Neural dynamics of error processing in medial frontal cortex

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Adaptive behavior requires an organism to evaluate the outcome of its actions, such that future behavior can be adjusted accordingly and the appropriate response selected. During associative learning, the time at which such evaluative information is available changes as learning progresses, from the delivery of performance feedback early in learning to the execution of the response itself during learned performance. Here, we report a learning-dependent shift in the timing of activation in the rostral cingulate zone of the anterior cingulate cortex from external error feedback to internal error detection. This pattern of activity is seen only in the anterior cingulate, not in the pre-supplementary motor area. The dynamics of these reciprocal changes are consistent with the claim that the rostral cingulate zone is involved in response selection on the basis of the expected outcome of an action. Specifically, these data illustrate how the anterior cingulate receives evaluative information, indicating that an action has not produced the desired result.

Keywords: Error processing; Motor learning; Cognitive control; ACC; Cingulate cortex; fMRI

Introduction

To survive in changing environments, an organism must be able to adapt its behavior to the situation at hand. This flexibility can be achieved by evaluating response outcomes and adjusting behavior accordingly (Dickinson, 1985). In this regard, error signals provide important evaluative information, since they indicate that a behavior was inadequate given the current context and that, in future, a different response needs to be selected (Holroyd and Coles, 2002).

Existing data on the neural substrates of action selection indicate that the medial frontal cortex plays a crucial role in selecting actions on the basis of their outcomes (Matsumoto and Tanaka, 2004) and subsequent monitoring of response outcomes (Holroyd et al., 2004a; Ridderinkhof et al., 2004). Rather than attributing a single role to this vast cortical expanse, recent studies have started to associate different functions to the different anatomical structures that lay within the medial frontal cortex (Picard and Strick, 2001; Rushworth et al., 2004). In this context, an anterior portion of the cingulate cortex, the rostral cingulate zone anterior (RCZa), has been specifically associated with processing of error information and selecting appropriate behavioral adjustments (Holroyd and Coles, 2002; Rushworth et al., 2004; Fiehler et al., 2004).

These inferences on the neural bases of error processing have been obtained in the context of a “static” experimental environment, in which the organism knows the behavior that is appropriate for the current situation. Thus, a given response can be evaluated immediately against an internal representation of the correct stimulus–response relationship. Should the response be incorrect, error information is available from an internal error-detection process at the time of the response (Gehring et al., 1993; Holroyd et al., 2005). However, in a novel environment, with as yet unknown stimulus–response associations, error information is not available until the delivery of external performance feedback. This implies that, during the learning of stimulus–response associations by trial and error, the time at which error information is available will change. Prior to learning, error information will not be available until external performance feedback is delivered, but after learning, error information will be available earlier from internal sources at the time of the response itself. Thus, a neural structure that adjusts behavior as a function of the evaluation of response outcomes should dynamically shift its responsivity as a function of learning, from external sources provided by error feedback to internal sources associated with the error response itself. We predicted that, following error feedback, activity in the anterior...
Materials and methods

Subjects

We studied eight right-handed male volunteers (mean age = 30.4 years, SD = 13.4) with normal or corrected-to-normal vision after obtaining informed consent according to institutional guidelines of the local ethics committee (CMO region Arnhem-Nijmegen, Netherlands). They were paid €10 per hour for their participation. Imaging data from 5 additional subjects were discarded, since these subjects either failed to learn the appropriate stimulus–response mappings adequately (2 subjects, less than 50% correct on post-scanning forced-choice recall task) or performed without any errors during the last part of the scanning session, indicating that the RT deadline was not tight enough for these subjects (3 subjects).

Experimental setup

Subjects lay supine in the scanner. Head movements were minimized by an adjustable padded head holder. Visual stimuli (visual angle of approximately 6°) were projected onto a mirror above the subjects’ heads. Motor responses were recorded via an MR-compatible keypad (MRI Devices, Waukesha, WI), positioned on the right side of the subject’s abdomen. Stimulus presentation and response collection were controlled by a PC running Presentation 0.51 (Neurobehavioral Systems, San Francisco, CA).

Behavioral procedure

Participants were asked to try to learn arbitrary associations between visual stimuli (black and white drawings of cars, airplanes, boats, etc.) and motor responses (pressing of one of four buttons with the fingers of the right hand) by trial-and-error using performance feedback (Fig. 1). We manipulated the degree of learning achieved during the experimental session by varying the number of times a visuomotor mapping was presented. For one condition (High Learning, HL), four distinct visuomotor mappings were presented 36 times each over the course of the scanning session, while for a control condition (Low Learning, LL), 24 different mappings were presented 6 times each. Trials enabling learning (HL) were pseudo-randomly intermixed and matched in number with trials in which learning was less likely to occur (LL). Participants received either performance feedback (green or red square) or neutral feedback (gray square, see Experimental timing) after each response, with a variable delay between these two events. To encourage error commission even during learned performance, a stringent reaction time deadline of 750 ms was enforced. When subjects responded after this deadline, immediate feedback (blue square) was provided and the trial ended. Subjects were instructed to try to avoid this at all costs. Subjects practiced the task in the scanner for 50 trials using a different stimulus set before the experimental session.

Following the scanning session, participants performed a forced choice recall test, in which all stimuli of the HL condition and a subset (50%) of the stimuli of the LL condition were presented 7 times each, randomly intermixed. Subjects were required to press the button corresponding to each stimulus, as during the scanning session. However, during the recall test, there was no reaction time deadline and no feedback was given, to allow for a reliable assessment of the learning of the stimulus–response mappings.

Imaging procedures

Images were acquired using a 1.5T Sonata scanner (Siemens, Erlangen, Germany). BOLD sensitive functional images were acquired using a single shot gradient EPI sequence (TR/TE 2.2s/40 ms, 28 transversal slices, interleaved acquisition, voxel size 3.5 × 3.5 × 3.5 mm). Following the experimental session, structural images were acquired using an MP-RAGE sequence (TR/TE/TI 2250 ms/3.93 ms/850 ms, voxel size 1 × 1 × 1 mm).
Experimental timing

Our design was aimed at dissociating response- and feedback-related neurovascular activities despite their temporal proximity. We achieved this by using an event-related fMRI design that has proven effective in dissociating transient responses time-locked to sensory and motor events (Thoenissen et al., 2002; Toni et al., 1999). We introduced a variable delay between response and feedback (3.9–5.2 s, uniform distribution) and between the trials (1.3–13.5 s). Also, we introduced neutral feedback on approximately half of the trials, to decorrelate the stimulus/response and feedback regressors. Furthermore, before actual scanning, we ran simulations in order to optimize the range and order of delay lengths, inter-trial intervals, and neutral feedback stimuli and to minimize correlations between the regressors describing the expected BOLD signal to response and feedback events (Friston et al., 1999). Following the scanning session, we verified the ability of our design to dissociate response and feedback-related activity by examining the evoked hemodynamic responses in V1 and M1. As expected, we found reliable BOLD responses to both the stimulus/response epoch and the feedback epoch in V1, but only response-related activation in M1 (data not shown).

Data analysis

Imaging data were analyzed using SPM2 (www.fil.ion.ucl.ac.uk/spm/). The first five volumes of each participant’s data set were discarded to allow for T1 equilibration. Prior to analysis, data were spatially realigned and corrected for differences in slice time acquisition using the middle slice in time as reference. Each participant’s structural image was coregistered to the first of the functional images. Images were then normalized onto the ICBM template (http://www.loni.ucla.edu/ICBM/) using linear transformations only. Finally, data were spatially smoothed using an isotropic 6 mm FWHM Gaussian kernel.

Using standard multiple regression procedures (Friston et al., 1995), we partitioned the sources of experimental variance in the fMRI time series into main effects of Condition (High Learning or Low Learning), Epoch (activity time-locked either to the response or to feedback presentation), and Outcome (correct or incorrect for response-related data; correct, incorrect, or neutral for feedback-related data). Model regressors were convolved with a canonical hemodynamic response function (Friston et al., 1998). Learning-dependent modulations of activity were modeled as first and second order parametric effects of time on the model regressors. Furthermore, before actual scanning, we ran simulations in order to optimize the range and order of delay lengths, inter-trial intervals, and neutral feedback stimuli and to minimize correlations between the regressors describing the expected BOLD signal to response and feedback events (Friston et al., 1999). Following the scanning session, we verified the ability of our design to dissociate response and feedback-related activity by examining the evoked hemodynamic responses in V1 and M1. As expected, we found reliable BOLD responses to both the stimulus/response epoch and the feedback epoch in V1, but only response-related activation in M1 (data not shown).

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Results

Behavioral data

Behavioral data indicated that our design was successful in manipulating the degree of learning achieved by the participants during the scanning session. Participants learned the stimulus–response mappings at a faster rate in the High Learning condition than in the Low Learning condition (Condition $\times$ Time interaction on Error Rate: $F_{(7,49)} = 3.2, P = 0.035$, Fig. 3a). Although participants never reached error-free performance during the scanning session in either condition (because of the RT deadline), a post-scanning forced choice recall test indicated that more associations were learned in the High Learning condition (HL: 91%; LL: 43%; $t(7) = 12.1, P < 0.001$).

RTs on correct trials (Fig. 3b) did not differ between the two conditions (main effect of Condition: $F_{(1,7)} = 0.446$, n.s.).

Over the course of the scanning session, RT increased to approach the RT cut-off (main effect of Time: $F_{(7,49)} = 5.493, P = 0.008$), but at a similar rate across conditions (Condition $\times$ Time interaction: $F_{(7,49)} = 1.514$, n.s.). The number of missed responses did not differ across conditions (HL: 19.7% [SD = 15.2]; LL: 18.4% [SD = 12.3]; $t(7) = 0.908$, n.s.).

Imaging data

We isolated BOLD signals satisfying our criteria by testing, in each of the ROIs, for time-dependent response-related increases and feedback-related decreases in activity during error trials. In addition, to distinguish genuine learning-related changes from mere time-related effects, we required this interaction to be stronger in the High Learning condition than in the Low Learning condition.

Our VOI analyses identified a region within the rostral cingulate zone anterior (RCZa, Table 1) which showed learning-related changes in activation elicited by incorrect responses and negative performance feedback. As illustrated in Fig. 4, this region showed greater feedback-related error activation during initial learning. During learning, this feedback-related activation decreased, while the response-related error signal showed a reciprocal increase (Fig. 4c).

To further characterize the activity evoked in this region, we calculated effect sizes for each main and time-related effect, normalizing the relevant parameter estimate of the multiple regression onto its standard error (see Materials and methods). The reciprocity of the dynamic modulation of activity in this cluster is indicated by the presence of significant learning-related effects, but no overall effects of response or feedback. The RCZa showed no significant activity on correct trials (Fig. 4d), indicating that activity in this region was specifically related to error processing. Furthermore, the time-related changes in activation on incorrect trials were stronger in the High Learning (feedback: $-3.17$; response: $2.60$, Fig. 4b) as compared to the Low Learning (feedback: $-2.70$; response: $0.15$) condition, providing evidence that these changes are not simply due to time-related effects (e.g., fatigue, habituation, sensitization), but are genuinely learning-related.

There were further clusters of activity in the pre-SMA VOI, although they did not survive the additional Bonferroni correction for multiple VOIs (Table 1). This region showed a clear modulation of activity as a function of learning, as illustrated in Fig. 5, for both response (effect size: $4.20$) and feedback (effect size: $-3.76$), and these modulations were not as strong in the LL condition (feedback: $-0.04$; response: $1.93$).

Fig. 3. Behavioral results. Error rates (a) and RT on correct trials (b) for the High Learning (blue) and Low Learning (red) conditions obtained during the scanning session. Curves are fitted first and second order polynomials; error bars indicate ± SEM. It can be seen that subjects learned the stimulus–response mappings at a faster rate in the High Learning condition than in the Low Learning condition (panel a). RT did not differ between the two conditions, approaching the RT cut-off (dashed horizontal line) at a similar rate (panel b).
Crucially, this region did not show the same reciprocity of effects seen in the anterior cingulate clusters, as indicated by a significant main effect of response (effect size: 6.74). Furthermore, the pre-SMA revealed response-related activity during correct trials (effect size: 4.55), an indication that this region is not exclusively driven by error signals.

Discussion

The present data indicate that, over the course of learning a set of arbitrary visuomotor mappings, a region along the cingulate sulcus (RCZa) shifts its responsiveness to different sources of error information as a function of learning. Error feedback-related activation decreases as learning proceeds, while error response-related activation increases, and these effects are reciprocal (Fig. 4). These results show not only that the anterior cingulate cortex responds to both internal (Carter et al., 1998; Garavan et al., 2002; Ullsperger and Von Cramon, 2001) and external (Holroyd et al., 2004b; Ullsperger and Von Cramon, 2003) sources of error information, but also that this cingulate region responds to the earliest source of error information available.

Furthermore, the present data argue against a unique cognitive contribution of the vast expanse of cerebral cortex labeled ‘medial frontal cortex’, confirming and detailing the functional heterogeneity of different anatomical portions of this region (Nachev et al., 2005; Rushworth et al., 2004). While the RCZa is activated in response to the first signal that an error has occurred, independent of the source of this information, pre-SMA shows response-related effects over and above learning-dependent modulations of activity on both correct and incorrect trials. These findings are consistent with the results of Akkal et al. (2002), showing that CMAr neurons are more likely to be modulated by performance feedback than pre-SMA neurons. This suggests that pre-SMA might be closer to motor aspects of the learning process than to the generation of error-related behavioral adjustments.

It could be argued that the differential time-related effects seen in the RCZa for the High Learning and Low Learning conditions reflect the putative role of the anterior cingulate region in controlling arousal (Critchley et al., 2003). However, our behavioral data and post-scanning forced choice recall test indicate that subjects learned the stimulus–response mappings in both learning conditions, although to a different extent (Fig. 3a). This result implies that subjects were evaluating stimuli and feedback during both Low and High Learning trials, although the rapid turnover of stimuli–response mappings in the former condition prevented them from learning as effectively as during the latter condition. Moreover, the two experimental conditions evoked overlapping reaction times profiles (Fig. 3b). These behavioral results are not immediately compatible with different arousal levels evoked by the High and Low Learning conditions.

Recently, Walton et al. (2004) have shown that the RCZa can be active not only on incorrect trials, but also on correct trials, provided that these trials convey behaviorally relevant information. In the current learning task, it is possible that the first correct trial associated with a specific mapping might have evoked ACC activity. Unfortunately, the current study was not designed to address this particular issue and we lack an adequate number of “first correct” trials to be able to provide a reliable estimate of ACC activity under these circumstances. This issue remains open for further investigation.

In this study, we have focused our search on RCZa on the basis of the role played by CMAr [the equivalent of RCZa in the macaque (Picard and Strick, 1996)] in reward-based action selection and evaluation (Shima and Tanji, 1998). Although recent meta-analyses (Ridderinkhof et al., 2004; Ullsperger and Von Cramon, 2004) have reported that error processing within the
anterior cingulate may encompass both RCZa and RCZp, the arm-subfields of these two areas are structurally and hodologically different (Picard and Strick, 1996), i.e. they are meant to have different functional properties. Indeed, an analysis on the error-processing characteristics of both RCZa and RCZp revealed a more caudal cingulate region (8, 10, 40) with response-related activation, but no strong effects of feedback. However, this result was obtained in an explorative whole-brain fixed-effects analysis. Given the strong hypothesis-driven nature of this report, we prefer to limit the inferences of this study to those regions for which we had explicit hypotheses.

Overall, these results are consistent with a series of recent studies showing that portions of the general region labeled ‘cingulate cortex’ are involved in action selection based on the expected outcome of an action (Bush et al., 2002; Hadland et al., 2003; Shima and Tanji, 1998), integrating information regarding a motor response and its potential outcome (Williams et al., 2004). Our results illustrate how a specific portion of the medial frontal cortex, the RCZa, might receive evaluative information, which can be used to adapt behavior accordingly (Holroyd et al., 2004a; Ridderinkhof et al., 2004). Conceptually, our results are also consistent with the notion that ACC activation during error trials is the result of an error in reward prediction, indicating that ongoing events are unexpectedly disadvantageous and that this information is subsequently used to guide action selection (Holroyd and Coles, 2002).

In conclusion, in this study, we have illustrated the dynamic characteristics of the interplay between external and internal sources of error information, emphasizing the contribution of a specific portion of medial frontal cortex (RCZa) to the selection of appropriate behaviors.

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