

- abundance (3). We normalized the AD values to "frequency" values by referring the AD values on each chip to a calibration curve constructed from the AD values for the 11 control transcripts with known abundances that were spiked into each hybridization (9). This "frequency normalization" allowed comparison of transcript measurements across multiple array experiments. Frequency values for each gene were expressed in number concentrations (transcripts per million, or ppm), under the assumptions described (9).
13. P. Tamayo *et al.*, *Proc. Natl. Acad. Sci. U.S.A.* **96**, 2907 (1999).
  14. Supplemental information is available at [www.sciencemag.org/feature/data/1053496.shl](http://www.sciencemag.org/feature/data/1053496.shl).
  15. I. L. Johnstone, J. D. Barry, *EMBO J.* **15**, 3633 (1996).
  16. M. C. Costanzo *et al.*, *Nucleic Acids Res.* **28**, 73 (2000). (The Proteome WormPD database is available at <http://www.proteome.com/databases/index.html>).
  17. C. E. Rocheleau *et al.*, *Cell* **90**, 707 (1997).
  18. P. W. Carter, J. M. Roos, K. J. Kemphues, *Mol. Gen. Genet.* **221**, 72 (1990).
  19. Additional details are available as supplemental information (14).
  20. J. H. Xin, B. P. Brandhorst, R. J. Britten, E. H. Davidson, *Dev. Biol.* **89**, 527 (1982).
  21. S. A. Chervitz *et al.*, *Science* **282**, 2022 (1998).
  22. G. M. Rubin *et al.*, *Science* **287**, 2204 (2000).
  23. To classify worm genes, we used the sequence comparison results of Rubin *et al.* (22). In the period between the date of our chip design (early 1999) and the date of this more recent sequence comparison (early 2000), some of the worm ORFs that had been included in the chip designs had been deleted or altered by worm sequence curators (for details on the curation of the Wormpep database, see [http://www.sanger.ac.uk/Projects/C\\_elegans/wormpep/](http://www.sanger.ac.uk/Projects/C_elegans/wormpep/)). To reconcile worm ORFs in the sequence comparison with worm ORFs on the arrays, we limited our analysis to worm ORFs on the arrays that were unchanged between the two data sets, using curation lists maintained at the Sanger Centre (the Wormpep.history file). Thus, we identified on the arrays 2905 worm genes shared between yeast, worm, and fly (core genes); 3150 worm genes shared between worm and fly, but not yeast (animal genes); and 8741 worm genes that were unique to the worm (worm genes).

Shared worm genes were defined as those that had significant homology to one or more genes in yeast or fly (BLASTP expectation values  $E < 1 \times 10^{-10}$ , as described by Rubin *et al.*) Discrepancies between these numbers of ORFs and comparable counts in Rubin *et al.* are accounted for by the facts (i) that 2% of worm ORFs were not monitored directly by the arrays and (ii) that, to be conservative, we excluded from consideration ORFs that had either been deleted or altered in the Wormpep database in the interval between the chip design and the sequence comparison of Rubin *et al.*

24. We thank M. Whitley for advice on array experiments; K. Griffiths for array bioinformatics support; E. Wilson, A. Velasco, and H. Horton for technical assistance; J. Freeman for bioinformatics support during the chip design process; G. Sherlock for sharing yeast and worm sequence comparison data; M. Yandell for sharing yeast, fly, and worm sequence comparison data; and D. Mootz for providing purified oocytes.

26 June 2000; accepted 29 September 2000

## Song Replay During Sleep and Computational Rules for Sensorimotor Vocal Learning

Amish S. Dave and Daniel Margoliash\*

Songbirds learn a correspondence between vocal-motor output and auditory feedback during development. For neurons in a motor cortex analog of adult zebra finches, we show that the timing and structure of activity elicited by the playback of song during sleep matches activity during daytime singing. The motor activity leads syllables, and the matching sensory response depends on a sequence of typically up to three of the preceding syllables. Thus, sensorimotor correspondence is reflected in temporally precise activity patterns of single neurons that use long sensory memories to predict syllable sequences. Additionally, "spontaneous" activity of these neurons during sleep matches their sensorimotor activity, a form of song "replay." These data suggest a model whereby sensorimotor correspondences are stored during singing but do not modify behavior, and off-line comparison (e.g., during sleep) of rehearsed motor output and predicted sensory feedback is used to adaptively shape motor output.

In reinforcement learning, systems learn through interaction with the environment by trying to optimize some measure of performance. Biological systems may experience a substantial delay between premotor activity and assessment of performance through sensory feedback (1). This delay poses the problem of how to reward or punish a premotor circuit when that circuit is participating in a different task by the time the reward or punishment is computed. Reinforcement learning is further complicated in systems such as vocal learning, where the mapping of sensory feedback (fundamentally represented as frequency versus time) onto motor output (mus-

cle dynamics) is of high dimensionality (a many-to-many dynamic mapping). Methods developed in the field of machine learning solve the problem of reinforcement learning with delayed reward (2), and a variety of biological solutions have been proposed to the problem of learning sequences of actions (3). Here, we report on neuronal data that represent a solution to the problem of sensorimotor mapping in the bird vocal-motor ("song") system. The physiological properties observed during sleep also suggest an algorithmic implementation for reinforcement learning of song.

Zebra finch songs are organized hierarchically, with one or more notes composing a syllable, and sequences of syllables forming a motif, which are repeated to form song. We investigated neurons in the forebrain nucleus robustus archistriatalis (RA), whose descending projections represent the output of the

forebrain song system. During singing, RA neurons exhibit short bursts of activity, whose identity varies with the note that immediately follows the burst (4). In awake birds, outside the context of vocalizations, RA neurons are regularly firing. RA neurons also prominently burst "spontaneously" and respond to sounds, but only during sleep (5). With the goal of comparing motor, auditory, and ongoing bursting activity, we recorded single neurons in the RA of singing male zebra finches, permitted the animals to fall asleep by turning off the lights, and then tested the same neurons' sensory and ongoing discharge properties (6, 7).

The spiking patterns of RA neurons in singing birds consisted of phasic patterns of premotor excitation superimposed over a background of profound inhibition (4) (Fig. 1, B and C). This premotor activity was virtually invariant for multiple occurrences of the same sound. After the lights were turned off, RA auditory responses were initially weak but gained strength with time, reflecting the gradual transition into sleep (5). Responses to playback of the bird's own song (BOS) also consisted of phasic patterns of excitation separated by inhibition that were similar for multiple occurrences of the same sound, differing mainly in the strength of response rather than pattern (8).

The timing of auditory responses to the BOS was very well aligned to the timing of premotor activity (Fig. 1F). The only exceptions were instances of silence following the end of a motif or the end of song, where the auditory response could include an additional burst that corresponded with the syllable that would have followed if the song had continued without pause. To compare motor and auditory activity, we analyzed the singing-related activity surrounding each syllable of song (4, 9). The spike patterns from the response to the BOS playback were then compared with the spike patterns from premotor activity derived from

Department of Organismal Biology and Anatomy, University of Chicago, 1027 East 57 Street, Chicago, IL 60637, USA.

\*To whom correspondence should be addressed. E-mail: [dan@bigbird.uchicago.edu](mailto:dan@bigbird.uchicago.edu)



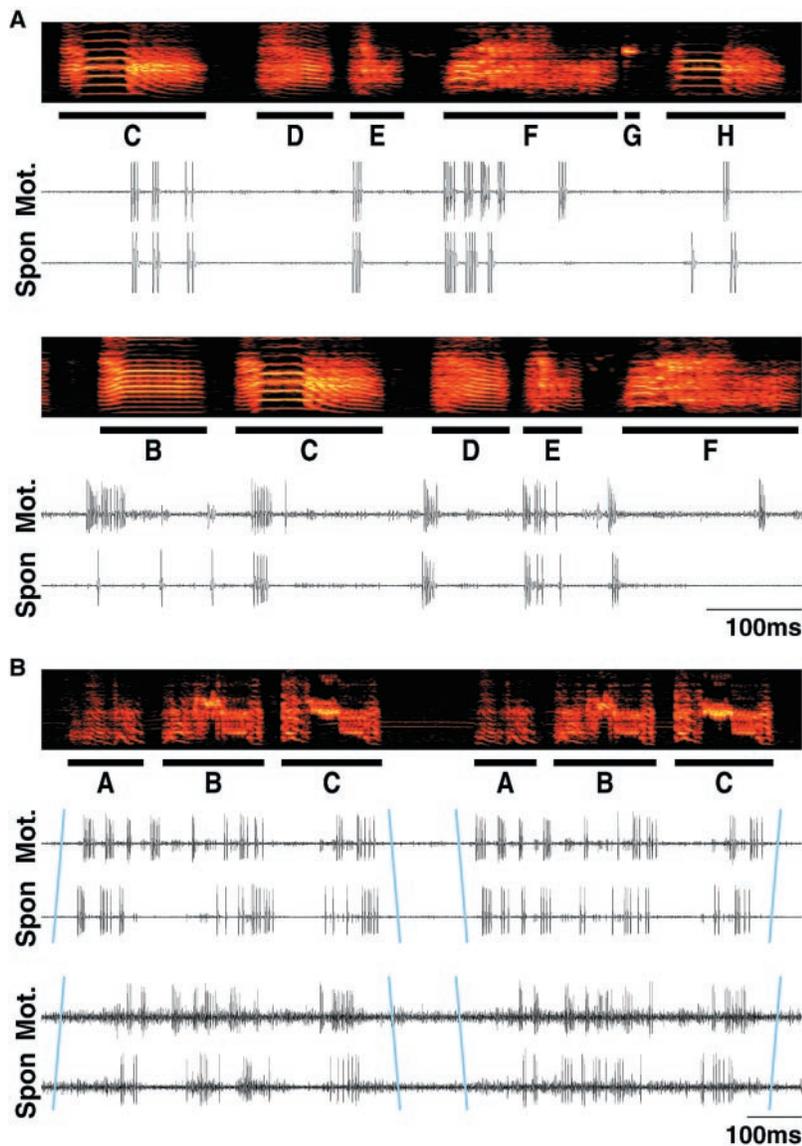
## REPORTS

the corresponding syllables, showing that the timing of excitation and inhibition during auditory stimulation was well aligned to such timing during singing (Fig. 1, D and E). A cross-correlation procedure revealed a strong, significant ( $P < 0.02$ ) correlation (10) between premotor and sensory spike patterns in all 17 neurons (from three birds) (mean normalized peak correlation =  $0.49 \pm 0.13$  SD). Thus, sensorimotor transformations in the song system result in a correspondence between temporally precise sensory and motor activity observed at the level of individual cells.

The auditory activity was only slightly delayed in relation to motor activity (by  $8 \pm 2$  ms; range, 4 to 13 ms). Because premotor activity in RA can lead the onset of syllables by up to  $\sim 40$  ms (4), this was surprising and suggested that the sensory patterns representing subsequent syllables were generated by responses to previous syllables. To characterize the extent of temporal integration in the auditory responses, we presented stimuli in which a syllable chosen at random from the final motif of the BOS was substituted by a background of equal duration, and we as-

essed the effect on the neuronal activity during the same or subsequent syllables (11). The deletion of a syllable substantially reduced the neuronal activity occurring one to three syllables later (Fig. 2A), up to  $\sim 250$  ms (8). This property was ubiquitous for all RA neurons that were auditory (14 neurons from three birds) (12) (Fig. 2B). These response properties are suggestive of temporal combination sensitivity, in which a sensory neuron's response is nonlinearly dependent on the temporal sequence of preceding syllables. Such responses have also been described for neurons in the nucleus HVC, which projects to the RA (13). Thus, in the RA as well as in the HVC (4, 13), the integration time of individual neurons appears to be considerably greater when in the sensory (auditory) state than during singing. Given the alignment of auditory and motor activity in the RA, one way of interpreting these results is that auditory responses to song syllables represent a prediction of subsequent premotor activity.

We also searched for similarities between ongoing bursting activity during sleep (5) and the sensorimotor patterns of RA neurons. For each cell, a visual inspection of samples of activity from long stretches (15 to 60 min) of undisturbed sleep identified repeated examples of one or more complex burst patterns, suggestive of the patterns that we had observed in the cell's premotor activity. To quantify this match, we developed a procedure to automate burst detection (14), considering only bursts of eight spikes or more to ease the computational burden and to allow for statistical analysis. By this procedure,  $7.1 \pm 5.3\%$  of all spikes (14 neurons from three birds) occurred in bursts, an average of  $175.4 \pm 144.6$  (range, 38 to 581) bursts per cell. For each cell, a measure of similarity between each burst and the single longest bout of the cell's premotor activity (4 to 8 s, consisting of several motifs or songs) was computed and tested for significance (15, 16). The results showed that  $15.3 \pm 6.5\%$  of bursts (range, 2.6 to 26.8%) significantly matched premotor activity. Only the cell with 2.6% matching bursts failed to exceed the 5% level expected by chance (16). Examples of matches between longer sequences of complex ongoing bursts and premotor activity were particularly compelling (Fig. 3A). In an exceptional case when two RA neurons were recorded simultaneously from different electrodes during sleep, both neurons commonly exhibited simultaneous bursting, with the different burst patterns for each neuron corresponding to the same sequences of syllables (Fig. 3B). This suggests that populations of RA neurons burst in a coordinated fashion during sleep. Bursts (and matching bursts) preferentially occurred during periods when the rate of ongoing discharge was lower and more variable (Fig. 4). Such modulation may



**Fig. 3.** Neuronal replay during undisturbed sleep. (A) Raw traces of neuronal activity (900 ms) during sleep ("Spon") in two different neurons for one bird. For each sample, a representative corresponding sample of premotor activity ("Mot.") and a color spectrograph of the song that the bird sang are shown. (B) Raw traces (1400 ms) of simultaneous recordings from two neurons ( $\sim 400$   $\mu\text{m}$  apart) in another bird. (The second neuron's activity is visible in the background of the first neuron's signal, an artifact of the pairing of signals used to achieve differential recordings resistant to movement-induced artifacts.) Both neurons simultaneously burst during sleep, with complex burst structures that match premotor activity. Apparent temporal expansion (first motif: A, B, and C) and compression (second motif: A, B, and C) is highlighted by the blue lines. This phenomenon has also been reported in population activity of hippocampal neurons (23).

correspond to specific phases of the sleep cycle.

In the sensorimotor phase of vocal learning, the mapping between auditory feedback and vocal output is the fundamental computational problem to be solved (17). A solution to this problem is reflected in the sensorimotor activity patterns of RA neurons. Precision of spike timing has been observed in a number of systems and provides evidence for temporally based neural codes in sensory processing, although only in a few cases has the behavioral relevance been directly demonstrated (18). The observed correspondence between auditory activity and vocal output demonstrates that, in the RA, sensorimotor mapping is based on a temporal code. This correspondence is likely to arise from auditory input recruiting similar components of the RA pattern-generating circuits as those recruited during singing. In the hierarchical organization of the song system (4, 19), the sensorimotor correspondence may first emerge at the single-cell level within the RA. The data suggest that, during vocal development, the song system learns to generate premotor commands by association with a prediction of future commands based on the timing of auditory feedback from preceding syllables. This can be interpreted as learning the match between the auditory response to a sequence of syllables with the premotor pattern for a subsequent syllable or as learning the match between the prediction of a sensory representation of a syllable with the premotor representation of the same syllable.

In the birdsong system, RA receives input from the HVC and from an anterior forebrain pathway (AFP) (Fig. 1A). Sensorimotor song learning could result in part from “online”

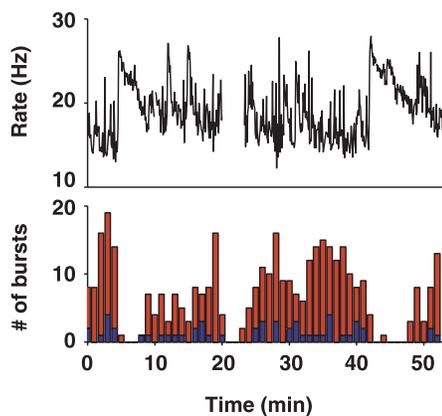
mechanisms, whereby during singing, HVC activity in response to auditory feedback from sequences of syllables is delayed through the AFP to produce a prediction of activity in the RA of a subsequent syllable. The data collected during sleep, however, also suggest “off-line” models for learning that address the problems of feedback delay and sequence generation. Such models share some similarities with temporal-difference models of reinforcement learning and sequence generation that are prominent in mammalian work on basal ganglia and the cerebellum, in that they reward or modify the system on the basis of its overall performance, not on the basis of the performance of individual components or movements (3). The AFP has been likened to a mammalian corticobasal ganglia-thalamocortical loop (20, 21).

In the vocal learning model motivated by the present data, signals that arise in the RA during singing train the AFP to generate a prediction of auditory feedback; during sleep rehearsal, the AFP’s predicted feedback provides reinforcement to RA neurons. During singing, sensorimotor efference copy signals (premotor output and expected auditory feedback) traverse the AFP, and via the lateral subdivision of the magnocellular nucleus of the anterior neostriatum (IMAN) projection onto area X, are compared with real auditory feedback arriving in area X from the HVC (Fig. 1A). Efference copy is brought into temporal register with auditory feedback using the long (~50 ms) synaptic delays observed in the medial subdivision of the dorsolateral nucleus of the thalamus (DLM) (21). This stimulates area X neurons that are sensitive to temporally coincident input. The output of the IMAN onto the RA has a reduced effect because the IMAN is not in temporal register with driving input from the HVC. During sleep, replay of song premotor patterns via ongoing bursting generates coherent activity throughout the song system that is similar to singing in the absence of actual sound production and perception. The output of the IMAN represents a prediction of the real auditory feedback that would have resulted from the burst-generated motor command, is in near coincidence with HVC bursts driving the RA, and is used to modify RA neurons that are sensitive to temporally coincident input.

The proposed algorithm for birdsong learning depends on circadian modulation of neuronal activity patterns (22). Our observation of neuronal replay of sensorimotor patterns during sleep is consistent with data from hippocampal studies suggesting that sleep is important for the consolidation of neuronal temporal codes for spatial memory (23, 24). The fundamental prediction of our model is that birdsong learning depends on sleep or other off-line computations.

References and Notes

1. In songbirds, there is a minimum delay of ~70 ms or longer between the production of a burst of neuronal activity contributing to song generation and the reception of processed auditory input from the resulting vocalization. In zebra finches, syllables are typically 50 to 200 ms in duration and may comprise one or several notes. The minimal delay is the sum of premotor lead (~50 ms) and sensory lag. Sensory lag is the sum of the minimal latency for response (~20 ms) and the sensory integration time, which, for song system neurons, can be tens to hundreds of milliseconds.
2. R. S. Sutton, A. G. Barto, *Reinforcement Learning: An Introduction* (MIT Press, London, 1998).
3. R. E. Suri, W. Schultz, *Exp. Brain Res.* **121**, 350 (1998); D. G. Beiser, J. C. Houk, *J. Neurophysiol.* **79**, 3168 (1998); G. S. Berns, T. J. Sejnowski, *J. Cogn. Neurosci.* **10**, 108 (1998); J. Brown, D. Bullock, S. Grossberg, *J. Neurosci.* **19**, 10502 (1999).
4. A. C. Yu, D. Margoliash, *Science* **273**, 1871 (1996).
5. A. S. Dave, A. C. Yu, D. Margoliash, *Science* **282**, 2250 (1998)
6. All neuronal data reported in this report are from single units. Eighteen neurons (10, 5, and 3) from three birds contributed to the data reported herein. Two additional neurons from a fourth bird exhibited a match between auditory responses and ongoing bursts. These data were the genesis for this study but did not include premotor data.
7. The chronic recording procedures and conduct of the experiments are described in detail on *Science Online* (24).
8. Additional procedures and data regarding the auditory responses of the neurons are described on *Science Online* (24).
9. The pattern of RA neuronal activity depends on notes, which are constituents of syllables. By definition, the same pattern of notes is repeated for a given syllable type. We analyzed RA activity at the syllable level to ease the analysis and to minimize the number of deletion stimuli required in subsequent experiments. Following established procedures, we identified each syllable that the bird sang, and the times of onset and offset were manually determined from spectrographs and oscillographs of the acoustic signal. The spectrographs of all exemplars for each syllable type were cross-correlated to establish optimal time shifts; these were then applied to the corresponding spike bursts. The aligned time-shifted spikes were the basis for further analysis.
10. Peak cross-correlation values were tested for significance by using a bootstrap procedure, described in detail on *Science Online* (24).
11. Multiple stimuli were derived from the BOS, one for each syllable deleted from the last complete motif of the BOS; these were presented in one block during the night while the bird was asleep. Deleted syllables were replaced with samples from silent intervals between motifs or syllables. Spike rates were computed over the duration of each syllable. For each syllable from the second to the penultimate, we compared the spike rates over the interval of the target syllable when a previous syllable or the target syllable itself was deleted (experimental data) against the spike rates for the target syllable when only a subsequent syllable was deleted (control data), using a Mann-Whitney *U* test at the 95% confidence level. The last syllable in the song was excluded because, by definition, control data were not available. (The response to the BOS, presented at the beginning of the night’s recordings as the bird transitioned into sleep, showed fluctuations in the strength of response but not in the timing of spikes. Thus, it was appropriate for correlation testing of sensorimotor comparisons but not for a comparison with spike counts of deleted syllable stimuli.)
12. The three birds presented with deletion stimuli had songs with eight, five, and three syllables per motif. All 10 neurons tested in the eight-syllable bird (Fig. 2B) and all 3 neurons tested in the three-syllable bird showed a loss of excitatory response when the target or a previous syllable was deleted. One of two neurons tested in the five-syllable bird showed no response to any syllable of the BOS; the other



**Fig. 4. (Top)** The firing rate during recordings of RA ongoing activity over almost 1 hour of sleep, estimated from a 100-point moving average of the interspike intervals (there was a gap in the data collection of ~3.25 min). **(Bottom)** A histogram (30-s bins) of the number of bursts identified by a burst-finding procedure. The number of bursts that significantly matched the premotor activity is shown in blue.

- neuron responded at the second syllable, and the response was suppressed by deletion of the first or second syllable.
13. D. Margoliash, *J. Neurosci.* **3**, 1039 (1983); D. Margoliash, E. S. Fortune, *J. Neurosci.* **12**, 4309 (1992).
  14. We gathered sufficient data from 14 neurons (from three birds) to permit quantitative comparisons between ongoing discharge and singing. Interspike interval histograms of the ongoing discharge of RA neurons during sleep are bimodal; the longer interval peak is related to nonbursting activity. Thus, bursts were defined as continuous sequences of interspike intervals falling outside of the normal (nonbursting) distribution of intervals.
  15. Two tests of significance were devised. For each neuron, a distance was computed between each ongoing burst and the exemplar stretch of activity during singing. Each best match (lowest distance) was tested for significance by using a bootstrap procedure that assessed the probability of occurrence of the exact sequence of intervals observed in the burst. In the second test, for the one bird

- with the greatest number of recordings, we also compared each neuron's ongoing discharges during sleep with the premotor data from other neurons. For seven out of eight neurons, there were more matches with the neuron's own premotor data than with the premotor data from the other neurons. The same neuron failed to achieve significance in both tests.
16. The procedures for identifying bursts in ongoing activity, matching bursts to premotor activity, and assessing the significance of the matches are described in detail on *Science Online* (24).
  17. M. Konishi, *Z. Tierpsychol.* **22**, 770 (1965).
  18. C. E. Carr, W. Heiligenberg, G. J. Rose, *J. Neurosci.* **6**, 107 (1986); *J. Neurosci.* **6**, 1372 (1986); A. Moiseff, M. Konishi, *J. Neurosci.* **1**, 40 (1981); C. E. Carr, M. Konishi, *J. Neurosci.* **10**, 3227 (1990); G. Laurent, M. Wehr, H. Davidowitz, *J. Neurosci.* **16**, 3837 (1996); R. C. deCharms, M. M. Merzenich, *Nature* **381**, 610 (1996); A. K. Engel, P. R. Roelfsema, P. Fries, M. Brecht, W. Singer, *Cereb. Cortex* **7**, 571 (1997); Y. Prut *et al.*, *J. Neurophysiol.* **79**, 2857 (1998).

19. E. T. Vu, M. E. Mazurek, Y.-C. Kuo, *J. Neurosci.* **14**, 6924 (1994).
20. S. W. Bottjer, F. Johnson, *J. Neurobiol.* **33**, 602 (1997).
21. M. Luo, D. J. Perkel, *J. Neurosci.* **19**, 6700 (1999).
22. G. E. Hinton, P. Dayan, B. J. Frey, R. M. Neal, *Science* **268**, 1158 (1995).
23. G. Buzsáki, *Neuroscience* **31**, 551 (1989); M. A. Wilson, B. L. McNaughton, *Science* **265**, 676 (1994); W. E. Skaggs, B. L. McNaughton, *Science* **271**, 1870 (1996); Z. Nádasdy, H. Hirase, A. Czurkó, J. Csicsvari, G. Buzsáki, *J. Neurosci.* **19**, 9497 (1999).
24. Supplemental data are available at [www.sciencemag.org/feature/data/1051099.shl](http://www.sciencemag.org/feature/data/1051099.shl).
25. We thank T. Q. Gentner, J.-M. Ramirez, P. S. Ulinski, and especially two anonymous reviewers for valuable comments on the manuscript. This work was supported by grants from the NIH (MH59831 and MH60276) to D.M. and (MH11615) to A.S.D.

6 April 2000; accepted 8 September 2000

## Structure of Murine CTLA-4 and Its Role in Modulating T Cell Responsiveness

David A. Ostrov,<sup>1,2</sup> Wuxian Shi,<sup>2</sup> Jean-Claude D. Schwartz,<sup>1</sup> Steven C. Almo,<sup>2\*</sup> Stanley G. Nathenson<sup>1,3\*</sup>

The effective regulation of T cell responses is dependent on opposing signals transmitted through two related cell-surface receptors, CD28 and cytotoxic T lymphocyte-associated antigen 4 (CTLA-4). Dimerization of CTLA-4 is required for the formation of high-avidity complexes with B7 ligands and for transmission of signals that attenuate T cell activation. We determined the crystal structure of the extracellular portion of CTLA-4 to 2.0 angstrom resolution. CTLA-4 belongs to the immunoglobulin superfamily and displays a strand topology similar to V $\alpha$  domains, with an unusual mode of dimerization that places the B7 binding sites distal to the dimerization interface. This organization allows each CTLA-4 dimer to bind two bivalent B7 molecules and suggests that a periodic arrangement of these components within the immunological synapse may contribute to the regulation of T cell responsiveness.

T cell-dependent immune processes require cell-surface interactions that mediate the initiation, modulation, and the ultimate course of the response. The specificity of T cell recognition is determined by the engagement of the T cell receptor (TCR) on T cells with cognate peptide-major histocompatibility complex (MHC) complexes presented by antigen-presenting cells (APCs) (1, 2). Additional signals are required to sustain and enhance T cell activity, the most important of which is provided by the engagement of CD28 on T cells with its ligands B7-1 (CD80) and B7-2 (CD86) on APCs (3, 4). In contrast, the interaction of B7 isoforms with

CTLA-4, a CD28 homolog (31% identity), provides inhibitory signals required for down-regulation of the response (5).

Unlike CD28, which is expressed on resting T cells, CTLA-4 is not detected on the cell surface until 24 hours after activation, peaking at 36 to 48 hours after activation (6). In addition, CTLA-4 exhibits an affinity for the B7 isoforms that is 10 to 100 times that for CD28 (7). On the basis of these differences in expression patterns and affinities, it is likely that CTLA-4 directly competes with CD28 for binding B7 and also directs the assembly of inhibitory signaling complexes that lead to quiescence or anergy (8). Consistent with the inhibitory role of CTLA-4, mice deficient in CTLA-4 die as a consequence of unchecked polyclonal T cell expansion, which results in fatal lymphoproliferative disorders (9). Thus, the balance between the opposing signals elicited by CD28 and CTLA-4 is central to the regulation of T cell responsiveness and homeostasis (10).

Because of its dominant role in modulating T cell activity, CTLA-4 has received considerable attention as a therapeutic agent (11). The soluble CTLA-4-immunoglobulin (CTLA-4-Ig) fusion protein acts as an inhibitor of CD28-B7 costimulation and has specific inhibitory effects in animal models of autoimmunity, transplant rejection, asthma, and allergy (3, 12). The efficacy of CTLA-4-Ig treatment of human disease has been demonstrated in clinical trials on patients with psoriasis vulgaris (13). This approach may well extend to a variety of T cell-mediated diseases including autoimmune diabetes, rheumatoid arthritis, systemic lupus erythematosus, and graft-versus-host disease (13). In contrast to strategies that interfere with the CD28-B7 association, reagents that interfere with the CTLA-4-B7 interaction intensify specific T cell responses. For example, blocking antibodies directed against CTLA-4 enhance rejection of preestablished tumors and protect against secondary challenge in animal models of prostate cancer and colon carcinoma (14).

The structure of the soluble extracellular domain of murine CTLA-4 (15) revealed two independent copies of the CTLA-4 dimer in the asymmetric unit of the crystal (Fig. 1, A and B, and Table 1) (16). The CTLA-4 monomer is a two-layer  $\beta$ -sandwich that exhibits the chain topology found in the immunoglobulin variable domains (Fig. 1A) (17). The front and back sheets, composed of strands A'G'FCC' and ABEDC', respectively, are connected by two intersheet disulfide bonds. The disulfide bond between the B and F strands is a signature for the immunoglobulin fold; the disulfide bond joining strands C' and D is unique to the CD28/CTLA-4 family (Fig. 1A) (17, 18).

The nuclear magnetic resonance (NMR) structure of a monomeric form of human CTLA-4 (18) shows the same overall topology as the murine homolog, with a root mean square (rms) deviation of 2.4 Å between equivalent C $\alpha$  atoms. The most significant

<sup>1</sup>Department of Microbiology and Immunology, <sup>2</sup>Department of Biochemistry, <sup>3</sup>Department of Cell Biology, Albert Einstein College of Medicine, Bronx, NY 10461, USA.

\*To whom correspondence should be addressed. E-mail: [almo@aecom.yu.edu](mailto:almo@aecom.yu.edu) or [nathenso@aecom.yu.edu](mailto:nathenso@aecom.yu.edu)

## Supplementary Material

All birds were sexually mature males >120 days of age, originally obtained from a commercial breeder. Before experiments, birds were outfitted with a chronic recording apparatus [A. Dave, A. C. Yu, J. J. Gilpin, D. Margoliash, in *Methods for Simultaneous Neuronal Ensemble Recordings*, M. Nicolelis, Ed. (CRC Press, Boca Raton, FL, 1999), pp. 101-120]. Subsequently, birds were maintained on a 16-/8-hour light/dark schedule, and experiments began near the end of the day to facilitate the bird's falling asleep. Electrodes were advanced or retracted by manually turning a screw while the bird was restrained until RA single units were isolated. The bird was then released and induced to sing by presenting a female in an adjoining half-cage or by broadcasting female calls. The lights were then turned off, and the bird fell asleep. Infrared video monitoring of sleeping birds facilitated assessment of the bird's sleeping state (including lack of movements, closed eyes, and low respiratory rate). BOS playback commenced soon after the lights were turned off, as this seemed to decrease the movements of the birds in the initial minutes of darkness, reducing the risk of losing unit isolation. After a variable number of presentations of the BOS (30 to 220 repetitions, 12 s per repetition), syllable-deleted versions of the BOS were presented in randomized order. After auditory stimulation, up to 60 min of ongoing activity was recorded in silence and darkness. When possible, thereafter, either additional repetitions of auditory stimuli were presented or the bird was awakened and more vocal motor activity was obtained until the unit signal isolation was lost.

Auditory responses of RA neurons were analyzed on the basis of the following procedure. Interspike interval histograms of ongoing discharge of RA neurons during sleep are bimodal; the longer interval peak is related to nonbursting activity [A. S. Dave, A. C. Yu, D. Margoliash, *Science* **282**, 2250 (1998)]. We fit a Gaussian to the longer interval peak using a nonlinear least squares curve-fitting procedure (Matlab, The Math Works). During playback of the BOS, a unit was considered responsive to a syllable when the distribution of interspike intervals occurring between the start of that syllable and the start of the next syllable was different from the Gaussian interval distribution of the nonbursting ongoing activity (Mann-Whitney  $U$  test,  $P < 0.05$ ). In 86% (93 out of 108) of syllables, units were responsive to two or more occurrences in the song of the same syllable. In 46% (43 out of 93) of these cases, the strength of response (number of spikes) differed significantly (paired  $t$  test,  $P < 0.05$ ) across occurrences of the same syllable in song (average difference =  $21\% \pm 14\%$ ). Each of 18 units analyzed showed this effect.

We compared premotor and auditory response data for each of 17 units (from three birds) by constructing vectors (histograms

binned at 1 ms) from auditory responses to song playback and from premotor activity. Cross-correlations of the two vectors were calculated with lags ranging over  $\pm 100$  ms. The cross-correlation was normalized by the square root of the product of the zero-lag autocorrelation values of the two vectors. Peak values ranged from 0.23 to 0.66. The peak of the cross-correlogram was tested for significance by randomly shuffling the values of bins in the auditory response vector and then repeating the cross-correlation (200 times). The resultant distribution of peak correlations defined a normal distribution, which was used to derive the probability of obtaining the actual peak value. For one site, deletion stimuli (see report) but not the intact BOS were presented because cell isolation began unexpectedly during stimuli presentation with a simultaneous recording site on another electrode. Thus, instead of the BOS, the response to an auditory stimulus with the last syllable of the BOS replaced with silence was compared with the premotor activity of that unit.

We also compared premotor and ongoing discharges of the single units. Bursts occurring in ongoing activity during sleep were defined as continuous sequences of intervals that fell outside of the 99% confidence interval of the distribution of the longer interspike intervals. To match ongoing bursts against syllable-level activity recorded during singing, we removed long intervals from the beginnings and ends of bursts. We also determined the longest duration between syllable onsets within a motif and split any bursts at intervals greater than this duration (185 to 220 ms). One site was excluded because it did not burst sufficiently (rate of ongoing discharge of only 2.46 spikes/s); another site required lowering the confidence threshold for finding burst from 99 to 95% to obtain any bursts. Sufficient data (at least 15 min of sleep, some bursts, and sufficient premotor activity) were therefore available for 14 units (from three birds). For each unit, we computed distances (0.5-ms resolution) between each burst from ongoing activity and the unit's premotor data. The distance  $D$  between the two spike trains at a given shift was defined as the average interval from each spike in one train to the nearest spike in the other train

$$D = \frac{\sum_{i=1}^{n_i} \text{Min}(5.0, \text{Min}_{j=1}^{n_j} |t_i - t_j|) + \sum_{j=1}^{n_j} \text{Min}(5.0, \text{Min}_{i=1}^{n_i} |t_j - t_i|)}{n_i + n_j}$$

### Supplemental Figure 1.

where  $i$  and  $j$  index the two spike trains and  $n_i$  and  $n_j$  are the number of bursts in each train. A ceiling of 5 ms was set to increase the robustness of the algorithm to outlier spikes. That is, the duration between any spike and the nearest spike in the other train, at any given shift, was allowed to range from 0 to

5 ms, with any values greater than 5 ms being treated as exactly 5 ms.

Two measures of significance were developed to determine the significance of the best match (lowest distance). Using a bootstrap procedure, we shuffled the intervals within an ongoing burst, and the resulting burst was similarly matched against the premotor activity. The distribution of distances from 1000 shuffled bursts was approximately normal. Across all neurons and all bursts, skewness values ranged from -1.5 to 0.5, with a mean of  $-0.4 \pm 0.3$ . Similarly, kurtosis values ranged from -1.6 to 7.4, with a mean of  $3.1 \pm 0.6$ . Bursts were considered to have a significant match to the premotor activity if there was less than a 5% probability of obtaining as strong a correlation by chance. This procedure relied on precise matches between the ongoing and the premotor activity to find matches and could not detect ongoing activity that matched premotor activity only with higher order statistics. The data also show evidence of compression/expansion of the time axis of ongoing discharge during sleep as compared to premotor patterns (see text and Fig. 3B of report). Time "warping" is a nonlinear phenomenon that our matching procedure was not designed to detect; hence, our estimate of matching between bursts and premotor activity is a lower bound estimate.

For the bird with the largest number of recordings, we also evaluated the significance of matches by matching a unit's bursts from ongoing discharge to the premotor data belonging to different units recorded from the same bird. Data from eight units provided seven of these pairings for each unit. For seven units, the greatest percentage of bursts matched the unit's own motor activity, with the difference between this percentage and the mean of the percentages for the heteronymous mappings being  $8.2 \pm 4.9\%$ . For the other site, the unit's own premotor data provided the worst percentage of matching bursts (average difference of 11.7%). The premotor activity of this unit was not apparently unusual; however, the ongoing discharge of this unit had the fewest bursts (in only 1.7% of time in undisturbed sleep did this unit exhibit bursts), the highest firing rate, and was the most regular (lowest standard deviation of the interspike interval distribution).