

1 **The N300: An Index For Predictive Coding Of Complex Visual Objects and Scenes**

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3 Manoj Kumar, Kara D. Federmeier, Diane M. Beck

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5 Corresponding Author: Manoj Kumar

6 **Email:** mk35@princeton.edu

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8 **Supplementary Materials**

9 **Experiment 1**

10 **N300 Distributional Analysis**

11 To compare the N300 in our experiment with its characterization in the existing literature, we
12 performed an ANOVA in the time-window 250-350 ms with the factor of representativeness
13 (Good/Bad) and dividing the 16 scalp channels into 3 additional factors: 2 levels of Hemisphere
14 (right and left scalp sites), 2 levels of Laterality (lateral and medial scalp sites), and 4 levels of
15 Anteriority (prefrontal, frontal, central/parietal, and occipital scalp sites). In addition to confirming
16 the main effect of Good vs. Bad (bad larger than good) ($F(1,19) = 11.97$; $p = 0.0026$, $E = 1$), there
17 were interactions of Good/Bad with both Laterality ($F(1,19) = 7.35$; $p = 0.014$; $E = 1$) and Anteriority
18 ($F(3,57) = 27.86$; $p < 0.0001$; $E = 0.5336$), as well as three-way interactions: Good/Bad x
19 Hemisphere x Anteriority ($F(3,57) = 3.13$; $p = 0.0324$; $E = 0.8353$), and Good/Bad x Laterality x
20 Anteriority ($F(3,57) = 6.79$; $p = 0.0005$; $E = 0.7571$). Overall, N300 responses were observed over
21 both left and right hemisphere sites and were largest over the front of the head and larger over
22 medial compared to lateral electrode sites. Over frontal sites, the effect of Laterality was more
23 pronounced and there was a tendency for larger effects over left compared to right hemisphere
24 sites. These results show that the topographic distribution of the N300 effect for natural scenes
25 has a similar distributional profile as the N300 for objects (Schendan & Kutas, 2002, 2003,
26 2007).

27 **Post N300 Components**

28 Analyses of the N400 and LPC are presented in Tables S1 and S2.

29 **Table S1.** The grand average mean values in the N400 time-window (350-500 ms), shown for
 30 11 frontal electrode sites along with t-test and Bayes factor values. The N400 for the bad
 31 exemplars is larger (more negative) than that for the good exemplars. The t- test and Bayes
 32 factor calculations compared the within-subject Good/Bad difference to 0.

Condition	N	Mean (μV)	Bad/Good Difference Mean (μV)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
Bad	20	-3.3 \pm 0.96	-1.14	-1.78 to -0.50	-3.74	0.0014	27.3
Good	20	-2.2 \pm 0.96					

33 Note: \pm values reflect the normed standard deviation within subjects.

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35 **Table S2:** The grand average mean values, in the LPC time-window (500-800 ms), shown for
 36 15 posterior electrode sites (LMCe, RMCE, LDCe, RDCe, MiCe, MiPa, LLTe, RLTe, LDPa,
 37 RDPa, LLOc, RLOc, LMOc, RMOc, MiOc) along with t-test and Bayes factor values. The LPC
 38 for bad exemplars has a smaller mean amplitude than for the good exemplars. The t- test and
 39 Bayes factor calculations compared the within subject Good/Bad difference to 0.

Condition	N	Mean (μV)	Bad/Good Difference Mean (μV)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
Bad	20	3.3 \pm 1.1	-1.17	-1.91 to -0.43	-3.3	0.0036	12.1
Good	20	4.5 \pm 1.1					

40 Note: \pm values reflect the normed standard deviation within subjects.

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42 **ERP Analysis Conditioned on Participants' Judgements**

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44 To examine the N300 effect conditionalized on participants' explicit judgements, we repeated
45 the Bayes factor analysis, but now only using trials wherein participants' judgments aligned with
46 the condition designation (i.e., good exemplars judged as good and bad exemplars judged as
47 bad). As can be seen in **Table S3**, using the same time window and same set of electrode sites,
48 we again find a good/bad N300 effect for this subset of trials (Bayes Factor = 5.4; $t = -2.89$, $p =$
49 0.009). The post-N300 components, the N400 and LPC, also show significant effects for this
50 subset of trials, albeit with reduced Bayes factors.

51 **Table S3.** The grand average mean values along with t-test and Bayes factor values, only for
52 trials in which participants marked good exemplars as good or bad exemplars as bad, for the
53 N300 and post N300 components. This analysis was carried out at identical electrode sites to
54 the corresponding analyses in Tables 1, S1, and S2. The t- test and Bayes factor calculations
55 compared the within-subject Good/Bad difference to 0.

56

ERP	Condition	N	Mean (μ V)	Bad/Good Difference Mean (μ V)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
N300	Bad	20	-6.4 \pm 1.03					
	Good	20	-5.4 \pm 1.03	-0.95	-1.63 to -0.26	-2.89	0.0094	5.4
N400	Bad	20	-3.2 \pm 1.51					
	Good	20	-2.1 \pm 1.51	-1.02	-2.01 to -0.02	-2.13	0.046	1.5

	Bad	20	3.2±1.67					
LPC	Good	20	4.4±1.67	-1.20	-2.3 to -0.09	-2.26	0.0359	1.8

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58

59 **Comparing the N300 to Bad Exemplars Judged as Good or Bad**

60 Bad exemplars (designated based on large-scale rating data) were explicitly judged to be bad
61 by participants in this study about half the time (mean = 56.2%, std. dev = 15.6%). We
62 compared the N300 amplitude to these exemplars based on participants' judgments and found
63 no evidence that the N300 differs for bad exemplar trials that subjects responded to as "bad"
64 (-6.35 μ V) vs. "good" (-6.37 μ V) (Bayes Factor = 0.23; $t(19) = 0.04$; $p = 0.97$). Participant
65 judgements also did not reliably modulate either the N400 (Bayes Factor = 0.23; $t(19) = 0.16$, p
66 = 0.88) or the LPC (Bayes Factor = 0.29; $t(19) = -0.71$; $p = 0.48$). Thus, we see no indication
67 that the ERP patterns were importantly affected by participants' explicit judgments, although we
68 note that this analysis is conducted on half the trials of the main analysis. Because judgments
69 for the good exemplars were much more consistent (mean judged to be "good" = 86.2%, std.
70 dev = 13.9%), there were insufficient trials to examine the impact of participant judgment for
71 these items.

72

73 **Experiment 2**

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75 **N300 Distributional Analysis with Cuing**

76 A novel feature of **Experiment 2** was the use of a verbal precue. To see how the distributional
77 properties of the N300 effect under cued conditions compare to that in prior work, we performed
78 an ANOVA of the Cueing (match/mismatch) and Good/Bad factors using the same electrode
79 sites as in the ANOVA analysis for **Experiment 1**, adding the identical electrode factors: 2

80 levels of Hemisphere (right and left scalp sites), 2 levels of Laterality (lateral and medial scalp
81 sites), and 4 levels of Anteriority (prefrontal, frontal, central/parietal, and occipital scalp sites).
82 We replicate the main effect of Good and Bad from **Experiment 1**, with larger N300 responses
83 to bad exemplars than to good; $F(1,19) = 15.34$; $p = 0.0009$, $E=1$. The topographic distribution of
84 this effect was also similar to that in **Experiment 1** and in the larger N300 literature: N300
85 effects were observed over both left and right hemisphere sites, with the left hemisphere
86 showing somewhat larger effects as compared to the right hemisphere (Good/Bad x
87 Hemisphere ($F(1,19) = 5$; $p = 0.0375$, $E=1$) and were larger over medial than lateral sites
88 (Good/Bad x Laterality ($F(1,19) = 21.1$; $p = 0.0002$, $E=1$) and larger over the front of the head
89 (Good/Bad x Anteriority ($F(3,57) = 7.13$; $p = 0.0004$, $E=0.4030$)).

90

91 We also found an interaction of Good/Bad x Cueing ($F(1,19) = 5.87$; $p = 0.0255$, $E=1$), with large
92 N300 effects when the stimuli matched the cue, but negligible effects in the mismatch condition.
93 There was a 3-way interaction of Good/Bad x Cueing x Laterality ($F(1,19) = 4.83$; $p = 0.0406$,
94 $E=1$), because the tendency for N300 effects to be larger over medial than lateral sites was
95 more apparent in the match condition, which showed strong N300 effects.

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97 **N400: The Effects of Cuing**

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99 The N400 is known to be affected by semantic expectancy. To confirm that pattern and examine
100 its interaction with Good/Bad status in **Experiment 2**, we performed an ANOVA analysis in the
101 N400 time-window (350-500 ms). Consistent with the larger literature, we found a main effect of
102 Cuing ($F = 5.43$; $p = 0.03$; $E = 1$) with a smaller N400 amplitude to stimuli that matched the cue
103 compared to the those that mismatched. We also found a significant interaction of Good/Bad x
104 Cuing ($F = 13.7$; $p = 0.0015$; $E = 1$), with the good exemplars in the match condition having the
105 smallest amplitude as compared to the good and bad exemplars in the match- and- mismatched

106 condition. Note that we do not claim that these effects in the N400 time window are completely
 107 independent of the preceding N300 effects in our experiment. Prior work has shown that the
 108 N300 and N400 are functionally dissociable (Federmeier & Kutas, 2002; Gratton et al., 2009),
 109 but their similar response pattern to, e.g., incongruent and congruent items, can make
 110 separating them challenging under some circumstances (Draschkow et al., 2018). We also
 111 computed the Bayes factor for the N400 in the match and mismatch conditions (**Table S4**) and
 112 we see strong evidence for the N400 in the match condition (Bayes factor 1287.4) as compared
 113 to the mismatch condition (Bayes factor 0.56).

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116 **Table S4.** The grand average mean values, in the N400 time-window (350-500 ms), shown for
 117 11 frontal electrode sites. The t- test and Bayes factor calculations compared the within subject
 118 Good/Bad difference to 0.

Condition	Cue	N	Mean (μ V)	Difference (μ V)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
Bad	Match	20	-5.3 \pm 0.98					
Good	Match	20	-3.5 \pm 1.41	-1.82	-2.5 to -1.15	-5.67	1.8E-05	1287.4
Bad	Mismatch	20	-4.7 \pm 1.65					
Good	Mismatch	20	-5.6 \pm 2.06	0.91	-0.42 to 2.25	-1.44	0.17	0.56

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120 Note: \pm values reflect the normed standard deviation within subjects.

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123 **LPC**

124 For the LPC, we examined differences in the 500-800 ms time window, encompassing the Late
 125 Positive Complex (LPC), over 15 posterior sites (identical to the sites for the analysis in
 126 **Experiment 1**). We replicate the main effect of Good/Bad with a larger LPC amplitude for good
 127 as compared to bad exemplars ($F = 4.84$; $p = 0.0403$; $E = 1$). The LPC is known to index
 128 confidence in decision making (Finnigan et al., 2002) and the main effect of Cuing in
 129 **Experiment 2** aligns with this understanding, with the match condition showing a larger LPC
 130 amplitude as compared to the mismatch condition ($F = 19.69$; $p = 0.0003$; $E = 1$). The interaction
 131 of Good/Bad x Cuing is also significant ($F = 19.82$; $p = 0.0003$; $E = 1$) with the good match having
 132 the largest LPC amplitude as compared to the Good mismatch, Bad match, and Bad mismatch
 133 conditions. We also computed the Bayes factor for the LPC in the match and mismatch
 134 conditions (Table S5) and we see strong evidence for the LPC in the match condition (Bayes
 135 factor 9014.4) as compared to the mismatch condition (Bayes factor 0.41).

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138 **Table S5.** The grand average mean values, in the LPC time-window (500-800 ms), shown for
 139 15 posterior electrode sites. The t- test and Bayes factor calculations compared the Good/Bad
 140 difference to 0.

Condition	Cue	N	Mean (μV)	Mean Bad/Good Difference (μV)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
Bad	Match	20	0.4±0.98	-2.24	-2.94 to -1.54	-6.69	2.1E-06	9014.4
Good	Match	20	2.7±1.09					
Bad	Mismatch	20	0.9±1.66	0.68	-0.56 to 1.93	1.15	0.26	0.41
Good	Mismatch	20	0.2±1.61					

141

142 Note: \pm values reflect the normed standard deviation within subjects.

143

144 ERP Analyses Conditioned on Participants' Judgements

145 As for Experiment 1, we also examined Good/Bad effects conditionalized on participants'
146 explicit judgements, only including exemplars on which subjects judgement was congruent with
147 the category cue; i.e., they responded to a cue congruent stimulus as 'Yes' and cue incongruent
148 stimulus as "No", for both good and bad exemplars. As can be seen in **Table S6**, we again
149 found that explicit judgments did not notably impact the ERP patterns.

150

151 **Table S6.** The grand average mean values along with t-test and Bayes factor values, only for
152 trials in which participants responded to a cue congruent stimulus as 'Yes' and cue incongruent
153 stimulus as "No", for both good and bad exemplars. This analysis was carried out at identical
154 electrode sites to the corresponding analyses in Tables 2, S1, and S2. The t- test and Bayes
155 factor calculations compared the within-subject Good/Bad difference to 0.

156

ERP	Condition	Cue	N	Mean (μ V)	Difference (μ V)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
N300	Bad	Match	20	-7.2 \pm 1.18					
	Good	Match	20	-5.1 \pm 1.24	-2.15	-2.98 to -1.33	-5.45	2.9E-05	835.8
	Bad	Mismatch	20	-6.5 \pm 1.68					
	Good	Mismatch	20	-5.9 \pm 1.48	-0.55	-1.72 to 0.63	-0.98	0.34	0.35
N400	Bad	Match	20	-5.3 \pm 1.18					
	Good	Match	20	-3.6 \pm 1.50	-1.71	-2.6 to -0.82	-4.02	0.0007	48.2

	Bad	Mismatch	20	-4.7±1.62					
	Good	Mismatch	20	-5.6±1.91	0.92	-0.37 to 2.21	1.49	0.15	0.6
LPC	Bad	Match	20	0.6±1.22					
	Good	Match	20	2.6±1.13	-2.05	-2.85 to -1.25	-5.35	3.6E-05	689.4
	Bad	Mismatch	20	1.0±1.76					
	Good	Mismatch	20	0.2±1.49	0.79	-0.42 to 2.01	1.36	0.19	0.5

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158 Note: ± values reflect the normed standard deviation within subjects.

159

160 **N300: Subsampling the Match Trials**

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162 The mismatch condition in **Experiment 2** had fewer trials and therefore possibly a lower signal-
 163 to-noise-ratio as compared to the match condition. To verify that we see evidence for the
 164 Good/Bad effect in the match condition even when trial numbers are equated to the mismatch
 165 condition, we randomly subsampled, in the match condition, 30 trials from the good exemplars
 166 and 30 trials from the bad exemplars and recomputed the statistics at identical frontal electrode
 167 sites. Even with just this subset of trials, we find that N300 amplitudes are larger for bad than for
 168 good exemplars in the match condition (Bayes Factor = 35.6; $t=-3.9$; $p=0.001$; **Table S7**).

169 Moreover, when we combine the sampled good data with the mismatch data, there is still an
 170 interaction between Cuing and the Good/Bad effect (Bayes Factor = 3.1), such that the
 171 Good/Bad effect is reduced under mismatch compared to match conditions.

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173

174 **Table S7.** The grand average mean values for the N300 and post N300 components for
 175 subsampled good and bad exemplars (30 trials each) in the match condition, computed at
 176 identical electrode sites as those used in **Table 2**, **S4** and **S5**. The t- test and Bayes factor
 177 calculations compared the within subject Good/Bad difference to 0.

178

ERP	Condition	Cue	N	Mean (μV)	Difference (μV)	Bad/Good Difference 95% C.I.	t(19)	p	Bayes Factor
N300	Bad	Match	20	-7.9 \pm 1.84	-2.42	-3.74 to -1.1	-3.9	0.001	35.6
	Good	Match	20	-5.5 \pm 1.77					
N400	Bad	Match	20	-5.8 \pm 1.53	-1.82	-3.03 to -0.62	-3.16	0.0051	8.99
	Good	Match	20	-4.0 \pm 1.89					
LPC	Bad	Match	20	0.3 \pm 1.79	-1.99	-3.18 to -0.81	-3.52	0.002	18.0
	Good	Match	20	2.3 \pm 1.61					

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180 Note: \pm values reflect the normed standard deviation within subjects.

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182 **Supplementary Materials: References**

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