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Conformational selection is a dominant mechanism of ligand binding†

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Abstract

Molecular recognition in biological macromolecules is achieved by binding interactions coupled to conformational transitions that precede or follow the binding step, two limiting mechanisms known as conformational selection and induced fit, respectively. Sorting out the contribution of these mechanisms to any binding interaction remains a challenging task of general interest in biochemistry. Here we show that conformational selection is associated with a vast repertoire of kinetic behaviors, can never be disproved *a priori* as a mechanism of ligand binding, and is sufficient to explain the relaxation kinetics documented experimentally for a large number of systems. On the other hand, induced fit features a narrow spectrum of kinetic behaviors and can be disproved in many cases where conformational selection offers the only possible explanation. This conclusion offers a paradigm shift in the analysis of relaxation kinetics, with conformational selection acquiring pre-eminence as a mechanism of ligand binding. The dominant role of conformational selection supports the emerging structural view of the macromolecule as a conformational ensemble from which the ligand selects the initial optimal fit to produce a biological response.

All biological processes proceed through macromolecular interactions that take place over a wide range of time scales and binding affinities. The interaction involves two components: specific binding of the ligand, L, to its macromolecular target, E, and linked conformational changes that precede or follow the binding step. The combination of binding steps and conformational transitions in any given mechanism of recognition generates the repertoire of kinetic behaviors accessible to experimental measurements. The challenge facing the experimentalist is to decipher the nature of conformational transitions involved in the recognition process from analysis of the transient behavior of the system relaxing to equilibrium (1, 2). Once the differential equations associated with the kinetic mechanism are defined, the time evolution of the concentration of each species involved assumes the general form

$$x(t) = A_0 + \sum_{i=1}^N A_i \exp(\lambda_i t) = A_0 + \sum_{i=1}^N A_i \exp(-k_{obs,i} t) \quad (1)$$

where the λ_i s are the N non-zero eigenvalues of the matrix of the set of differential equations and the A_i s are the corresponding eigenvectors (1–3). Each λ_i defines a value of k_{obs} for the relaxation observed experimentally. A system of N independent species gives rise to N independent eigenvalues and relaxations. However, some of the relaxations may be too fast

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to measure by conventional stopped-flow techniques or may be spectroscopically silent. Analysis of a ligand binding interaction must therefore rely only on the relaxations accessible experimentally and interpretation of the underlying mechanism should be based on the simplest kinetic scheme consistent with the observations.

Two basic mechanisms of ligand recognition have been widely used in the analysis of relaxation kinetics (2, 4). Both mechanisms envision a conformational isomerization that either precedes or follows the binding step. In the first case (Scheme 1, Figure 1), originally discussed by Eigen as pre-equilibrium (1, 2) and more recently referred to as conformational selection (5, 6), the macromolecule is assumed to undergo a transition between multiple conformations from which one is selected for binding by the ligand. In this scheme, the number of conformations accessible to the macromolecule decreases upon ligation as the ligand singles out the one with optimal fit from the pre-existing ensemble. In the second case (Scheme 2, Figure 1), introduced by Koshland as induced fit (7), the macromolecule is assumed to first bind the ligand and then change its conformation in the bound form to optimize the complex. Here, the number of conformations accessible to the macromolecule increases with ligation. In the case of conformational selection, the macromolecule has a pre-existing flexibility to accommodate the ligand. In the case of induced fit, such feature is unnecessary because the ligand induces the optimal fit with the binding site after the initial complex is formed. When available, structural information on the free and bound forms of the macromolecule is of diagnostic value: evidence of multiple conformations in the free form supports conformational selection (8) and multiple conformations in the bound form support induced fit (9). When structural information is not available or inconclusive, the distinction between the two mechanisms must be made from the kinetic signatures of ligand binding. Kinetic data are also necessary to derive information on the rate constants governing the recognition process.

The mechanisms in schemes 1 and 2 (Figure 1) contain three species, of which only two are independent and relax to equilibrium according to two independent exponentials or k_{obs} values that can be derived analytically (10). The largest k_{obs} is linked to the binding step in both mechanisms and increases asymptotically as $k_{\text{on}}[\text{L}]$ (Figure 1). The smaller k_{obs} converges to a finite asymptote and conveys information about the conformational transitions. When the smaller k_{obs} is accessible to experimental measurements, the following can be concluded (10): if k_{obs} decreases with or is independent of $[\text{L}]$, then the macromolecule behaves according to conformational selection (Scheme 1, Figure 1) and induced fit (Scheme 2, Figure 1) is ruled out unambiguously; if k_{obs} increases with $[\text{L}]$, then the macromolecule behaves according to induced fit or conformational selection and neither mechanism is ruled out unambiguously. The latter conclusion refutes the long held notion that an increase of k_{obs} with $[\text{L}]$ is unambiguous proof of induced fit (2, 4, 11, 12), which is true only when conformational transitions are rate-limiting (10). The important finding emerged from the analytical solution of schemes 1 and 2 in the general case (10) is that conformational selection, unlike induced fit, can never be disproved *a priori* as a mechanism of ligand binding. Indeed, conformational selection is sufficient to account for all possible behaviors of k_{obs} and becomes necessary when k_{obs} decreases with $[\text{L}]$.

How rich is the functional repertoire of conformational selection in the general case? When the number of species involved in the kinetic mechanism of ligand binding increases, the analytical solution of the differential equations becomes algebraically intractable. That limits our ability to assess the general validity of conclusions drawn from the analysis of simple kinetic schemes. Here we take an alternative approach and explore the asymptotic behavior of the underlying relaxations for more complex kinetic schemes both analytically and computationally. Conformational selection emerges from this analysis as a dominant

mechanism of ligand binding that is sufficient to explain the kinetic behavior of practically all systems reported in the literature thus far.

Methods

Simulations were performed using Mathematica (version 8). Values for independent kinetic rate constants were randomly assigned using the RandomVariate function according to 10^{RV} , where RV is a random number following a normal distribution with $\mu=1.5, 3.0$ or 4.5 and $\sigma=0.25, 0.5, 1.5$ or 3.0 . The overall properties of the various schemes were found to be independent of the values of μ and σ used in the simulations. The results in Figures 4 and 5 refer to $\mu=3$ and $\sigma=0.5$ to best reproduce experimental data and involved rate constants spanning a range of 5 orders of magnitude. Dependent rate constants due to detailed balancing were defined as $k_{14}=k_{12}k_{23}k_{34}k_{41}/k_{21}k_{32}k_{43}$, $k_{32}=k_{31}k_{12}k_{23}/k_{21}k_{13}$, and $k_{24}=k_{42}k_{23}k_{34}/k_{43}k_{32}$ for schemes 3, 4, and 5 (Figure 1), respectively. The characteristic equation was solved numerically for each set of randomly assigned rate constants at $[L]=0$ and $[L]=10^{51}$ and a total of 10^6 sets of solutions were generated for each mechanism. The k_{obs} for each limiting case ($[L]=0$ and $[L]=\infty$) were ordered and paired up such that the largest, second largest, and third largest k_{obs} when $[L]=0$ were associated with the largest, second largest, and third largest k_{obs} , respectively, when $[L]=\infty$. The logarithm of the ratio of k_{obs} values for $[L]=\infty$ and $[L]=0$ was used to assess the magnitude and direction of change for k_{obs} with increasing $[L]$. Simulations revealed that the largest k_{obs} for all schemes increased according to $k_{on}[L]$. For the general linkage scheme (Scheme 3, Figure 1), the second largest k_{obs} also increased according to $k_{on,2}[L]$, where $k_{on,2}$ is the smaller of the two second-order rate constants for ligand association. All remaining k_{obs} were found to be several orders of magnitude smaller. These observations guided the assumptions used to derive the analytical expressions for k_{obs} in the limit where $[L]=\infty$. In particular, $k_{obs,1} \gg k_{obs,2}, k_{obs,3}$ for schemes 1 and 2 and 4, 5 and 6 (Figure 1), while $k_{obs,1}, k_{obs,2} \gg k_{obs,3}$ for scheme 3 (Figure 1). Additionally, when $[L]=\infty$, only the highest order $[L]$ terms were considered significant. No further assumptions were required to derive the expressions presented in Figure 1.

Results

Conformational selection (Scheme 1, Figure 1) and induced fit (Scheme 2, Figure 1) exhaust the repertoire of two-step kinetic mechanisms linking ligand binding to conformational transitions when a single saturable k_{obs} is measured experimentally. Improved resolution of stopped-flow spectrometers has made it increasingly common to detect two saturable relaxations. In this case, four possible scenarios should be considered when $[L]$ increases: both k_{obs} increase, both k_{obs} decrease, the larger k_{obs} increases and the smaller k_{obs} decrease, the larger k_{obs} decreases and the smaller k_{obs} increases. Figure 2 displays these four scenarios as the quadrants of a plot where the range of smaller k_{obs} values is displayed as a function of the range of larger k_{obs} values. The range is defined as the ratio of the values of k_{obs} when $[L]=\infty$ and $[L]=0$. The points in the plot map experimental data taken from a variety of systems, namely the immunoglobulins IgE (13) and IgG (14), protein kinase A (15), DnaC (16), CheA (17), histone deacetylase-like amidohydrolase (18), polymerase X (19), 3-hydroxybenzoate 6-hydroxylase (20), 3-chloroacrylic acid dehalogenase (21), proline utilization A protein (22), ACTR and CREB-binding protein (23), G-quadruplex folding (24) and DnaB (25). With the exception of DNP-Ser binding to IgE SPE7 (13), these points fall in all quadrants but the one in the upper left corner where the faster saturable relaxation decreases and the slower saturable relaxation increases with $[L]$. What mechanism explains the behavior of such systems?

The existence of two saturable k_{obs} requires at least two steps in the kinetic scheme associated with conformational transitions. In their simplest version, both conformational selection (Scheme 1, Figure 1) and induced fit (Scheme 2, Figure 1) do not meet this requirement and cannot explain the behavior of the systems in Figure 1. The two mechanisms are special cases of a general linkage scheme (Figure 3, Scheme 3, Figure 1) where the macromolecule is assumed to exist in two alternative conformations each capable of binding the ligand at a single site (2, 11). The equilibrium properties of this scheme are well known and form the basis of linkage thermodynamics (26, 27). The steady state properties have also been discussed in great detail first by Botts and Morales (28) in connection with the action of a modifier on enzyme activity and then by Hill (3) in the context of energy transduction. Recent discussions of the linkage scheme have been presented in terms of mutational analysis (29), fluxes (30) and diffusion-controlled reactions (31). The four species in the linkage scheme are numbered sequentially to simplify notation for the kinetic rates. Here, k_{14} ($\text{M}^{-1}\text{s}^{-1}$) is the second-order association rate constant for ligand binding to species 1 and k_{23} ($\text{M}^{-1}\text{s}^{-1}$) is the second-order association rate constant for ligand binding to species 2. Both rate constants are analogous to the k_{on} used in the simpler schemes 1 and 2 (Figure 1). k_{41} (s^{-1}) is the first-order rate constant for dissociation of the complex (species 4) into species 1, while k_{32} (s^{-1}) is the dissociation rate constant from species 3 into species 2. Again, both rates constants are analogous to k_{off} used in the simpler schemes 1 and 2 (Figure 1). k_{12} (s^{-1}) and k_{21} (s^{-1}) are the first-order rate constants for conformational isomerization between the free species and k_{34} (s^{-1}) and k_{43} (s^{-1}) are the analogous rate constants for isomerization between the bound species. Of the four species in the general linkage scheme, only three are independent and associated with three distinct relaxations or k_{obs} . The behavior of the various relaxations for finite $[\text{L}]$ is algebraically intractable but easier to discuss in the relevant limits $[\text{L}]=0$ and $[\text{L}]=\infty$ where the macromolecule is either free or bound to the ligand. When $[\text{L}]=0$, two of the three relaxations depend on the properties of the bound ensemble (k_{34} , k_{43} , k_{41} , and k_{32} , Figure 1), while one relaxation equals the sum $k_{12}+k_{21}$, which gives the rate of equilibration between the two possible conformations of the macromolecule in the absence of ligand. The precise ordering of the three k_{obs} , from largest to smallest, depends on the relative magnitudes of the rate constants. When $[\text{L}]=\infty$, two of the three independent k_{obs} increase linearly with $[\text{L}]$, consistent with the presence of two independent binding steps in the general linkage scheme. The third and smallest k_{obs} plateaus at $k_{34}+k_{43}$, which corresponds to the rate of equilibration between the two possible conformations of the macromolecule in the bound form. Although the scheme has two steps reflecting conformational transitions, it is associated with only one saturable relaxation that may increase or decrease with $[\text{L}]$ depending on whether $k_{34}+k_{43}$ is greater or smaller than the value for $[\text{L}]=0$. The possible range of such k_{obs} values is plotted in Figure 4 for a large number of simulated cases and is reproduced almost entirely by conformational selection.

In addition to being unable to generate two saturable relaxations as required by the experimental data in Figure 2, the general linkage scheme features a kinetic repertoire no richer than that of its special case conformational selection (Figure 4). A kinetic mechanism where conformational transitions are more numerous than binding steps becomes of interest. Consider the extension of the basic conformational selection scheme to include a third conformation in the free form (Scheme 4, Figure 1). Of the three relaxations in the scheme, one reflects the binding step and increases linearly with $[\text{L}]$ when $[\text{L}]=\infty$, but the other two are saturable and can either increase or decrease with $[\text{L}]$ yet never in a combination where the larger decreases when the smaller increases. Remarkably, that matches the distribution of k_{obs} values observed experimentally with the exception of a single case (Figure 5). The population distributions predicted for this scheme depend on the standard deviation used to generate the rate constants in the simulation. As the range of kinetic rate constants increases, the scheme predicts that both k_{obs} will increase in 49% of the simulations (52% observed

experimentally), the larger k_{obs} will increase and the smaller k_{obs} will decrease in 32% of simulations (24% observed experimentally), and both k_{obs} will decrease in 19% of the simulations (19% observed experimentally). In none of the simulations is a decrease in the larger k_{obs} coupled with an increase in the smallest k_{obs} . An analogous extension of the basic induced fit scheme to include a third conformation in the bound form (Scheme 5, Figure 1) also produces three relaxations. One reflects the binding step and increases linearly with $[L]$ when $[L] = 0$, but the other two saturable relaxations always increase with $[L]$ and populate only the upper right quadrant of the plot (Figure 5). The inability of induced fit to produce a relaxation that decreases with $[L]$ appears to be a general property that limits application of this mechanism in the analysis of ligand binding data. On the other hand, conformational selection is more versatile and offers a sufficient explanation of the kinetics of ligand binding for a large number of different systems.

It is of interest to explore what kinetic mechanism would produce relaxations that populate the plot in Figure 2 in its entirety. Going back to the general linkage scheme (Figure 3, Scheme 3, Figure 1), we note that elimination of one binding step keeps the number of relaxations unchanged but increases the saturable k_{obs} from one to two (Scheme 6, Figure 1). The transformation produces a linearized scheme with a pre-existing equilibrium between two forms, allowing selective binding to one of the forms and subsequent isomerization to a second bound form. The resulting mechanism is a mixture of conformational selection and induced fit and its properties have been discussed previously under certain approximations (16, 32). It can be shown mathematically that when $[L]=0$ two of the relaxations from this mixed scheme (Scheme 6, Figure 1) are equivalent to the two asymptotic limits of induced fit (Scheme 2, Figure 1) and depend on the rates of isomerization between the bound conformations (k_{34} , k_{43}) and the rate of ligand dissociation (k_{32}). The third relaxation is equal to $k_{12}+k_{21}$, i.e., the equilibration rate between the two conformations in the free form. These asymptotes are minimally altered from those of the general linkage scheme (Scheme 3, Figure 1). In the limit where $[L] = \infty$, the largest k_{obs} grows as $k_{23}[L]$ (non-saturable relaxation) while the remaining two k_{obs} plateau at the finite values of k_{12} and $k_{34}+k_{43}$, corresponding to the asymptotes of the saturable relaxations in schemes 1 and 2, respectively. Linearization of the general linkage scheme generates two saturable relaxations that may either increase or decrease with increasing $[L]$ independent of each other (Figure 5), thereby covering the entire spectrum of possible kinetic behaviors for two saturable relaxations. In addition, five of the six asymptotic values for the three relaxations contain information on the rates of conformational isomerization (Scheme 6, Figure 1).

Discussion

The main conclusion of our analysis is that relaxation kinetics cannot disprove conformational selection as a mechanism of ligand binding when one or two saturable relaxations are accessible to experimental measurements. We would like to propose as a conjecture that this conclusion applies generally to any number of saturable relaxations. Indeed, extension of scheme 4 (Figure 1) to include four free species produces three saturable k_{obs} , all of which able to either increase or decrease with $[L]$. The argument that conformational selection is disproved by enzymes with lid-gated active sites such as ribokinase (33), adenosine kinase (34) and glucokinase (9), where the conformational change may only follow the binding step (35), is easily refuted. The closed conformation of the enzyme may pre-exist in the absence of ligand, no matter how energetically unfavorable, making the mixed scheme (Scheme 6, Figure 1) ideally suited to describe the relaxation properties of such enzymes.

The preponderance of systems relaxing to equilibrium with a single k_{obs} increasing with $[L]$ to a saturable level has been used for decades as evidence of induced fit as a dominant mechanism of ligand binding (1, 2, 4, 11, 12). We now know that this conclusion is incorrect: conformational selection easily accounts for a k_{obs} increasing with $[L]$ to a saturable level when the value of k_{off} for ligand dissociation is slower than the rates of isomerization of the pre-existing conformations of the macromolecule (10). Furthermore, the number of systems directly disproving induced fit with k_{obs} decreasing with $[L]$ is rapidly expanding and includes alkaline phosphatase (32), glucokinase (36), several trypsin-like proteases (10, 12, 37–42), the immunoglobulin IgE (43), and DNA in its B to Z transition (44). The richer repertoire of kinetic behaviors associated with conformational selection is not limited to the case of a single saturable relaxation, but extends to the case of two saturable relaxations so abundantly represented in the literature (13–25). With a single exception, the entire landscape of kinetic properties displayed by these systems (Figure 2) is captured by an extension of conformational selection to include a third conformation in the ensemble of free forms available to the macromolecule (Figure 5). As the number of pre-existing conformations increases, so does the number of saturable relaxations, thereby making conformational selection capable of accounting for even more complex ligand binding kinetics. A paradigm shift emerges where conformational selection acquires pre-eminence as a mechanism of ligand binding.

The combination of conformational selection and induced fit produces a mixed scheme (Scheme 6, Figure 1) whose properties expand those of conformational selection alone and explain the peculiar case of a fast relaxation decreasing when the slow relaxation increases (Figure 5). Although this scenario has so far occurred in a single case reported in the literature (13), its interpretation requires addition of induced fit to the mechanism of ligand binding. The richness of kinetic behaviors associated with conformational selection, either alone (Schemes 1 and 4, Figure 1) or in combination with induced fit (Scheme 6, Figure 1), offers functional support to the emerging view of the macromolecule as a conformational ensemble (6, 45, 46) promoted by NMR (47) and x-ray crystallography (8). Thermal fluctuations generate a large number of interconverting states compatible with the allowed folds from which the ligand selects the most complementary one(s) for optimal fit. The ensemble view of the macromolecule agrees well with the fact that conformational selection can never be disproved as a mechanism of ligand binding, which supports the conclusion that the pre-existence of alternative conformations is a fundamental property of all macromolecules. This conclusion is not incompatible with induced fit providing an additional mechanism for optimization of the complex and the scenario envisioned by the mixed scheme (Scheme 6, Figure 1) is likely to become more common as our ability to explore the conformational landscape of the macromolecule improves. Indeed, maltose binding protein shows structural evidence of conformational transitions preceding and following the binding step (47, 48).

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Figure 1.

Mechanisms of ligand binding discussed in the text with corresponding asymptotic values for k_{obs} in the limits $[L]=0$ and $[L]=\infty$. Schemes 1 and 2 correspond to the well-known conformational selection and induced fit mechanisms of binding. Scheme 3 is the general linkage scheme, of which schemes 1 and 2 represent special cases (see also Figure 3). Schemes 4 and 5 represent generalized extensions of schemes 1 and 2, respectively. Scheme 6 is a special case of scheme 3, where the binding interaction with species 1 is negligible. Special notations are $C=k_{12}+k_{21}+k_{13}+k_{31}+k_{23}+k_{32}$ and $S=k_{12}k_{23}+k_{12}k_{31}+k_{12}k_{32}+k_{13}k_{21}+k_{13}k_{23}+k_{13}k_{32}+k_{21}k_{31}+k_{21}k_{32}+k_{23}k_{31}$ (Scheme 4), $I=k_{23}+k_{32}+k_{24}+k_{42}+k_{34}+k_{43}$ and $F=k_{24}k_{32}+k_{23}k_{34}+k_{24}k_{34}+k_{23}k_{42}+k_{32}k_{42}+k_{34}k_{42}+k_{23}k_{43}+k_{24}k_{43}+k_{32}k_{43}$ (Scheme 5).

Figure 2.

Four-quadrant plot with experimental data. The range of the smaller k_{obs} is plotted as a function of the range of the larger k_{obs} for the following experimental systems: IgE (green squares), IgG (green circle), protein kinase A (red circles), DnaC (yellow circle), CheA (yellow triangle), histone deacetylase-like amidohydrolase (blue square), polymerase X (yellow square), 3- hydroxybenzoate 6-hydroxylase (green triangle), 3-chloroacrylic acid dehalogenase (red square), proline utilization A protein (blue circle), ACTR and CREB-binding protein (red diamond), K^+ -mediated G-quadruplex folding (red triangles), and DnaB (blue triangles). The range is defined as the ratio of the values of k_{obs} for $[\text{L}] = \text{---}$ and $[\text{L}] = 0$. Multiple points using the same color and shape indicate different ligands binding to the same macromolecule. Points falling in the upper right quadrant indicate that both k_{obs} increase, points in the lower right quadrant indicate a larger k_{obs} that increases coupled to a smaller k_{obs} that decreases, points in the lower left quadrant indicate that both k_{obs} decrease, and points in the upper left quadrant indicate a larger k_{obs} that decreases coupled with a smaller k_{obs} that increases - a scenario confined exclusively to DNP-Ser binding to IgE SPE7 (13).

Figure 3. General linkage scheme (Scheme 3, Figure 1) with conformational selection and induced fit as special cases. Species 1 and 2 represent two species with differing affinities for the ligand L. Deletion of species 4 yields conformational selection (Scheme 1, Figure 1) while the removal of species 2 produces induced fit (Scheme 2, Figure 1).

Figure 4.

Distribution of the range of saturable k_{obs} , computed as the ratio between the values for $[L]=$ and $[L]=0$, for conformational selection (Scheme 1, Figure 1), induced fit (Scheme 2, Figure 1) and the general linkage scheme in Figure 3 (Scheme 3, Figure 1). Negative values indicate simulations in which k_{obs} decreases with increasing $[L]$, while positive values indicate simulations in which k_{obs} increases with $[L]$. Individual rate constants in these simulations varied independently over a range of 5 orders of magnitude.

Figure 5.

Four-quadrant plot with the simulations of extended conformational selection (Scheme 4, Figure 1), extended induced fit (Scheme 5, Figure 1), and mixed scheme (Scheme 6, Figure 1). Simulations were performed as described under Methods and the results are plotted in the form of a density map. Experimental data displayed in Figure 2 (black circles) are overlaid on the simulations for comparison. The best fits of the experimental data to the extended conformational selection and mixed models were obtained by increasing the standard deviation used to generate the rate constants ($\sigma = 3.0$). Removing the single point of DNP-Ser binding to IgE SPE7 in the upper left quadrant allows scheme 4 to completely explain the observed experimental distributions shown in Figure 2 ($\chi^2 = 2.32$, $\sigma = 5.99$). On the other hand, scheme 6 gives a statistically significant worse fit of the observed distribution ($\chi^2 = 20.25$, $\sigma = 7.82$).