers per vesicle drop due to competition with danglers for receptor sites. These danglers form into bridges. The fraction agglutinated drops off when the antibody concentration gets too high. This is caused by all of the vesicle’s receptors being tied up, until finally there are no open receptors on either vesicle, and therefore no bridging.

Figure 16 shows the effects of number of sites per vesicle (n) upon the fraction agglutinated. This curve is relevant since the computer time used increases rapidly with increasing n and realistic numbers of receptor sites would take too long to calculate, if possible. If the effects were to form a smooth curve the effects could be extrapolated to a realistic number of receptors per vesicle.

This plot uses an antibody concentration of $10^{-12}$ M, to set a low value to try and approximate the threshold detection value for different methods. The light scattering techniques in Cohen and Benedek’s papers [17,18] rely on agglutination for detection. Their papers describe laser light and quasi-elastic light scattering from the early stages of agglutination for low level
Agglutination Model

Figure 16

\[ Ab^0 = 10^{-12} \]
detection of ligands. The plot could also be related to the immunolysis experiment which relies on the self-quenching fluorescent compound carboxyfluorescein, and the lowest level of light detectable by the light meter to show a response. This diagram used an arbitrary fraction of agglutination of $2 \times 10^{-9}$, which relates to $8.4 \times 10^{-22}$ M of the vesicles to be lysed, to estimate the detection threshold. At this lower limit of agglutination detection, the vesicles would need at least 675 receptor sites to detect an antibody concentration of $10^{-12}$ M.

V. NEW DERIVATIONS

Two more models have been derived, but have not been programmed. These models are extensions of the two which have been presented above.

A. Antibody Multivalency Effects on the Sandwich Model

The sandwich model can be made more realistic by taking into account antibody multivalency. As before, the IgG will be represented ideally and the IgM will be limited to an effective binding number of two to limit the mathematical complexity.
This model again uses the vesicle with n receptor sites which can be involved first with the linking ligand, then the antibody. The degeneracy of this system is

\[ d_{f,s,t} = \frac{n!}{[f!s!t!2^t(n-f-s-2t)!]} \]  

(36)

Figure 17 shows this model where f receptor sites have a link attached, s with an antibody bound to a link singularly, in a similar manner as a dangler, and t with links bound by the same antibody.

The Law of Mass Action equations necessary to derive the equilibrium equations are

\[ K_t \quad t \text{ times} \]

\[ V'_f,s,t = V'_f,s+1,t-1 \quad \Rightarrow \quad V_{f,s+t,o} \]  

(37)

\[ K_s \quad s+t \]

\[ V'_f,s+t,o = V'_f+1,s-1,o + a \quad \Rightarrow \quad V_{f+s+t,o,o+a} \quad \text{times} \]  

(38)

\[ K_f \quad f+s+t \]

\[ V'_f+s+t,o,o = V'_f+s+t-1,o,o + l \quad \Rightarrow \quad V_{o,o,o+l} \quad \text{times} \]  

(39)
Sandwich Model
Monovalent Receptor-
Bivalent Ligand

Figure 17
These equations taken to the limits shown above yield the equilibrium equation

\[ V_{f,s,t} = d_{f,s,t}(K_t/\alpha)^{t}(K_s/\alpha)^{s+t}(K_f/\alpha)^{f+s}V_{0,0,0} \]  
(40)

\( \lambda \) is the number of receptor sites on each antibody and \( \alpha \) is the geometric factor which takes local concentration effects into account.

Defining \( x = K_f/\lambda \), \( y = K_s/\alpha \), and \( z = K_t/\alpha \) the following conservation equations can be derived.

\[ V^0 = \sum_{t=0}^{n-f-s/2} \sum_{s=0}^{n-f} \sum_{f=0}^{n} d_{f,s,t}x^{f+s}y^{s+t}z^{t}V_{0,0,0} \]  
(41)

\[ \text{Ab}^0 = \sum_{t=0}^{n-f-s/2} \sum_{s=0}^{n-f} \sum_{f=0}^{n} d_{f,s,t}(s+t)x^{f+s}y^{s+t}z^{t}V_{0,0,0} \]  
(42)

\[ L^0 = \sum_{t=0}^{n-f-s/2} \sum_{s=0}^{n-f} \sum_{f=0}^{n} d_{f,s,t}(f+s+2t)x^{f+s}y^{s+t}z^{t}V_{0,0,0} \]  
(43)

Using Equation 9 and its derivative the conservation equations can be simplified into
\[ V^o = (1+x)^n(1+xy)^{n-f}H(t) \]  
(44)

\[ Ab^o = a \cdot 2Ksa \ell + K^2sa^2\ell + \]
\[ (n-f)xy(l+x)_n(1+xy)^{n-f-1}H(t)V_{o,o,o} \]
\[ +(1+x)^n(1+xy)^{n-f}tH(t)V_{o,o,o} \]  
(45)

\[ Lo = \ell \cdot 2Ksa \ell + K^2sa^2\ell + nx(1+x)^{n-1}(1+x)^{n-f}H(t)V_{o,o,o} \]
\[ + (1+x)^n(n-f)xy(1+xy)^{n-f-1}H(t)V_{o,o,o} \]
\[ + (1+x)^n(1+xy)^{n-f}2tH(t)V_{o,o,o} \]  
(46)

where

\[ H(t) = \sum_{t=0}^{n-f-s/2} \frac{(n-f-s)!/[t!2^t(n-f-s-2t)!]}{(xyz)^t} \]  
(47)

The results from this model would be expected to be related to the monovalent ligand sandwich model much in the same manner as Waite and Chang's papers on monovalent direct binding and multivalency effects are related [14,15]. It would be expected that these effects include geometric factors for local concentration of receptors after one end of the antibody is attached. It is also logical that the number of sandwiches with both sites on
the ligand being attached having a peak when compared to initial antibody concentration. This would be caused by the competition of the "two-footed sandwiches" with the danglers for receptor sites with links attached.

B. Bivalent Receptor/Bivalent Ligand Model

The bivalent receptor, bivalent ligand model is the most complex model presented. The vesicle surface is covered with n surface immunoglobulins (sIg) which each have two receptor cites. This system is shown in Figure 9.

Theoretically a B-cell has antibody chains from surface Ig to sIg which are up to twenty receptors long. These chains can form loops upon themselves, long continuous chains, or large clusters. The number of possible arrangements is mathematically intractable. To make this model's derivation somewhat straightforward this model limits the chaining effects to two sIg. The possible arrangements of two receptor cites is also shown on Figure 9.

The degeneracy of this system need to take this $2^\text{n}$ factor into account, but this falls out since all possible combinations of antibiotic attachment and empties also have this $2^\text{k}$ factor.
Bivalent Receptor–Bivalent Ligand Model

Possible Arrangements

Figure 18
\[ d_{a-g} = n!/[a!b!c!d!e!f!g!(n-a-b-c-2d-2e-2f-2g)!] \]

(48)

The Law of Mass action equations are shown diagrammatically on Figures 19a and 19b.

The breakdown of each structure is not always a singular path, but this system falls under the same rules as a state function, thus any one pathway may be selected for each combination. The one pathway chosen must be consistent for all the Mass Action equations. The equilibrium equation for any specific combination of antibodies is given by

\[
V_{a-g} = d_{a-g}(K_{a}ab)^{a+b+c+d+2e+2f+2g}(K_{b}ab)^{b}(K_{c}c)^{c}(K_{d}d)^{d}

(K_{e}e+f+g(K_{f}f)(K_{g}ab)^{g}

(49)

This equation uses ab to represent free antibody concentration to eliminate confusion between it and receptor-ligand combination a. The binding constants \( K_b \) through \( K_g \) all have geometric local concentration effects tied into their values.
Bi-Bi Model  Law of Mass Action "Equations"

\[ a \rightarrow Y \]
\[ b \rightarrow Y + Y \rightarrow Y + Y \]
\[ c \rightarrow Y \rightarrow Y \]
\[ d \rightarrow Y + Y \rightarrow Y + Y \]
\[ e \rightarrow Y + Y \rightarrow Y + Y \rightarrow Y + Y \]

Figure 19a
Figure 19b
The conservation equation for the vesicles is

\[ v^0 = \sum_{g=0}^{\xi} \sum_{f=0}^{\phi} \sum_{e=0}^{\epsilon} \sum_{d=0}^{\omega} \sum_{c=0}^{\omega} \sum_{b=0}^{\omega} \sum_{a=0}^{\omega} d_{a-g}(K_{a}^{ab})^a(K_{b}^{ab^2})^b \]

\[ (K_{c}^{ab})^c(K_{d}^{ab})^d(K_{e}^{ab^2})^e(K_{f}^{ab^2})^f(K_{g}^{ab})^g \]

\[ \xi = n-a-b \]
\[ \omega = n-a-b-c/2 \]
\[ \epsilon = n-a-b-c-2d/2 \]
\[ \phi = n-a-b-c-2d-2e/2 \]
\[ \xi = n-a-b-c-2d-2e-2f/2 \]

(50)

Models have been derived with similar arrangements [21] of bivalent ligands and bivalent receptors, but again, this model is receptor-carrier based [10]. This will allow questions about linked receptors per vesicle to be answered.
VI. CONCLUSION

The purpose of this paper was to describe in detail the derivation of a theoretical model for antigen-antibody interaction and how that interaction relates to the immune response, immunoassay methods of detection, and the use of immunoadhesins for AIDS therapy. These models were receptor carrier based which allows the questions of number of interactions per vesicle or cell to be addressed. The results may be used to discover parameters for these systems. With help from experimental immunologists to relate real known binding constants and concentrations, boundary limits and interactions on the cell/vesicle surface may be described. These models can be used as building blocks for describing more complex systems.
REFERENCES


7 J.N. Weinstein, E. Ralston, L.D. Leserman, R.D.


19  J.F. LePage, private communication.


APPENDIX 1

Computer programming of these equations does pose some interesting problems. The precision for the VAX is normally $e^{-88}$ and double precision is $e^{-300}$. One of the problems occurs when adding a series of numbers of greatly different orders of magnitude. Even when working in natural logarithms, the numbers must be changed back to their non logarithmic form to add them. This problem can be avoided by using a series of steps to make the numbers more manageable.

Desired quantity $\ln (x_1 + x_2 + x_3)$

(a) Find: $\ln x_1, \ln x_2, \ln x_3$
(b) Find largest quantity from (a)
(c) Subtract (b) from all values in (a)
(d) Take the exponential of values in (c)
(e) Sum numbers in d (magnitudes close)
(f) Take ln (e)
(g) Add ln (f) + ln (b)
(h) Equals ln $(x_1 + x_2 + x_3)$

i.e. Desired $10^4 + 10^{-1}$ ($\log_{10}$ will be used)

(a) $\log 10^4 = 4, \log 10^{-1} = -1$
(b) largest = 4
(c) 0, -5
(d) $10^0 = 1, 10^{-5} = .00001$
(e) 1.00001

(f) \log 1.00001 = 4.3 \times 10^{-6}

(g) 4 + 4.3 \times 10^{-6} = 4.0000043

(h) 10^{(g)} = 10000.1 \text{ as expected}

The other complication arises when the factorial for large numbers is needed. This arises in all the degeneracy calculations since \( n! \) is a very large number. This can be approximated using Stirling's approximation which says:

\[
\ln(n!) = (n+1-1/2)\ln(n+1) - (n+1) + 1/2 \ln(2\pi) + 1/12(n+1) - 1/360(n+1)^3 + 1/1260(n+1)^5 \ldots
\]

This approximation gives very exact answers without overloading the computer. Most numbers would overflow normal precision since 100! = 9.3 \times 10^{157}, and many would overflow double precision, therefore Stirling's approximation is used whenever \( n \) is greater than seven.
Appendix 2

This computer program is used to develop the plots and results for the sandwich model. The loops fill a two-dimensional array for all $V_f, s$ combinations. This array is forced into a one-dimensional array where the summation of the fraction can be found. The data is then passed to data files to be interpreted.
PROGRAM: SAND.FOR

To run type:
fortran/g_floating sand.for
link sand
run sand

implicit real*8 (a-h,o-z)
dimension Vfs(200,200)
dimension temp(40000)
dimension temp2(40000)

akf=1.d5
aks=1.d5
ictr=1
nsand=1

sum=0

Vo=dlog(4.2d-13)
do 5 n=100,101,10
do 10 ih=900,901,2
a=8.3*10.**(-ih/100.)
do 15 im=900,300,-50
al=10.**(-im/100.)

x = aks*a
y = akf*al
w = aks*a*al

Voo = Vo-n*dlog(1+x*y+y)
do 20 if = 0,n
do 25 is = 0,n-if

Dfs=fact(n)-fact(if)-fact(is)-fact(n-is-if)
Vfs(if+1,is+1)=Dfs+(if+is)*dlog(y)+is*dlog(x)+Voo

25 continue
20 continue
nn=n-1

yy=1+x*y

aLo=Voo+dlog(n*y)+nn*dlog(yy)
alLo=Voo+dlog(n*x*y)+nn*dlog(y+yy)
a2Lo=dexp(alLo)+al+w+dexp(alLo)
goto 11
Abo=dlog(n*x*y)+nn*dlog(y+yy)+dlog(1-y/yy)+Voo
Abo2=a+w+dexp(Abo)

11 continue
  ictr=1
  do 100 ia=0,n
       do 95 ib=0,n-ia
          if (ib.lt.nsand) goto 95
          Temp(ictr)=Vfs(ia+1,ib+1)
          ictr=ictr+1
  95 continue
 100 continue
     do 110 ic=1,(ictr-1)
        big=temp(1)
        if (temp(ic).gt.big) big=temp(ic)
 110 continue
  sum=0
  do 120 id=1,(ictr-1)
     temp2(id) = temp(id) - big
     temp2(id)=dexp(temp2(id))
     sum=temp2(id)+sum
 120 continue
  sum=sum*dexp(big)
  write(25,999) n
 999 format(’ ’,’ current n value is: ’,i5)
  write(25,1000) Abo2
 1000 format(’ ’,’ computed value of Abo is: ’,d16.8)
  write(25,1001) a2Lo
 1001 format(’ ’,’ computed value of Lo is: ’,d16.9)
     fraction=sum/dexp(vo)
  write(25,1002) nsand,fraction
 1002 format(’ ’,’ fraction with
'i5,'or more sandwiches is: ',d16.8)
write(20,1003) dlog10(a2Lo),fraction,dlog10(fraction)
write(21,1004) n,fraction,dlog10(fraction)
1003 format(3e16.8)
1004 format(i5,2e16.8)
15 continue
10 continue
5 continue
stop
end
real*8 function fact(n)
implicit real*8 (a-h,o-z)
t1=.5*dlog(2.d0*3.14159265358979d0)
if (n.gt.6) goto 30
if (n.eq.0) fact=0.0
if (n.eq.1) fact=0.0
if (n.eq.2) fact=dlog(2.d0)
if (n.eq.3) fact=dlog(6.d0)
if (n.eq.4) fact=dlog(24.d0)
if (n.eq.5) fact=dlog(120.d0)
if (n.eq.6) fact=dlog(720.d0)
return
30 z=n+1
fact=(z-.5)*dlog(z)-z+t1+1./((12.*z)-1./((360.*z**3)+1./((1260.*z**5)-1./((1680.*z**7))
return
end
Appendix 3

This program is more complex and layered to solve the agglutination model. The functions $S(i)$, $G(n)$ and others which are used more than once have their own arrays to make the calculations easier.
ccc PROGRAM: BEST.FOR

to run, type
fortran/g_floating [wai.te.inter]best.f0r
link best
run best
type for000.dat

Philosophy of program:
sums of large numbers are handled with the
logarithmic, reduce by biggest, exp, sum,
log, increase by biggest method.
the quadratic formula is implemented with similar
consideration given to large/small numbers in
function f
Sterling’s formula is implemented in function fact

The summary of variables possessing the various functions is as
follows:

(nsize)
s(nsize+1,k+1) is the s(i)
g(nsize+1) is the g(nsize)
j(nsize+1) is the j(nsize)

hn is the h(n)
wn is the w(n)

pn is involved with the m or more bridges calculation

implicit real*8 (a-h,o-z)
dimension t1(300),bigest(300),sum1(300),s(300,300)
dimension zz(300,300),zzn(300,300),sum(300),tut(300)
dimension uu(300,300),uun(300,300),sumi(300),sl(300)
dimension s2(300), s3(300)
zero=0.0
mbrdg=3
do 987 qw=-5,7,1
ak1=10.**(qw)
ak2=1.d5
aki=1.d3
akz=1.d5
val=2.0
alpha=7.d5
beta=7.d5
y=ak2/alpha
z=akz/beta
vo=4.2d-13
write(0,2323)
2323 format('',k1,k2,ki,kz,val')
write(0,1212) ak1,ak2,aki,akz,val
1212 format('',5d16.8)
write(0,3434)
3434 format('','alpha,beta,vo')
write(0,1212) alpha,beta,vo
do 7860 n=100,101,10
write(0,8642)
8642 format('')
write(0,4545) n
4545 format('','current value for n is:',i6)
do 5 nsize=0,n
do 15 k=0,nsize/2
do 25 j=0,k
t1(j+1)=fact(nsize)-fact(k-j)
1-fact(j)-fact(nsize-k-j)+j*dlog(y/2.)
25 continue
bigest(k+1)=t1(1)
do 35 j=0,k
if(t1(j+1).gt.bigest(k+1)) bigest(k+1)=t1(j+1)
35 continue
sum1(k+1)=0.0
do 45 j=0,k
t1(j+1)=t1(j+1)-bigest(k+1)
if(t1(j+1).lt.-300.) goto 102
t1(j+1)=dexp(t1(j+1))
goto 101
102 t1(j+1)=0.0
101 sum1(k+1)=sum1(k+1)+t1(j+1)
45 continue
s(nsize+1,k+1)=dlog(sum1(k+1))+bigest(k+1)
15 continue
5 continue
do 666 i=800,300,-5
a=10.**(-i/100.)
write(0,8642)
write(0,5676) a
5676 format(’ ’,’ value for free antibody a is :’,d16.8)
x=a*val*ak1
do 135 nsize=0,n
do 145 k=0,nsize/2
t11=k*dlog(x)
t2=(nsize-k)*dlog(x)
if(k.eq.0) t3=-10000.
if(k.eq.0) goto 17
ak=k
t3=t11+dlog(ak)
17 if(nsize.eq.k) t4=-10000.
if(nsize.eq.k) goto 18
anmk=nsize-k
t4=t2+dlog(anmk)
18 if((nsize/2)*2.eq.nsize) goto 3
goto 333
3 if(k.eq.(nsize/2)) t11=t11-dlog(2.d0)
  if(k.eq.(nsize/2)) t2=t2-dlog(2.d0)
  if(k.eq.(nsize/2)) t3=t3-dlog(2.d0)
  if(k.eq.(nsize/2)) t4=t4-dlog(2.d0)
333 zz(nsize+1,k+1)=s(nsize+1,k+1)+t11
  zzn(nsize+1,k+1)=s(nsize+1,k+1)+t2
  uu(nsize+1,k+1)=s(nsize+1,k+1)+t3
  uun(nsize+1,k+1)=s(nsize+1,k+1) + t4
145 continue
  big=zz(nsize+1,1)
  bigu=uu(nsize+1,1)
  do 56 k=0,nsize/2
    if(zz(nsize+1,k+1).gt.big) big=zz(nsize+1,k+1)
    if(zzn(nsize+1,k+1).gt.big) big=zzn(nsize+1,k+1)
    if(uu(nsize+1,k+1).gt.bigu) bigu=uu(nsize+1,k+1)
    if(uun(nsize+1,k+1).gt.bigu) bigu=uun(nsize+1,k+1)
  56 continue
  do 67 k=0,nsize/2
    zz(nsize+1,k+1)=zz(nsize+1,k+1)-big
    zzn(nsize+1,k+1)=zzn(nsize+1,k+1)-big
    uu(nsize+1,k+1)=uu(nsize+1,k+1)-bigu
    uun(nsize+1,k+1)=uun(nsize+1,k+1)-bigu
67 continue
  sum(nsize+1)=0.0
  sumu(nsize+1)=0.0
  do 78 k=0,nsize/2
    if(zz(nsize+1,k+1).lt.-300.) goto 578
    zz(nsize+1,k+1)=dexp(zz(nsize+1,k+1))
  goto 478
578 zz(nsize+1,k+1)=0.0
478 continue
  if(zzn(nsize+1,k+1).lt.-300.) goto 579
  zzn(nsize+1,k+1)=dexp(zzn(nsize+1,k+1))
  goto 479
479 continue
  if(uu(nsize+1,k+1).lt.-300.) goto 679
  uu(nsize+1,k+1)=dexp(uu(nsize+1,k+1))
goto 680
679 uu(nsize+1,k+1)=0.0
680 continue
  if(uun(nsize+1,k+1).lt.-300.) goto 779
  uun(nsize+1,k+1)=dexp(uun(nsize+1,k+1))
goto 780
779 uun(nsize+1,k+1)=0.0
780 continue
  sum(nsize+1)=sum(nsize+1)+zz(nsize+1,k+1)
  1+zzn(nsize+1,k+1)
  sumu(nsize+1)=sumu(nsize+1)+uu(nsize+1,k+1)
  1+uun(nsize+1,k+1)
78 continue
  sum(nsize+1)=dlog(sum(nsize+1))+big
  sumu(nsize+1)=dlog(sumu(nsize+1))+bigu
135 continue
  do 601 k=1,n
    s1(k)=2.*fact(n)+k*dlog(x)+(k-1)*dlog(z)+dlog(aki)-fact(k)-
    1 2.*fact(n-k)
    ak=k
    s2(k)=s1(k)+dlog(ak)+2.*sum(n-k+1)
    s3(k)=s1(k)+dlog(2.d0)+sum(n-k+1)+sumu(n-k+1)
    s1(k)=s1(k)+dlog(2.d0)+2.*sum(n-k+1)
  601 continue
  bl=s1(1)
  b2=s2(1)
  do 602 k=1,n
    if(s1(k).gt.bl) bl=s1(k)
    if(s2(k).gt.b2) b2=s2(k)
    if(s3(k).gt.b2) b2=s3(k)
  602 continue
$s1(k) = s1(k) - b1$
$s2(k) = s2(k) - b2$
$s3(k) = s3(k) - b2$

603 continue
hn=0.0d0
wn=0.0d0
pn=0.0d0
do 604 k=1,n
if(s1(k) .lt.-300.) goto 720
s1(k)=dexp(s1(k))
goto 721
720 s1(k)=0.0
721 continue
if(s2(k) .lt.-300.) goto 722
s2(k)=dexp(s2(k))
goto 723
722 s2(k)=0.0
723 continue
if(s3(k) .lt.-300.) goto 724
s3(k)=dexp(s3(k))
goto 725
724 s3(k)=0.0
725 continue
hn=hn+s1(k)
wn=wn+s2(k)+s3(k)
if(k.lt.mbrdg) pn=pn+s1(k)
604 continue
pn=hn-pn
hn=dlog(hn)+b1
wn=dlog(wn) + b2
pn=dlog(pn)+b1
bbb=sum(n+1)
ccc=dlog(vo)
aaa=hn
bog=aaa
if(bbb.gt.bog) bog=bbb
if(ccc.gt.bog) bog=ccc
aaaa=aaa-bog
bbbb=bbb-bog
cccc=ccc-bog
v00=f(aaaa, bbbb, cccc)
8889 if(v00.gt.-300.) write(0,1111) dexp(v00)
   if(v00.le.-300.) write(0,1111) zero
term1=dlog(a)
term2=sumu(n+1)+v00
term3=wn+2.*v00
big=term1
if(term2.gt.big) big=term2
if(term3.gt.big) big=term3
term1=term1-big
term2=term2-big
term3=term3-big
stum=0.0
if(term1.lt.-300.0) goto 851
term1=dexp(term1)
goto 852
851 term1=0.0
852 continue
   if(term2.lt.-300.) goto 853
term2=dexp(term2)
goto 854
853 term2=0.0
854 continue
   if(term3.lt.-300.) goto 855
term3=dexp(term3)
goto 856
855 term3=0.0
856 continue
856 continue
abo=term1+term2+term3
abo=dlog(abo)+big
if(abo.gt.-300.) write(0,1113) dexp(abo)
if(abo.le.-300.) write(0,1113) zero
write(0,1114) dlog10(dexp(abo))
guppy=hn+2.*v00-ccc
if(guppy.gt.-300.) write(0,222) dexp(guppy)
if(guppy.le.-300.) write(0,222) zero
write(0,242) mbrdg,dexp(patch)
if(patch.gt.-300.) write(0,242) mbrdg,dexp(patch)
if(patch.le.-300.) write(0,242) mbrdg,zero
write(1,111) dlog10(dexp(abo)),dexp(guppy)
write(2,111) dexp(abo),dexp(guppy)
write(3,111) dlog10(dexp(abo)),dexp(patch)
write(4,1121) n,dexp(guppy)

242 format('',' fract. with ',i5,' or more
bridges : ',d16.8)
111 format('','3e16.8)
1121 format('','i5,3e16.8)
1111 format('',' computed value of v00 is: ',d16.8)
1113 format('',' computed value of abo is: ',d16.8)
1114 format('',' computed value of log(abo) is: ',d16.8)
222 format('',' f(agglut)= ',d16.8)
666 continue
7860 continue
987 continue
stop
end
real*8 function fact(n)
implicit real*8 (a-h,o-z)
t1=.5*dlog(2.d0*3.14159265358979d0)
if(n.gt.6) goto 10
if(n.eq.0) fact=0.0
if(n.eq.1) fact=0.0
if(n.eq.2) fact=dlog(2.d0)
if(n.eq.3) fact=dlog(6.d0)
if(n.eq.4) fact=dlog(24.d0)
if(n.eq.5) fact=dlog(120.d0)
if(n.eq.6) fact=dlog(720.d0)
return
10 z=n+1
   fact=(z-.5)*dlog(z)-z+t1+1./(12.*z)-1./
   1(360.*z*z*z)+1./(1260.*
   1z*z*z*z*z) -1./(1680.*z*z*z*z*z*z*z)
   return
end
real*8 function f(a,b,c)
implicit real*8 (a-h,o-z)
x=dlog (4.d0)+a+c-2.*b
111 format(' ',3d16.8)
   if(x.gt.0.0) goto 15
   if(x.lt.-5.0)goto 5
   x=dexp(x)+1.
   x=dsqrt(x)*dexp(b)-dexp(b)
   f=dlog(x)-dlog(2.d0)-a
   return
5 tl=dexp(a+c-b-b)
   f=c-b-tl+2.*tl*tl*tl-5.*tl*tl*tl+14.*tl**4-42.*tl**5
   return
15 y=2*b-a-c
   if(y.lt.-300.)goto 35
   y=dexp(y)
   goto 36
35 y=0.0
36 continue
   z=(-dsqrt(y)+dsqrt(y+4.d0))/2.
   f=dlog(z)+.5*c-.5*a