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Surprise Signals in the Supplementary Eye Field: Rectified Prediction Errors Drive Exploration-Exploitation Transitions

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Running Head: SEF Surprise Signals Drive Behavioral Transitions

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32 **Abstract**

33 Visual search is coordinated adaptively by monitoring and predicting the environment. The
34 supplementary eye field (SEF) plays a role in oculomotor control and outcome evaluation. However,
35 it is not clear whether the SEF is involved in adjusting behavioral modes based on the preceding
36 feedback. We hypothesized that the SEF drives exploration–exploitation transitions by generating
37 “surprise signals” or rectified prediction errors, which reflect differences between predicted and
38 actual outcomes. To test this hypothesis, we introduced an oculomotor two-target search task in
39 which monkeys were required to find two valid targets among four identical stimuli. After they
40 detected the valid targets, they exploited their knowledge of target locations to obtain a reward by
41 choosing the two valid targets alternately. Behavioral analysis revealed two distinct types of
42 oculomotor search patterns: exploration and exploitation. We found that two types of SEF neurons
43 represented the surprise signals. The error-surprise neurons showed enhanced activity when the
44 monkey received the first error feedback after the target-pair change, and this activity was followed
45 by an exploratory oculomotor search pattern. The correct-surprise neurons showed enhanced
46 activity when the monkey received the first correct feedback after an error trial, and this increased
47 activity was followed by an exploitative, fixed-type search pattern. Our findings suggest that
48 error-surprise neurons are involved in the transition from exploitation to exploration, and that
49 correct-surprise neurons are involved in the transition from exploration to exploitation.

51 **Introduction**

52 Dynamic control of exploration and exploitation is essential for searching optimally in an uncertain
53 environment (Gittins 1979; Krebs et al. 1978). During a search for a new object, we explore the
54 visual field based on predictions about the location of the object and the behavioral outcomes. After
55 we find the object, we memorize the location and exploit the knowledge in the next search. This
56 process involves the use of eye movements to search for an object. Thus, the higher-order
57 oculomotor center is expected to be involved in adaptive searches using eye movements.

58 The supplementary eye field (SEF) plays a crucial role in the control of higher-order eye
59 movements. The SEF, first identified as an oculomotor area in the dorsomedial frontal cortex
60 (Schlag and Schlag-Rey 1987), is involved in the control of anti-saccades (Schlag-Rey et al. 1997)
61 and the temporal organization of multiple saccades (Berdyeva and Olson 2010; Isoda and Tanji
62 2003; Lu et al. 2002). The SEF has recently been found to exert non-motor supervisory functions in
63 oculomotor control (Schall et al. 2002). Specifically, distinct groups of SEF neurons were activated
64 after errors, after the successful withholding of a prepared eye movement, or in association with
65 reinforcement in an oculomotor countermanding task (Stuphorn and Schall 2002; Stuphorn et al.
66 2000). In addition, the SEF is related to prediction and detection of reward (Amador et al. 2000) and
67 to the evaluation of the amount of reward (Roesch and Olson 2003; Uchida et al. 2007).
68 Furthermore, the SEF is suggested to have executive control over saccade generation by
69 monitoring behavioral consequences (Schall and Boucher 2007; Stuphorn and Schall 2006).
70 These studies show that the SEF is involved in both the motor and cognitive aspects of eye

71 movement control. However, it remains unclear how the SEF contributes to adaptive oculomotor
72 search based on performance monitoring. Thus, we attempted to investigate how the error- and
73 correct-outcome signals represented in the SEF are used to adjust subsequent searches. We
74 hypothesized that the SEF controls transitions between exploration and exploitation by producing
75 two types of surprise signals in response to unexpected erroneous and unexpected correct
76 outcomes (Fig. 1). To test this hypothesis, we introduced an oculomotor two-target search task in
77 which monkeys were required to explore valid targets with eye movements and to exploit
78 memorized target locations to select the next valid target. To classify SEF neurons, we constructed a
79 reinforcement learning model in which the predictions of correct and error feedback were updated
80 via surprise signals, defined as the differences between prediction and outcomes. We found that two
81 types of surprise neurons responded to unexpected correct and error feedback, respectively. Their
82 post-feedback activities were related to the shift between oculomotor search patterns in the
83 following trials, implying that the SEF plays a supervisory role in the exploration–exploitation
84 transitions based on performance monitoring.

85 **Methods**

86 *Subjects and apparatus*

87 Two Japanese monkeys (*Macaca fuscata*: monkey K, 6.5 kg; and monkey M, 6.1 kg) were trained to
88 perform the task in this study. They were cared for in accordance with the National Institutes of
89 Health guidelines and the guidelines of Tohoku University. The monkeys were kept in individual
90 primate cages in an air-conditioned room with food available *ad libitum*. During the experiments, the
91 monkeys sat in a primate chair with their heads restrained and faced a screen on which visual stimuli
92 were presented. Eye position was monitored with an infrared corneal reflection system sampling at
93 250 Hz.

94

95 *Behavioral task*

96 We introduced an oculomotor search task in which the monkeys were required to find two valid
97 targets among four identical targets and to make saccades alternately to the two targets (Fig. 2A).
98 Because the positions of the valid targets were not shown explicitly, the monkeys had to search for
99 the targets by trial and error. After a predetermined number of correct trials, the valid-target pair was
100 changed without instruction; thus, the monkeys had to change their search strategy (i.e., exploration
101 and exploitation) after receiving error feedback. The events in a trial are shown in Figure 2A.
102 Monkeys were allowed to choose one target in each trial. First, the monkeys were required to fixate
103 on a central fixation point for 500–700 ms. The onset of fixation was signaled when the color of the
104 central fixation point changed from white to red. When the monkeys fixated on the point for 500–700
105 ms successfully, the color was changed from red to green; this was followed by the appearance of

four identical visual targets: $2 \times 2^\circ$ white points presented 8.5° from the central fixation point. If the monkeys maintained fixation on the central fixation point for an additional 500–700 ms, the fixation point disappeared (Go signal), and the monkeys were allowed to make a saccade to one of the four peripheral targets. When the monkeys fixated on a target, the color of the selected target turned from white to red, signaling the start of fixation. The monkeys had to maintain fixation for 700 ms to receive feedback, which indicated whether the selected target was valid. There were two types of visual feedback. When the monkey fixated on an invalid target, the four dots disappeared (error feedback). When the monkeys fixated on the valid target, the color of the selected stimulus turned from red to green (correct feedback), and the four stimuli turned off 500 ms after the color change. The reward was then provided 500 ms after the disappearance of the four stimuli. After an interval of 2,000–2,500 ms, the next trial began. After the monkeys selected the valid target, the currently valid target was shifted to the other one within the currently valid pair (Fig. 2A, see the changes between “Bottom-right” and “Bottom-left”). If the fixation was interrupted before the feedback, the four peripheral targets were turned off, and the next trial began without reward (fixation-break trial). The fixation-break trials were excluded from our analyses.

We defined four types of valid pairs, each of which consisted of two adjacent targets among the four presented (Fig. 2B). The valid pair was changed in a pseudorandom manner after a predetermined number of consecutive correct choices (6–10 trials) or total number of correct choices (10–20 correct trials) after each pair change. Because the current valid pair was shifted without instruction, the monkeys had to search the valid paired targets based on behavioral

126 outcomes.

127 We classified the oculomotor search pattern into three types: “Adjacent,” “Diagonal,” and
128 “Repeat” (Fig. 2C). These patterns were defined by the relative location of the target chosen in the
129 current trial (n) compared with that of the target chosen in the previous trial ($n-1$). For example, if the
130 subject selected a target that was adjacent to the previously selected one, the search was classified
131 as an “Adjacent” search pattern. The adjacent search patterns were further divided into two types
132 based on whether the target selected in the current trial was included in the current valid pair or was
133 not. “Diagonal” patterns were defined as searches in which the subject selected the target diagonal
134 to the previously selected one. “Repeat” patterns were defined as searches in which the same target
135 selected in the previous trial was selected again.

136 We distinguished two phases of search behavior: exploration and exploitation. We defined the
137 exploration phase as the trials from the first error trial after a valid-pair change to the first correct trial
138 in which the monkey found the first valid target. The exploitation phase was defined as the
139 consecutive correct trials after the trial in which the monkey found both targets of the current valid
140 pair. If the subject made an error during the consecutive correct trials, the trials after the error were
141 not categorized into the exploitation phase. Based on these definitions, we inferred the internal
142 behavioral mode of the monkeys from their oculomotor search patterns and determined the timing of
143 the behavioral transitions. Examples of behaviors and feedback are presented in Figure 2D. After
144 the valid targets were changed without instruction, the monkeys made errors and began to search
145 for new valid targets.

146

147 *Computational model*

148 *General characteristics.* To categorize the task-related neurons, we developed a model for an
149 evaluation system to monitor behavioral outcomes, predict forthcoming feedback, and generate
150 “surprise signals”. The following two general assumptions underpinned this model.

151 (1) We assumed that the evaluation system detects changes in the environment by calculating the
152 differences between predictions and actual outcomes. The product of such comparisons were
153 so-called “prediction errors,” which can be represented as either signed or unsigned values. A
154 signed prediction error was calculated by subtracting the prediction value from the actual outcome.
155 Thus, a prediction error was either positive or negative, depending on the two values. Dopamine
156 neurons are well-known examples of signed prediction errors (Schultz 1997; 2006). In our analysis,
157 signed prediction errors were rectified, and only the positive components were used as surprise
158 signals that encoded the occurrence of an unexpected event or the non-occurrence of expected one
159 (Alexander and Brown 2011). Specifically, when an event occurred unexpectedly, the surprise signal
160 was calculated as a rectified prediction error by subtracting the prediction value from the actual
161 outcome value. Based on Alexander’s original model, we hypothesized two types of rectified
162 prediction error signals: correct and error. The correct type responded to correct feedback with an
163 increasing firing rate, and the error type responded to error feedback with an increasing firing rate.
164 Although the two types of rectified prediction errors are two aspects of the signed prediction error,
165 this issue of whether the surprise signals are encoded in the brain as single or multiple prediction

errors is not trivial, because the rectified prediction errors could lead to different behaviors.

(2) We assumed that the evaluation system would be able to predict future events such as the occurrences of correct and error feedback. Furthermore, the two signals were followed by different behavioral adjustments as shown in the results of the behavioral analyses. It is plausible that two separate representations of correct and error outcomes enable flexible behavioral adjustments by facilitating the prompt switch to the next search.

Based on these assumptions, we defined two types of prediction errors: surprise signals for correct and error feedback. The mathematical details are presented in the Appendix.

Surgical and recording methods

After the monkeys were trained to select targets at a rate of 90% correct during the exploitation phase, acrylic recording chambers (40 × 30 mm) were implanted in their skulls under aseptic conditions. Using Horsley–Clarke's stereotactic coordinates, markers indicating reference points were placed on the skull. Anesthesia was induced using ketamine hydrochloride (8 mg/kg i.m.) plus atropine sulfate, followed by pentobarbital sodium (30 mg/kg i.m.). Antibiotics and analgesics were used to prevent postsurgical infection and pain, respectively.

Standard electrophysiological techniques for single-cell recording were used to record the activity from the SEF (Mushiake et al. 2006). We located the SEF using previously established criteria to differentiate the SEF from adjacent areas, such as the supplementary motor area, based on responses to intracortical microstimulation (22–42 cathodal pulses of 0.2-ms duration at 333 Hz)

(Isoda and Tanji 2003; Wang et al. 2005). When eye movements were elicited by the stimuli at currents of $<50 \mu\text{A}$ in response to more than half the stimuli, the neuron was judged to be in the SEF. The cells were also assessed by location (2–8 mm rostral to the genu of the arcuate sulcus (AS genu) and within a few millimeters of the interhemispheric cleft) (Moorman and Olson 2007). Neuronal activity was recorded using glass-insulated Elgiloy electrodes (impedance of 0.8–1.0 M Ω) and parylene-insulated tungsten microelectrodes (impedance of 1.0–1.8 M Ω). An electrode was inserted through the dura using a hydraulic microdrive (Narishige, Tokyo, Japan). Neuronal activity was amplified and band-pass filtered (1–6 kHz), and single-unit potentials were discriminated using a template-matching algorithm (MCP-Plus and MSD; Alpha Omega Engineering, Nazareth, Israel). Data were stored in a microcomputer for off-line analysis.

Analysis of single-unit activity

Task-related neurons. If neuronal activity changed during the feedback or saccade period, the neurons were judged to be task-related. Specifically, if neuronal activity during a pre-feedback (0–300 ms before feedback) or post-feedback (0–300 ms after feedback) period differed significantly from that during the control period (200–500 ms after onset of fixation to the central fixation point, Mann–Whitney, $p < 0.05$), the neuron was judged to be feedback-related (Fujii et al. 2002). If the activity during a saccade period (0–300 ms after the Go signal) differed significantly from that during the control period (Mann–Whitney, $p < 0.05$), the neuron was judged to be “eye movement-related”.

206 *Regression analysis.* We hypothesized that SEF neurons would play evaluative roles in the
 207 oculomotor search task. Therefore, we tested whether the neurons represented the values in the
 208 evaluation system described above. To classify the neurons, we performed regression analyses
 209 containing a variable representing each value in the aforementioned model. We determined the
 210 value that fits best for each neuron by comparing the coefficients of determinants. We performed
 211 regression analyses on a trial-by-trial basis as

$$212 \quad \text{Firing rate} = \alpha_i \text{Value}_i + \varepsilon_i \quad (i = 1..10),$$

213 where Value_i denotes the explanatory variables that were the values described above (O_c , V_c , SS_c ,
 214 $\text{signed } PE_c$, O_e , V_e , SS_e , $\text{signed } PE_e$) and by the feedback in the current and previous trials. α_i are
 215 regression coefficients, and ε_i are intercepts for the explanatory variables. We calculated the
 216 coefficient of determination (R^2) for each model for each neuron. We defined a neuron as
 217 feedback-related when the R^2 was higher than 0.1. Based on this analysis, we selected the model
 218 with the higher R^2 as the best one to characterize neuronal activity.

219 To eliminate the simple visual-stimuli-responsive neurons from the feedback-related neurons, we
 220 compared the neuronal activity following the visual stimuli that signaled a correct or an error
 221 outcome from those that signaled only task phases. If the neuron did not show a significant change
 222 in activity after the outcome signals ($p < 0.05$), we excluded that neuron from further analyses.
 223 Specifically, to eliminate a neuron that responded to the change from red to green, we compared the
 224 activities during the periods after fixation onset with those after correct feedback (both stimuli
 225 consisted of a change in the color of the fixated target from red to green). We excluded the neuron if

it did not exhibit significantly different activities during the two periods. Additionally, to exclude neurons related to the absence of stimuli and those related to eye movements, we compared the activities during the periods in the error and correct trials in which stimuli were disappeared. We excluded the neuron if it did not show a significant difference between the activities in error and correct trials.

Directional selectivity. To examine whether task-related neurons were modulated by saccade direction, we performed regression analyses using the following equation:

$$\text{Firing rate} = \alpha_1 H + \alpha_2 V + \varepsilon,$$

where H denotes the horizontal location of the selected target, left (0) or right (1), V denotes the vertical location, bottom (0) or top (1). α_1 and α_2 are coefficients, and ε is an intercept. If the effect of either H or V were significant ($p < 0.05$) during the saccade period (0–300 ms after Go signal), the neuron was classified as having directional selectivity.

Normalized firing rate. The firing rate for each neuron was normalized in a 20-ms window by constructing a z-score of activity for each neuron in each bin. The z-score was computed across all correct and error trials by normalizing by the standard deviation in the mean activity of each neuron (Buschman and Miller 2009).

246 *Statistical analyses of the neuronal activity and behavior:* To investigate the activity pattern of the
 247 surprise neurons, we examined the changes in the activity during the post-feedback period during
 248 trials that were crucial to oculomotor target selection. Specifically, we calculated the firing rates for
 249 the following four types of trials: the first error trial after the change in the valid-target pair (1st E), the
 250 first correct trial after the pair change (1st C), the second correct trial after the pair change (2nd C),
 251 and the correct trials after more than two consecutive correct trials (CC). We calculated the
 252 averaged firing rates of the surprise neurons in these index trials. We used ANOVA with Tukey's
 253 HSD correction (factor: trial types, $p < 0.05$) to assess the differences in neuronal activity between
 254 the activities in these trial types. For other statistical tests, we used paired or simple t -tests for
 255 comparisons of two values. To investigate the relationship between the activity and oculomotor
 256 search patterns, we categorized trials by the oculomotor search pattern in the following trials
 257 (diagonal or adjacent) and calculated the mean post-feedback activity in the trials. In this analysis,
 258 the exploration phase was analyzed for the error-surprise and error-outcome neurons, and the
 259 exploitation phase was analyzed for the correct-surprise and correct-outcome neurons.

260

261 *Choice probability:* To determine the extent to which the activity of a neuron predicted the
 262 oculomotor search pattern (adjacent or diagonal), we calculated the choice probabilities of
 263 task-related neurons (Britten et al. 1996). In this situation, the choice probability was an estimate of
 264 the probability that an observer correctly predicted whether the monkey would engage in the
 265 adjacent or diagonal search pattern in the subsequent trial. To compute the choice probability, a

266 receiver operating characteristic curve was computed from the distributions of neuronal activities for
267 each of the two search patterns of the animal. The area under the curve was integrated to produce
268 the choice probability (Yang and Heinen 2014).

269

270

271 **Results**

272 ***Behavioral results***

273 After training, the accuracy of the two monkeys reached ~90% in the exploitation phase (monkey K:
 274 $91 \pm 3\%$, monkey M: $88 \pm 4\%$, mean \pm SD). After a valid-pair change, they began to explore new
 275 targets and found two targets in a new pair within an average of 4.2 trials (K: 4.2 ± 1.2 , M: 4.3 ± 1.3).
 276 We first examined the monkeys' oculomotor search patterns in the exploration and exploitation
 277 phases in terms of the three search patterns described in the Methods section: diagonal, adjacent,
 278 and repeat. During the exploitation phase, the adjacent search pattern was dominant (K: $97 \pm 2\%$,
 279 M: $96 \pm 2\%$), and the diagonal search was observed only rarely (K: $1 \pm 1\%$, M: $1 \pm 1\%$) (Fig. 4A). In
 280 contrast, during the exploration phase, the probability of the diagonal search increased (K: $54 \pm 5\%$,
 281 M: $48 \pm 5\%$), and the adjacent search decreased (K: $39 \pm 5\%$, M: $47 \pm 4\%$). This increase in the
 282 probability of the diagonal search during the exploration phase compared with the exploitation
 283 phase was statistically significant (paired t -test; $p < 0.001$). In the exploration phase, 4.2 trials were
 284 required to find the current pair of targets after a pair change. However, this raised the question of
 285 whether the 4.2 trials were optimal. To answer this, we performed the following analyses.

286 To investigate whether the subsequent oculomotor search pattern was dependent on behavioral
 287 history, we analyzed the search patterns in the next trials ($n+1$) based on the outcomes in the current
 288 (n) and previous ($n-1$) trials. The results showed that the frequency of diagonal search increased
 289 after the monkey made errors, especially when the error occurred after correct trials (Fig. 4B, error
 290 \rightarrow correct). However, adjacent searches were dominant when correct feedback occurred in the

current trial (n), irrespective of the outcome in the previous trial ($n-1$). We observed that the monkeys choose the diagonal target in the specific situation in which the subject made an error after selecting the target that was adjacent to the previous correct target. This pattern was included in the correct →error in Figure 4B.

Next, we analyzed the dependency of the monkeys' behavior on the history of the outcomes of the two preceding trials. Specifically, we compared the probability of a diagonal search occurring between the trials in which the two preceding trials ($n-2$) were correct and erroneous, respectively. The outcome of the two-preceding trials ($n-2$) did not have a significant effect on the search patterns in the next trials ($n+1$) (paired t -test, $p > 0.001$). Importantly, this suggests that the monkeys chose the search pattern based on the outcomes of the current (n) and previous ($n-1$) trials and did not use the information provided before the previous trial. This result implied that the monkeys showed suboptimal behavior during our task. Based on the finding that the monkeys used a combination of information from the current and previous trials, we categorized the neurons by the outcomes of the current and previous trials.

In addition, to examine whether the search pattern was affected by the direction of the chosen target, we analyzed the relationship between the position of the target selected in the current trial and the search patterns in the subsequent trials. The results revealed no significant relationship between search patterns and saccade directions (ANOVA, factor: direction of stimuli, $p = 0.17$). These results suggest that distinct search strategies operated during the exploration and exploitation phases.

311

312 ***Neuronal database***

313 We recorded 295 neurons in two monkeys (monkey K: 268, monkey M: 27). Of these SEF neurons,
 314 200 cells (K:184, M: 16) were found to be feedback-related, and 212 cells (K: 197, M: 15) were
 315 eye-movement-related. Among the 212 eye-movement-related neurons, 105 (K: 100, M: 5) showed
 316 significant directional selectivity ($p < 0.05$) during the saccade period. We present the various types
 317 of neurons in Table 1.

318 In the following section, we initially discuss outcome neurons that responded to correct or error
 319 feedback. We next address the surprise neurons that showed enhanced activity when the feedback
 320 was unexpected. Finally, we investigate the relationship between neuronal activities and behavioral
 321 patterns and present data showing that the enhanced activity of the surprise neurons was related to
 322 shifts between oculomotor search patterns.

323

324 ***Outcome neurons***

325 We identified SEF neurons showing elevated activity in response to error or correct feedback. The
 326 error-outcome neurons responded to error feedback in the current (n) trial, irrespectively of the
 327 outcomes of previous trials (Fig. 5A, B). The error-outcome neurons were categorized in terms of
 328 encoding the outcome values of error feedback (OV_e). We identified 47 error-outcome neurons. On
 329 the other hand, the correct-outcome neurons were divided into two types. The first type of neuron
 330 showed elevated activity in response to correct feedback, irrespectively of the outcomes of previous

trials (Fig. 5C). The second type of neuron showed distinctly decreasing activity in response to error feedback (Fig. 5D). We categorized 29 cells in the correct-outcome neurons, 21 of which were categorized as the first type of neuron, and the remaining eight of which were categorized as the second type. These correct-outcome neurons were viewed as encoding the outcome values of correct feedback (OV_c). Population-level activity was observed to increase slightly in the correct trials and to decrease slightly in the error trials (Fig. 5E). These neurons were not classified as prediction-error neurons but as outcome neurons, because they responded differentially depending on outcome history.

Surprise neurons

The surprise neurons showed enhanced activity when the unpredicted-outcome signals were fed back. There were two types of surprise neurons. First, error-surprise neurons increased in activity in response to the error feedback after a correct trial. Second, the correct-surprise neurons increased in activity in response to the correct feedback after an error trial. Afterwards, we investigated how these neurons were related to the behavioral transitions between exploration and exploitation.

The first example was an error-surprise neuron, which showed elevated activity in response to the error feedback in the current (n) trial when correct feedback was received in the previous ($n-1$) trial (Fig. 6A, correct→error). However, the activity was moderate when the error feedback occurred repeatedly and was predicted (Fig. 6A, error→error). Because the outcome of the current trial differed from that of the previous trial, the current-error outcome was surprising, and we refer to

351 these neurons as “error-surprise neurons”. These neurons were viewed as encoding the surprise
 352 signal as error feedback (SS_e). A total of 15 SEF neurons were placed in this group. Population-level
 353 analysis showed the highest activity when the error feedback was received in the current (n) trial
 354 after correct feedback had been received in the previous (n-1) trial (Fig. 6B, correct→error). There
 355 was no response when the correct feedback was received in the current (n) trial (correct→correct
 356 and error→correct). The second example was a correct-surprise neuron, which exhibited elevated
 357 activity when correct feedback was given in the current (n) trial after error feedback was received in
 358 the previous (n-1) trial (Fig. 6C, error→correct). The post-feedback activity decreased when correct
 359 trials occurred successively and were predicted (correct→correct). Because the outcome of the
 360 current trial differed from that of the previous trial, the current correct outcome was surprising, and
 361 we refer to these neurons as “correct-surprise neurons.” These neurons were viewed as encoding
 362 the surprise signal for correct feedback (SS_c). A total of 14 neurons were placed in this group.
 363 Population-level analysis showed similar activity when correct feedback was received in the current
 364 (n) trial following error trials (Fig. 6D, error→correct). Notably, these error- and correct-surprise
 365 neurons differed from neurons reflecting conventional prediction errors (*signed PE_e* and *signed PE_c*),
 366 because the surprise neurons encoded “*rectified*” positive parts of the prediction errors.

367 To confirm the validity of the categorization of the surprise neurons, we compared the R^2 values of
 368 the surprise neurons in terms of the different values in the model. Of the 15 error-surprise neurons,
 369 11 had R^2 values higher than 0.1 for signed prediction errors (*signed PE_e*), and 10 had R^2 values
 370 higher than 0.1 for error outcomes (OV_e) (Table 2). However, the R^2 values for error-surprise signals

371 were significantly higher than those of the other two groups (paired t -test, SS_e vs. *signed PE_e* , $p =$
 372 0.009 ; SS_e vs. OV_e , $p = 0.002$). On the other hand, among the 14 correct-surprise neurons, nine had
 373 R^2 values higher than 0.1 for signed prediction errors (*signed PE_c*), and seven had R^2 values higher
 374 than 0.1 for correct outcomes (OV_c) (Table 2). However, the R^2 values for correct-surprise signals
 375 were significantly higher than those in other groups (paired t -test, SS_c vs. *signed PE_c* , $p = 0.001$; SS_c
 376 vs. OV_c , $p = 0.0007$).

377 Finally, to investigate whether the surprise neurons were related to specific target directions, we
 378 used regression analyses to examine the directional selectivity of the neurons during saccade
 379 periods. We found that 10 error-surprise neurons (10/15) and seven correct-surprise neurons (7/14)
 380 showed directional selectivity during the 0–300-ms period after the feedback signals (ANOVA;
 381 factor: directions of stimuli; $p < 0.01$). On the other hand, 38 error-outcome neurons (38/47) and 15
 382 correct-outcome neurons (15/29) showed directional selectivity. This result is consistent with a
 383 previous finding that half the SEF neurons showed directional selectivity (Roesch and Olson 2003).

384

385 ***Surprise neurons and history of behavioral outcomes***

386 We investigated the relationship between the neuronal activity of surprise neurons and the outcome
 387 history. We identified two types of surprise neurons in the SEF that exhibited enhanced activity
 388 when an unpredicted outcome occurred. First, we investigated how the surprise neurons reflected
 389 changes in outcomes after pair shifts. Specifically, we analyzed the averaged firing rates of the
 390 surprise neurons in the following trials: the first error trial after a pair change (1st E), the first correct

trial after a pair change (1st C), the second correct trial after a pair change (2nd C), and consecutive correct trials (CC). These trials were crucial because they provided clues about the valid targets by tracking the history of behavioral outcomes. As a result, the error-surprise neurons showed significantly enhanced responses in the first error trial after the pair shifts compared with those in the consecutive correct trials (Fig. 7A: 1st E vs. CC, $p = 0.012$). The activity in the first error trial was significantly higher compared with those in the first and second correct trials (Fig. 7A: 1st E vs. 1st C, $p = 0.029$; 1st E vs. 2nd C, $p = 0.025$). We also found that the neuronal activity level of the error-surprise neurons was higher in the first error trial than in the second error trial (1st vs. 2nd, 7.0 ± 1.2 vs. 4.8 ± 0.90 , $p = 0.013$). We did not find a significant change after the third try due to the low number of trials under this condition (< 3 trials per neuron). The correct-surprise neurons showed significantly higher activity in the first correct trials compared with the preceding first error trial (Fig. 7B, 1st E vs. 1st C, $p = 0.040$). Furthermore, the activity in the first correct trial of correct-surprise neurons was larger in the first try than in the second try after a pair change, but this difference was not statistically significant (1st vs. 2nd try, 6.0 ± 0.8 vs. 4.4 ± 0.5 , $p = 0.16$). We did not find a significant change in the third try, because of the scarcity of trials performed under this condition (< 3 trials per neuron). These findings suggest that surprise neurons play a role in detecting the unexpected trials after the valid pair changes. This result is consistent with a previous study that found that neurons in the frontomedial region showed enhanced activity during the first correct trials after a change in the task condition (Quilodran et al. 2008).

Surprise neurons and exploration–exploitation transitions

To investigate whether the surprise neurons in the SEF were related to the exploitative or exploratory oculomotor search pattern, we analyzed the relationship between the post-feedback activities and search patterns in the subsequent trials. According to our behavioral analysis, there were two main oculomotor search patterns: diagonal and adjacent. The use of the diagonal pattern increased in the exploration phase, but the adjacent pattern was dominant in the exploitation phase (Fig. 4A). Thus, we investigated the relationship between the post-feedback activities and the oculomotor search patterns in the following trials. Specifically, we compared the neuronal activities in the trial followed by a diagonal search with those in the trials followed by an adjacent search. The results showed that error-surprise neurons had significantly higher activity levels in trials followed by a diagonal search than those followed by an adjacent search (Fig. 7C; $p = 0.040$). Furthermore, correct-surprise neurons showed significantly higher activity levels in trials followed by an adjacent search than in those followed by a diagonal search (Fig. 7D; $p = 0.010$). In contrast, the activity levels of error- and correct-outcome neurons were not significantly correlated with their activities or search patterns in subsequent trials (Fig. 7E and F; error-outcome neuron: $p > 0.05$, correct-outcome neuron: $p > 0.05$). These results suggest that the increased activity of the surprise neurons are related to the shifts between the oculomotor search patterns, but that the activity of the simple outcome neurons is not.

To investigate the influence of post-feedback activity on subsequent behavioral adjustments further, we calculated the choice probability for surprise and outcome neurons (Fig. 7G). In this

analysis, we computed the extent to which we could predict whether the subsequent search pattern was diagonal or adjacent. According to the results, the choice probability of the error-surprise neurons was higher than that of the error-outcome neurons ($p = 0.0028$). The choice probability of the correct-surprise neurons was slightly higher than that of the correct-outcome neurons, but this difference was not significant ($p = 0.33$).

In summary, the error-surprise neurons exhibited increased responses in the first error trials, and this activity was related to the behavioral shift from exploitation to exploration in the following trials. The correct-surprise neurons showed significant responses in the first correct trials, and this activity was related to the shift from exploration to exploitation in the following trials. We plotted the distribution of these SEF neurons and found no clear segregation (Fig. 8B and C).

Discussion

We introduced a two-target search task in which monkeys had to find two adjacent targets by trial and error. Behavioral analysis revealed two distinct oculomotor search patterns during the exploration and exploitation phases and revealed that the monkeys used their memory from only the two preceding trials. We found two types of surprise neurons: error-correct surprise neurons and correct-surprise neurons. These neurons showed elevated responses to unexpected error and correct feedback, respectively. Specifically, the error-surprise neurons selectively showed increasing activity when the monkeys made the first error after a pair change. The correct-surprise neurons showed enhanced activity when the monkeys found the first correct target after a pair change. Furthermore, the surprise neurons were related to the exploration–exploitation transitions. Specifically, the error-surprise neurons showed enhanced activity preceding the transition from exploitative to explorative search pattern. In contrast, the correct-surprise neurons showed enhanced activity preceding the transition from explorative to exploitative search pattern. Our results suggest that the SEF is involved in behavioral transitions between exploration and exploitation by encoding the surprise signals (Fig. 1).

Behavioral shifts during the two-target search task

The unique feature of our search task was the absence of an external cue for valid targets and the attendant requirement that monkeys search for the targets by trial and error. Previous studies used visual search tasks in which monkeys had to find a target among multiple identical stimuli (Amiez et al. 2012; Mirpour and Bisley 2012; Procyk et al. 2000; Quilodran et al. 2008). In contrast, our

oculomotor search task required monkeys to make alternate saccades to two valid targets during the exploration phase to motivate them to change the search pattern after a pair change. Thus, the subjects had to maintain their memory of previous choices and feedback to choose subsequent behavioral patterns.

Interestingly, the behavioral analyses revealed that the monkeys remembered the outcomes of only the two preceding trials: the current (n) and the previous trial ($n-1$). In a previous study, Quilodran et al. (2008) demonstrated that monkeys could remember the outcomes of three preceding trials according to their ability to choose the correct target among four possibilities after three consecutive error trials (Quilodran et al. 2008). This discrepancy between Quilodran's and our findings may be explained not by a limit in the monkeys' working memory but rather by their search strategy. In the study by Quilodran et al., the monkeys had to find one correct target among four targets, but our study required the monkeys to select two valid targets alternately. Our task was more demanding, and the monkeys used the two preceding outcomes to choose the subsequent search patterns.

Surprise signals in the SEF

Our results revealed surprise neurons in the SEF. These were defined as neurons that encoded the unsigned, rectified values of the prediction error, the difference between the prediction and the outcome. These surprise neurons engaged in activity that was distinct from that displayed by outcome neurons and reward-prediction-error neurons. Their pattern of activity was similar to that of

the surprise neurons in the ACC and prefrontal cortex (PFC) in that they responded to unpredicted outcomes (Asaad and Eskandar 2011; Hayden et al. 2011). Learning-selective activity has been observed in the SEF during an associative learning task, and this type of activity appeared preferentially when a novel stimulus was introduced (Chen and Wise 1995). One recent study revealed a group of neurons that encoded valence-related reward-prediction errors (So and Stuphorn 2012). In this context, we possibly demonstrated that SEF neurons encode a different aspect of the prediction error and that distinct types of surprise neurons represent the rectified prediction errors for correct and error feedback, respectively. Our finding supports the notion that the surprise signal promotes adaptive learning in a changing world (Courville et al. 2006; Pearce and Hall 1980).

The SEF encodes two types of prediction errors: one is based on the amount of reward and the other on the subjective probabilities of feedback. The former type of prediction error was identified using a behavioral task that involved changing the amount of the reward systematically (So and Stuphorn 2012). On the other hand, we investigated the latter type of prediction error as a surprise signal. We computed surprise signals based on the subjective probability of feedback, because we assumed that the SEF was able to evaluate outcomes based on the subjects' behavioral history. This is consistent with a recent study in which SEF neurons showed metacognition-related activity that reflected the subjective probability of correct choices (Middlebrooks and Sommer 2012).

We found that a portion of the surprise neurons showed directional selectivity during saccade periods, implying that they integrated directional information and outcome signals. We assumed that

the directional information in the saccade period and the outcome information during the post-feedback period would be combined to adjust subsequent behavior. On the other hand, a previous report showed that the reward value and directional information were represented separately in the SEF, but these signals were expected to be used in concert to adjust behavior (So and Stuphorn 2012). Future studies should address how information regarding action and outcome is integrated to adjust subsequent behavior.

Surprise signals in the frontal areas

Surprise signals, responses to unpredicted outcomes, were also observed in other frontal areas, such as the ACC, in single-cell recording studies (Amiez et al. 2005; Hayden et al. 2011; Shidara and Richmond 2002). Correct/error-related activity was reported by single-cell recording studies (Ito et al. 2003; Matsumoto et al. 2007) as well as by event-related potential studies (Emeric et al. 2008; Gamba et al. 1986). Neurons in the lateral prefrontal cortex (LPFC) were also reported to reflect positive or negative prediction errors (Asaad and Eskandar 2011; Seo and Lee 2007). The SEF is connected anatomically to the ACC and LPFC (Huerta and Kaas 1990; Wang et al. 2005). These functional and anatomical studies suggest that surprise signals are encoded in the frontal network in the SEF, ACC, and LPFC.

Furthermore, human electroencephalogram studies have reported event-related potentials in the medial frontal region in response to unexpected negative, unexpected positive, or novel stimuli (Falkenstein et al. 1991; Gehring et al. 1993; Holroyd et al. 2008; Ridderinkhof et al. 2004; Wessel et

al. 2012). Indeed, the relationship between error-related responses and neuronal activity should be investigated in future research.

Surprise signals and exploration–exploitation transitions

We found that the elevated activities of the surprise neurons in the SEF were followed by specific search patterns. Combined with the behavioral analysis that showed distinctive explorative and exploitative search patterns, our results suggest that the SEF neurons are related to the transitions between exploration and exploitation in visual searches (Fig. 1). As a frontomedial region, the ACC is functionally significant in adaptive, exploratory behavior (Daw et al. 2006; Procyk et al. 2000; Shima and Tanji 1998; Walton et al. 2004; Williams et al. 2004). A recent study showed that the SEF was more active during a matching-pennies task that required exploration than it was during a simple saccade task (Donahue et al. 2013). A previous study demonstrated that ACC neurons encoded the first correct feedback and contributed to the end of exploration in a target-search task (Quilodran et al. 2008). Our study extends their findings by showing that the SEF may contribute not only to the transition from exploration to exploitation but also to that from exploitation to exploration. Thus, the question arises regarding the neuronal mechanisms that underlie exploration–exploitation transitions after detection of surprise signals in the SEF. Cohen et al. proposed that cortical and subcortical areas contribute differently to the management of exploration and exploitation (Cohen et al. 2007). According to their model, cortical areas such as the ACC and the orbitofrontal cortex (OFC) generate evaluation signals and send them to subcortical regions such as the locus

coeruleus (LC) or dopaminergic network. The subcortical systems regulate decision-making processes (Doya 2002; Yu and Dayan 2005), and this activity has been related to behavioral transitions (Barnes et al. 2005; Graybiel 2008). Cortical and subcortical contributions to exploration and exploitation have also been supported by studies evaluating genetic factors in humans (Frank et al. 2009). Based on these findings, our data suggest that, along with the ACC and OFC, the SEF plays an evaluative role. Indeed, anatomical studies have shown that the SEF is connected to the LC (Arnsten and Goldman-Rakic 1984; Shook et al. 1990) and the ACC (Huerta and Kaas 1990; Wang et al. 2005). Thus, the SEF is expected to manage the balance between exploration and exploitation by sending surprise signals to the subcortical systems, such as the LC.

Exploration–exploitation transition as set shifting

Our task resembles the Wisconsin Card Sorting Test (WCST) in that it requires cognitive flexibility in switching mental “sets” (Milner 1963; Stuss et al. 2000). Functional MRI and single-cell recordings from monkeys performing an analog of the WCST have showed that the medial and lateral prefrontal cortices are involved in set shifting (Kamigaki et al. 2012; Konishi et al. 2010; Nakahara et al. 2002). Another study showed that ACC neurons responded to correct or error feedback and changed activity depending on the outcome of the previous trial (Mansouri et al. 2006). Compared with the WCST, our task requires understanding the spatial combinations of the targets, but it investigates how the oculomotor system is involved in the task.

From a clinical perspective, our finding is consistent with a study of a patient with a focal SEF lesion

(Husain et al. 2003). This patient demonstrated difficulty adapting to unexpected rule changes in a rule-reversal task. According to our findings, the patient may have been impaired in the ability to switch between exploration and exploitation strategies appropriately.

Surprise signals from the perspective of predictive coding and timing

The surprise neurons showed increased activity when the outcome was unexpected and decreased activity when the same outcome was repeated. This pattern of responses fits well with predictive coding theory, which holds that the brain infers the cause of sensory inputs based on a generative model and that prediction errors are used to adjust the model until the error is minimized (Friston 2005). The framework of the predictive coding explains the neuronal mechanism for repetition suppression, in which stimulus-evoked neural activity is attenuated when the same stimulus is repeatedly presented and activity is enhanced when the stimulus is unpredictable (Summerfield et al. 2008; Todorovic and de Lange 2012). Thus, the surprise signal in our task corresponds to the prediction error in predictive coding theory (Friston 2005).

We also found that a group of SEF neurons showed increased activity until the predicted event occurred. This type of activity is consistent with predictive timing theory, which posits that the brain estimates the timing of events and holds that prediction errors are generated when expected events do not occur (Arnal and Giraud 2012; Nobre et al. 2007). The predictive timing theory may explain various types of activity in SEF neurons, such as those that responded to predictable target motion (Heinen and Liu 1997) and error-feedback (Stuphorn et al. 2000), reward-predicting (Amador et al.

2000), event-predicting, and event-detecting neurons (Campos et al. 2010). Additionally, the pre-supplementary motor area has been reported to encode elapsed time (Mita et al. 2009). Combined with these data, our results provide evidence that the SEF is involved in both predictive coding and predictive timing. Furthermore, our study also suggests that the signals generated according to these theories are important for making subsequent behavioral adjustments.

Effects of feedback signals on outcomes and surprise neurons

We found a substantial difference in the number of neurons and the activity patterns of the error- and correct-outcome neurons. Two possibilities for this difference are discussed below. The first possibility is that these differences were due to variations in the visual stimuli that signal error and correct feedback. Error feedback was signaled by the disappearance of visual stimuli, whereas correct feedback was signaled by a change in the color of a fixation target. The second possibility involves the difference in behavioral context. In our task, error feedback was critical when exploitation was switched to exploration; in contrast, correct feedback was related to the maintenance of current exploitation and the switch from exploration to exploitation. The context-dependent difference in response to error or correct feedback is consistent with the previous finding that error/correct-related activities were influenced by the likelihood of error in the behavioral tasks (Jessup et al. 2010; Wessel et al. 2012).

We found 15 error-surprise neurons (10% of all feedback-related neurons) and 14 correct-surprise neurons (10%). These numbers seem small compared with the numbers reported in previous

studies showing reward-prediction errors in the ACC (Hayden et al. 2011) and LPFC (Asaad and Eskandar 2011; Seo and Lee 2007). However, our result is comparable with other studies evaluating the SEF (So and Stuphorn 2012) and ACC (Matsumoto et al. 2007). The differences in the numbers of neurons that encode reward-prediction errors may be explained by the type of feedback. Like the present study, the research conducted by So's and Matsumoto's groups investigated neuronal activities during the post-feedback period, and feedback was provided in the form of visual signals. In contrast, the feedback in the former studies was an actual reward. Thus, differences in the type of feedback may be a key contributor to the responsiveness of cortical neurons to unexpected outcomes.

Limitations in the original model and alternatives to the model

We found that the correct-surprise and correct-outcome neurons were suppressed strongly in response to error feedback. These neurons were categorized as correct-outcome or surprise neurons because we fitted each model to the post-feedback activity according to its relative firing rates in correct and error trials. This result raises the possibility that SEF neurons change their response preferentially to behavioral errors rather than to correct feedback. This preference for error responses in the SEF is consistent with findings of the dominance of error-related negativity by event-related potential studies (Falkenstein et al. 1991; Gehring et al. 1993; Holroyd and Coles 2002).

The correct-surprise neurons showed seemingly “*signed*” responses to the feedback (Fig. 6C,D),

because they exhibited error-related suppression. However, this activity was explained better by the model of “*unsigned*” surprise signals. Indeed, the difference between actual and expected outcomes was reflected only in an increase in the activity and not in a reduction in the activity of the correct-surprise neurons. To resolve this apparent discrepancy, we hypothesized an additional model of correct-surprise signals based on a combination of the two types of outcome activity (Fig. 9). In this model, the correct-outcome value was increased in response to correct feedback, and the error-outcome value was reduced in response to error feedback. Previous studies have demonstrated that SEF neurons reflect both types of outcome values, as predicted by this model (Amador et al. 2000; Roesch and Olson 2003; So and Stuphorn 2012; Stuphorn et al. 2000). New types of unsigned surprise signals should be investigated in the future.

Conclusion

We found two types of surprise neurons in the SEF during the two-target oculomotor search task. Error-surprise neurons responded to unexpected error feedback and were involved in the transition from exploitation to exploration. Correct-surprise neurons responded to unexpected correct feedback and were involved in the transition from exploration to exploitation. Our results suggest that the SEF contributes to the control of exploration–exploitation transitions by detecting unexpected behavioral outcomes.

642 **Appendix**

643 In this section, we present the values used in and theoretical background of our model.

644 *Outcome values.* Outcome values represent whether a specific outcome was fed back. Outcome
 645 values at time t were divided into two types: the outcome value of correct feedback ($OV_{c,t}$) and the
 646 outcome value of error feedback ($OV_{e,t}$). We set the $OV_{c,t}$ as 1 and the $OV_{e,t}$ as 0 for correct
 647 feedback. In contrast, we set the $OV_{c,t}$ as 0 and the $OV_{e,t}$ as 1 for error feedback.

648

649 *Prediction values.* Prediction values represent the probability that a specific outcome occurred at
 650 time t . Prediction values were divided into two types: the prediction value of correct feedback ($PV_{c,t}$)
 651 and the prediction value of error feedback ($PV_{e,t}$). $PV_{c,t}$ and $PV_{e,t}$ represented the subjective
 652 probability of correct and error feedback at time t , respectively. These values were updated at each
 653 time t using a temporal difference algorithm (Alexander and Brown 2011; Sutton and Barto 1998).
 654 Each iteration (dt) represents 10 ms of real time. The prediction values at time t ($PV_{i,t}$) were
 655 computed as

$$656 \quad PV_{i,t} = \sum X_t \times W_t$$

657 where X_t is a set of task-related stimuli modeled as a temporal delay chain. W_t is a learned weight
 658 and constrained by $W_t \rightarrow 0$. W_t was computed as follows:

$$659 \quad W_{t+1} = W_t + \alpha \delta_t \varepsilon_t$$

660 where α is the learning rate ($0.1 \leq \alpha \leq 0.3$) and ε_t is an eligibility trace at time t . $\delta_{c,t}$ and $\delta_{e,t}$ were the
 661 temporal difference errors in correct or error feedback, respectively, at time t and were defined as

662 follows:

$$663 \quad \delta_{c,t} = OV_{c,t+1} + \gamma PV_{c,t+1} - PV_{c,t}$$

$$664 \quad \delta_{e,t} = OV_{e,t+1} + \gamma PV_{e,t+1} - PV_{e,t}$$

665 where $OV_{c,t}$ and $OV_{e,t}$ represent occurrences of correct and error feedback, respectively, at time t .

666 $PV_{c,t}$ and $PV_{e,t}$ are the prediction values of correct and error feedback, respectively, at time t . γ is a

667 temporal discounting factor ($0 \leq \gamma \leq 1$). $PV_{c,t}$ and $PV_{e,t}$ reflect the subjective probability of correct and

668 error feedback based on the histories of outcomes, regardless of actions.

669

670 *Multiple prediction errors.* In this model, signed and unsigned prediction errors were computed

671 based on differences between predicted and actual outcomes. First, we defined signed prediction

672 errors for correct and error feedback (*signed $PE_{c,t}$* and *signed $PE_{e,t}$*) as follows:

$$673 \quad \text{signed: } PE_{c,t} = OV_{c,t} - PV_{c,t}$$

$$674 \quad \text{signed: } PE_{e,t} = OV_{e,t} - PV_{e,t}$$

675 where $OV_{c,t}$ and $OV_{e,t}$ represent occurrences of correct and error feedback, respectively, at time t .

676 $PV_{c,t}$ and $PV_{e,t}$ are prediction values for correct and error feedback, respectively, at time t .

677 Next, we defined surprise signals (SS) as rectified unsigned prediction errors. The surprise signals

678 were divided into correct- or error-responsive types. An error-surprise signal ($SS_{e,t}$) was generated

679 when an unexpected error outcome was received at time t . A correct-surprise signal ($SS_{c,t}$) was

680 generated when an unexpected correct outcome was received at time t . Thus, these two values

681 were produced to respond to the unpredicted outcome of correct and error feedback at time t . The
 682 values were defined as follows:

$$683 \quad SS_{c,t} = [OV_{c,t} - PV_{c,t}]^+$$

$$684 \quad SS_{e,t} = [OV_{e,t} - PV_{e,t}]^+,$$

685 where $[OV_{c,t} - PV_{c,t}]^+$ and $[OV_{e,t} - PV_{e,t}]^+$ are the rectified positive parts of the prediction errors. $OV_{c,t}$
 686 and $OV_{e,t}$ represent occurrences of correct and error feedback, respectively, at time t . $PV_{c,t}$ and $PV_{e,t}$
 687 are prediction values of correct and error feedback, respectively, at time t . The $SS_{c,t}$ and $SS_{e,t}$ are
 688 high when the prediction values are low and unpredicted feedback occurs. These values were
 689 shown in Figure 3. The values were emphasized for visualization purpose.

690

Figure legends

Figure 1. Hypothetical function of SEF neurons in the exploration–exploitation transitions during an oculomotor search

We hypothesized that the SEF not only detects the feedback signals but contributes to the subsequent behavioral adaptation. During exploitative oculomotor search, the subject predicts correct feedback. When unpredicted-error feedback occurs, the SEF surprise neuron detects an unpredicted error and produces an error-surprise signal to encourage the subject to shift from exploitative to exploratory behavior. During exploratory search, the subject predicts error feedback. When unexpected correct feedback is delivered, the SEF neuron produces a correct-surprise signal, leading the subject to engage in exploitative behavior.

Figure 2. Oculomotor two-target search task

A: Sequence of events in two consecutive trials. The monkey chose one of four identical targets (Top-left [TL], Top-right [TR], Bottom-left [BL], Bottom-right [BR]) by making a saccade to and fixating on it. When a valid target was selected, it turned green and a reward was given. When an invalid target was selected, the four targets disappeared, and the same trial was repeated until the valid target was selected. The monkeys were required to choose two targets alternately. In this figure, the valid-target pair consisted of the two targets, “Bottom-right” and “Bottom-left”. The monkey had to select the target “Bottom-right” (upper row) and then select the target “Bottom-left” in the next trial (lower row). **B:** Four types of valid-target pairs. The current valid pair (two stimuli enclosed by the

dotted line) was switched to the other one in a pseudorandom manner. **C:** Three types of oculomotor search patterns. The patterns were defined by the relative location of the currently chosen stimulus compared with the location of the stimulus that was chosen in the previous trial. **D:** Examples of behaviors before and after a valid-pair change. The monkey made alternate saccades to two valid targets and received rewards in consecutive trials (exploitation phase). When the valid pair changed, the monkey made an error and began to search for new valid targets (exploration phase). Within several trials, the monkey found the valid pair of targets and began to select the two stimuli alternately to gain the reward. Adj: adjacent search pattern. Diag: diagonal search pattern.

Figure 3. Model of the error- and correct-surprise signals

The surprise signal (SS) was calculated by the two basic values, the prediction value (PV) and the outcome value (OV). The OV represents whether preferred feedback occurred (framed by a gray line). The PV represents the subjective probability of the preferred feedback. We categorized the PV into two types: higher and lower probability. The OV shows transient increases during feedback, and the PV shows increases before feedback and decreases after feedback. Dotted lines indicate the timing of the feedback. Both outcome values and surprise signals were divided into two types depending on the preference for correct or error feedback. **A:** Simulated value of the error-surprise signal (SS_e). The value is the rectified (positive) portion of the prediction error produced by subtracting PV_e from OV_e . The SS_e is high when the error feedback was not expected but an error occurred (blue, framed by a black line). The post-feedback response is lower when the error

feedback occurred repeatedly (sky blue). **B:** Correct-surprise signal (SS_e). The value is the rectified portion of the prediction error produced by subtracting PV_c from OV_c . The SS_c is high when the correct feedback was expected but an error occurred (red, framed by a black line). The response was lower when correct feedback was expected and occurred repeatedly (magenta).

Figure 4. Behavioral results

A: The oculomotor search patterns depended on exploration and exploitation phases. The relative incidences of the specific search patterns of two monkeys were plotted for each phase. The adjacent search pattern was dominant during the exploitation phase. The diagonal search pattern was common during the exploration phase. **B:** The search patterns in subsequent trials ($n+1$) depended on the combination of the outcomes in the current (n) and previous ($n-1$) trials. The diagonal search pattern was frequent after an error was fed back in the current (n) trial, especially after feedback to a correct response was given in the previous ($n-1$) trials (correct \rightarrow error). The adjacent search pattern was dominant after feedback to a correct response was given in the current (n) trial, irrespective of the outcomes in the previous ($n-1$) trials (correct \rightarrow correct and error \rightarrow correct). The adjacent search pattern was divided into groups according to whether the selected target was included in the current pair (other pair).

Figure 5. Outcome neurons

A: An example of an error-outcome neuron in the SEF. This neuron responded to error feedback, regardless of the outcome of the preceding trial. This neuron did not respond to correct feedback. **B:**

753 Population-level activity of the error-outcome neurons. **C**: An example of a correct-outcome neuron.
 754 This neuron showed increasing activity in response to correct feedback, irrespectively of the
 755 outcomes in the previous trials. **D**: The second example of correct-outcome neuron. This neuron
 756 showed reduced activity in response to error feedback. **E**: Population-level activity of the
 757 correct-outcome neurons. All firing rates were normalized, and the average z-scores were aligned to
 758 error or correct feedback (FB, gray dotted lines), respectively. Colored dotted lines in **B** and **E**
 759 indicate the standard errors of the mean z-scores.

760

761 **Figure 6. Surprise neurons**

762 **A**: An example of error-surprise neurons in the SEF. This neuron showed elevated activity when an
 763 unpredicted error was fed back (correct→error). This post-feedback response decreased when the
 764 error was repeated in successive trials (error→error). **B**: Population-level activity of the
 765 error-surprise neurons. **C**: An example of a correct-surprise neuron. This neuron showed the
 766 highest activity when unpredicted correct feedback was fed back (error→correct). This activity
 767 decreased when correct feedback was repeated (correct→correct). **D**: Population-level activity of
 768 the correct-surprise neurons. Conventions are the same as in Figure 5.

769

770 **Figure 7. Relationship between neuronal activity and search patterns**

771 **A**: Averaged activity of the error-surprise neurons. The averaged firing rates in the following trials
 772 were plotted as follows: the first error trial after a valid-target change (1st E), the first correct trial after

773 a pair change (1st C), the second correct trial after a pair change (2nd C), and the correct trial after
 774 consecutive correct trials (CC). Dotted lines show the timings of valid-pair changes. The activity was
 775 high at the start of exploration after the valid-pair changes (1st E vs. CC, $*p = 0.012$). The activity in
 776 the first error trial was higher than was that in the first correct and the second correct trials (1st E vs.
 777 1st C, $*p = 0.029$; 1st E vs. 2nd C, $*p = 0.025$). **B**: Averaged activity of the correct-surprise neurons.
 778 The activity level of correct-surprise neurons was high in the first correct trial after the valid-pair
 779 changes ($*p = 0.040$). **C–G**: Post-feedback neuronal activities and the subsequent search patterns.
 780 **C**: Enhanced activity of the error surprise-neurons was related to the diagonal search pattern in the
 781 next trials (paired *t*-test; $*p = 0.040$). **D**: A higher level of activity of the correct-surprise neurons was
 782 related to the adjacent search pattern in the subsequent trials (paired *t*-test; $*p = 0.010$). **E,F**: The
 783 activities of the error-outcome (**E**) and correct-outcome neurons (**F**) were not related to the search
 784 patterns in the next trials (paired *t*-test; $p \rightarrow 0.05$). All error bars indicate the standard error of the
 785 mean. **G,H**: Choice probabilities for error-surprise and error-outcome neurons (**G**) and for
 786 correct-surprise and correct-outcome neurons (**H**). Bar graphs show the mean choice probability
 787 value of each group. The choice probability of the error-surprise neurons was significantly higher
 788 than that of the error-outcome neurons ($**p = 0.0028$).

789

790 **Figure 8. Distribution of task-related neurons in the SEF**

791 Data are shown in the dorsal view of the frontal lobe. **A**: Task-related neurons. The positions of the
 792 gray dots indicate the locations of the task-related neurons of two monkeys relative to the gross

morphological landmarks of the midline and the genu of the arcuate sulcus (AS Genu). The sizes of the dots reflect the number of neurons recorded at each location. The dashed horizontal lines indicate the frontal level of the AS genu . **B**: Error-surprise neurons. The dots indicate the location of error-surprise neurons (orange) and error-outcome neurons (gray). **C**: Correct-surprise neurons. The dots indicate correct-surprise neurons (green) and correct-outcome neurons (grey). The relationship between the size of dots and the number of neurons is the same as in **B**.

Figure 9. Hypothetical model of correct-surprise signals

This surprise signal may explain the correct-surprise signals observed in Figure 6**C,D**. In this model, a differential response to correct and error feedback was hypothesized, and the surprise signals were associated with increasing activity in response to correct feedback and decreasing activity in response to error feedback. FB: feedback.

808 **Table 1. Categorized neurons (Monkey K+ monkey M).**

809

	Error	Correct
Outcome neuron	47 (44 + 3)	29 (28 + 1)
Surprise neuron	15 (13 + 2)	14 (13 + 1)
Prediction-error neuron	3 (2 + 1)	8 (8 + 0)
Prediction neuron	20 (20 + 0)	10 (8 + 2)
Subtotal	85 (79 + 6)	61 (57 + 4)
Total	146	

810

811

812 **Table 2. Categorized neurons with overlap**

813

	Outcome value	Surprise signal	Prediction error
Error-outcome neuron	(47)	35	36
Error-surprise neuron	10	(15)	11
Correct-outcome neuron	(29)	7	9
Correct-surprise neuron	12	(14)	1

814

815 The figures in parentheses represent the total numbers of categorized neurons in Table 1.

816

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- 970
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Figure 1

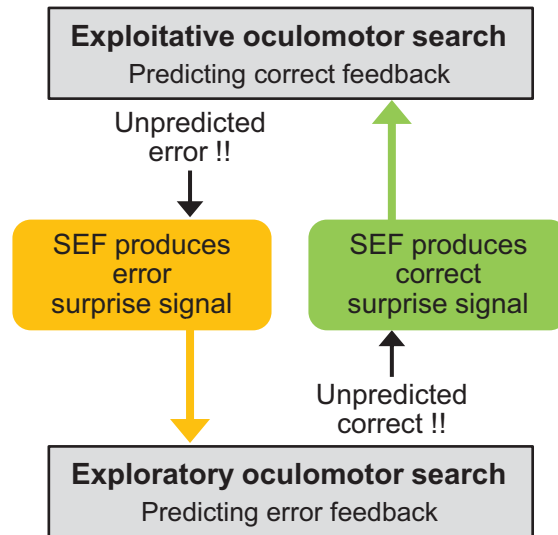


Figure 2

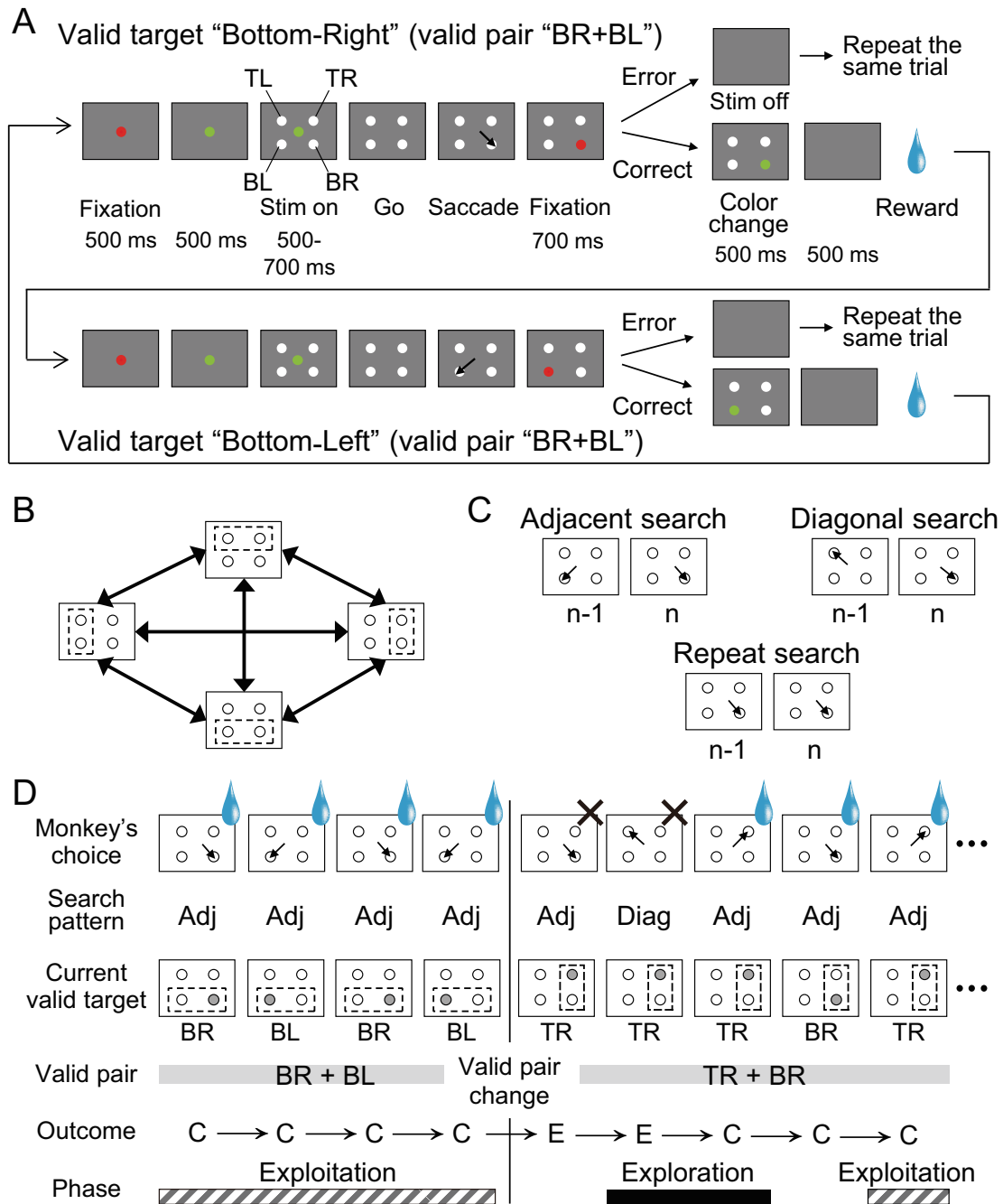
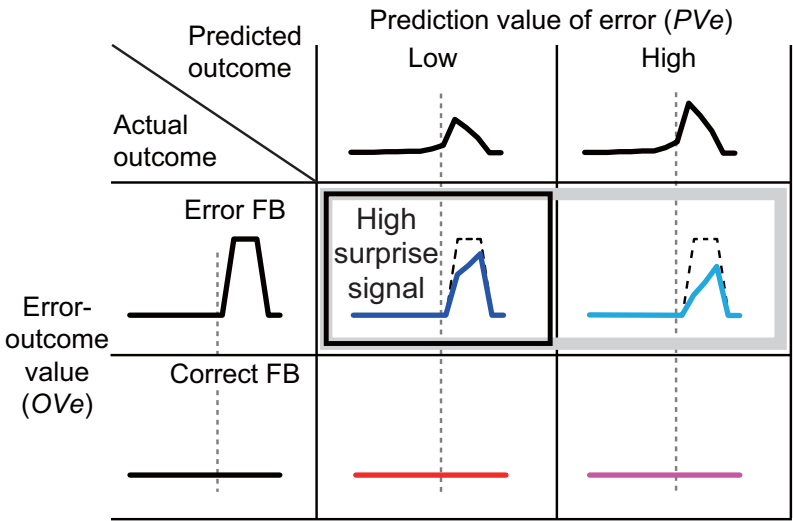


Figure 3

A Error-surprise signal



B Correct-surprise signal

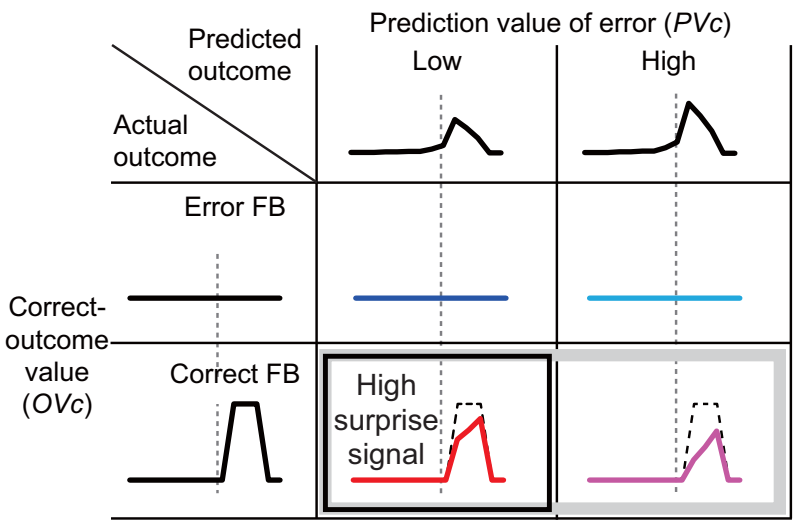


Figure 4

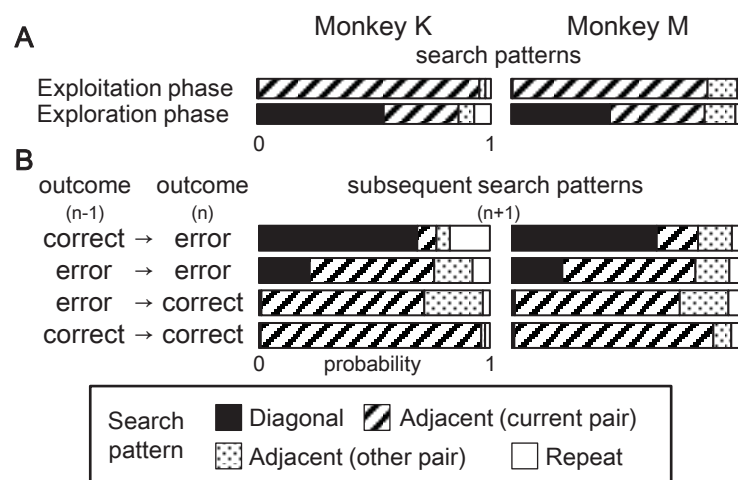


Figure 5

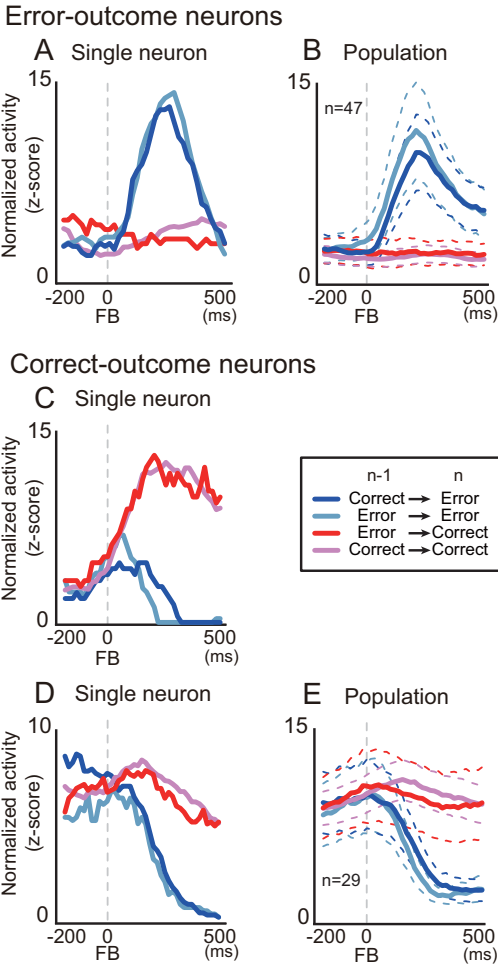


Figure 6

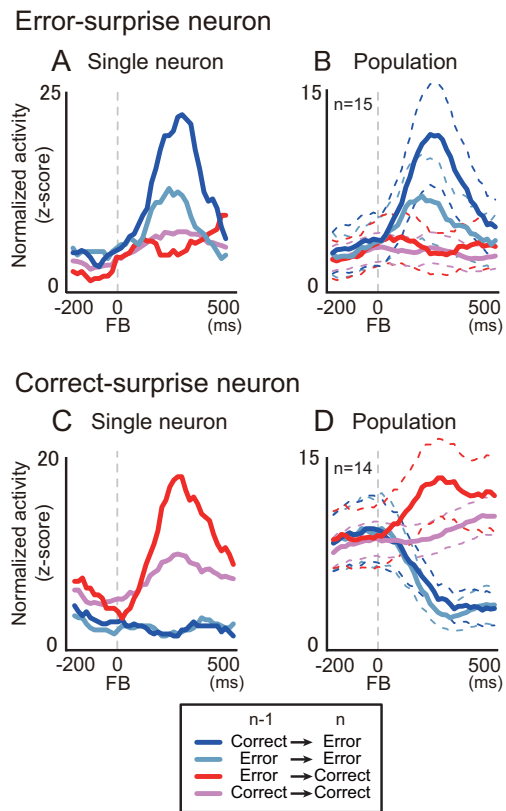


Figure 7

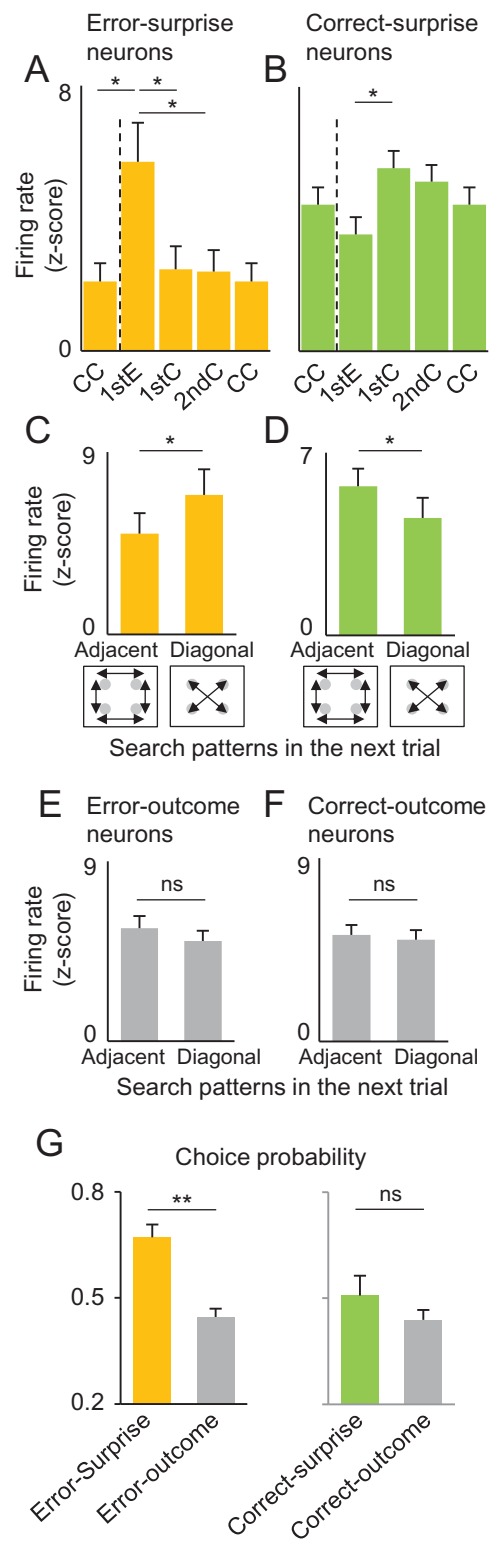


Figure 8

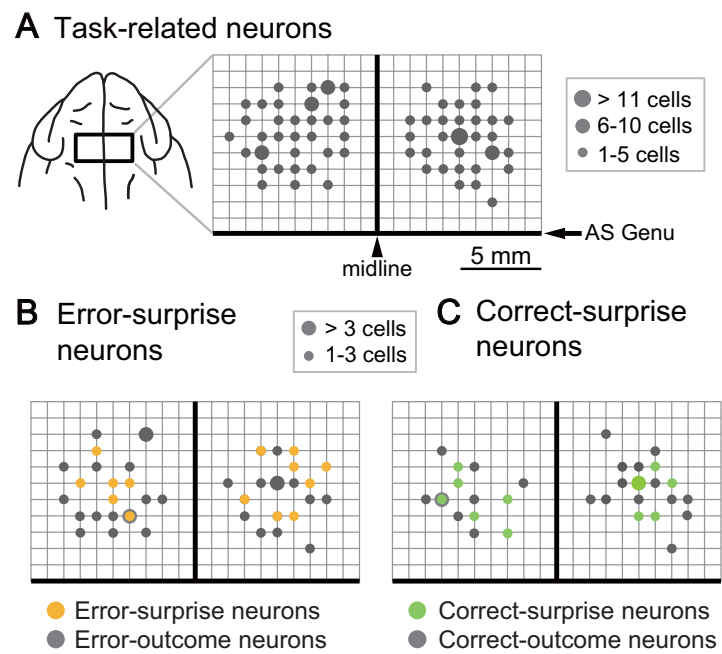


Figure 9

