**ABSTRACT**

Molecular recognition is central to biological processes, function, and specificity. Proteins associate with ligands with a wide range of association rate constants, with maximal values matching the theoretical limit set by the rate of diffusional collision. As less is known about RNA association, we compiled association rate constants for all RNA/ligand complexes that we could find in the literature. Like proteins, RNAs exhibit a wide range of association rate constants. However, the fastest RNA association rates are considerably slower than those of the fastest protein associations and fall well below the diffusional limit. The apparently general observation of slow association with RNAs has implications for evolution and for modern-day biology. Our compilation highlights a quantitative molecular property that can contribute to biological understanding and underscores our need to develop a deeper physical understanding of molecular recognition events.

Keywords: molecular recognition; binding kinetics; RNA structure

**INTRODUCTION**

Molecular recognition underlies all of life. The individual molecules that determine the accuracy of what complexes form and what reactions occur are responsible for the fidelity of biological responses and ultimately for the cellular and organismal order that is emblematic of life. Nevertheless, the most accurate process is of no value to an organism if it cannot be carried out rapidly, to overcome stray chemical and physical processes and to outcompete other organisms.

The rate of association between molecules sets a speed limit for metabolic functions and regulatory responses, with a theoretical maximum set by the rate of diffusional collision (see Eigen and Hammes 1963; Berg and von Hippel 1985; Fersht 1999; Schreiber et al. 2009 and below). Underscoring the importance of speed, nature has evolved clever mechanisms to overcome the physical limit of three-dimensional diffusion, for example, allowing proteins to slide in one dimension along DNA or hop between DNAs and harnessing electrostatic interactions to speed associations (von Hippel and Berg 1989; Halford and Marko 2004; Schreiber et al. 2009). On the other end of the spectrum, association rates can fall many orders of magnitude below diffusion control, and in some cases slow association can contribute to regulation and specificity (e.g., Wu et al. 2002; James and Tawfik 2005).

Given the importance of molecular recognition and its speed, and the generally adopted model of an early RNA world that was later overcome by a protein-dominant world (Gilbert 1986; Cech 2009), we wondered whether there might be differences in molecular recognition between RNA and proteins at the level of association rates. Such differences could have contributed to the course of evolution and may impact current-day biology. On the molecular level, differences in association rates may reflect atomic-level properties critical in molecular recognition and may provide a unique window into understanding these processes.

**RESULTS**

To broadly assess the kinetics of association, we gathered all of the RNA/ligand association rate constants we could find in the literature ($n = 64$) (Supplemental Tables S1, S2). For comparison, we gathered a large number of protein/ligand association rate constants ($n = 207$) (Supplemental Table S3). We also examined a third category of association events, those between proteins and RNA molecules ($n = 22$) (Supplemental Table S4). While we expect that our RNA/ligand list is nearly complete, there are too many protein/ligand measurements to fully assemble (see Materials and Methods). Nevertheless, we expect our list to be reasonably representative and suitable for revealing trends and stimulating ideas and discussion about their physical origins and biological implications.

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RNA and protein ligand association rate constants

Figure 1 shows rate constants for ligands binding to structured RNAs and to proteins (leftmost and middle columns, respectively). We expect a generic upper limit corresponding to diffusional collision at $\sim 10^9$ M$^{-1}$ sec$^{-1}$ (Eigen and Hammes 1963; Berg and von Hippel 1985; Fersht 1999; Schreiber et al. 2009). For protein ligands, the values span a wide range of over 10$^6$-fold and approach and even exceed the diffusional limit, as electrostatic attraction between the protein and ligand can increase their rate of association (e.g., Lambeth et al. 1980; Wallis et al. 1995; Schreiber and Fersht 1996; Radic et al. 1997; Polticelli et al. 1998; and see Schreiber et al. 2009 and references therein). There is also a large range observed for RNA/ligand association rate constants. However, the highest values are distinctly lower than those for proteins, falling well below the diffusional limit (Fig. 1, compare left and middle columns).

As a crude test of whether there is some type of intrinsic limitation to the rate of associations involving RNA, we separately plotted rate constants for the association between RNA and proteins (Fig. 1, rightmost column). The values exceeded those for RNA with non-protein binding partners (Fig. 1, compare left and right columns) and reach the upper range observed for other protein/ligand associations (Fig. 1, compare middle and right columns), indicating that there is no inherent limitation to association involving RNA.

To further probe association rates involving RNA we parsed its association rates (Fig. 2), first separating out RNA/RNA associations (“RNA ligands”) from those with non-RNA ligands (“Small molecule ligands”), and, for the RNA/RNA group, further demarcating subgroups that involve formation of base pairs only (“BP”), formation of base pairs and additional tertiary interactions (“BP + 3°”), and formation of non-Watson–Crick tertiary interactions only (“3°”). RNA binding to structured RNAs that involve base-pair formation is significantly faster than binding that does not (Fig. 2, compare “BP” and “3°”), although the number and diversity of RNAs binding via only tertiary contacts is small. Interestingly, the binding of structured RNAs via base-pairing occurs at rates similar to those for simple duplex formation between short RNA oligomers (Fig. 2, compare “BP” and “RNA duplex”; “RNA duplex” values are from Turner et al. 1990 and are listed in Supplemental Table S5), suggesting that there are no general constraints or accelerants to base-pairing in structured RNAs. Non-RNA ligands bind to structured RNAs with rate constants extending up to those for recognition via base-pair formation (Fig. 2, compare “Small molecule ligands” and “BP”), although on average the values are somewhat lower and a wide range of values is observed. Interestingly, the measured values for naturally occurring and in vitro–selected RNAs have similar means and span similar ranges (Fig. 2, compare “Natural” and “In vitro” columns above “Small molecule ligands”).

Implications and mechanisms of slow RNA association

We consider the assembled RNA association data from biological and then physical perspectives. Our goal is to stimulate discussion about potential implications and origins of the observed trends and, ultimately, experiments that reveal the physical bases for the observed behavior and the ensuing biological manifestations.

Implications of slow RNA recognition in early evolution

We previously proposed that nonspecific RNA chaperones provided the first foothold into the RNA world for peptides and then proteins (Tsukihashi et al. 1993; Herschlag 1995), and it is possible that association rates played a role in these transitions. According to this model, short peptide sequences that nonspecifically bind RNA acted as chaperones; short

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4 For RNAs that form non-Watson–Crick tertiary interactions only (3°), $k_{on}$ values were obtained for tertiary contact partners physically separated in two molecules (see Supplemental Table S2 and references therein).
polymers and nonspecific interactions would be more likely to arise evolutionarily than long polymers and specific interactions, and RNA’s tendency to fall into kinetic traps (Gartland and Sueoka 1966; Lindahl et al. 1966; Herschlag 1995; Uhlenbeck 1995; Russell 2008) would allow a selective advantage from nonspecific chaperones that could aid RNAs’ escape from these traps and thereby yield a higher fraction of RNA molecules in functional conformations (Fig. 3).

Considering association rates, it is also possible that early peptides provided selective advantages by increasing association rates, as has been seen in model RNA/peptide and protein systems (e.g., Kumar and Wilson 1990; Munroe and Dong 1992; Herschlag et al. 1994; Portman and Dreyfuss 1994).

The next evolutionary step, according to this model, was the transition to longer peptides and specific functions. Once peptides gained a foothold there would be advantages and successive selection for longer and more accurate peptide production, presumably leading to specific binding and structured proteins (Fig. 3, “RNA + protein”) (Herschlag 1995; Cech 2009). But why would proteins have supplanted RNAs in functional roles so resoundingly? Common and reasonable ideas include an ability of proteins to create more specific folds and binding pockets (Herschlag 1995; Russell 2008) and to carry out more efficient and a broader range of chemical catalysis with their >20 amino acid side chains than the four canonical RNA side chains (Herschlag 1995; Narlikar and Herschlag 1997). Instead or additionally, the diversity of protein side chains instead of RNA side chains may have arisen because of the constraint on RNA to also serve as genetic material, because of more facile routes to synthesize and incorporate additional amino acid side chains (e.g., Huber et al. 2003; Leman et al. 2004) and/or because of limits imposed on RNA folding, molecular recognition, and catalytic function by its uniformly charged backbone (Narlikar and Herschlag 1997).

The observations herein suggest that an additional driving force could have been operative: faster molecular recognition by proteins than by RNA (Figs. 1, 3). Especially if the metabolic milieu in the RNA world was rather complex, with efficient ribozymes limited by their recognition speed, increasing association rates and thus enzyme efficiency beyond that possible with ribozymes could have provided a selective advantage for proteins. Indeed, the potential for faster recognition by proteins may also have provided some of the selective advantage for additional amino acid side chains with different “shapes” that can better preorganize protein structures and thereby speed ligand binding (see below).

**Implications of slow RNA recognition in modern-day biology**

The simplest model for the presence of complex structured RNAs that carry out functions in modern biology is that these are vestiges from an earlier RNA world (Gilbert 1986; Cech 2009). Indeed, the RNA core of the ribosome provides the strongest single argument for functional RNAs predating proteins (Cech 2000; Nissen et al. 2000). While it is possible
that all modern-day RNAs are evolutionary vestiges that would be eliminated in a perfectly engineered—rather than evolved—organism, it is also possible that certain processes benefit from utilization of RNA, and that some of this utilization is connected to how RNA recognizes ligands. Base-pairing is a highly effective and highly specific means to recognize and discriminate between RNAs, and is highly modular. While there are cases of modular proteins that bind individual bases, each protein module involves multiple amino acids and thus requires a longer and less probable an evolutionary search to switch recognition specificity to another base than the single base swap needed for RNA. For this reason, recognition by RNA may be more evolvable (Houle 1992; Wagner and Altenberg 1996; Kirschner and Gerhart 1998; Wagner 2005; Bloom et al. 2006). It may also be harder to achieve near-uniform affinity for any base accompanied by high specificity for that base with protein modules than with individual RNA bases. Finally, base pairs in RNA duplexes are themselves accessible and can therefore be recognized by additional RNA elements (or proteins) to increase (or decrease) base-pair affinity and specificity (e.g., Leontis et al. 2002; Carlson et al. 2003; Arya 2011; Masliah et al. 2013 and Houglund et al. 2006 and references therein).

When else might RNA prove advantageous? Folding studies have demonstrated that RNA readily adopts stable alternative structures—and we and others have argued and shown that these alternative structures can be inactive and deleterious (Gartland and Sueoka 1966; Lindahl et al. 1966; Herschlag 1995; Uhlenbeck 1995; Russell 2008); however, riboswitches are biological RNAs that use such metastable states for regulation (Haller et al. 2011; Serganov and Patel 2012; Furtig et al. 2015), and ribosome and spliceosome function rely on multiple conformational states (Guthrie and Patterson 1988; Madhani and Guthrie 1994; Voorhees and Ramakrishnan 2013; Chen and Moore 2014; Warnasooriya and Rueda 2014; Rodnina et al. 2017). Thus, kinetically trapped metastable states could have been co-opted through evolution to provide a series of functional conformational steps, with these steps controlled by proteins such as ATPases and GTPases in pre-mRNA splicing and translation (Staley and Guthrie 1998; Chang et al. 2013; Jarmoskaiate and Russell 2014; Bourgeois et al. 2016).

Finally, the slow observed ligand recognition by RNA provides an opportunity for kinetic control of specificity when binding is rate limiting (Karbsttein and Herschlag 2003; Gleitsman and Herschlag 2014). In this mechanism, referred to as “$k_{on}$ specificity,” added neighboring interactions can increase the lifetime of an initial encounter complex and thereby provide an attached ligand multiple chances at productive binding prior to diffusing away and thus a higher $k_{on}$. For RNA, adding neighboring interactions can be as simple as adding residues that can base pair with the structured RNA, and indeed there is evidence for this mechanism with the Tetrahymena group I intron (Karbstlein and Herschlag 2003). There, residues neighboring the guanosine at the 3′-intron–exon junction form base pairs with intron sequences near the guanosine binding site and make association of that guanosine faster and more probable than other G residues. Similarly, positively charged functional groups on ligands can provide an “anchor” that increases residence times and
new therapeutics that target RNA. Bound state, and it may be possible to co-opt this strategy for inhibitable state and need not even be maintained in the final form receptor-ligand complex (R-L) represented as simple association processes governed by a single set of association and dissociation rates. When association is slower than diffusion-limited collision, additional steps are needed to describe binding. (B) Induced fit and conformational capture models of ligand binding. These models account for slower-than-diffusional binding. The tertiary capture model invokes conformational changes prior to the encounter between R and L, while the induced fit model represents the opposite extreme in which conformational changes occur after the initial encounter. (C) Complex model of ligand binding considering an ensemble of conformational states and series of conformational changes to the receptor during the association process. Even the complex representation in the figure is a vast simplification, as it depicts a linear order of conformational changes (and a limited number of conformational steps) rather than a larger and more realistic multi-dimensional energy landscape that can account for multiple orders for the local conformational changes and multiple possible binding pathways.

thus association rate constants. There is evidence for such an effect for a positively charged guanosine analog binding to a group I intron (RN Sengupta and D Herschlag, unpubl.) and analogous effects may be operative in aminoglycoside antibiotics binding to specific ribosome states (Vicens and Westhof 2003; Trylska and Kulik 2016) and other structured RNAs (Stage et al. 1995; Walter et al. 1999). These interactions can allow a ligand to “wait” for the formation of an inhibitable state and need not even be maintained in the final bound state, and it may be possible to co-opt this strategy for new therapeutics that target RNA.

**Physical origins of slow RNA recognition**

Molecular binding events are typically represented as simple association processes (Fig. 4A). The widespread awareness that the predominant conformational states of free and bound RNAs often differ has led to considerable discussion about induced fit or conformational capture models in which conformational changes occur after the initial encounter or prior to the encounter, respectively (Fig. 4B) (e.g., Leulliot and Varani 2001; Rose and Weeks 2001; Webb et al. 2001; Boehr et al. 2009; Daniels et al. 2014). Most kinetic experiments cannot distinguish these pathways (Chakraborty and Di Cera 2017), but even more critically, these alternatives represent limiting models for a wide range of possible pathways (Fig. 4C); labeling a process by one of these two designations does not reveal the myriad of molecular rearrangements and atomic-level forces that make up the actual pathway and determine the efficiency of association.

The nucleic acid association process that we know the most about is duplex formation (Fig. 5; see Bloomfield et al. 2000 and references therein). Early work on simple, short DNA oligos showed that duplex formation was much slower than diffusion and revealed a large temperature dependence (Ross and Sturtevant 1960; Wetmur and Davidson 1968), something not observed for diffusional processes (Berg and von Hippel 1985), and similar rates and temperature dependences are observed for RNA duplexes (Pörschke and Eigen 1971; Pörschke et al. 1973; Nelson and Tinoco 1982). The observed enthalpies of activation (AH‡) derived from these temperature dependences match enthalpies for the formation of 2–3 base pairs, leading to a zipper model involving nucleation via the formation of 2–3 bp, after which subsequent base-pair formation was favored over the dissociation of the 2–3 bp nucleation complex (Fig. 5; Pörschke 1977). More recent optical trap experiments with DNA hairpins provided strong support for this model (Woodside et al. 2006).

While the above results provide a compelling model, they do not provide the detailed timing and extent of the individual conformational transitions that together constitute the binding process and, correspondingly, they do not explain “why” the process has its maximal barrier subsequent to the formation of two base pairs. Indeed, even duplex formation is extraordinarily complex, involving many bond rotations, changes in solvation, and the breaking and making of multiple hydrogen bonds, van der Waals, and stacking interactions that cannot be captured by these approximate models (Chakraborty and Di Cera 2017).
interactions. As the steps involved in association represent basic RNA conformational transitions, learning more about association processes has the potential to deepen our fundamental understanding of RNA. More generally, the observation of slower association with structured RNAs than structured proteins provides a perspective for considering the question of the molecular features involved in RNA conformational dynamics and ligand association. Below we briefly consider molecular properties that differ between proteins and RNA and may contribute to their different association rates.

- **Limited conformational preorganization**: The limited number and diversity of RNA side chains and the greater number of backbone rotatable bonds, relative to protein side chains and its backbone, have been suggested to limit RNA packing and preorganization, and electrostatic repulsion from the uniformly charged RNA backbone may provide additional limitations (Narlikar and Herschlag 1997). Limited preorganization in the conformational substates matching that for the bound complex introduce the need for rearrangements prior to and/or accompanying binding and thus additional free energy barriers to binding. Indeed, it has been suggested that protein–RNA complexes recognize RNA or DNA by base-pairing faster than naked RNA because of preorganization of the bound RNA strand, although electrostatic attraction introduced with the protein partner could also or instead be responsible (Gorski et al. 2017 and references therein; see also below). Finally, we note that intrinsically disordered proteins (IDPs) can bind ligands faster than observed for structured RNAs (Shammas et al. 2016). This faster binding is consistent with significant kinetic barriers for RNA rearrangements to binding-competent conformations that are absent for the IDPs. Nevertheless, electrostatic and/or hydrophobic IDP–ligand interactions could also or instead increase residence times and thus association rates, as described above for k_{on} specificity, underscoring the complexity of molecular association events and the need for careful and systematic investigation.

- **Hydrophobic interactions**: Hydrophobic interactions, which occur in many protein–ligand binding processes (Fersht 1999), are less directional than hydrogen bonds and less constrained than stacking interactions and thus may allow facile exploration in early association complexes to “find” the additional binding interactions needed to favor formation of the final bound state. Hydrophobic groups may also shed solvating water more readily than polar and charged groups to allow native contacts to form. Nevertheless, protein ligands range in their hydrophobicity, polarity, and charge, so this strategy is unlikely to account for all fast protein associations; different association events are likely to differ substantially.

- **Hydrogen bonds**: The ability of RNA bases to form two or three hydrogen bonds with interaction partners would seem to be an advantage for association by increasing the residence time of precursor complex states. Perhaps these hydrogen bonds are weaker and short-lived, relative to at least some protein–ligand hydrogen bonds. Alternatively, hydrogen bonds (and other interactions) stabilizing the alternative state of the RNA may also be more stable or more numerous and thus slower to resolve, thereby leading to more dissociation of encounter complexes prior to formation of a stably bound state.

- **Solvation and electrostatics**: Macromolecular motions require or are coupled to solvent motions, and the charged nature of RNA may result in greater coupling and more restricted motion. It is also possible that reorganization of the ion atmosphere that surrounds RNA as a polyelectrolyte
(Lipfert et al. 2014) must also readjust to facilitate motions and engenders an additional impediment to rearrangement.

The myriad interactions and possibilities outlined above highlight that we have a long way to go to understand association involving both RNA and proteins. We hope that comparing and contrasting these different biomolecules has helped illuminate underlying factors; nevertheless, the factors are too numerous and varied to expect mechanistic clarity from these rudimentary comparisons alone. Understanding RNA and protein associations at the atomic level will require an integration of information-rich experiments that provide detailed information at high time resolution and, given the large number of interactions and degrees of freedom in these systems, tight feedback between these experiments and predictions from atomic-level molecular dynamics models. Finally, we have not proven that RNA cannot bind ligands faster; it will be fascinating to explore this question through SELEX-type experiments that increase the selective pressure for fast and not just strong binding.

Closing remarks
Nature is powered by but also limited by the physical and chemical properties of its constituent materials and by the laws of physics. The slow association observed for ligands binding to structured RNAs may reflect basic physical properties of RNA and may have impacted the course of evolution and the molecules and molecular behaviors seen in modern-day biology. We hope that the observations and initial discussion presented herein will stimulate discussions, models, and new experiments that deepen our understanding across the interface of physics and biology.

MATERIALS AND METHODS
Compiling kinetic data from the literature
Association rate constants were assembled from peer-reviewed literature through an extensive internet search using both Google Scholar and the PubMed database. Search terms included "association rate constant," "on rate," "kon," "binding rate," "ligand binding kinetics," "association kinetics," "fast association rate," and "slow association rate." For nucleic acids, the terms "duplex formation," "duplex association," and "base-pair association" were also used. Papers identified through this search were manually scanned for relevant information including the association rate constant and the identity of the binding partners and their classification (RNA, DNA, protein, small molecule). For RNA binding processes, we also compiled the method used for determining the association rate constant (e.g., surface plasmon resonance, NMR, etc.).

Experimental conditions (buffer, temperature, etc.), and, when available, the dissociation rate constant ($k_{off}$) and equilibrium dissociation constant ($K_d$). This information was used to create Supplemental Tables S1–S5.

Our aim was to create a database that was representative of the repertoire of macromolecular association rate constants. To avoid overrepresentation, for papers that contained multiple mutants of a single protein, typically only the wild-type association rate constant was entered. A notable exception comes from papers where the mutations were specifically aimed at altering binding kinetics; in these cases typically the extrema and wild-type values were included (e.g., work on barnase/barstar [Schreiber and Fersht 1996]). Similarly, for binding experiments where a large number of binding conditions were explored, association rate constants for every condition may not be included in the table; however, the conditions for the listed association rate constant are indicated.

Potential sources of bias
Whereas the protein data set contains different types of proteins from various organisms and biochemical pathways (Supplemental Table S3), the RNA data set (Supplemental Tables S1, S2) is less diverse, having a few predominant RNA types. Further, many of these RNAs have been artificially truncated to facilitate handling and biochemical analyses and these alterations have the potential to affect activity (e.g., Uhlenbeck 2003) and $k_{on}$ (e.g., Soulière et al. 2013).

Second, while in principle molecular constants are independent of the technique used to obtain them, there are on occasion technique-specific effects. For example, surface plasmon resonance (SPR) involves association at a surface, so surface effects and accessibility can affect the rate of binding (e.g., Myszka 1997). Also, care is needed in applying this and other techniques and in analyzing results to prevent errors in measurement and interpretation (e.g., Fried 1989; Hellman and Fried 2007; Suddala and Walter 2014; Vander Meulen et al. 2016). For the RNA data presented, there were no systematic differences in association rate constants for SPR versus the other methods (not shown).

Finally, binding interactions with a strong electrostatic component will be highly sensitive to salt concentration. For example, the $k_{on}$ value for the theophylline RNA aptamer increases by two orders of magnitude from 0 to 10 mM MgCl$_2$ (Latham et al. 2009). Similarly, the Barnase–Barstar association rate constant increases by four orders of magnitude from 5 mM to 2 M NaCl (Schreiber and Fersht 1996). These studies underscore the importance of varying reaction conditions to evaluate the sensitivity of $k_{on}$ toward salt concentration, and the ability to learn about this component of binding through such studies.

SUPPLEMENTAL MATERIAL
Supplemental material is available for this article.

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