The pulvinar nucleus of the thalamus is suspected to have an important role in visual attention, based on its widespread connectivity with the visual cortex and the fronto-parietal attention network. However, at present, there remain many hypotheses on the pulvinar’s specific function, with sparse or conflicting evidence for each. Here we characterize how the human pulvinar encodes attended and ignored objects when they appear simultaneously and compete for attentional resources. Using multivoxel pattern analyses on data from two functional magnetic resonance imaging (fMRI) experiments, we show that attention gates both position and orientation information in the pulvinar: attended objects are encoded with high precision, while there is no measurable encoding of ignored objects. These data support a role of the pulvinar in distractor filtering—suppressing information from competing stimuli to isolate behaviourally relevant objects.
The pulvinar nucleus of the thalamus is believed to have an important, integrative function in the mammalian brain, likely related to visual attention. While pinpointing the pulvinar's specific role in vision has proven notoriously difficult, one line of evidence suggests the pulvinar may be important in isolating behaviourally relevant objects from surrounding distractors. When Desimone et al. deactivated the macaque pulvinar by muscimol injection, visual discrimination was impaired in the visual field contralateral to the injection site, but only when distractors were present; performance was unaffected in the absence of distractors. Recently, a similar pattern was reported in human subjects with pulvinar lesions due to stroke: patients had compromised visual discrimination in the contralesional visual field, but only when salient distractors were present. These results point towards an involvement of the pulvinar in resolving competition between visual targets and distractors. However, other studies based on lesions have failed to reinforce this notion. In a flanker interference task and a global/local interference task, subjects with pulvinar lesions showed no impairment in suppressing the distracting dimensions.

These results reveal two missing pieces of the puzzle with regard to distractor filtering in the pulvinar. First, lesion studies still paint a mixed picture of whether intact pulvinar function is necessary for normal distractor suppression. This may be due partly to the wide variety of stimuli and tasks used across the studies. Second, and more importantly, the current evidence for the pulvinar's role in distractor suppression comes almost exclusively from monkey and human lesion studies. To establish a role of the human pulvinar in distractor suppression, it is critical to show that the pulvinar performs such a function in healthy humans.

Here, we tested for distractor filtering in the human pulvinar by characterizing how the pulvinar encodes targets and distractors when they compete for attentional resources. We capitalize on a newly emerging understanding of the human pulvinar's organization. Specifically, the presence of precise, lateralized, spatial maps in both hemispheres. Using a multi-voxel pattern analysis, we measured the precision with which attended and ignored stimuli are coded in the pulvinar when both are present simultaneously within the same visual hemifield. Our results show that attention gates both spatial and featural information in the pulvinar: the positions and orientations of attended targets were coded with high precision in pulvinar responses, while there was no detectable encoding of distractor positions or orientations, even though the targets and distractors differed only in their behavioural relevance. These data support the hypothesis that the pulvinar is involved in filtering distracting visual information and highlighting behaviourally relevant targets.

**Results**

**Attentional gating of spatial information in the pulvinar.** Subjects viewed four visual stimuli at a time; two were attended and two were ignored. The stimuli were Gabor patches in the four visual quadrants (Fig. 1a); in each 10-s stimulation block, they appeared at one of five possible eccentricities ranging from 8.4° to 9.6° (Fig. 1b). The top two Gabors were fixed to have the same eccentricity as each other, as were the bottom two, but the eccentricities for the upper and lower Gabors were random with respect to each other. On alternating runs, subjects attended to the Gabors in either the upper or the lower visual field (Fig. 1c), watching for slight contrast decrements that occurred an average of five times per block. Attention was manipulated between the upper and lower visual fields (rather than left versus right) so that an attended and an ignored stimulus were always present within the same visual field, and hence represented in the pulvinar in the same hemisphere. At the end of each block, the subject made a response indicating whether there were more contrast decrements on the left, right, or the same number on the left and right (3 AFC; sensitivity (d') for the task was 1.59, indicating that the task was challenging but above threshold). The contrast decrements also occurred in the ignored stimuli, such that they were not just irrelevant but also distracting; subjects were motivated to ignore those stimuli as completely as possible. We tracked the positions of three of the five subjects during scanning (Fig. 2).

Independently manipulating the positions of the upper and lower Gabors allowed us to separately measure position coding at the attended and ignored locations within the same functional run. We extracted patterns of blood oxygenation level-dependent (BOLD) responses corresponding to the attended and ignored stimuli from the same set of voxels by running two separate general linear models: one in which the five predictors were coded according to the positions of the attended stimuli, and one in which they were coded according to the positions of the ignored stimuli (Fig. 3). In each general linear model (GLM), we separately contrasted the five predictors against the fixation baseline condition to obtain five maps of t-values. The result was a set of maps corresponding to the five attended positions and another set corresponding to the five ignored positions. Because the positions of the attended and ignored stimuli varied independently of each other, information about one of the dimensions was randomly distributed in the maps corresponding to the other dimension. Our subsequent pattern analysis tested for systematic variation in these activity maps as a function of stimulus position.

We used a cross-correlation approach to test for position selectivity in the BOLD response in the pulvinar. The analysis (Fig. 4) tested whether stimuli presented closer together in space produced more similar patterns of activity. The centre (position 3) map was used as a baseline, and we compared the four other positions with it by cross-correlating the activity patterns (Fig. 4a). Cross-correlating all of the activity patterns for all stimulus positions against each other yielded similar results (Supplementary Fig. S1), but using the central map as a baseline is more conservative because it avoids making assumptions about the linearity of the relationship between physical stimulus separation and correlation, as well as the independence of points in the plot (Supplementary Methods).

In the plots in Fig. 4a, each point represents one voxel within the region of interest (ROI), its value from one BOLD map plotted on the x-axis (t units), and its value from a second BOLD map plotted on the y-axis. The correlation across voxels between a given pair of
BOLD maps served as a measure of their similarity within that ROI; we plotted the correlation between each pair as a function of the distance between the corresponding stimuli in visual space (Fig. 4b). A significant negative slope on this plot indicates precise discrimination of the stimulus positions—the multivariate pattern of response within the ROI in question changed systematically with changes in stimulus position. This analytical approach allowed us to separately measure the encoding of the attended and ignored stimuli within the pulvinar without making any assumption about the nature of the underlying topographic organization there. We performed this analysis in pulvinar ROIs defined based on the Talairach Atlas, but fit to each subject’s individual anatomy (Fig. 4c).

Figure 5 shows position discrimination for attended objects (red data) and ignored objects (blue data) in the pulvinar. Data from the five subjects are plotted together and fit with a regression model that included a random effect of subject. Both the left and right
besides position coding in the pulvinar. We also sought to verify that attention gates types of stimulus information thought to have a high density of driving input to the pulvinar (difference in the pulvinar versus visual cortex). Of particular note, the pulvinar displayed a complete gating of the ignored object positions, and attentional modulation of position coding was significantly stronger in the pulvinar than in the visual cortex (FDR control for multiple comparisons). In contrast to the pulvinar and early visual cortex separately for the attended and ignored stimuli, which were simultaneously present at all times. To measure orientation coding, we used a support vector machine (SVM) pattern classifier to predict which orientation had been presented on each block. We used SVM in the second experiment rather than the correlation analysis because the power of the latter is diminished when discriminating between just two stimulus categories rather than a multi-level parametric stimulus manipulation (in the analysis of the first experiment, though we paired stimuli according to two possible separations, we utilized all five stimulus conditions); SVM, however, is well suited to this case of binary classification.

Classification results are shown for the pulvinar and early visual cortex in panels b and c of Fig. 6, respectively (data are collapsed across the three subjects). In the pulvinar, we found robust encoding of the attended stimulus orientations (54.7% correct; P < 0.001; 7.8 s.d. above the mean of the bootstrapped chance distribution; n = 192 training blocks and 48 test blocks per classification iteration per subject; reported performance is an average across 1,000 classification iterations) but no evidence of encoding of ignored orientations (49.9% correct; P = 0.62; 0.34 s.d. below bootstrapped chance; n = 192 training blocks and 48 test blocks per subject). Classification of the attended orientations was significantly better than classification of ignored orientations (P < 0.001; non-parametric bootstrap test for whether the attended-chance difference was larger than the ignored-chance difference; n = 192 training blocks and 48 test blocks per subject in the classification of both the attended and ignored stimulus orientations).

In the visual cortex, in contrast to the pulvinar, we found significant classification of both attended and ignored stimulus orientations besides position coding in the pulvinar. We also sought to verify that attentional gating in the pulvinar happens for task-relevant stimulus features (in experiment 1, subjects attended at the locations of the cued stimuli, but were not required to make a position-related judgment). Three subjects participated in this second experiment. The experimental design was similar to the first experiment, but now subjects were required to detect orientation changes in the cued Gabors. Each Gabor was oriented at either −45° or +45° relative to vertical, and the orientations in the upper and lower visual fields varied independently of each other from block to block (Fig. 6a). The left and right Gabors within the upper or lower visual field always had opposite orientations. An average of eight times per block, one of the four Gabors changed in orientation by 8°, for 200 ms. At the beginning of each run, subjects were instructed to attend to the orientations of either the upper two or lower two Gabors, and respond any time they detected an orientation change in one of the cued Gabors but to ignore orientation changes in the other two. Sensitivity (d’) for the orientation change detection task was 1.15; the task was challenging but above threshold.

We measured orientation coding within the pulvinar and visual cortex separately for the attended and ignored stimuli, which were encoded in the multi-voxel pattern of activity within the pulvinar. If the attentional gating of spatial information we observed in the pulvinar is related to filtering out distractors, we would expect measurable encoding of the ignored stimulus positions at earlier stages of visual processing, including in the input signals to the pulvinar. We repeated the same analysis in seven independently localized visual areas: V1, V2, V3, V3a, VP, V4v and MT+... Although the pulvinar showed significant coding of the attended stimulus positions, as indicated by significant negative slopes in the regression fits (Fisher z = −0.49, P = 0.011 in left pulvinar and z = −0.67, P = 0.001 in right pulvinar; n = 20 for each test; P-values were assessed using a non-parametric bootstrap test described in the Methods). All subjects showed the same trend individually. However, there was no detectable encoding of ignored objects (z = 0.11 in left pulvinar and z = 0.05 in right pulvinar, both P > 0.5; n = 20 for each test; non-parametric bootstrap test). Figure 5b shows the same analysis performed in bilateral pulvinar ROIs; discrimination of attended positions was significantly better than discrimination of ignored positions (P = 0.01; n = 20 points in each plot; bootstrap test for a difference in model fits between attended and ignored). It is important to note that the data in the attended and ignored plots came from the same functional runs, analysed within the same set of voxels, with the same stimuli present; no difference in signal-to-noise ratio, retinotopically specific adaptation, or other aspects of signal quality could explain the difference in coding precision we found for attended and ignored objects in the pulvinar (Supplementary Methods). These data demonstrate an essentially absolute gating of spatial information in the pulvinar by attention: only attended positions are encoded in the multi-voxel pattern of activity within the pulvinar.

If the attentional gating of spatial information we observed in the pulvinar is related to filtering out distractors, we would expect measurable encoding of the ignored stimulus positions at earlier stages of visual processing, including in the input signals to the pulvinar. We repeated the same analysis in seven independently localized visual areas: V1, V2, V3, V3a, VP, V4v and MT+. Figure 5c shows position discrimination plots, constructed in the same way as those shown for the pulvinar, for V1: V1 exhibited precise coding of both the attended and the ignored stimuli (attended: left−z = 1.34, right−z = 1.46; ignored: left−z = 1.07, right−z = 1.23; all P < 0.001; n = 20 for each test; non-parametric bootstrap test). In contrast to the pulvinar, each of the seven comparison visual areas showed significant discrimination of the ignored stimulus positions (non-parametric bootstrap test with n = 20 for each test; all P < 0.02; significant with FDR control for multiple comparisons23 with q = 0.05). Only the pulvinar displayed a complete gating of the ignored object positions, and attentional modulation of position coding was significantly stronger in the pulvinar than in the visual cortex (P = 0.008; n = 20 points in each plot; bootstrap test for a greater attended–ignored difference in the pulvinar versus visual cortex). Of particular note is the significant encoding of ignored positions in MT+, which is thought to have a high density of driving input to the pulvinar24.

**Attentional gating of orientation information in the pulvinar.** We sought to test whether attention gates types of stimulus information besides position coding in the pulvinar. We also sought to verify that...
Collectively, our results show a gating of ignored information within the pulvinar that generalizes across at least two stimulus features (position and orientation) and is distinct from the pattern of attentional modulation we observed in the visual cortex.

**Discussion**

Our results show that stimulus encoding in the human pulvinar is gated by attention: when behaviourally relevant visual stimuli compete with distractors for attentional resources within the same visual field, the pulvinar precisely represents attended but not ignored objects. Thus, activity in the pulvinar reflects an isolated representation of attended targets, which can serve to selectively confer further processing benefits on important stimuli.

Previous studies have reported significant functional magnetic resonance imaging (fMRI) responses in the pulvinar to...
unattended stimuli. For example, when Smith et al.29 presented subjects with optic flow stimuli and examined the BOLD response in the pulvinar when attention was directed either towards or away from the stimuli, they found significant pulvinar activation in both the attended and unattended cases, with an ~20% amplitude modulation with attention. How do our present results square with these previous findings? First, it is important to distinguish between the amplitude of the raw BOLD response and information content in fMRI patterns of activity, which can be dissociated26,30,31; our study takes the important step of decoding position and orientation information within the multivariate pattern of activity in the pulvinar, providing converging evidence for attentional gating from two information-based MVPA approaches. Equally important, though, is a key difference in experimental design between the above-mentioned studies and our study. The spatial attention manipulations in both Smith et al.29 and Schneider12 had attended and ignored stimuli on opposite sides of the horizontal meridian, and thus in different hemispheres in the pulvinar. Given the evidence for at least partially independent attentional resources in the left and right hemispheres32,33, one possibility is that a left/right attentional manipulation induces less competition between attended and unattended stimuli than if both stimuli fell in the same visual field. Thus, a key aspect of our design is the upper/lower manipulation of attention, forcing targets and distractors to compete for attentional resources within the same visual field and same hemisphere.

Previously13, we reported position sensitivity in the pulvinar under passive viewing conditions (attention was engaged at the fixation point by a task unrelated to the stimuli). Critically, the stimuli used to measure spatial discrimination were neither targets nor distractors—they were irrelevant to the fixation task that subjects performed. Other studies of retinotopy in the healthy human pulvinar have used undivided attention14,15 or featural attention12, but our study is the first to test how the healthy human pulvinar responds to competing stimuli in the same visual field. We introduced competition between simultaneously presented stimuli, which resulted in the ignored objects being gated out from encoding in the pulvinar. Unlike putative attentional source regions such as the IPS areas34, which show only weak responses to passive visual stimulation, the pulvinar precisely encodes passively viewed stimuli13 but not distracting ones.

Figure 5 | Attention gates spatial encoding in the pulvinar. (a) Correlation-based analysis for left and right pulvinar, using the BOLD maps for the attended positions (red data) and the ignored positions (blue data). Each subject contributes four points to each plot—two independently computed correlations at each separation. Because multiple points were plotted for each subject, we fit the data with a regression model that included a random effect of subject22. Correlations are presented as Fisher z-scores for the sake of linear comparison. In both hemispheres, there was significant discrimination of the attended positions (Fisher z = −0.49, P = 0.011 in left pulvinar; z = −0.67, P = 0.001 in right pulvinar; n = 20 for each test; non-parametric bootstrap test), but no discrimination of the ignored positions (z = 0.11, P = 0.69 in left pulvinar; z = 0.05, P = 0.57 in right pulvinar; n = 20 for each test; non-parametric bootstrap test). (b) Position discrimination analysed in bilateral pulvinar ROIs. Discrimination of attended positions was significantly better than discrimination of ignored positions (P = 0.01; n = 20 points in each plot; bootstrap test for a difference in model fits between attended and ignored). (c) In primary visual cortex (V1), there was significant discrimination of both the attended and ignored stimulus positions, consistent with previous results20. Attentional modulation of position discrimination (zatt − zignore) was significantly greater in the pulvinar than in V1 (P = 0.04; n = 20 points in each plot; bootstrap test for a greater attended−ignored difference in the pulvinar versus V1).
We intentionally analysed ROIs that encompassed the entire pulvinar. Although the primate pulvinar has been classically divided into several subnuclei based on cytoarchitecture, no structural or functional homology between the monkey and human pulvinar has been firmly established. Further, interactions between attentional signals and visual maps in the pulvinar likely involve more than a single subregion. While the lateral and inferior portions of the monkey pulvinar are the most visually responsive,3,36, the pulvinar’s connections with areas in the fronto-parietal attention network project from the medial portion37. We therefore considered the multivariate pattern of activity within the pulvinar as a whole.

Our results dovetail with recent reports that visual awareness is correlated with variability in the strength of responses in the pulvinar38–39. For example, Padmala et al.38 used affective conditioning to increase the salience of otherwise neutral stimuli. Then, presenting conditioned and non-conditioned stimuli in a rapid serial visual presentation task during scanning, they found that stimulus detection was correlated with larger responses in the pulvinar; this effect was much larger for affectively conditioned stimuli. Our present results are complementary and demonstrate the flip side of the coin: we show that distracting stimuli are gated out from encoding in the pulvinar, while Padmala et al. demonstrated that the more salient an attended target is, the more robust a response it produces in the pulvinar, particularly if it has conditioned affective significance. In future work, it will be important to study whether and how affective significance interacts with the attentional gating effect we report here.

Our orientation classification results complement and extend previous work on decoding orientation information from the human visual cortex26–28. Earlier work has shown that attending to an orientation biases signals in the visual cortex to preferentially represent that orientation, yielding a boost in classification accuracy27,28. This attention-related enhancement can occur selectively for task-relevant features40, and is carried at least in part by a boosted response within the subpopulation of voxels that is tuned to the attended orientation27. In experiment 2, we similarly found that attention significantly improved the precision of orientation coding in early visual cortex (Fig. 6c). We extended the classification framework to measure orientation-selective responses in the human pulvinar for the first time, and found a significantly larger attentional modulation of orientation information in the pulvinar than in the visual cortex (Fig. 6b), such that there was no measurable information about the ignored orientations, even though the ignored and attended stimuli differed only in their task relevance. The gating of orientation information in the pulvinar by attention sets the pulvinar apart from the pattern of results found in the visual cortex, and lends support to the hypothesis that the pulvinar has an important role in distractor filtering.

Selective attention reduces the deleterious influences of distracting information on perception and behaviour,9,40–42 and is therefore a critical mechanism for many visuomotor and cognitive functions. The pulvinar has long been suspected to be a neural interface where attention signals can gate out distractors, but this filtering function is yet to be demonstrated in the human pulvinar. For example, Padmala et al.38 found that distracting stimuli were gated out from encoding in the pulvinar, particularly if they had conditioned affective significance. In future work, it will be important to study whether and how affective significance interacts with the attentional gating effect we report here.

Methods

Subjects. Seven healthy subjects, ranging in age from 22 to 31 years, participated in the study. All subjects provided written consent before participation, and all experimental procedures were approved by the UC Davis and UC Berkeley institutional review boards.

Experimental design. Before each scanning run, subjects were cued to attend to the Gabors in either the upper or lower visual field, and the subject attended at that location for the duration of the entire run. Attend-upper and attend-lower runs were interleaved so that no trends across the scanning session could differentially affect the two conditions. Within a run, the Gabors positions in the upper and lower visual fields were randomised with respect to each other, and the ordering of the conditions within the upper and lower visual fields was pseudo-randomized by shuffling within a group of the five position conditions plus a baseline condition, and then presenting six such shuffled groups in sequence. The Gabors (0.4 cycles per degree, peak contrast 87% (Michelson), 1.66° s.d. envelope, 8 Hz counterphase flikcer)
appeared at five possible eccentricities, ranging from 8.4° to 9.6° in increments of 0.3°. Attend-upper and attend-lower runs were constructed in a completely identical fashion; the only difference between the two run types was the attended location. This upper/lower attention manipulation allowed the positions of the attended and ignored Gabors to vary independently of each other, and for both attended and ignored stimuli to be represented in each hemisphere. Stimuli were presented in 10-s blocks. We used a blocked design, rather than a rapid event-related design, to maximize the power in the position discrimination analysis (Supplementary Methods).

In experiment 2, we measured orientation discrimination rather than position discrimination. Subjects viewed four Gabors (0.4 cycles per degree, 87% peak Michelson contrast, 1.66° s.d. envelope, 8 Hz counterphase flicker), one in each visual quadrant (Fig. 6a). The orientations of the Gabors in the upper and lower visual fields varied independently (Fig. 6a), and the positions of the Gabors were randomly jittered from block to block with a s.d. of 0.47° (mean 9.0° eccentricity), independently in the upper and lower visual fields and independently of the orientations, to ensure that orientation decoding generalized across changes in position and was not simply capitalizing on localized luminance differences between stimulation. Successful performance on the orientation change detection task required attending to both Gabors in the cued location (upper or lower visual field), while ignoring the Gabors in the non-cued field as completely as possible. As in experiment 1, the stimuli in attend-upper and attend-lower runs were constructed in a completely identical fashion; the only difference was the verbal instruction at the beginning of the run.

fMRI acquisition. Scanning was conducted on a Siemens Trio 3T system at the UC Davis Imaging Research Center. Functional images were collected with a gradient-recalled echo EPI sequence using an eight-channel phased-array head coil. Whole-brain anatomical images were acquired with a high-resolution (1 mm³) Turbo Spin Echo protocol. fMRI acquisition parameters were TR = 2,000 ms, TE = 26 ms, FA = 90°, FOV = 22×22 cm², voxel size = 1.528×1.528×2.5mm³. The 24-slice imaging volume was centred on the thalamus and run through the occipital pole. Runs were 360 s long (180 volumes). Using a Digital Projection Mercury 5000HD projector, stimuli were back-projected at 75 Hz onto a semi-transparent screen from outside the bore. Subjects were able to see the screen and stimuli via a mirror angled at 45°, located 9.5 cm directly above their eyes. Each subject participated in eight main experimental runs and one additional retinotopic localizer run.

Data analysis. fMRI data preprocessing, ROI definition and general linear models were performed in BrainVoyager QX v. 2.0; all other analysis was performed in MatLab R2008a. Data preprocessing consisted of slice scan time correction with cubic spline interpolation, 3D motion correction with trilinear interpolation and high-pass filtering with a cutoff of two cycles per run. The data were not otherwise spatially smoothed. The data were normalized to Talairach space to allow for atlas-assisted definition of the pulvinar ROIs (Fig. 4c).

Position discrimination. For each functional run, we constructed two general linear models, one with the five attended stimulus positions as five predictors, and the other with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of GLM recoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attended stimulus positions as five predictors, and one with the five ignored stimulus positions as predictors. In addition, there was a sixth predictor for the baseline (fixation only) blocks. Figure 3 depicts the GLM encoding procedure. For each GLM, we separately contrasted each of the five stimulus position predictors against the fixation baseline to generate five maps of linear models, one with the five attends.
support Vector Machine on functional MRI data. *Neuroimage* **28**, 980–995 (2005).

26. Jehee, J. F. M., Brady, D. K. & Tong, F. Attention improves encoding of task-relevant features in the human visual cortex. *J. Neurosci.* **31**, 8210–8219 (2011).

27. Serences, J. T., Saperoa, S., Scolari, M., Ho, T. & Muftuler, T. Estimating the influence of attention on population codes in human visual cortex using voxel-based tuning functions. *Neuroimage* **44**, 223–231 (2009).

28. Kamitani, Y. & Tong, F. Decoding the visual and subjective contents of the human brain. *Nat. Neurosci.* **8**, 679–685 (2005).

29. Smith, A. T., Cotton, P. L., Bruno, A. & Moutsiana, C. Dissociating vision and visual attention in the human pulvinar. *J. Neurophysiol.* **101**, 917–925 (2009).

30. Haynes, J. D. & Rees, G. Decoding mental states from brain activity in humans. *Nat. Rev. Neurosci.* **7**, 523–534 (2006).

31. Tong, E. & Pratte, M. S. Independent hemispheric attentional systems mediate visual search in split-brain patients. *Nature* **342**, 543–545 (1989).

32. Silver, M. A., Ress, D. & Heeger, D. J. Topographic maps of visual spatial attention in human parietal cortex. *J. Neurophysiol.* **94**, 1358–1371 (2005).

33. Kaas, J. H. & Lyon, D. C. Pulvinar contributions to the dorsal and ventral streams of visual processing in primates. *Brain Res. Rev.* **55**, 285–296 (2007).

34. Hardy, S. G. & Lynch, J. C. The spatial distribution of pulvinar neurons that project to two subregions of the inferior parietal lobule in the macaque. *Cereb. Cortex* **2**, 217–230 (1992).

35. Wilke, M., Mueller, K. M. & Leopold, D. A. Neural activity in the visual thalamus reflects perceptual suppression. *Proc. Natl Acad. Sci.* **106**, 9465–9470 (2009).

40. Carrasco, M. Visual attention: the past 25 years. *Vision Res.* **51**, 1484–1525 (2011).

41. Mangun, G. R. Neural mechanisms of visual selective attention. *Psychophysiology* **32**, 4–18 (1995).

42. Treisman, A. M. Strategies and models of selective attention. *Psychol. Rev.* **76**, 282–299 (1969).

43. Lin, C.- J. & Chang, C. C. LIBSVM: a library for support vector machines. *ACM Transactions on Intelligent Systems and Technology* **2**, 27:1–27:27 (2011).

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Author contributions

J.F. and D.W. designed the experiments. J.F. carried out the experiments and analysed the data. J.F. and D.W. wrote the paper.

Additional information

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