SUPPLEMENTARY MATERIAL

TITLE: "Segmenting surface boundaries using luminance cues"

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	BIC: Additive		BIC: D	oivisive	BICD - BICA		
	None	Lapse	None	Lapse	None	Lapse	
CJD	-427.510	-430.820	-419.665	-423.020	7.845	7.800	
ERM	-448.073	-451.383	-438.469	-441.779	9.605	9.605	
KNB	-424.487	-427.797	-411.348	-414.658	13.139	13.139	

Table S1: Bayes Information Criterion (BIC) for fits of Additive and Divisive SDT models to data

 from **Experiment 3**, with (*Lapse*) and without (*None*) lapse rates.

Observer	neu	con	inc	χ^2	р
CJD	0.80	0.83	0.74	4.575	0.102
ERM	0.70	0.78	0.67	5.742	0.057
KNB	0.78	0.83	0.81	1.287	0.525
MCO	0.76	0.82	0.78	2.612	0.271
NRB	0.89	0.92	0.89	1.115	0.573
RCL	0.78	0.86	0.81	3.842	0.146
JCO	0.92	0.95	0.95	1.974	0.373
HAP	0.90	0.87	0.86	1.603	0.449
EMW	0.88	0.93	0.88	3.598	0.166
POOLED	0.82	0.86	0.82	15.319	<0.001***

* p < 0.05, ** p < 0.01, *** p < 0.001

Table S2: Statistical tests (Pearson's χ^2 , df = 2) of the hypothesis that the proportion correct (*Pc*) is identical for the *neutral* (**neu**), *congruent* (**con**), and *incongruent* (**inc**) conditions of **Experiment 4a**.

Observer	neu	con-0	χ^2	р	con-180	χ^2	р
CJD	0.80	0.90	4.800	0.028*	0.75	0.982	0.322
ERM	0.70	0.83	5.905	0.015*	0.72	0.129	0.720
KNB	0.78	0.84	1.500	0.221	0.81	0.362	0.548
MCO	0.76	0.83	2.185	0.139	0.81	1.15	0.283
NRB	0.89	0.96	4.579	0.032*	0.87	0.142	0.706
RCL	0.78	0.86	2.736	0.098	0.85	2.068	0.150
JCO	0.92	0.97	3.241	0.072	0.92	0.022	0.883
HAP	0.90	0.88	0.280	0.597	0.86	1.061	0.303
EMW	0.88	0.98	8.422	0.004**	0.88	0.000	1.000
POOLED	0.82	0.89	24.383	<0.001***	0.83	0.288	0.592

* p < 0.05, ** p < 0.01, *** p < 0.001

Table S3: Statistical comparisons (Pearson's χ^2 , df = 1) of observer performance in the neutral (**neu**) case of **Experiment 4a** with each sub-condition of the congruent trials: Aligned-phase (**con-0**) and opposite-phase (**con-180**).

Observer	con-0	con-180	χ^2	р		
CJD	0.90	0.75	7.792	0.005**		
ERM	0.83	0.72	3.470	0.063		
KNB	0.84	0.81	0.312	0.577		
MCO	0.83	0.81	0.136	0.713		
NRB	0.96	0.87	5.207	0.022*		
RCL	0.86	0.85	0.040	0.841		
JCO	0.97	0.92	2.405	0.121		
HAP	0.88	0.86	0.177	0.674		
EMW	0.98	0.88	7.689	0.006**		
POOLED	0.89	0.83	15.732	<0.001***		
* p < 0.05, ** p < 0.01, *** p < 0.001						

Table S4: Statistical tests (Pearson's χ^2 , df = 1) of the hypothesis that proportion correct is identical for phase-aligned (**con-0**) and opposite-phase (**con-180**) stimuli for the congruent case in **Experiment 4a**.

Observer	isolated	masked
CJD	0.0041 [0.0035, 0.0055]	0.0100 [0.0083, 0.0165]
KNB	0.0043 [0.0039, 0.0048]	0.0083 [0.0072, 0.0098]
MXD	0.0056 [0.0048, 0.0066]	0.0113 [0.0094, 0.0170]

Table S5: Median LSB segmentation thresholds and non-parametric 95% confidence intervals (200 bootstraps) when the LSB is presented in isolation (**isolated**) and in the presence of an uninformative LTB masker (**masked**) having no boundary (equal number of B and W micropatterns on each side). We see slightly higher LSB segmentation thresholds in the presence of the uninformative LTB masker, rising from about 0.5% to 1% Michelson Contrast.

Observer	neu	con	inc	χ^2	р
CJD	0.87	0.83	0.73	13.189	0.001**
KNB	0.86	0.85	0.83	0.933	0.627
MXD	0.90	0.90	0.87	0.829	0.661
POOLED	0.88	0.86	0.81	10.786	0.005**
* p < 0.05, ** p < 0.01, *** p < 0.001					

Table S6: Statistical tests (Pearson's χ^2 , df = 2) of the hypothesis that the proportion correct (*Pc*) is identical for the *neutral* (**neu**), *congruent* (**con**), and *incongruent* (**inc**) conditions of **Experiment 4b**.

Observer	neu	con-0	χ^2	р	con-180	χ^2	р
CJD	0.87	0.95	4.605	0.032*	0.70	12.69	< 0.001***
KNB	0.86	0.89	0.531	0.466	0.80	1.786	0.181
MXD	0.90	0.94	1.648	0.199	0.85	1.278	0.258
POOLED	0.88	0.93	5.552	0.018*	0.78	12.794	<0.001***
* p < 0.05, ** p < $\overline{0.01}$, *** p < 0.001							

Table S7: Statistical comparisons (Pearson's χ^2 , df = 1) of observer performance in the neutral (**neu**) case of **Experiment 4b** with each sub-condition of the congruent trials: aligned-phase (**con-0**) and opposite-phase (**con-180**).

Observer	con-0	con-180	χ^2	р	
CJD	0.95	0.70	21.645	< 0.001***	
KNB	0.89	0.80	3.092	0.079	
MXD	0.94	0.85	4.310	0.038*	
POOLED	0.93	0.78	24.857	<0.001***	
* p < 0.05, ** p < 0.01, *** p < 0.001					

Table S8: Statistical tests (Pearson's χ^2 , df = 1) of the hypothesis that proportion correct is identical for phase-aligned (**con-0**) and opposite-phase (**con-180**) stimuli for the congruent case in **Experiment 4b**.

Observer	BIC ₁	BIC ₂	$BIC_2 - BIC_1$	IE ratio	DC response
CJD	-334.993	-316.758	18.235	0.00	26.7473
ERM	-375.381	-367.651	7.730	0.20	8.8932
KNB	-344.449	-309.400	35.049	0.25	4.4296
MCO	-353.603	-325.880	27.723	0.30	-0.0339
NRB	-277.090	-208.152	68.938	0.30	-0.0339
RCL	-312.702	-299.927	12.775	0.25	4.4296
JCO	-161.696	-156.019	5.6770	0.05	22.2837
HAP	-273.753	-236.640	37.113	0.20	8.8932
EMW	-267.175	-208.309	58.866	0.30	-0.0339

Table S9: Bayes Information Criteria (BIC) for model fits to data from Experiment 4a, for both one-stage (BIC1) and two-stage (BIC2) models. The two rightmost columns show the optimal values of the IE ratio for the first-stage filters in the two-stage model, as well as the DC response of this first-stage filter to uniform unit luminance.

Observer	BIC ₁	BIC ₂	$BIC_2 - BIC_1$	IE ratio	DC response
CJD	-320.289	-287.982	32.307	0.05	22.2837
KNB	-318.614	-270.836	47.778	0.25	4.4296
MXD	-287.426	-222.493	64.933	0.20	8.8932

Table S10: Same as Table S9 but for Experiment 4b.

SUPPLEMENTARY FIGURE CAPTIONS

Figure S1: Effects of density on LTB segmentation thresholds

Effects of micro-pattern density on segmentation thresholds for experienced observers (author CJD and naïve observers KNB, ERM). Plotted are means (circles) and 1 SEM (lines) obtained from N = 200 bootstrapped fits of the SDT psychometric function. We see that thresholds are slightly higher for 16 micro-patterns, and that similar performance is obtained for 32 and 64 micro-patterns.

Figure S2: Thresholds estimated with and without lapse rates

(a) Thresholds from **Experiment 1a**. We see nearly identical threshold estimates whether or not lapse rates are included in our PF definitions. (b) Thresholds from **Experiment 1b**.

Figure S3: Contrast thresholds for LSB segmentation

(a) Fits of SDT psychometric function to LSB segmentation performance for same observers shown in Fig. 3a. (b) Histogram of threshold for all observers.

Figure S4: Fits of one-stage model to LSB segmentation performance

Same as **Fig. 4b** in main text, but shows fits of the one-stage model to LSB segmentation data from **Experiment 1b**.

Figure S5: Fits of additive and divisive one-stage model

(a) Same as **Fig. 6b**, but with lapse rates estimated. (b) Divisive model fit to **Experiment 3** data accurately predicts performance in **Experiment 2**.

Figure. S6: Fits of models to Experiment 4a data

Same as Fig. 8b in main text, but for remaining observers.

Fig S7: Fits of the two-stage model (Fig. 8a) to Experiment 3 data

Fits of both additive and divisive instances of the two-stage model shown in **Fig. 8a** to psychophysical data obtained in **Experiment 3**. As with the one-stage model (**Fig. 6**), we find much better fits with the divisive two-stage model.

Fig S8: Results from Experiment 4b

- (a) Same as Fig. 7a in main text for Experiment 4b.
- (b) Same as Fig. 7b in main text for Experiment 4b.
- (c) Same as Fig. 8b in main text for Experiment 4b.















