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Supplementary Materials

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Figure S1. Behavioral evaluation including multiple word lengths. (a) Response times aggregated 1396 across participants from the British lexicon (BLP) project (Keuleers et al., 2012) for the word 1397 lengths 4-8. The left panel shows the word/non-word by orthographic prediction error (oPE) 1398 1399 interaction and the right panel shows the word/non-word by number of pixels (Npixel) interaction for each word length separately. In addition, the upper panel shows letter strings that are correctly 1400 categorized in nearly all cases (accuracy > .95) and the lower panel shows the response times to 1401 the items, which were less accurately processed (i.e., accuracy < .95). The median split resulted 1402 in a subset of the BLP (i.e., the easy words) which are roughly comparable to words used in the 1403 previous experiments (e.g. see Fig. 3), as the BLP study includes a large number of very rare 1404 words (median log. word frequency per million is .3). Bluish colors represent non-words (N) and 1405 greenish colors represent words (W), while the hue of the colors reflects word length (i.e., bright 1406 1407 to dark reflects short to long letter strings). For both effects, we first estimated linear regression models with either the oPE or the Npixel effect and allowing interactions with word/non-word 1408

1409 status, word length, and accuracy. Note that the oPE in this first analysis was based on lengthspecific predictions (i.e., for the estimation of the oPE of four-letter words, all four-letter words 1410 of the lexicon were included in the prediction). For the oPE model, a significant four-way 1411 1412 interaction was found (estimate = -1.078e-04; SE = 4.199e-05; t = -2.567). Separating hard vs. easy words allowed us to disentangle the four-way interaction: In easy words/non-words, we 1413 found a consistent (i.e., across length levels) oPE by word/non-word interaction (estimate = 1414 1.530e-04; SE = 4.047e-05; t = 3.780) in the same direction as previously shown (positive effect 1415 1416 for words and a negative effect for non-words). For hard words/non-words, we found that the oPE by word/non-word interaction was inconsistent across letter length levels, which was 1417 1418 indicated by a significant oPE and letter length interaction (estimate = -3.530e-05; SE = 8.092e-06; t = -4.363). In addition, for the hard words both the oPE by word/non-word interaction 1419 1420 (estimate = -1.685e-04; SE = 6.905e-05; t = -2.440) and the main effect of oPE were reversed (estimate = 2.828e-04; SE = 5.802e-05; t = 4.874 compare to estimate = -1.000e-04; SE = 2.440e-041421 1422 05; t = -4.101, for easy words). For the Npixel model, no four-way interaction and no Npixel interaction or main effect were found. In sum, in this analysis we showed that the oPE by 1423 1424 word/non-word interaction shown previously for word lengths of five letters (see main text) is consistent for easy-to-process English items with word lengths from 4-8 letters. Secondly, the 1425 word/non-word by orthographic prediction error interaction was also reliable when the prediction 1426 1427 included all words of all letter lengths from the English lexicon (see part b of this Figure) and the orthographic prediction error estimation was based on this length-unspecific prediction (estimate: 1428 0.02; SE=0.007; t=3.349). (b) Letter-length unspecific prediction for English based on ~60,000 1429 1430 English words from the SUBTLEX database (Heuven et al., 2014).

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Figure S2. Dutch lexical decision behavior and prediction using a proportional script. (a) Effect 1433 of the orthographic prediction error parameter, (b) number of pixels parameter and (c) showing 1434 the same model comparisons as implemented in Figure 3 for the data from the first Dutch lexicon 1435 project (DLP1; (Keuleers, Diependaele, & Brysbaert, 2010); 4,305 five-letter stimuli; 39 1436 1437 participants) and the same effects and model comparisons for the second Dutch lexicon project (DLP2; (Brysbaert, Stevens, Mandera, & Keuleers, 2016); 3,145 five-letter stimuli; 81 1438 participants) are presented in (d,e,f). Before going into the details of the two studies one has to 1439 1440 note that the patterns we have found in the data in relation to our parameters of interest do not 1441 replicated within these two Dutch studies and, in addition, do not replicate with the findings from German, English, and French shown in Figure 3. In general, this is difficult for the interpretations 1442 1443 of the results. For the DLP1 pattern we found a significant interaction of the orthographic prediction error with word/non-words and no significant effect of number of pixels. The 1444 interaction pattern in contrast to the findings in other languages (Fig. 3a), however, was 1445 qualitatively different as it showed a negative orthographic prediction error effect for words and a 1446 positive effect for non-words. The pattern is exactly the inverse from all other languages. Still 1447 model comparisons highlighted that the orthographic prediction error was relevant for the model 1448 1449 fit since the predictor increased the model fit with no further increase of fit when the number of pixel parameter was included. None of these findings could be replicated in the DLP2 dataset, 1450 1451 showing no significant fixed effects or interactions and no substantial changes in model fit

- relation to the null model. (g) Prediction image from a PEMoR implementation using five-letter words with a proportional Times New Roman script. 1452
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Figure S3. Detailed description of significant activation clusters in the EEG study for (a) the
orthographic prediction error; (b) word/non-word effect; (c) interaction of word/non-word and the
orthographic prediction error. On the left, the effect sizes from regression ERPs are presented as
time courses for each sensor and time-point (color coding reflects scalp position). This part of the
Figure reproduces Figure 5. The right column displays time courses with one line per channel,
masked by significance using cluster statistics (see Methods for details; Maris & Oostenveld,
2007).

Table S1. Results from linear mixed model regression analysis (with the exception of the British data including multiple word lengths was estimated based on word aggregated data) for the behavioral lexical decision tasks (LDT) and handwriting analyses.

	E	SE	t	
German LDT N°1: Orthographic p	rediction	error bas	ed on	
word length specific prediction				
Intercent	6.49	0.023	288	
mercept	-0.03	0.004	65	
Orthographic prediction error (oPE)	-0.03	0.004	0.5	
Number of pixels (Npixel)	-0.007	0.004	1.8	
Word/non-word (Lex)	0.33	0.009	33.1	
Word frequency	-0.12	0.004	33.5	
Error	-0.03	0.005	6.2	
oPE X Lex	0.03	0.006	5.0	
Npixel X Lex	0.000	0.006	0.1	

German LDT N°1: Orthographic prediction error based on word length general prediction

Intercept	6.48	0.023	288.3
Orthographic prediction error (oPE)	-0.03	0.004	6.3
Number of pixels (Npixel)	-0.01	0.004	1.7
Word/non-word (Lex)	0.33	0.010	33.2
Word frequency	-0.12	0.004	35.5
Error	-0.03	0.005	6.2
oPE X Lex	0.03	0.006	4.5
Npixel X Lex	-0.00	0.006	0.0

German LDT N°1: Orthographic prediction error based on word length specific prediction including orthographic Levenshtein distance and word frequency

Levenshtein distance and v	vord freq	uency	
Intercept	6.66	0.023	237.1
Orthographic prediction error (oPE)	-0.02	0.004	4.3
Number of pixel (Npixel)	-0.00	0.004	0.2
Word/non-word (Lex)	0.29	0.011	27.0
Error	-0.03	0.005	6.2
Orthographic Levenshtein distance	-0.08	0.008	10.5
Word frequency	-0.12	0.004	35.5
oPE X Lex	0.03	0.006	5.2
Npixel X Lex	-0.00	0.005	0.6

German LDT N°2 including noise: 0%			
Intercept	6.32	0.024	263.9
Orthographic prediction error (oPE)	-0.02	0.016	1.4
Number of pixels (Npixel)	-0.00	0.015	0.2
Word/non-word (Lex)	0.27	0.05	5.4

Word frequency	-0.07	0.02	4.9
Error	-0.07	0.010	6.8
oPE X Lex	0.05	0.02	2.3
Npixel X Lex	-0.02	0.021	1.2

German LDT N°2 including noise: 20%

Intercept	6.45	0.026	245.4
Orthographic prediction error (oPE)	-0.06	0.017	3.3
Number of pixels (Npixel)	-0.00	0.013	0.3
Word/non-word (Lex)	0.37	0.049	7.5
Word frequency	-0.14	0.02	6.1
Error	-0.14	0.010	5.4
oPE X Lex	0.04	0.022	1.6
Npixel X Lex	0.02	0.022	0.7

German LDT N°2 including noise: 40%

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Intercept	6.84	0.042	162.9
Orthographic prediction error (oPE)	-0.02	0.021	1.0
Number of pixels (Npixel)	-0.08	0.018	4.1
Word/non-word (Lex)	0.14	0.049	2.8
Word frequency	-0.11	0.06	1.9
Error	-0.00	0.010	0.1
oPE X Lex	-0.00	0.028	0.1
Npixel X Lex	0.08	0.026	2.9

British LDT

Intercept	6.39	0.013	507.1
Orthographic prediction error (oPE)	-0.007	0.001	5.3
Number of pixels (Npixel)	0.008	0.001	6.7
Word/non-word (Lex)	0.12	0.003	46.2
Word frequency	-0.067	0.001	58.0
oPE X Lex	0.008	0.002	4.2
Npixel X Lex	-0.003	0.002	1.9

British LDT 4-8 Letters: Length specific prediction

8	1	1	
Intercept	6.26	0.157	39.7
Orthographic prediction error (oPE)	-0.001	0.000	5.0
Number of letters (Nletters)	0.062	0.027	2.3
Word/non-word (Lex)	0.155	0.162	0.3
Error	0.043	0.165	0.8
oPE X Lex	-0.001	0.000	4.5
oPE X Nletters	-0.001	0.000	3.3
oPE X Error	-0.002	0.000	5.1
Nletters X Lex	-0.006	0.028	0.8
Nletters X Error	-0.245	0.172	1.4

0 0 0 2 8	13
0.020	0.4
01 0.000	2.4
02 0.000	5.0
01 0.000	3.2
03 0.030	0.1
001 0.000	26
	036 0.028 01 0.000 02 0.000 01 0.000 03 0.030 001 0.000

British LDT 4-8 Letters: Lengtl	n general	predictio	n
Intercept	5.25	0.421	12.5
Orthographic prediction error (oPE)	0.002	0.000	3.7
Number of letters (Nletters)	0.250	0.061	4.1
Word/non-word (Lex)	1.064	0.438	2.4
Error	1.264	0.443	2.9
oPE X Lex	-0.002	0.001	3.1
oPE X Nletters	-0.000	0.000	3.6
oPE X Error	-0.002	0.001	4.0
Nletters X Lex	-0.183	0.065	29
Nletters X Error	-0.002	0.005	4.0
Lex X Error	-1.426	0.467	3.1
oPE X Lex X Nletters	0.001	0.000	2.9
oPE X Lex X Error	0.002	0.001	3.6
oPE X Nletters X Error	0.001	0.000	4.0
Nletters X Lex X Error	0.228	0.068	3.5
oPE X Lex X Nletters X Error	-0.001	0.000	3.3

British LDT 4-8 Letters: Number of pixel

Intercept 6.590 0.157 42.0

Number of pixel (Npixel)	0.000	0.001	0.3
Number of letters (Nietters)	0.092	0.028	3.2
Word/non-word (Lex)	-0.124	0.162	0.8
Error	-0.309	0.165	1.9
Npixel X Lex	0.000	0.001	0.2
Npixel X Nletters	0.000	0.001	1.4
Npixel X Error	0.000	0.001	0.4
Nletters X Lex	0.000	0.001	0.4
Nletters X Error	-0.057	0.02)	2.0
Lex X Error	-0.090	0.030	3.0
Naivel V I ex V Metters	0.035	0.171	0.2
Npixel A Lex A Metters	0.000	0.001	0.9
Npixel X Lex X Error	0.000	0.001	0.1
Npixel X Nletters X Error	0.000	0.001	1.2
Nletters X Lex X Error	0.000	0.021	2.2
Npixel X Lex X Nletters X Error	0.009	0.031	1.2
French LDT			
Intercept Orthographic prediction error (oPE)	6.63 -0.002	0.005	1,333 2.0
Number of pixels (Npixel)	0.002	0.001	1.3
Word frequency	-0.042	0.003	34.1
oPE X Lex	0.005	0.002	2.0
Npixel X Lex	-0.007	0.002	3.0

Dutch LDT			
Intercept	6.45	0.019	348.1
Orthographic prediction error (oPE)	0.005	0.002	3.2
Number of pixels (Npixel)	0.001	0.002	0.6
Word/non-word (Lex)	0.101	0.004	23.8
Word frequency	-0.061	0.002	36.9
oPE X Lex	-0.016	0.002	6.6
Npixel X Lex	0.002	0.002	1.0

Dutch LDT2

Intercept	6.35	0.016	391.11 ₁₆₁		
Orthographic prediction error (oPE)	0.002	0.002	1.1		
Number of pixels (Npixel)	-0.001	0.002	0.6		
Word/non-word (Lex)	0.048	0.005	9.4		
Word frequency	-0.023	0.001	26.9		
oPE X Lex	-0.003	0.003	1.3		
Npixel X Lex	0.003	0.003	0.5		
Handwriting: Script based orthographic prediction error Intercept 1.465 0.010 154.3					
Mean prediction strength	0.052	0.007	7.4		
Number of pixels with a prediction	0.015	0.008	2.1		
Letter case	0.039	0.012	3.2		
Handwriting: Readability ratings					
Intercept	11.5	1.4	8.1		
Mean prediction strength	-5.9	1.0	6.2		

Note. E: Estimate; SE: Standard error; *t*: t-value. All t's >2 are considered a significant effect.

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