

Commentary

Michael S. C. Thomas

Mark H. Johnson

Developmental^{Q1} Neurocognition
Laboratory, School of Psychology
Birkbeck College
University of London
Malet Street, Bloomsbury
London WC1E 7HX, UK
E-mail: m.thomas@bbk.ac.uk

The Computational Modeling of Sensitive Periods

INTRODUCTION

In the continuing debate on sensitive periods, Tyler (this issue) argues for a mechanistic explanation of sensitive periods in development, rather than simply deriving a relationship between plasticity and age. Armstrong et al. (2005) endorse convergent approaches to assess types of plasticity, including the use of behavioral evidence, neurophysiological evidence, functional magnetic resonance imaging, event related potentials, and an appeal to evolutionary perspectives. In this contribution, we propose that a computational level of analysis is a key component in understanding the mechanisms through which functional plasticity alters in the cognitive system. To support our case, we discuss three examples of specific computational models that exhibit reductions in plasticity, and show how these models relate to Johnson (2005) three proposals for how sensitive periods might end: *endogenous*, *self-terminating*, and *stabilization*. Typically, we will find that implemented computational models of sensitive periods demonstrate multiple influences at work when functional plasticity reduces. Further, we suggest that computational modeling will allow us to understand how different factors interact to result in a functional reduction of plasticity in different cases.

THE IMPORTANCE OF COMPUTATIONAL IMPLEMENTATION

Implementation serves to evaluate the assumptions contained within theoretical proposal. It may be as straight-

forward as demonstrating that, in a given cognitive domain, turning down a "learning rate" parameter in a model of development is sufficient to capture the behavioral data indicating a sensitive period. Models are a concrete way to ask, does the theory really work? However, more often multiple assumptions are contained within any theory, and models serve as an exploration and explication of how these factors may interact in driving the functional plasticity of a system. Further, models may generate novel, testable predictions for how plasticity can be increased or decreased in the system. Most importantly, implementation forces the modeler to make decisions about hidden assumptions within verbally specified theories.

Three issues are of particular relevance for sensitive periods in functional brain development: (1) what is the actual nature of the *representations* used to encode the problem domain? It turns out that both overlap between the representations generated by old and new experiences, and systematicity within problem domains can both be influential in determining functional plasticity. (2) What is the *frequency* with which the system encounters various experiences? It turns out that under some conditions, frequency can overcome changing conditions of internal plasticity. (3) What level of *processing resources* is available to the system? It turns out that under some conditions, changes in resources can be directly equivalent to changes in plasticity, particularly in parallel processing systems, and further that competition for limited resources can account for many instances of reduced functional plasticity.

Let us consider the last of these three points. Processing resources are of particular relevance where recovery from damage is used as a metric of plasticity. Evidence of "crowding effects" in children who have suffered brain damage indicates that capacity limitations can influence cognitive development (Anderson, Northam, Hendy, & Wrennall, 2001). A crowding effect describes the situation where after recovery, there is a generalized depression of neuropsychological functions rather than specific cognitive deficits, as if the remaining system has the computational

Received 31 January 2006; Accepted 31 January 2006

Correspondence to: M. S. C. Thomas

Contract grant sponsor: MRC Career Establishment

Contract grant number: G0300188

Published online in Wiley InterScience

(www.interscience.wiley.com). DOI 10.1002/dev.20134

© 2006 Wiley Periodicals, Inc.

properties but not the capacity to follow the normal course of development. It has been argued that children's ability to recover from brain damage depends, to some extent, on their premorbid level of processing resources, termed cerebral or cognitive "reserve" (Dennis, 2000; Stern, 2002). The greater the premorbid level of resources, the better the prospect for recovery. A focus on resources prompts the following conclusion: *one cannot interpret a developmental failure to recover from brain damage as a lower level of plasticity unless the domain(s) in question can definitely be acquired with the reduced level of resources*, were this reduced level to be present at the start of development. Thus, when de Schonen, Mancini, Camps, Maes, and Laurent (2005) observe in children with pre-, peri-, or postnatal brain damage a failure to later acquire face recognition expertise, the authors interpret this in terms of "poor postlesional face-processing plasticity" (p. 184); yet it may be that the remaining processing resources available to the child were simply insufficient to acquire the normal level of expertise whatever the level of plasticity.

Alternative explanations of this nature derive from the requirement to make decisions about resources when building a model. Implementation, for example, would force a modeler to make a decision on what is happening inside a learning system during a period of sensory deprivation. However, the fact that models of development employ analytically derived learning algorithms itself leads to new candidate explanations of changes in functional plasticity. Take the well-known example of Hebbian learning. Within the brain, Hebbian learning can be grossly characterized as cells that fire together, wire together. More specifically, the change in the connection strength between two neurons is held to be proportional to the product of their correlated activity. More formally,

$$\Delta w_{ij} = \varepsilon a_i a_j \quad (1)$$

where a_i is the activation of the sending unit and a_j is the activation of the receiving unit, w_{ij} is the connection strength between them, Δ is the change in strength, and ε is the "learning rate parameter" (see, e.g., O'Reilly & Munakata, 2000, equation 4.2). The learning rate parameter is employed when multiple associations are to be learnt in the same network. Its value is typically set at less than 1 to prevent wild oscillations between different connection strengths after each training experience and instead encourage the network to converge on a compromise value that will accommodate all associations. Clearly, the plasticity of a system using this algorithm can be manipulated just by altering the "learning rate parameter." But less obviously, increases in the activation of either the sending or receiving unit themselves increase plasticity. That is, under the terms of the Hebbian algorithm, simply *a more activated system will be more*

plastic one.¹ It is not clear whether this candidate mechanism for altering plasticity has relevance for brain development. Event-related potential studies of brain activity indicate that voltage potentials are of greater amplitude earlier in development (see, e.g., Nelson & Monk, 2001, Fig. 9.5), though other factors such as skull thickness and conductivity may partially explain this. Brain metabolism measured through PET shows a rising then falling profile across development, with a peak in mid-childhood, though synaptic density appears to peak around 1 year of age (Chugani, Phelps, & Mazziotta, 1987; Huttenlocher, 2002). In fMRI, the BOLD response in children and adolescents appears to be similar to that in adults in time course and peak amplitude (Casey, Davidson, & Rosen, 2002), although on individual tasks, brain activations in children have been found to be more widespread than in adults (e.g., Casey et al., 1997). The extent to which these neurophysiological measurements relate to the working computational learning algorithm in the brain, and their changes during development, may be a promising novel line of enquiry in developmental cognitive neuroscience.

We now turn to some examples of implemented models, where the impact of factors such as representational overlap, frequency, and resource level becomes apparent. First, let us recap Johnson (2005) three classes of explanation for the end of sensitive periods. These are that (a) the termination arises from endogenous factors controlled by maturation or an external environmental "trigger," (b) learning is self-terminating, in that the system drives itself into a representational state where it is no longer responsive, and (c) underlying plasticity does not actually reduce but the constraints on plasticity (such as environmental inputs) become stable. The following three examples all exhibit sensitive periods that come to an end, and each appeals to one of the above explanations. Note that all examples will use algorithms that contain a "learning rate" parameter but in all cases, that parameter is held constant throughout training.

EXAMPLE 1: CHICK IMPRINTING AND THE SELF-TERMINATING SENSITIVE PERIOD

O'Reilly and Johnson (1994) constructed a model of filial imprinting in the chick brain. When chicks are exposed to visual stimuli early in life, they can develop a strong preference for a given object. This imprinting can only be established in a specific period of life, is relatively unaffected by subsequent exposure to different objects, and is self-terminating in that the sensitive period is experience driven rather than based on strict chronologi-

¹See Mareschal and Bremner (2006) for an application of this idea to infant behavioral development.

cal age. O'Reilly and Johnson (1994) neurocomputational model was based on the known neuroanatomy of the chick forebrain and contained several features, including the development of translation invariance for objects presented on its retina. Here we will just concentrate on how its representations developed, simplifying the dynamics of the model somewhat. The model was self-organizing, in that it developed representations on an output layer based on exposure to patterns presented on an input layer. In the simulations capturing the closing of the sensitive period, the model was trained on Object A for 100 presentations. It was then trained on an entirely dissimilar Object D. After 150 presentations of D, the network switched its preference from A to D, where preference was assessed by the total activation on the output layer produced by each object. However, if the model was initially trained for only 25 presentations longer (125 presentations of Object A), its preference did not switch to D even after 900 presentations of Object D. Experience-dependent self-organization led to the closing of the sensitive period at 125 presentations of A. This provides an example of how self-termination of plasticity might work.

It is instructive to consider how this process worked in terms of underlying computations. Increased training on Object A led to further recruitment of units on the output layer to represent this input pattern. After 125 presentations of Object A, the majority of units on this layer were now representing Object A. Since Object D was dissimilar to A (their representations were nonoverlapping), it could only activate and, therefore, attempt to recruit different output units to those activated by A. That is, it could not impinge on the units already recruited by A due to the lack of similarity. As a result, however, much learning took place on D, there only remained a minority of the output units that could become selective for this stimulus. Given that the model's stimulus preference was driven by total activation engendered on the output layer, D could never become the preferred stimulus once A had recruited a majority of the output units. There were insufficient resources left to permit this (see O'Reilly & Johnson, 1994, p.374).

Therefore, although this is clearly an instantiation of a *self-terminating* sensitive period, it arises due to *competition for limited resources* and a *lack of representational overlap between new and old experiences* in this implementation.

EXAMPLE 2: NON-NATIVE PHONEME DISCRIMINATION AND THE SENSITIVE PERIOD ENDED BY STABILIZATION

Monolingual Japanese speakers have difficulty discriminating the English /r/ and /l/ sounds despite repeated

exposure to words containing them, consistent with reduced functional plasticity for the acquisition of nonnative phonemic contrasts in second language learners. However, if exaggerated versions of /r/ and /l/ phonemes are presented to monolingual Japanese speakers, they can learn to distinguish both these phonemes and subsequently normal exemplars of the /r/ and /l/ phonemes (McCandliss, Fiez, Protopapas, Conway, & McClelland, 2002). McClelland, Thomas, McCandliss, and Fiez (1999) constructed a neurocomputational model to explore how this reduction in plasticity might take place in monolingual speakers. The model used a self-organizing architecture, with an input layer on which the phonemes were presented and an output layer that had to develop the relevant categories. Two versions of the model were trained. A "Japanese" model learned a single category of phonemes in the /l/-/r/ region of input space and learned a single output category, while an "English" model was presented with two partially overlapping input categories standing for tokens of /l/ and /r/ and learned two output categories. In the transfer condition, "adult Japanese" networks with 300 epochs of training were exposed to the English-like environment with separate /l/ and /r/ tokens. None subsequently reorganized their output layer into two output categories. However, when "exaggerated" tokens of /l/ and /r/ were used for the two input categories, all "adult Japanese" networks learned to discriminate these stimuli within only a few epochs of their introduction into the training set and this discrimination then extended to the original exemplars.

Again, it is instructive to consider the exact representations used. Each phoneme was represented by a 3×3 square on a grid-like input layer. The single "Japanese" /l/-/r/ input was a 3×3 square in the center of the input layer. After training, a single output category came to represent this input pattern. The "English" /l/ and /r/ categories were represented by two 3×3 squares on the input layer that overlapped by one row. Their representations had three squares in common and six squares separate. In the "English" condition, the six nonoverlapping squares were sufficient to drive the development of two separate output categories. When the "Japanese" net was exposed to "English" input, the two "English" phoneme categories overlapped the single "Japanese" category by two rows each, that is, each shared six squares with the single "Japanese" category and differed by only three.

Consider a trained "Japanese" network with its one output category. It is now presented with the two novel "English" input categories. When either novel input is presented, the network receives activation from six squares that fall within its original input category and only three that fall outside. The final output state is the result of a competition, in which the six old inputs defeat

the three new: the novel input is assimilated to the original single category, and the network does not register that it has seen something new. In order for plastic change to occur, new units must win the competition on the output layer. The exaggerated tokens of the “English” /l/ and /r/ categories are created so that they only overlap with the single “Japanese” phoneme by three squares; six squares fall outside the old category. Now the network receives signals from six squares that the input is something new and only three that it is old. Different units win the competition to become activate on the output layer, and this causes reorganization into two output categories. These categories can then also be activated by the original /l/ and /r/ tokens, since these overlap their exaggerated versions by six squares.

In this model, then, the sensitive period of the self-organization ended because its input had stabilized. Although different tokens appeared in its environment corresponding to the shift to “English” input, the representational overlap between old and new experiences was so great that the learning system was essentially “blind” to the change. Only when the difference between exemplars was artificially increased was the latent plasticity of the system revealed and reorganization triggered. Here is an example of Johnson’s *stabilization* class of termination, but one that crucially depends on *representational overlap* for its implementation.

EXAMPLE 3: THE EMERGENCE OF SPECIALIZED FUNCTIONAL STRUCTURE AND THE SENSITIVE PERIOD ENDED BY ENDOGENOUS FACTORS

The preceding examples have focused on sensitive period effects in self-organizing systems. Research has also explored sensitive periods in associative systems that are required to learn input–output mappings. These have included research on sensitive periods for recovery from damage (Marchman, 1997) and age-of-acquisition effects (Ellis & Lambon Ralph, 2000; Lambon Ralph & Ehsan, [in press](#)^{Q2}), both in the domain of language acquisition. In this section, we briefly discuss some results from our own simulation work extending the findings of Marchman (1997).

Marchman (1997) employed the English past tense as a test domain to study acquisition, loss, and recovery in associative networks. The English past tense is of note because it is characterized by a predominant rule (e.g., talk-talked, drop-dropped, etc.) that extends to novel stems (e.g., wug-wugged), but also contains exception verbs (go-went, hit-hit, sing-sang). This aspect of grammar has been much studied because of the problems its dual regular/irregular structure presents for children

during language acquisition. It has even been proposed that different brain areas become specialized for the processing of regular and irregular verbs (see, e.g., Tyler, Marslen-Wilson, & Stamatakis, 2005). The English past tense is of interest here because it is possible to simulate the emergent specialization of regular and irregular verbs to different pathways in an associative network (Thomas & Karmiloff-Smith, 2002; Thomas & Richardson, 2006). The problem can, therefore, additionally serve as a test domain with which to explore sensitive periods in the emergence of specialized functional structure. This issue is important because plenty of evidence suggests that children suffering unilateral brain damage can reorganize their systems to achieve a functional structure sufficient to generate behavior in the normal range, while adults who suffer similar damage exhibit persisting deficits. Aphasia after left hemisphere damage is one example (see Bates & Roe, 2001). Such evidence implies a sensitive period for when functional structures can be reorganized after damage.

Our simulations used an associative network with two pathways, trained using the backpropagation algorithm. The architecture is shown in Figure 1. The input layer is connected to output layer either directly or via a layer of intermediate processing units. During training, the direct route is more suited to learning regular past tenses and the general rule, while the indirect route comes to specialize in exception mappings that require its additional computational power (see Thomas & Karmiloff-Smith, 2002, for details of this model). We assessed the functional plasticity of this system by measuring its recovery from damage at different points in training. In the normal condition, a network is trained for 500 epochs. A lesion occurring at 490 epochs would only, therefore, give the

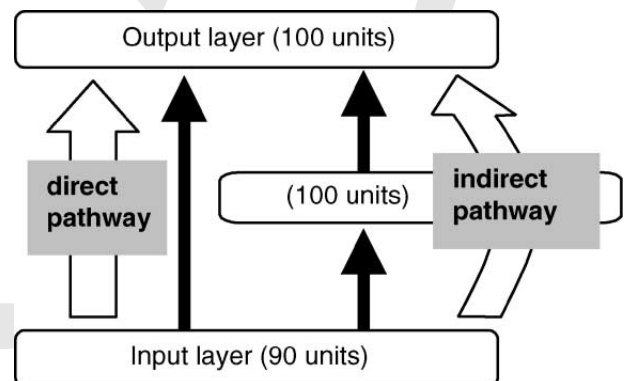


FIGURE 1 Architecture of the associative network trained on the English past tense problem. Rectangles represent layers of simple neuron-like processing units, and black arrows represent matrices of connections between layers. Verbs stems were coded on the input layer and past tenses on the output layer using phonological features.

network 10 epochs to recover. The confound of lesion age and recovery time can of course suggest poorer plasticity later in training. However, we can control for this artifact by extending training beyond the normal period, so that each network has 500 epochs to recover from damage irrespective of when the damage occurred. The network was damaged either prior to training, or after 10, 50, 100, 250, 400, 450, or 490 epochs of training by lesioning 75% of the connections in both pathways. Its ability to recover was then assessed.

Crucially, the network was also given an *endogenous reduction* in its plasticity. From 100 epochs onwards, any network connection below a given threshold had a small probability of being pruned away (i.e., set to zero for the remainder of training), implementing the idea that the network is initially over-resourced but then prunes away unnecessary connections (Huttenlocher, 2002). With fewer connections, the network's ability to learn is reduced. The 100-epoch onset presumed an endogenous trigger for pruning in the model.

Figure 2 shows the normal endstate performance (gray bars) for the regulars, rule generalization, and three types of exception verb (labeled EP1, EP2, EP3f). It also demonstrates the level of endstate recovery achieved following damage at different points during training. Performance levels are shown both for recovery at the completion of 500 epochs, where later lesions will have had shorter recovery times (white bars), and following a fixed recovery period of 500 epochs post lesion (black bars). Regulars and rules indicated little evidence of sensitive periods in this associative system, with similar

levels of recovery whenever the damage occurred. Regular patterns and rule generalization retained their functional plasticity because of the high type frequency and systematicity amongst regular past tenses in the training set (see Lambon Ralph & Ehsan, *in press*^{Q3}, and Seidenberg & Zevin, 2006, for discussions of the influence of systematicity and frequency on age-of-acquisition effects). Regulars are best positioned to use the remaining resources after damage. By contrast, all three types of exception pattern exhibited sensitive periods. In the case of EP3f exception patterns, the sensitive period declined in a roughly linear fashion. These verbs have arbitrary input–output mappings but high token frequency in the training set, and their high token frequency allows the best recovery of the exception patterns. For EP1 and EP2 exception patterns, the decline in recovery with age was steeper; perhaps one might call these “critical” rather than sensitive periods. Overall, the results show that *within the same architecture, sensitive and critical periods can appear in some parts of the problem domain but not others*, depending on the nature of the mapping problem and on frequency effects.

Figure 3 plots the proportion of connections remaining in one of the pathways of the network and depicts the gradual reduction through pruning as well as the sudden drop after lesion is applied at an early and a late point in training. Importantly, although pruning was an endogenous process, it was also influenced by activity-dependent changes in the network. When a lesion occurred early in training, the network was able to take advantage of the remaining resources and *fewer connections were pruned*.

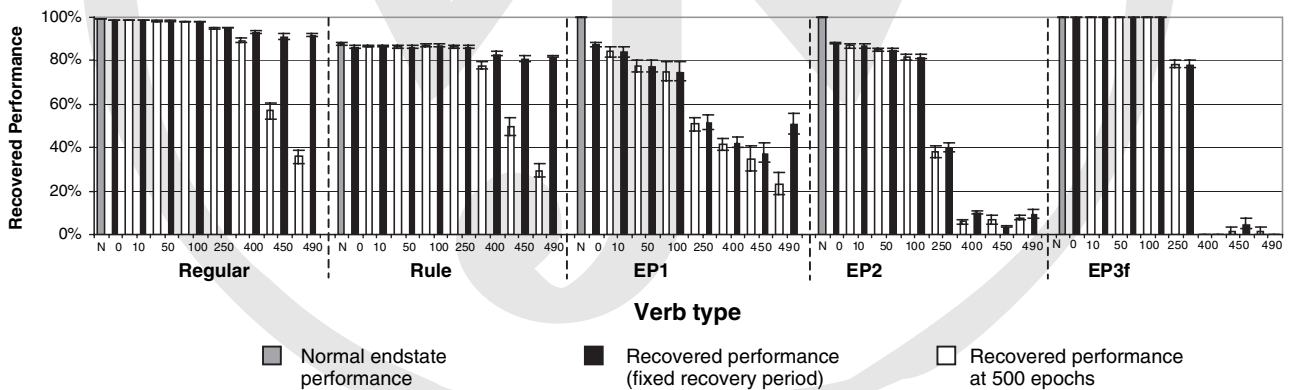


FIGURE 2 Performance of the network at the end of training (500 epochs) for five pattern classes within past tense: *Regular* (e.g., talk-talked), *Rule* (wug-wugged), *EP1* (hit-hit), *EP2* (sing-sang), and *EP3f* (go-went). The *EP* numbers mark increasing degrees of inconsistency with regular mappings and the *f* registers the high token frequency of this class. Gray bars show normal performance. Black bars show recovery after lesions at different points in training (0, 10, 50, 100, 250, 400, 450, and 490 epochs) with a fixed period of 500 epochs of training post-lesion. White bars show the recovered level of performance at the end of normal training (e.g., a lesion at 490 epochs will have only 10 epochs of training post lesion). Error bars depict standard errors over six replications with different initial random seeds.

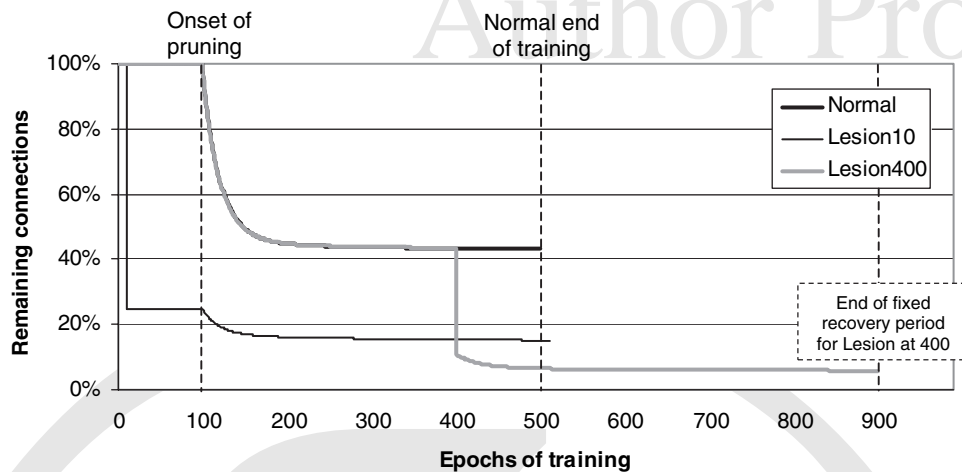


FIGURE 3 The proportion of connections remaining in the direct pathway with the combined effects of pruning (onset 100 epochs) and lesions, shown for the normal case, and for lesions after 10 epochs or 400 epochs of training. Similar functions were found for connections in the indirect pathway. (A connection was pruned with 5% probability each epoch if its absolute value was less than 0.5. Lesions probabilistically removed 75% of connections in both pathways).

It turned out that both resources and pruning were key in generating the sensitive periods observed in this model. When the model was trained with fewer resources (units) in the indirect pathway, sensitive periods appeared for all pattern classes. When the normal network was trained without pruning, none of the pattern classes exhibited sensitive periods.

Figure 4(a) shows the relative functional specialization of each pattern type to the direct (+ve) or indirect (−ve) pathways of the associative network. Figure 4(b) focuses on one specific contrast in emergent specialization, rule formation versus EP3f patterns, assessed across a fixed period of recovery after damage. In the normal condition, regular and rule generalization revealed partial specialization to the direct pathway, while exception patterns showed differing degrees of specialization to the indirect pathway. If both the routes of the network were damaged prior to training, this immediately changed how each pattern class used the two pathways. The indirect pathway was relied upon more heavily. However, as damage occurred later in training, this pattern progressively changed, with increasing reliance on the direct pathway to drive recovery (even when recovery time was controlled). Two points are of note: first, we see here sensitive periods for the emergence of specialized functional structure, with different functional structures arising depending on the time of damage. Second, for the exception verbs, the sensitive period for functional structure corresponded with a sensitive period for behavior (i.e., the alternate functional structure was less able to support recovery); but for regular verbs and rule generalization, the sensitive period for functional struc-

ture had no corresponding sensitive period in behavior. *The sensitive periods for functional structure and for behavior could, therefore, dissociate.*

The explanation for these effects involves several factors. Broadly, the results depend first on how well different pattern types can exploit the resources remaining at different points in training, based on their frequency and similarity. Later lesions cause more reduction in resources because they come on top of losses through pruning. Early damage can retard the endogenous pruning process. Systematicity, high type frequency, and high token frequency, all advantage a pattern class in making use of remaining resources. Second, the two pathways have different plasticity at an algorithmic level. It takes more training to alter the two sets of connections arranged in series in the indirect route than it takes to alter the single set in the direct route. Third, later in training, connections in each pathway become larger and if these connections are not useful for driving behavior after damage, they take longer to reset (an effect called “entrenchment”). These three factors interact to determine which pattern classes will recover and how the two pathways will be used.

This simulation is useful because it can begin to explore the relationship (and possible mismatch) between sensitive periods in behavior and in the emergence of specialized functional structure, but once more, similarity, frequency, and resources mediated the effects. In terms of Johnson (2005) proposals for how sensitive periods end, this model implemented an *endogenous* process of pruning. Yet even these endogenous factors *interacted with activity-dependent processes* in fashioning the final shape of the sensitive periods in plasticity.

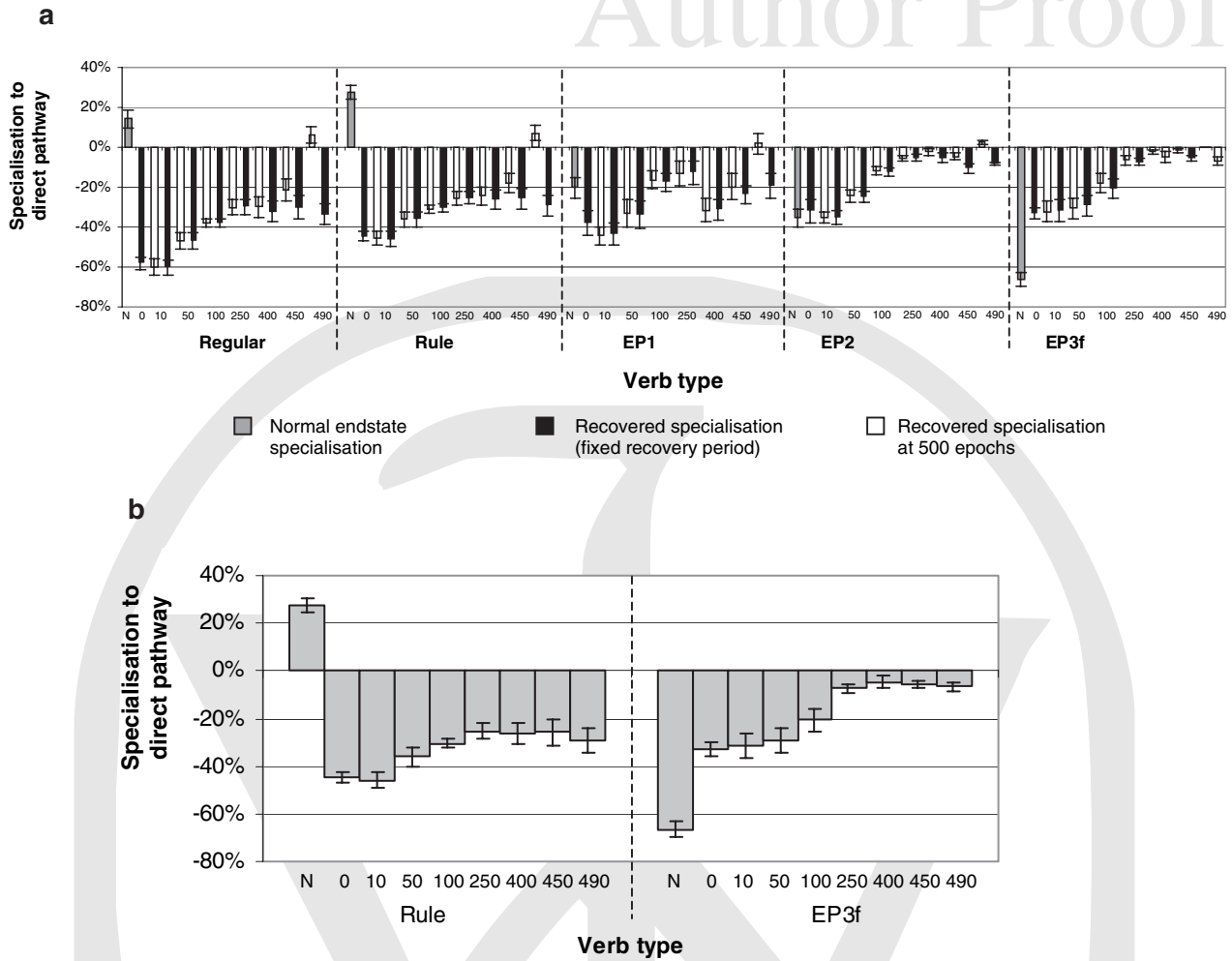


FIGURE 4 (a) Relative specialization of each pattern class to the direct (+ve) or indirect (–ve) pathway at the end of training, for the normal network and networks recovering from damage at different points in training. Specialization was assessed using the dissociation methodology of traditional cognitive neuropsychology. [If a pattern class is more specialized to the direct than indirect pathway, it should show a bigger deficit when the direct pathway experiences a further lesion than when the indirect pathway is similarly lesioned. The figure shows the difference in the size of the deficit for each pathway (see Thomas & Karmiloff-Smith, 2002, for details)]; (b) A single comparison drawn from the above data, depicting the relative specialization of *rule* versus *EP3f* patterns after a fixed recovery period following damage.

CONCLUSION

We began by endorsing the importance of specifying the underlying computational mechanisms of plasticity change in order to turn descriptions of sensitive periods into explanations, and by arguing for the utility of implemented neurocomputational models in this endeavor. Implementation forces clarity, reveals hidden assumptions, and generates new candidate explanations and testable hypotheses. In three examples, we illustrated implementations of Johnson (2005) proposals for how

sensitive periods might end. In each case, implementation demonstrated multiple additional factors at play that interacted with the closing of sensitive periods, including the similarity between representations, the frequency with which certain experiences occurred, and resource levels within the system. We believe that discovery of the full repertoire of mechanisms through which functional plasticity is modulated must rely on a program of computational modeling integrated within the multidisciplinary exploration of sensitive periods in development.

NOTES

This research was supported by MRC Career Establishment Grant G0300188 to Michael Thomas.

REFERENCES

- Anderson, V., Northam, E., Hendy, J., & Wrennall, J. (2001). *Developmental neuropsychology: A clinical approach*. East Sussex, UK: Psychology Press.
- Armstrong, V. L., Brunet, P. M., He, C., Nishimura, M., Poole, H. L., & Spector, F. J. (2005). What is so critical? A commentary on the re-examination of critical periods. *Developmental Psychobiology*
- Bates, E., & Roe, K. (2001). Language development in children with unilateral brain injury. In C. A. Nelson & M. Luciana (Eds.), *Handbook of developmental cognitive neuroscience* (p. 281–307). Cambridge, Mass: MIT Press.
- Casey, B. J., Davidson, M., & Rosen, B. (2002). Functional magnetic resonance imaging: Basic principles of application to developmental science. *Developmental Science*, 5, 301–309.
- Casey, B. J., Trainor, R. J., Orendi, J. L., Schubert, A. B., Nystrom, L. E., Giedd, J. N., Castellanos, F. X., Haxby, J. V., Noll, D. C., Cohen, J. D., Forman, S. D., Dahl, R. E., & Rapoport, J. L. (1997). A developmental functional MRI study of prefrontal activation during performance of a go-nogo task. *Journal of Cognitive Neuroscience*, 9, 835–847.
- Chugani, H. T., Phelps, M. E., & Mazziotta, J. C. (1987). Positron emission tomography study of human brain functional development. *Annals of Neurology*, 22, 487–497.
- Ellis, A. W., & Lambon Ralph, M. A. (2000). Age of acquisition effects in adult lexical processing reflect loss of plasticity in maturing systems: Insights from connectionist networks. *Journal of Experimental Psychology: Learning, Memory and Cognition*, 26, 1103–1123.
- de Schonen, S., Mancini, J., Camps, R., Maes, E., & Laurent, A. (2005). Early brain lesions and face-processing development. *Developmental Psychobiology*, 46, 184–208.
- Dennis, M. (2000). Childhood medical disorders and cognitive impairment: Biological risk, time, development, and reserve. In K. O. Yeates, M. D. Ris, & H. G. Taylor (Eds.), *Pediatric neuropsychology: Research theory and practice* (pp. 3–22). New York: Guilford.
- Huttenlocher, P. R. (2002). Morphometric study of human cerebral cortex development. In M. H. Johnson, Y. Munakata, & R. Gilmore (Eds.), *Brain development and cognition: A reader* (2nd edition) (pp. 117–128). Oxford: Blackwell.
- Johnson, M. H. (2005). Sensitive periods in functional brain development: Problems and prospects. *Developmental Psychobiology*, 46, 287–292.
- Lambon Ralph, M. A., & Ehsan, S. (*in press*^{O4}). Age of acquisition effects depend on the mapping between representations and the frequency of occurrence: Empirical and computational evidence. *Visual Cognition*
- Marchman, V. A. (1997). Constraints on plasticity in a connectionist model of English past tense. *Journal of Cognitive Neuroscience*, 5, 215–234.
- Mareschal, D., & Bremner, A. J. (2006). When do 4-month-olds remember the “what” and “where” of hidden objects? In Y. Munakata & M. Johnson (Eds.), *Attention and Performance XXI: Processes of Change in Brain and Cognitive Development*. Oxford: Oxford University Press.
- McCandliss, B. D., Fiez, J. A., Protopapas, A., Conway, M., & McClelland, J. L. (2002). Success and failure in teaching the [r]–[l] contrast to Japanese adults: Tests of a Hebbian model of plasticity and stabilization in spoken language perception. *Cognitive, Affective, and Behavioral Neuroscience*, 2, 89–108.
- McClelland, J. L., Thomas, A. G., McCandliss, B. D., & Fiez, J. A. (1999). Understanding failures of learning: Hebbian learning, competition for representational space, and some preliminary experimental data. In J. A. Reggia, E. Rupp, & D. Glanzman (Eds.), *Disorders of brain, behavior, and cognition: The neurocomputational perspective*. Oxford: Elsevier. pp 75–80.
- Nelson, C. A., & Monk, C. S. (2001). The use of event-related potentials in the study of cognitive development. In C. A. Nelson & M. Luciana (Eds.), *The Handbook of developmental cognitive neuroscience* (p. 125–147). Cambridge, MA: MIT Press.
- O’Reilly, R. C., & Johnson, M. H. (1994). Object recognition and sensitive periods: A computational analysis of visual imprinting. *Neural Computation*, 6, 357–389.
- O’Reilly, R. C., & Munakata, Y. (2000). *Computational explorations in cognitive neuroscience: Understanding the mind by simulating the brain*. Cambridge, Mass: MIT Press.
- Seidenberg, M. S., & Zevin, J. D. (2006). Connectionist models in developmental cognitive neuroscience: Critical Periods and the paradox of success. In Y. Munakata & M. Johnson (Eds.), *Attention & Performance XXI: Processes of Change in Brain and Cognitive Development*. Oxford: Oxford University Press.
- Stern, Y. (2002). What is cognitive reserve? Theory and research application of the reserve concept. *Journal of the International Neuropsychological Society*, 8, 448–460.
- Thomas, M. S. C., & Karmiloff-Smith, A. (2002). Are developmental disorders like cases of adult brain damage? Implications from connectionist modelling. *Behavioral and Brain Sciences*, 25, 727–788.
- Thomas, M. S. C., & Richardson, F. (2006). Atypical representational change: Conditions for the emergence of atypical modularity. In Y. Munakata & M. Johnson (Eds.), *Attention and Performance XXI: Processes of Change in Brain and Cognitive Development*. Oxford: Oxford University Press.
- Tyler, L. K., Marslen-Wilson, W. D., & Stamatakis, E. A. (2005). Differentiating lexical form, meaning, and structure in the neural language system. *Proceedings of the National Academy of Sciences*, 102, 8375–8380.
- Tyler, A. N. (*xxxx*^{O5}). When is a description not an explanation? *Developmental Psychobiology*

[Q1](#): Please check the affiliation.

[Q2](#): Please update.

[Q3](#): Please update.

[Q4](#): Please update.

[Q5](#): Please update.

Author Proof





111 RIVER STREET, HOBOKEN, NJ 07030

ELECTRONIC PROOF CHECKLIST, DEVELOPMENTAL PSYCHOBIOLOGY

*****IMMEDIATE RESPONSE REQUIRED*****

Please follow these instructions to avoid delay of publication.

☐ **READ PROOFS CAREFULLY**

- This will be your only chance to review these proofs.
- Please note that the volume and page numbers shown on the proofs are for position only.

☐ **ANSWER ALL QUERIES ON PROOFS** (Queries for you to answer are attached as the last page of your proof.)

- Mark all corrections directly on the proofs. Note that excessive author alterations may ultimately result in delay of publication and extra costs may be charged to you.

☐ **CHECK FIGURES AND TABLES CAREFULLY** (Color figures will be sent under separate cover.)

- Check size, numbering, and orientation of figures.
- All images in the PDF are downsampled (reduced to lower resolution and file size) to facilitate Internet delivery. These images will appear at higher resolution and sharpness in the printed article.
- Review figure legends to ensure that they are complete.
- Check all tables. Review layout, title, and footnotes.

☐ **COMPLETE REPRINT ORDER FORM**

- Fill out the attached reprint order form. It is important to return the form even if you are not ordering reprints. You may, if you wish, pay for the reprints with a credit card. Reprints will be mailed only after your article appears in print. This is the most opportune time to order reprints. If you wait until after your article comes off press, the reprints will be considerably more expensive.

RETURN

- ☐ **PROOFS**
☐ **REPRINT ORDER FORM**
☐ **CTA (If you have not already signed one)**

RETURN WITHIN 48 HOURS OF RECEIPT VIA FAX TO Alinefsk AT 201-748-6052

QUESTIONS?

Alyson Linefsky, Senior Production Editor
Phone: 201-748-6723
E-mail: alinefsk@wiley.com
Refer to journal acronym and article production number
(i.e., DEV 00-001 for *Developmental Psychobiology* ms 00-001).

COPYRIGHT TRANSFER AGREEMENT

Date:

To:

Production/Contribution

ID# _____

Publisher/Editorial office use only

Re: Manuscript entitled _____ (the "Contribution")
for publication in _____ (the "Journal")
published by Wiley-Liss, Inc., a subsidiary of John Wiley & Sons, Inc. ("Wiley").

Dear Contributor(s):

Thank you for submitting your Contribution for publication. In order to expedite the publishing process and enable Wiley to disseminate your work to the fullest extent, we need to have this Copyright Transfer Agreement signed and returned to us as soon as possible. If the Contribution is not accepted for publication this Agreement shall be null and void.

A. COPYRIGHT

1. The Contributor assigns to Wiley, during the full term of copyright and any extensions or renewals of that term, all copyright in and to the Contribution, including but not limited to the right to publish, republish, transmit, sell, distribute and otherwise use the Contribution and the material contained therein in electronic and print editions of the Journal and in derivative works throughout the world, in all languages and in all media of expression now known or later developed, and to license or permit others to do so.
2. Reproduction, posting, transmission or other distribution or use of the Contribution or any material contained therein, in any medium as permitted hereunder, requires a citation to the Journal and an appropriate credit to Wiley as Publisher, suitable in form and content as follows: (Title of Article, Author, Journal Title and Volume/Issue Copyright © [year] Wiley-Liss, Inc. or copyright owner as specified in the Journal.)

B. RETAINED RIGHTS

Notwithstanding the above, the Contributor or, if applicable, the Contributor's Employer, retains all proprietary rights other than copyright, such as patent rights, in any process, procedure or article of manufacture described in the Contribution, and the right to make oral presentations of material from the Contribution.

C. OTHER RIGHTS OF CONTRIBUTOR

Wiley grants back to the Contributor the following:

1. The right to share with colleagues print or electronic "preprints" of the unpublished Contribution, in form and content as accepted by Wiley for publication in the Journal. Such preprints may be posted as electronic files on the Contributor's own website for personal or professional use, or on the Contributor's internal university or corporate networks/intranet, or secure external website at the Contributor's institution, but not for commercial sale or for any systematic external distribution by a third party (e.g., a listserve or database connected to a public access server). Prior to publication, the Contributor must include the following notice on the preprint: "This is a preprint of an article accepted for publication in [Journal title] © copyright (year) (copyright owner as specified in the Journal)". After publication of the Contribution by Wiley, the preprint notice should be amended to read as follows: "This is a preprint of an article published in [include the complete citation information for the final version of the Contribution as published in the print edition of the Journal]", and should provide an electronic link to the Journal's WWW site, located at the following Wiley URL: <http://www.interscience.Wiley.com/>. The Contributor agrees not to update the preprint or replace it with the published version of the Contribution.

2. The right, without charge, to photocopy or to transmit online or to download, print out and distribute to a colleague a copy of the published Contribution in whole or in part, for the Contributor's personal or professional use, for the advancement of scholarly or scientific research or study, or for corporate informational purposes in accordance with Paragraph D.2 below.
3. The right to republish, without charge, in print format, all or part of the material from the published Contribution in a book written or edited by the Contributor.
4. The right to use selected figures and tables, and selected text (up to 250 words, exclusive of the abstract) from the Contribution, for the Contributor's own teaching purposes, or for incorporation within another work by the Contributor that is made part of an edited work published (in print or electronic format) by a third party, or for presentation in electronic format on an internal computer network or external website of the Contributor or the Contributor's employer.
5. The right to include the Contribution in a compilation for classroom use (course packs) to be distributed to students at the Contributor's institution free of charge or to be stored in electronic format in datarooms for access by students at the Contributor's institution as part of their course work (sometimes called "electronic reserve rooms") and for in-house training programs at the Contributor's employer.

D. CONTRIBUTIONS OWNED BY EMPLOYER

1. If the Contribution was written by the Contributor in the course of the Contributor's employment (as a "work-made-for-hire" in the course of employment), the Contribution is owned by the company/employer which must sign this Agreement (in addition to the Contributor's signature), in the space provided below. In such case, the company/employer hereby assigns to Wiley, during the full term of copyright, all copyright in and to the Contribution for the full term of copyright throughout the world as specified in paragraph A above.
2. In addition to the rights specified as retained in paragraph B above and the rights granted back to the Contributor pursuant to paragraph C above, Wiley hereby grants back, without charge, to such company/employer, its subsidiaries and divisions, the right to make copies of and distribute the published Contribution internally in print format or electronically on the Company's internal network. Upon payment of the Publisher's reprint fee, the institution may distribute (but not resell) print copies of the published Contribution externally. Although copies so made shall not be available for individual re-sale, they may be included by the company/employer as part of an information package included with software or other products offered for sale or license. Posting of the published Contribution by the institution on a public access website may only be done with Wiley's written permission, and payment of any applicable fee(s).

E. GOVERNMENT CONTRACTS

In the case of a Contribution prepared under U.S. Government contract or grant, the U.S. Government may reproduce, without charge, all or portions of the Contribution and may authorize others to do so, for official U.S. Government purposes only, if the U.S. Government contract or grant so requires. (U.S. Government Employees: see note at end).

F. COPYRIGHT NOTICE

The Contributor and the company/employer agree that any and all copies of the Contribution or any part thereof distributed or posted by them in print or electronic format as permitted herein will include the notice of copyright as stipulated in the Journal and a full citation to the Journal as published by Wiley.

G. CONTRIBUTOR'S REPRESENTATIONS

The Contributor represents that the Contribution is the Contributor's original work. If the Contribution was prepared jointly, the Contributor agrees to inform the co-Contributors of the terms of this Agreement and to obtain their signature to this Agreement or their written permission to sign on their behalf. The Contribution is submitted only to this Journal and has not been published before, except for "preprints" as permitted above. (If excerpts from copyrighted works owned by third parties are included, the Contributor will obtain written permission from the copyright owners for all uses as set forth in Wiley's permissions form or in the Journal's Instructions for Contributors, and show credit to the sources in the Contribution.) The Contributor also warrants that the Contribution contains no libelous or unlawful statements, does not infringe on the rights or privacy of others, or contain material or instructions that might cause harm or injury.

CHECK ONE:☐ Contributor-owned work_____
Contributor's signature_____
Date_____
Type or print name and title_____
Co-contributor's signature_____
Date_____
Type or print name and title**ATTACH ADDITIONAL SIGNATURE PAGE AS NECESSARY**☐ Company/Institution-owned work
(made-for-hire in the
course of employment)_____
Company or Institution (Employer-for-Hire)_____
Date_____
Authorized signature of Employer_____
Date☐ **U.S. Government work****Note to U.S. Government Employees**

A Contribution prepared by a U.S. federal government employee as part of the employee's official duties, or which is an official U.S. Government publication is called a "U.S. Government work," and is in the public domain in the United States. In such case, the employee may cross out Paragraph A.1 but must sign and return this Agreement. If the Contribution was not prepared as part of the employee's duties or is not an official U.S. Government publication, it is not a U.S. Government work.

☐ **U.K. Government work (Crown Copyright)****Note to U.K. Government Employees**

The rights in a Contribution prepared by an employee of a U.K. government department, agency or other Crown body as part of his/her official duties, or which is an official government publication, belong to the Crown. In such case, the Publisher will forward the relevant form to the Employee for signature.



WILEY

Publishers Since 1807

DEVELOPMENTAL PSYCHOBIOLOGY

Telephone Number:

• Facsimile Number:

To: Ms. Alyson Linefsky

Fax: 201-748-6052

From: Dr.

Date: _____

Re: Developmental Psychobiology, ms #

Dear Ms. Mary,

Attached please find corrections to ms# _____. Please contact me should you have any difficulty reading this fax at the numbers listed below.

Office phone:

Email:

Fax:

Lab phone:

I will return color figure proofs (if applicable) once I have checked them for accuracy.

Thank you,

Dr.

E-proofing feedback comments:



REPRINT BILLING DEPARTMENT • 111 RIVER STREET • HOBOKEN, NJ 07030
PHONE: (201) 748-6723; FAX: (201) 748-6052; E-MAIL: reprints@wiley.com

Please complete this form even if you are not ordering reprints.

This form MUST be returned with your corrected proofs and original manuscript. Your reprints will be shipped approximately 4 weeks after publication. Reprints ordered after printing are substantially more expensive.

JOURNAL: DEVELOPMENTAL PSYCHOBIOLOGY Volume: _____ Issue: _____

TITLE OF MANUSCRIPT: _____

MANUSCRIPT NUMBER: _____ No. pages: _____ 1st Author: _____

REPRINTS 8 1/4 X 11

No. of pages	100 Reprints \$US	200 Reprints \$US	300 Reprints \$US	400 Reprints \$US	500 Reprints \$US
1-4	336	501	694	890	1052
5-8	469	703	987	1251	1477
9-12	594	923	1234	1565	1850
13-16	714	1156	1527	1901	2273
17-20	794	1340	1775	2212	2648
21-24	911	1529	2031	2536	3037
25-28	1004	1707	2267	2828	3388
29-32	1108	1894	2515	3135	3755
33-36	1219	2092	2773	3456	4143
37-40	1329	2290	3033	3776	4528

REPRINTS ARE ONLY AVAILABLE IN LOTS OF 100. IF YOU WISH TO ORDER MORE THAN 500 REPRINTS, PLEASE CONTACT US AT (201) 748-6723 FOR A PRICE QUOTE.

COVERS

100 Covers:	\$90	200 Covers:	\$145	300 Covers	\$200
400 Covers:	\$255	500 Covers:	\$325	Additional 100s	\$65

☐ Please send me _____ reprints of the above article at \$ _____

☐ Please send me _____ Generic Covers of the above journal at \$ _____

Please add appropriate State and Local Tax {Tax Exempt No. _____} _____

Please add 5% Postage and Handling _____

TOTAL AMOUNT OF ORDER** _____

***International orders must be paid in US currency and drawn on a US bank*

Please check one: ☐ Check enclosed ☐ Bill Me ☐ Credit Card

If credit card order, charge to: ☐ American Express ☐ Visa ☐ MasterCard ☐ Discover

Credit Card No. _____ Signature: _____ Exp. Date: _____

BILL TO:

Name: _____

Address/Institution: _____

SHIP TO:

Name: _____

Address/Institution: _____

Phone: _____ Fax: _____

Purchase order no: _____ E-mail: _____

Softproofing for advanced Adobe Acrobat Users - NOTES tool

NOTE: ADOBE READER FROM THE INTERNET DOES NOT CONTAIN THE NOTES TOOL USED IN THIS PROCEDURE.

Acrobat annotation tools can be very useful for indicating changes to the PDF proof of your article. By using Acrobat annotation tools, a full digital pathway can be maintained for your page proofs.

The NOTES annotation tool can be used with either Adobe Acrobat 4.0, 5.0 or 6.0. Other annotation tools are also available in Acrobat 4.0, but this instruction sheet will concentrate on how to use the NOTES tool. Acrobat Reader, the free Internet download software from Adobe, DOES NOT contain the NOTES tool. In order to softproof using the NOTES tool you must have the full software suite Adobe Acrobat 4.0, 5.0 or 6.0 installed on your computer.

Steps for Softproofing using Adobe Acrobat NOTES tool:

1. Open the PDF page proof of your article using either Adobe Acrobat 4.0, 5.0 or 6.0. Proof your article on-screen or print a copy for markup of changes.
2. Go to File/Preferences/Annotations (in Acrobat 4.0) or Document/Add a Comment (in Acrobat 6.0) and enter your name into the "default user" or "author" field. Also, set the font size at 9 or 10 point.
3. When you have decided on the corrections to your article, select the NOTES tool from the Acrobat toolbox and click in the margin next to the text to be changed.
4. Enter your corrections into the NOTES text box window. Be sure to clearly indicate where the correction is to be placed and what text it will effect. If necessary to avoid confusion, you can use your TEXT SELECTION tool to copy the text to be corrected and paste it into the NOTES text box window. At this point, you can type the corrections directly into the NOTES text box window. **DO NOT correct the text by typing directly on the PDF page.**
5. Go through your entire article using the NOTES tool as described in Step 4.
6. When you have completed the corrections to your article, go to File/Export/Annotations (in Acrobat 4.0) or Document/Add a Comment (in Acrobat 6.0).
7. When closing your article PDF be sure NOT to save changes to original file.
8. To make changes to a NOTES file you have exported, simply re-open the original PDF proof file, go to File/Import/Notes and import the NOTES file you saved. Make changes and re-export NOTES file keeping the same file name.
9. When complete, attach your NOTES file to a reply e-mail message. Be sure to include your name, the date, and the title of the journal your article will be printed in.