Binocular integration of retinal motion information underlies optic flow processing by the cortex

Highlights

- Translation- and rotation-selective neurons are abundant in higher visual areas
- Optic-flow-selective neurons in V1 and RL/A rely on retinal direction selectivity
- Retinal direction selectivity controls functional segregation between V1 and RL/A
- Binocular integration of retinal motion information underlies optic flow selectivity

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In Brief

Locomotion creates different patterns of optic flow, but how these are encoded by the cortex is unclear. By mapping cortical activity in awake mice, Rasmussen et al. demonstrate that binocular integration of retinal motion direction selectivity causally influences optic flow selectivity of neurons residing in distinct areas of the visual cortex.
Binocular integration of retinal motion information underlies optic flow processing by the cortex

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SUMMARY

Locomotion creates various patterns of optic flow on the retina, which provide the observer with information about their movement relative to the environment. However, it is unclear how these optic flow patterns are encoded by the cortex. Here, we use two-photon calcium imaging in awake mice to systematically map monocular and binocular responses to horizontal motion in four areas of the visual cortex. We find that neurons selective to translational or rotational optic flow are abundant in higher visual areas, whereas neurons suppressed by binocular motion are more common in the primary visual cortex. Disruption of retinal direction selectivity in Frmd7 mutant mice reduces the number of translation-selective neurons in the primary visual cortex and translation- and rotation-selective neurons as well as binocular direction-selective neurons in the rostrolateral and anterior visual cortex, blurring the functional distinction between primary and higher visual areas. Thus, optic flow representations in specific areas of the visual cortex rely on binocular integration of motion information from the retina.

INTRODUCTION

The action of moving through an environment produces patterns of visual motion, known as optic flow, on the retina, which animals rely on to guide their behavior. Animal locomotion is largely described by a combination of forward-backward movements and left-right turning. Forward and backward movements induce translational optic flow (nasal-to-temporal or temporal-to-nasal motion in both eyes, respectively), whereas turning induces rotational optic flow (nasal-to-temporal motion in one eye and temporal-to-nasal in the other; Figures 1A and 1B). However, despite the increasing use of mice to study vision, it is unknown how these distinct optic flow patterns are encoded by the rodent cortex.

An extensive body of research has shown that neurons residing in brain areas involved in optic flow processing have complex receptive fields, often receive binocular inputs, and respond to both translational and rotational optic flow stimuli. Examples include the fly lobula plate (involved in course control),1,2 the zebrafish pretectal nuclei,3,4 the avian and mammalian accessory optic system (involved in gaze stabilization),5,6 and both the dorsomedial region of the medial superior temporal area and posterior parietal cortex (PPC) of monkeys (involved in spatial navigation).7–9 The mouse visual cortex contains a primary visual cortex (V1) and more than a dozen distinct higher visual areas (HVAs), each with unique sensitivities to visual features.10,11 The V1 receives retinal inputs via the lateral geniculate nucleus and distributes functionally specialized signals to different HVAs.12–14 Based on their anatomy, multi-sensory processing, and roles in spatial navigation, the rostrolateral (RL), anterior (A), and anteromedial (AM) HVAs are considered part of the PPC in mice,15–18 raising the possibility that they contain neurons sensitive to binocular optic flow.

In rodents, visual motion computations are not exclusive to the cortex and start in the retina. The retina contains mosaic arrangements of direction-selective (DS) cells that preferentially respond to motion in one of the four cardinal directions (nasal, temporal, dorsal, and ventral).19–21 These cells fall into two canonical classes: ON DS cells (which project to the nuclei of the accessory optic system and mediate the optokinetic reflex) and ON-OFF DS cells (which project to the lateral geniculate nucleus and the superior colliculus).22,23 Although the role of ON DS cells for mediating gaze-stabilizing eye movements is well established, the functional role of ON-OFF DS cells is unclear. The Frmd7 mutant mouse (Frmd7tm), a model of congenital nystagmus,24 is a valuable experimental tool for studying the contribution of ON-OFF DS cells to visual cortical processing. Importantly, this mouse is characterized by impaired horizontal direction selectivity in both ON and ON-OFF DS cells, as a result of transition from asymmetric to symmetric inhibitory inputs from starburst amacrine cells.25 At this time, a handful of studies have tested cortical activity in response to monocular visual motion stimulation in the Frmd7tm mouse.13,25,26 One study found that a specific form of direction selectivity in layer...
2/3 of V1, tuned to higher stimulus speeds and with a preference to posterior motion, was disrupted in Frmd7tm mice. Subsequent work expanded on this finding by showing that responses to posterior motion in layer 2/3 of the RL area, but not in the posteromedial (PM) area or layer 4 of V1, is also affected in this mouse. These data are suggestive of a segregated cortical pathway for processing signals originating from horizontally tuned ON-OFF DS cells. What might be the functional role of such a visual motion processing stream from the retina to the cortex? An intriguing hypothesis is that information from ON-OFF DS cells in the left and right eyes is systematically integrated in the cortex to create areas with distinct sensitivity to translational and rotational optic flow patterns. However, this has yet to be experimentally tested, and the cortical areas that might combine optic flow information from the left and right eyes remain unknown.

Here, we systematically map the responses of individual neurons across the visual cortex using two-photon calcium imaging.
during monocular and binocular optic flow stimulation within the monocular visual field of awake mice. We test the contribution of retinal horizontal direction selectivity to visual cortical activity using the Fmr1mice. Our data demonstrate that the mouse visual cortex contains an abundance of neurons that encode translational or rotational optic flow. Furthermore, our results suggest that information from retinal DS cells in each eye is integrated in the cortex as early as in V1, where it establishes response selectivity to backward translational optic flow, but that binocular retinal DS signaling for establishing selectivity to rotational optic flow is first integrated in the higher areas RL and A. These results support the hypothesis that retinal ON-OFF DS cell mosaics are specialized for detecting translational and rotational optic flow.

RESULTS

Discrete neuronal responses to monocular and binocular motion stimuli can be imaged in the visual cortex of awake mice
To identify individual areas of mouse visual cortex, we used intrinsic signal optical imaging. We first generated visual field sign maps from retinotopic maps, allowing us to identify V1 as well as the higher areas RL, A, AM, and PM (Figures 1C and S1). We chose to combine areas RL and A (RL/A), as these areas could not be clearly distinguished from each other in our dataset. For binocular animals to reliably detect different optic flow patterns, the brain must integrate motion signals from each eye. We therefore investigated the neuronal responses underlying binocular optic flow processing by presenting moving gratings to mice using a stimulus protocol that tests the repertoire of horizontal motions. The eight stimulus conditions in the protocol were generated by presenting gratings moving in a nasal or temporal direction to one eye at a time and then to both eyes to simulate the rotational (ipsiversive and contraversive) or translational (forward and backward) optic flow that the mouse would experience during locomotion (Figures 1A and 1B; Video S1; see STAR methods). To unambiguously probe the interaction of left and right retinal information in the cortex, the stimuli were presented only to the monocular visual fields and not to the frontal binocular visual field (Figure 1B). Our stimulus protocol did not effectively trigger the optokinetic reflex (Figure S2), likely due to the use of a low spatial frequency (0.03 cycles/°).

The tuning properties of individual layer 2/3 neurons were characterized in awake mice by transfecting cortical neurons with the genetically encoded calcium sensor GCaMP6f (expression driven by the synapsin promoter) and measuring changes in two-photon fluorescence during stimulus presentation (Figures 1D and 1E). A typical field of view contained ~100–150 neurons, and somatic calcium responses showed diverse but consistent patterns, depending on the eye being stimulated and the direction of motion (Figure 1E). Tuning curves for individual neurons were generated by plotting trial-averaged fluorescence changes as a function of stimulus conditions (Figure 1F). We systematically classified neurons into distinct functional types according to their tuning curves using regressor-correlation analysis. First, we generated a regressor map consisting of all possible all-or-none response combinations to the eight stimulus conditions, which resulted in 256 profiles (Figure 1G; see STAR methods). Next, the tuning curve for each neuron was assigned to the regressor with the highest correlation (Figures S3A–S3C). All tuning curves had high correlations with their assigned regressor (mean correlation coefficient; 0.91 ± 0.05; n = 28,712 neurons from 17 mice). These data confirm that we can reliably elicit responses to monocular and binocular motion stimuli, presented within the monocular visual field, in the visual cortex of awake mice and also robustly classify neurons into discrete response types.

The RL/A area of the visual cortex is enriched with optic-flow-selective neurons
We sought to investigate the response specificity of visual cortex neurons by sampling thousands of consistently responsive neurons in multiple areas of the visual cortex of nine mice (3,010 in V1, 4,165 in RL/A, 4,006 in AM, and 3,059 in PM; Table S1) and assigning them to regressors (Figures 1G and S4). To characterize the monocular and binocular optic flow coding properties of these neurons, we initially focused on three response classes: simple; translation or rotation selective; and binocular suppressed (Figures 1H and 2). The simple class comprised three groups that were characterized by their direction selectivity: monocular DS; binocular DS; and non-DS neurons. Translation- and rotation-selective neurons comprised four groups that were characterized by their response selectivity to either forward translational, backward translational, contraversive rotational, or ipsiversive rotational optic flow (Figures 1A, 1H, and 2). Binocular-suppressed neurons were characterized by a suppressed response during binocular motion stimulation and were further divided according to their DS or non-DS responses to monocular motion (Figures 1H and 2; see STAR methods). Within these three classes, the majority of neurons assigned to the same regressor had similar Ca2+ response time courses across stimulus conditions, and correlation strength distributions were generally unimodal, indicating no clear signal of further neuronal subpopulations (Figures S3D and S3E), although we did note minor response variability within the same regressor, which could result from heterogeneous spatiotemporal receptive field properties of the sampled neurons. Thus, to fully resolve whether subsets of neurons may have distinct response time course kinetics to certain stimulus conditions, future work should exhaustively probe the spatiotemporal receptive field properties of neurons within the three response classes.

For all visual cortical areas, we counted neurons assigned to each regressor and ranked regressors according to their frequency (Figures 3A and S5). To identify significantly overrepresented regressors, we performed shuffling and bootstrap analyses (see STAR methods). Interestingly, in contrast to previous work in zebrafish, the most abundant neurons in V1 were binocular-suppressed neurons, which have been described in the primary V1. These neurons constituted as much as 38.6% of all responsive V1 neurons and 56% of the significantly overrepresented regressors (Figures 3A and S5B). In contrast, simple and translation- or rotation-selective neurons constituted only 8.8% and 19.7% of all responsive neurons, respectively. Neurons that could not be assigned to these three classes were considered unclassified and not investigated further.

The abundance of neuronal classes was different in the HVAs (Figure 3A). Translation- or rotation-selective neurons were the most abundant response class in the RL/A area—24% of all
neurons and 46% of the significantly overrepresented regressors—whereas simple and binocular-suppressed neurons comprised only 2.5% and 14.8% of neurons, respectively. In area AM, translation- or rotation-selective neurons were again abundant and simple neurons were sparse (22.9% and 4% of neurons, respectively), but there was a higher proportion of binocular-suppressed neurons than in the RL/A area (17.8%). The PM area was characterized by an equal proportion of translation- or rotation-selective and binocular-suppressed neurons, constituting 24.4% and 25.3% of neurons, respectively.

These data establish that different areas of mouse visual cortex contain distinct distributions of monocular and binocular optic-flow-encoding neurons. In particular, the RL/A area is enriched with neurons encoding translational and rotational optic flow, whereas V1 is enriched with neurons activated by monocular motion but suppressed by binocular motion.

Retinal direction selectivity contributes to binocular optic flow processing in V1 and RL/A

To determine whether retinal direction selectivity contributes to the processing of optic flow in the visual cortex, we repeated our neuronal mapping in Frmd7 mutant (Frmd7tm) mice, which lack horizontal direction selectivity in the retina. Consistently responsive neurons were sampled in different areas of the visual cortex of eight mice (2,925 in V1, 3,125 in RL/A, 3,375 in AM, and 3,047 in PM; Table S1). This revealed a difference in the overall distribution of response classes in certain areas of Frmd7tm mice compared to wild-type mice (Figures 3A and 3B), which prompted us to examine the effects of direction selectivity on the proportions of monocular- and binocular-responsive neurons in each functional group or response type (Figures 4A–4D and S6A–S6D). In V1, the proportions of monocular DS and backward translation-selective neurons were reduced in Frmd7tm mice (Figure 4A). More strikingly, all groups of translation- or rotation-selective neurons, as well as binocular DS neurons, were reduced in the RL/A area of Frmd7tm mice (Figure 4B). In the AM area, only monocular and binocular DS neurons were reduced (Figure 4C). The proportion of DS and non-DS binocular-suppressed neurons was increased in both RL/A and AM areas of Frmd7tm mice (Figures 4B and 4C). Finally, none of the nine functional groups were significantly altered in the PM area of Frmd7tm mice (Figure 4D), underscoring previous work showing that motion processing in the PM area is independent of retinal DS signaling.

Together, these data show that simple and translation- or rotation-selective responses, but not binocular-suppressed responses, are impaired by disrupting retinal direction selectivity. Furthermore, we conclude that retinal direction selectivity contributes to binocular optic flow responses in the V1 and RL/A areas of the visual cortex.

Retinal direction selectivity establishes functional segregation between V1 and RL/A

Individual HVAs form distinct subnetworks, each of which represents a different information stream.10,32,33 We sought to find out how visual cortical areas are functionally organized with respect to their composition of optic-flow-sensitive neurons and whether retinal direction selectivity is involved in creating such an organization. To probe this, we used the mean proportion of neurons assigned to our functional response types to create an optic flow fingerprint for each visual area in wild-type and Frmd7tm mice and then we performed hierarchical clustering and correlation analyses (Figures 5A–5C; see STAR methods).

Hierarchical segregation (Figure 5A) together with a rather high correlation between optic flow representations (mean correlation coefficient, 0.81 ± 0.06; Figures 5B and 5C) were evident between the cortical areas of wild-type mice. In particular, V1 was noticeably separated from the RL/A, AM, and PM areas, suggesting functional specialization between V1 and the HVAs.10 In addition, the PPC areas (RL/A and AM) branched from both V1 and PM, indicating that the PPC has a distinct role in optic flow processing (Figure 5A). In contrast, there was little hierarchical segregation, and even more correlated optic flow representations, between visual areas in Frmd7tm mice (mean correlation coefficient, 0.93 ± 0.01; Figures 5A–5C). Notably, optic flow responses in area RL/A were remarkably
Figure 3. The RL/A area of the visual cortex is enriched with optic-flow-selective neurons in wild-type mice
Top: ranked distribution of the 50 most abundant response types and response classes in the V1, RL/A, AM, and PM areas of wild-type mice (A) and Frmd7tm mice with disrupted retinal direction selectivity along the horizontal axis (B). Dotted line denotes chance level obtained from averaging distributions from shuffled response profiles generated by bootstrapping (500 samples). Error bars are mean ± SEM. Middle: p values indicating the probability of proportions being higher in the shuffled than in the original dataset are shown. Values above the green line are not significant (n.s.) (p < 0.05). Bottom: corresponding ranked regressor profiles are shown (white, active; black, inactive). Inset: pie chart shows proportion of neurons within response classes for the significantly overrepresented (p < 0.05) response types. See also Table S1 and Figures S3–S5.

similar to those in V1 in Frmd7tm mice (correlation coefficient 0.58 and 0.98 for wild-type and Frmd7tm mice, respectively; p < 0.001, Fischer’s transformation; Figure 5C), abolishing any functional segregation between these areas. In contrast, the PM area of both wild-type and Frmd7tm mice appeared on the same branch (Figure 5A), supporting the notion that motion processing in this area is independent of retinal direction selectivity.

To further investigate area specialization, we assessed the proportion of monocular- versus binocular-driven functional groups within each visual area and quantified the relationship with a selectivity index (Figure 5C; see STAR methods). In wild-type mice, the bias toward monocular or binocular motion differed between visual areas to the extent that RL/A emerged as a specialized area for binocular optic flow processing (binocular optic flow index, −0.39 for V1, 0.21 for RL/A, 0.059 for AM, and −0.11 for PM). In contrast, this functional diversity was absent in Frmd7tm mice, and monocular-driven neurons were overrepresented across the visual areas (binocular optic flow index, −0.38 for V1, −0.44 for RL/A, −0.18 for AM, and −0.19 for PM).

From these data, we conclude that retinal direction selectivity contributes to functional segregation and response specialization between the different areas of the visual cortex. The most striking effect of retinal direction selectivity disruption in Frmd7tm mice is the transformation of optic flow responses in the RL/A area into responses reminiscent of responses in V1, indicating a specific role for the RL/A area in binocular integration of motion information originating from retinal DS cells.

DISCUSSION

Our study provides four major insights into how the processing of optic flow within the monocular visual field is functionally organized in the visual system of mice. First, translation- and rotation-selective neurons are abundant in areas RL/A, AM, and PM, whereas neurons suppressed by binocular motion are common in V1. Second, translation-selective neurons in V1 and translation- and rotation-selective neurons in the RL/A, but not AM and PM areas, rely on direction selectivity that is computed in the retina. Third, binocular-suppressed neurons, which would be efficiently activated by monocularly restricted motion but suppressed by self-motion-induced optic flow, do not rely on retinal direction selectivity. Fourth,
Retinal direction selectivity contributes to the functional segregation of optic flow responses between V1 and RL/A. Our results, therefore, demonstrate a causal link between retinal motion computations and optic flow representations in specific areas of the visual cortex. Furthermore, they establish a critical role for retinal direction selectivity in the cortical processing of whole-field optic flow, thereby answering a previously proposed hypothesis.

The altered optic flow representations in Frmd7tm mice imply potential functional circuits to link retinal horizontal DS cells and cortical layer 2/3 neurons with distinct optic flow response preferences (Figure 6). Our results suggest that information from retinal DS cells, tuned to motion in either the nasal or temporal direction, is propagated to layer 2/3 of the contralateral V1, where it contributes to establishing monocular DS responses tuned to horizontal motion. In turn, a fraction of backward translation-selective responses in V1 are likely synthesized from these monocular DS inputs, converging from V1 in both hemispheres via interhemispherically projecting neurons. In addition, a fraction of rotation-selective responses in area RL/A are likely synthesized from monocular nasal- and temporal-motion-preferring DS inputs converging from V1 in the same and opposite hemisphere, respectively. These hypotheses could be tested by functionally characterizing the presynaptic network of individual translation- or rotation-selective neurons using rabies-virus-based trans-synaptic tracing. Our data also suggest that translation- and rotation-selective neurons in V1 and RL/A are suppressed by visual motion in non-preferred directions on either retina (Figure 6). Such response suppression could be mediated by inhibitory monocular DS neurons or inhibitory
interneurons activated by excitatory monocular DS neurons. In the present study, we sampled both excitatory and inhibitory neurons, but future studies could clarify this issue by genetically assigning imaged neurons into excitatory and inhibitory cell types.

Our results also offer insights into the cortical pathways that process visual motion independently of direction selectivity computed in the retina. Our analyses reveal that neuronal responses suppressed by binocular motion are common in V1 and HVAs and that these do not rely on retinal direction selectivity. This suggests that the V1 circuitry associated with binocular-suppressed neurons is functionally segregated from the circuitry processing retinal direction selectivity.12,13,25,26 Interestingly, the majority of binocular-suppressed neurons in V1 had a preference for motion in the ipsilateral eye (Figure 1H), suggesting that these neurons may combine the following two distinct types of input: (1) DS or non-DS excitatory inputs originating from non-DS cells in the ipsilateral eye via interhemispherically projecting neurons in the contralateral V1 and (2) non-DS inhibitory inputs driven by the activity of the contralateral eye. Our analyses also detected retinal DS cell-independent binocular optic flow responses in layer 2/3 of the visual cortex (Figures 3 and S6). Prior work in monkeys showed that binocular-suppressed and binocular-facilitated responses of monocular V1 neurons can be observed in the main visual input layer (layer 4).31 In mice, one form of de novo direction selectivity emerges in layer 4.37 Hence, it is plausible that retinal direction-selectivity-independent forms of binocular-suppressed and binocular-facilitated DS responses may arise in layer 4 from binocular interactions of DS signals originating from cortically computed direction selectivity. This idea is consonant with previous work in mice demonstrating that layer 4 neurons in V1 generate directionally tuned responses independent of inputs from retinal DS cells.13

Accumulating evidence suggests that areas RL and A are part of the PPC in mice15–18—a key nexus of sensorimotor integration that is involved in decision making during spatial navigation,38 the encoding of body posture,39 global motion analysis,40,41 and representations of spatial information.42 Intriguingly, more than 50% of neurons in the RL area are multi-sensory in mice,43 integrating both tactile and visual sensory inputs.43 To advance our understanding of the behavioral function of area RL/A, it will thus be important to determine whether translation- and
rotation-selective neurons display multi-sensory representations of self-motion (for example, whether they encode the direction of whisker deflections). Moreover, identifying the specific projection targets of these neurons might provide insight into how sensory self-motion information feeds into, for example, neuronal circuits for movement control. We speculate that area RL/A, as defined in our experiments, may be the functional correlate of the ventral intraparietal area of the PPC in monkeys, where multi-sensory representation of self-motion is utilized for goal-directed movements.44 Thus, an intriguing question that emerges from our results is whether responses to binocular optic flow in the PPC of monkeys rely on retinal direction selectivity, as they do in the RL/A area in mice. A first step toward addressing this would be to determine whether retinal DS cells exist in non-human primates, making it possible to define common principles of visual motion processing as well as the modifications that have occurred throughout the course of evolution.

Figure 6. Proposed circuit model for translational and rotational optic flow processing

Left: FT optic flow activates nasal motion-preferring DS cells in the left and right retinas, mediating activity in nasal (N) motion-preferring MoDS (N-MoDS) neurons in V1 of both hemispheres, and subsequently their combination in FT-selective neurons in area RL/A. Activity in N-MoDS neurons also inhibits BT-selective neurons. BT optic flow activates temporal (T) motion-preferring DS cells in the left and right retinas, mediating activity in temporal motion-preferring MoDS (T-MoDS) neurons in V1 of both hemispheres, and subsequently their combination in BT-selective neurons in V1 and RL/A. Activity in T-MoDS neurons also inhibits FT-selective neurons. Right: IR optic flow activates temporal and nasal motion-preferring DS cells in the left and right retinas, respectively, mediating activity in N- and T-MoDS neurons in V1 of the left and right hemispheres, respectively. The signals from these V1 neurons, in turn, combine at IR-selective neurons in RL/A of the left hemisphere, and their activity inhibits CR-selective neurons in the left hemisphere. CR optic flow activates nasal and temporal motion-preferring DS cells in the left and right retinas, respectively, mediating activity in T- and N-MoDS neurons in V1 of the left and right hemispheres, respectively. The signals from these V1 neurons, in turn, combine at CR-selective neurons in RL/A of the left hemisphere, and their activity inhibits IR-selective neurons of the left hemisphere. The wiring diagram is expected to be mirror symmetric in relation to the midline.

STAR+METHODS

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SUPPLEMENTAL INFORMATION

Supplemental Information can be found online at https://doi.org/10.1016/j.cub.2020.12.034.

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## STAR METHODS

### KEY RESOURCES TABLE

| REAGENT or RESOURCE | SOURCE | IDENTIFIER |
|---------------------|--------|------------|
| **Bacterial and virus strains** | | |
| AAV2/1-Syn-GCaMP6f-WPRE | UPenn Vector Core | AV-1-PV2822 |
| **Chemicals, peptides, and recombinant proteins** | | |
| Isoflurane (IsoFlo vet) | Zoetis | Cat# 199112 |
| Fentanyl | Hameln | Cat# 621062 |
| Midazolam (Dormicum) | Hameln | Cat# 516081 |
| Medetomidine (Dormitor) | Orion | Cat# 068824 |
| Flumazenil (Anexate) | Hameln | Cat# 55081 |
| Atipamezole (Antisedan) | Orion Pharma | Cat# 1639405 |
| Chlorprothixene | Sigma | Cat# C1671-1G |
| Jet Denture Repair Powder | Lang Dental | Item# 1230CLR |
| Super Glue Precision | Loctite | Cat# 2062278 |
| Ultrasound gel (NeurGel) | Spes Medica | Cat# NEURGEL250V |
| Silicone oil (10,000 molecular weight) | Lrp Hitemp | Cat# 68130 |
| **Deposited data** | | |
| Original datasets on GitHub | This paper | https://github.com/Neurune/ OpticFlowCortex |
| **Experimental models: organisms/strains** | | |
| Mouse: C57BL/6J | Janvier Labs | C57BL/6JRj |
| Mouse: FRMD7tm1a(KOMP)Wtsi | KOMP Repository | Project ID: CSD48756 |
| Mouse: FRMD7tm1b(KOMP)Wtsi | KOMP Repository | Project ID: CSD48756 |
| Mouse: Edil3Tg(Slox2-cre)1Amc/J | Jackson Laboratory | Stock # 004783 |
| **Software and algorithms** | | |
| MATLAB | The MathWorks | https://ch.mathworks.com/products/matlab |
| LabView | National Instruments | https://www.ni.com/labview/d/ |
| Psychophysics Toolbox | | http://psychtoolbox.org/ |
| SciScan v1.3 | Scientifica | https://sciscan.sciencifica.uk.com/ |
| EyeLoop | | https://github.com/simonarvin/eyeloop |
| Spherical stimulus correction for mice | Spencer Smith, Labrigger | https://labrigger.com/blog/2012/03/06/ mouse-visual-stim/ |
| Suite2p | | https://github.com/cortex-lab/Suite2P |
| Quine and McCluskey algorithm | MathWorks File Exchange | https://se.mathworks.com/matlabcenral/ fileexchange/37118-mintruthable-flags |
| **Other** | | |
| Borosilicate glass micropipettes | Sutter Instruments | Item# BF100-50-10 |
| Picospritzer III | Parker | Cat# 051-0530-900 |
| Feedback-controlled heating pad | World Precision Instruments | Item# ATC2000 |
| Titanium imaging chamber | This paper | Custom |
| Gelfoam sponges | Pfizer | Item# G50825 |
| Glass coverslips (0.15 mm thickness) | Warner Instruments | Cat# 64-0700 |

## RESOURCE AVAILABILITY

### Lead contact

Further information and requests for resources should be directed to and will be fulfilled by the Lead Contact Keisuke Yonehara (keisuke.yonehara@dandrite.au.dk).
Materials availability
This study did not generate new unique reagents.

Data and code availability
Original data and code have been deposited to GitHub (https://github.com/Neurune/OpticFlowCortex).

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Mice
All experimental procedures were approved by the Danish National Animal Experiment Committee (2020-15-0201-00452) and were performed in compliance with the Guide for the Care and Use of Laboratory Animals. Wild-type mice (C57BL/6J) were obtained from Janvier Labs. Fmd7tm1a(KOMP)Wtsi mice, obtained as Fmd7tm1a(KOMP)Wtsi from the Knockout Mouse Project (KOMP) Repository14,24,25: Exon 4 and the neo cassette flanked by loxP sequences were removed by crossing with female Cre-deleter Edil3Tg(Sox2-cre)1Amc1/J mice (The Jackson Laboratory: stock 4783), as confirmed by PCR of genome DNA, and maintained in a C57BL/6J background. Experiments were performed on 9 male and female wild-type mice, and 8 female and male Fmd7tm1a mice. All mice were 12–18 weeks old during imaging experiments. Mice were kept on a reversed 12 h dark/light cycle and housed in groups of up to four littermates per cage.

Chronic cranial windows
Mice were anaesthetized with an intraperitoneal injection of a Fentanyl (0.05 mg/kg body weight; Hameln), Midazolam (5.0 mg/kg body weight; Hameln), and Medetomidine (0.5 mg/kg body weight; Domitor, Orion) mixture. To prevent neural edema during or after surgery, dexamethasone (0.2 mg/kg body weight; Dexam, Bimeda) was injected subcutaneously. Body temperature was maintained using a feedback-controlled heating pad (ATC2000, World Precision Instruments) and eyes were protected from dehydration with eye ointment (Viscotears, Novartis). The scalp overlying the skull was removed, and a custom head-fixing imaging head-plate, using a feedback-controlled heating pad (ATC2000, World Precision Instruments). The stimulated contralateral eye was kept lubricated by a thin layer of silicone oil. A 5 mm glass coverslip (0.15 mm thickness, Warner Instruments) was placed onto the brain to shield and gently compress the underlying cortex. The cranial window was sealed using a cyanoacrylate-based glue (Super Glue Precision, Loctite) and dental cement (Jet Denture Repair Powder). The center of the head-plate was positioned above V1 (stereotaxic coordinates: 2.5 mm lateral, 1 mm anterior of lambda). A 5 mm craniotomy was made in the center of the head-plate. After removing the skull flap, the cortical surface was kept moist with Ringer’s solution (in mM): 110 NaCl, 2.5 KCl, 1 CaCl2, 1.6 MgCl 2, 10 glucose, and 2.5 mm lateral, 1 mm anterior of lambda). A 5 mm craniotomy was made in the center of the head-plate. After removing the skull flap, the cortical surface was kept moist with Ringer’s solution (in mM): 110 NaCl, 2.5 KCl, 1 CaCl2, 1.6 MgCl 2, 10 glucose, and 22 NaHCO3. A 5 mm glass coverslip (0.15 mm thickness, Warner Instruments) was placed onto the brain to shield and gently compress the underlying cortex. The cranial window was sealed using a cyanoacrylate-based glue (Super Glue Precision, Loctite) mixed with black dental cement (Jet Denture Repair Powder mixed with iron oxide powdered pigment), to prevent light contamination from the visual display. In addition, a black O-ring was mounted on top of the head-plate to further prevent any light contamination during imaging. Mice were administered subcutaneous analgesia (0.1 mg/kg body weight; Temgesic, Indivior) and returned to their home cage after anesthesia was reversed with an intraperitoneal injection of a Flumazenil (0.5 mg/kg body weight; Hameln) and Atipamezole (2.5 mg/kg body weight; Antisedan, Orion Pharma) mixture.

Virus injections
Mice were anaesthetized with an intraperitoneal injection of a Fentanyl (0.05 mg/kg body weight; Hameln), Midazolam (5.0 mg/kg body weight; Hameln), and Medetomidine (0.5 mg/kg body weight; Domitor, Orion) mixture. To prevent neural edema during or after the surgery, dexamethasone (0.2 mg/kg body weight; Dexam, Bimeda) was injected subcutaneously. Three small 0.4 mm diameter craniotomies were made and ~100–150 nL AAV2/1-Syn-GCaMP6f-WPRE (2.13 × 1013 vg/ml, Penn Vector Core #AV-1-PV2822) slowly injected (5 min/injection) at a depth of ~275 μm below the dura. By using the pan-neuronal promoter, synapsin, this viral vector drives GCaMP6f expression in both excitatory and inhibitory neurons. Injections were made using a borosilicate glass micropipette (30 μm tip diameter) and a pressure injection system (Picospritzer III, Parker). The micropipette was advanced at a 20° angle relative to vertical to minimize compression of the brain. To prevent backflow during withdrawal, the micropipette was kept at the injection site for 10 min before it was slowly retracted. The skin was sutured shut and postoperative analgesia was administered subcutaneously (0.1 mg/kg body weight; Temgesic, Indivior). Mice were returned to their home cage after anesthesia was reversed with an intraperitoneal injection of a Flumazenil (0.5 mg/kg body weight; Hameln) and Atipamezole (2.5 mg/kg body weight; Antisedan, Orion Pharma) mixture.

Intrinsic signal retinotopic mapping
Before two-photon calcium imaging, cortical visual areas of each mouse were identified by intrinsic signal optical imaging as previously described13. Mice were anaesthetized with isoflurane (2–3% induction) and head-fixed in a custom holder. Chlorprothixene was administered intraperitoneally (2.5 mg/kg body weight; Sigma) as a sedative13, and isoflurane reduced to 0.5–1% during visual stimulation. Core body temperature was maintained at 37–38°C using a feedback-controlled heating pad (ATC2000, World Precision Instruments). The stimulated contralateral eye was kept lubricated by a thin layer of silicone oil. A 2 × air-objective (Olympus, 0.08 NA) was mounted on our Scientifica VivoScope, equipped with a CMOS camera (HD1-D-D1312-160-CL-12, PhotonFocus). The camera was connected to a Matrox Solios (eCL/XCL-B) frame-grabber via Camera Link. The microscope was defocussed 400–600 μm down from the pial surface, where intrinsic signals were excited using a red LED (KL1600, Schott) delivered through a 610 nm long-pass filter (Chroma). Reflected light was captured through a 700 ± 50 nm band-pass filter (Chroma) positioned in front of
of the camera, and images were collected at 6 frames per second. The 47.65 × 26.87 cm (width × height) display was angled 30° from the midline of the mouse and the perpendicular bisector was 10 cm from the bottom of the display, centered on the display left to right, and 10 cm from the eye. This resulted in a visual field coverage from −41.98° to 60.77° (total 102.75°) in elevation, and from −67.23° to 67.23° (total 134.46°) in azimuth. Retinotopic maps were generated by sweeping a spherically corrected full-field bar across the display (see Key Resources Table). The bar contained a flickering black-and-white checkerboard pattern on a black background. The width of the bar was 12.5° and the checkerboard square size was 25°. Each square alternated between black and white at 4 Hz. In each trial, the bar was drifted ten times in each of the four cardinal directions, moving at 8°/s. Usually, two to four trials resulted in well-defined retinotopic maps. From the raw image data, we used the response time course for each pixel and computed the phase and magnitude of the Fourier transform at the visual stimulus frequency. The phase maps were then converted into retinotopic coordinates from the geometry of our setup. From this, we identified visual area borders based on the visual field sign maps and superimposed those borders with the anatomical blood-vessel images to accurately localize visual cortical areas.

**Two-photon calcium imaging**

Imaging was initiated two weeks after virus injections. Mice were awake during all imaging sessions as previously described. To habituate mice to handling and the experimental conditions, one week after cranial window implantation, each mouse was head-fixed onto the imaging stage with its body restrained in a cylindrical cover, reducing struggling and overt body movements. The habituation procedure was repeated for at least four times for each mouse at durations of 15, 30, and 60 min on days one, two, and three, respectively. At the end of each session, mice were rewarded with chocolate paste. Imaging session lasted 1–2 h. The area targeted for two-photon imaging was localized by previous intrinsic signal optical imaging. Imaging was performed from layer 2/3, 120–275 μm below the dura, using a Scientifica VivoScope with a 7.9 kHz resonant scanner running SciScan, and with dispersion-compensated 940 nm excitation provided by a mode-locked Ti:Sapphire laser (MaiTai DeepSee, Spectra-Physics) through an Olympus 25 × (1.05 NA) objective. The emitted fluorescence photons were reflected off a dichroic mirror (525/50 nm) and collected using a GaAsP photomultiplier tube (Scientifica). Clear ultrasound gel (NeurGel, Spes Medica) was used as immersion medium. To prevent light leakage from the visual stimulation, the objective was shielded with black tape, in addition to the O-ring mounted on top of the head-plate, and black cloth covered the microscope. Average excitation laser power varied from 40 to 65 mW. Images had 512 × 512 pixels, at 0.2 μm per pixel, and were acquired at 30.9 Hz using bidirectional scanning. We observed no sign of GCaMP6f bleaching during experiments. Each mouse was imaged repeatedly over the course of 2–3 weeks.

**Visual stimulus for two-photon calcium imaging**

For visual stimulation during two-photon calcium imaging experiments, two 47.65 × 26.87 cm (width × height) displays were angled 30° from the midline of the mouse on the left and right side; each display subtending 115.61° in azimuth and 80.95° in elevation (Figure 1B). The visual stimulus protocol employed was adapted from a previous study. Full-field vertical sinusoidal gratings (100% contrast; spatial frequency of 0.03 cycles/°) with a spherical correction to simulate projection onto a virtual sphere moved horizontally at speeds of 10 or 40°/s. The horizontal transition consisted of eight separate conditions (6 s each, interspersed with 4 s of gray screen between conditions): 1) Left nasal, 2) Left temporal, 3) Right nasal, 4) Right temporal, 5) contraversive, 6) ipsiversive, 7) forward, 8) backward (Video S1). Conditions 1–4 and were thus monocular, and conditions 5–8 binocular, simulating the rotational and translational optical flow experienced during turning and straight movements, respectively. The sequence of eight conditions was repeated in six trials. The mouse’s binocular visual field (central 40°) did not contain the visual stimulus, to ensure only stimulation of the monocular visual field.

**Eye movement tracking**

In a subset of experiments, we tracked eye movements in awake mice during presentation of our visual stimulus protocol (Figure S2). We employed an eye-tracking system developed in our laboratory and recently described in detail. Briefly, a small 45° hot mirror was aligned above a CCD camera (Guppy Pro F-031, AlliedVision) lateral to the position of the mouse. The camera was positioned below the visual field. Behind the visual stimulus display, a near-infrared light source (SLS-02082-B, Mightex Systems) was angled at 45° to illuminate the recorded eye. The camera was connected to a PC via a dedicated frame grabber (FIW62, ADLINK) and images were collected at ~65 frames per second. Using the eye-tracking software, EyeLoop, images were processed, and pupil and corneal reflection coordinates were computed. From these, the angular eye coordinates (x and y) were calculated. Horizontal eye speed was obtained by taking the first derivative of the horizontal eye coordinates (Vy), and low pass filtering Vy, with a 1 s moving average filter. Saccades were identified as events with a speed > 20°/s. Stimulus-triggered horizontal eye speed and saccade rate traces were obtained by averaging over all trials.

**Data analysis**

**Preprocessing of two-photon calcium imaging data**

Imaging data were excluded from analysis if motion along the z axis was detected. Raw two-photon imaging movies were corrected for in-plane motion using a piecewise non-rigid motion correction algorithm implemented in MATLAB (Mathworks). To detect regions of interest (ROIs) we used the MATLAB implementation of Suite2p. ROIs were automatically detected using the motion-corrected frames and afterward manually curated using the Suite2p graphical user interface. From the motion-corrected movies and detected ROIs, we extracted the fluorescence time courses within each ROI. To correct the calcium traces for contamination...
from the surrounding neuropil, we also extracted the fluorescence of the surrounding neuropil for each ROI. The time series of the neuropil decontaminated calcium trace, $F_d(t)$, was described by:

$$F_d(t) = F(t) - \alpha \times N(t)$$

where $F(t)$ is the somata calcium trace, $N(t)$ is the neuropil trace, and $\alpha$ is the contamination factor. The contamination factor was determined for each ROI as previously $^{47}$. Briefly, $F$ and $N$ traces were first low pass filtered using the 8th percentile in a 180 s moving window, yielding $F_s$ and $N_s$, respectively. These were then used to establish $F_{d1}(t) = F(t) - N_{d1}(t)$ and $N_{d1}(t) = N(t) - N_s(t)$. $F_1$ and $N_1$ were then used to determine $\alpha$ as previously described $^{47,51}$. Using the neuropil decontaminated calcium trace, baseline calcium fluorescence, was computed for each stimulus condition as the mean during the pre-stimulus period $^{10}$. Fluorescence values were then converted to relative change compared to baseline according to: $\Delta F/F = (F_d - F)/F$, where $F_d$ is the instantaneous neuropil decontaminated calcium trace and $F$ is the baseline calcium fluorescence. The mean neuronal responses were computed as the average response during the visual stimulus, and the mean and standard deviation across trials for each stimulus condition was computed for each neuron. To identify neurons for further in-depth analysis we used three inclusion criteria: 1) Neurons were defined as visually reliable if their mean $\Delta F/F$ to the preferred stimulus condition exceeded 10%; 2) A response reliability index, $\delta$, was computed for each neuron as:

$$\delta = \frac{\mu_{\text{pref}} - \mu_{\text{blank}}}{\sigma_{\text{pref}} + \sigma_{\text{blank}}}$$

where $\mu_{\text{max}}$ and $\sigma_{\text{max}}$ are the mean and standard deviations of the response to the preferred stimulus condition respectively, and $\mu_{\text{blank}}$ and $\sigma_{\text{blank}}$ are the mean and standard deviations of the response to a blank stimulus respectively $^{10}$. Neurons with $\delta$ exceeding 0.6 were defined as reliable; and 3) A signal-to-noise ratio (SNR) was computed for each neuron as:

$$\text{SNR} = \frac{\mu_{\text{pref}}}{\sigma_{\text{SD}}}$$

where $\mu_{\text{pref}}$ is the mean of the response to the preferred stimulus condition and $\mu_{\text{SD}}$ is the mean of the standard deviation of the fluorescence trace during the baseline period (0.5 s before stimulus onset) for each trial $^{52}$. Neurons with SNR exceeding 0.5 were defined as robustly responding. Only neurons that fulfilled all inclusion criteria at both stimulus speeds were included for further analysis procedures.

**Response profile classification**

In order to classify the response of individual neurons into separate functional groups, representing distinct response profiles, we employed regression analysis similar to previously described $^1$. First, we summarized the response of each neuron by a tuning curve, including the mean $\Delta F/F$ for each of the eight stimulus conditions. We compiled this tuning curve for both stimulus speeds, and we determined the speed in which the highest mean $\Delta F/F$ was evoked; noted as the preferred speed of the neuron. By considering the response selectivity of a neuron to the eight stimulus conditions, we assumed that the response profile regressors could be described by an indicator function, $R$, as follows:

$$R(x) : = \begin{cases} 
1 & \text{if responsive to } x \\
0 & \text{if not responsive to } x 
\end{cases}$$

where $x$ is the stimulus condition, and $2^8$ (i.e., 256) possible regressors exist for $R$ (Figure 1G). These 256 regressors correspond to the possible response combinations from the monocular and binocular stimulations in the nasal and temporal directions. For each neuron we then computed the linear Pearson’s correlation for its tuning curve at the preferred speed against each of the 256 regressors and determined the regressor with the highest correlation (Figures S3A–S3C). All neuronal tuning curves had high correlation with its assigned response regressor (mean correlation coefficient, 0.91 ± 0.05, $n = 26712$ neurons from 17 mice). The response regressors were functionally described using a MATLAB implementation of the Quine and McCluskey algorithm (see Key Resources Table), in which the Boolean functions were minimized to find the logical function for each response profile that use only a small number of logical operations $^7$. Here, we focused on the simple (MoDS, BiDS, and non-DS), binocular-suppressed (DS, and non-DS), and translation-selective or rotation-selective (FT, BT, CR, and IR) response classes (Figure 2). The simple and translation- or rotation-selective response classes are responsive to both monocular and binocular motion stimulation, and these were identified and described in detail previously $^7$. In this work, we identified the binocular-suppressed response class, characterized by only responding to monocular motion stimulation, in a DS or non-DS manner (Figure 2). The binocular-suppressed functional groups (regressor IDs) were described by the following Boolean logical operations:

$$\#6 = (\neg NL \cap \neg NR) \cup (\neg TR \cap NR)$$

$$\#7 = (\neg TL \cap TR) \cup (TR \cap \neg NR)$$

$$\#8 = (TL \cap \neg TR) \cup (TL \cap \neg NL)$$
Response profile shuffling analysis

regressor (Figure S3D).  #56: right temporal, contraversive rotational, and backward translational) as the peak amplitude in the cross-correlation (same regressor for each of the stimulus conditions that evoked responses in the particular regressor profile (for example, regressor #56: right temporal, contraversive rotational, and backward translational) as the peak amplitude in the cross-correlation (Crrrnψsi, qψsi):  

\[ Crrrnψsi, qψsi = \int F_p(t) * F_q(t - \tau) \, d\tau \]

where \( F_p \) and \( F_q \) indicate Ca\(^{2+} \) transients (\( \Delta F / F \)) during stimulus condition \( s_i \) and \( s_j \) in neuron \( p \) and \( q \), respectively, which are both assigned to regressor \( r_n \). From these correlation strength values we obtained distribution histograms of correlation strength for each regressor (Figure S3D).

Response profile shuffling analysis

To evaluate the similarity in response time courses of neurons assigned to their respective regressors, we computed the correlation of Ca\(^{2+} \) transients across the stimulus conditions. First, we computed the correlation strength between pairs of neurons assigned to the same regressor for each of the stimulus conditions that evoked responses in the particular regressor profile (for example, regressor #56: right temporal, contraversive rotational, and backward translational) as the peak amplitude in the cross-correlation (Crrrnψsi, qψsi):

where \( F_p \) and \( F_q \) indicate Ca\(^{2+} \) transients (\( \Delta F / F \)) during stimulus condition \( s_i \) and \( s_j \) in neuron \( p \) and \( q \), respectively, which are both assigned to regressor \( r_n \). From these correlation strength values we obtained distribution histograms of correlation strength for each regressor (Figure S3D).

Response similarity analysis

To examine similarities and disparities in response class distributions across visual areas in wild-type and Frmd7tm mice, we carried out a hierarchical clustering analysis. For this, we used the mean proportion of neurons assigned to each of the 33 response types (i.e., regressors) within the simple, translation- and rotation-selective, and binocular-suppressed response classes to create a monocular and binocular motion flow “fingerprint” for each visual area in wild-type and Frmd7tm mice. To create a hierarchical cluster tree, we used the linkage function in MATLAB, and visualized the result in a dendrogram (Figure 5A). For quantifying similarities and disparities across visual areas within wild-type and Frmd7tm mice (Figures 5B and 4C), we computed Pearson’s correlation coefficients using the motion flow fingerprint of each visual area. To quantify the proportions of monocular versus binocular functional groups within each visual cortical area, we computed a binocular optic flow index (BOFI). For this, we determined the proportion of monocular driven (simple MoDS and non-DS, and binocular-suppressed DS and non-DS) and binocular driven (simple BiDS, and FT, BT, CR, and IR) functional groups, and computed the BOFI as:

\[ \text{BOFI} = \frac{\% \text{ binocular} - \% \text{ monocular}}{\% \text{ binocular} + \% \text{ monocular}} \]

with a BOFI of 1 indicating that only binocular driven functional groups are represented, while a BOFI of −1 indicates only monocular driven groups are represented.

Comparison of response classes and functional groups among visual cortical areas

To compare eye movement speeds, or saccade rates, before and during visual stimulation or between wild-type and Frmd7tm mice, we carried out statistical investigations. First, we obtained shuffled response tuning curves for each neuron by randomly shuffling the mean \( \Delta F / F \) values recorded during the eight stimulus conditions, added with noise from a normal distribution:

\[ T'_k = t'_k + n \sim N(M, SD) \]

where \( T'_k \) is the de novo generated tuning curve of neuron \( k \) from the shuffled pattern of \( s_i \), \( t'_k \) is the original tuning curve of neuron \( k \) with the original pattern of \( s_i \), and \( n \) is a noise value obtained from a normal distribution, \( N \), with a mean, \( M \), and standard deviation, \( SD \), calculated from the original tuning curve of neuron \( k \). Next, we determined the frequency of each regressor profile using the shuffled response tuning curves. Using this, we estimated the false positive probability (\( P_{\text{false positive}} \)) for each regressor by comparing the frequency of assigned neurons in the shuffled dataset to that in the original dataset using bootstrap sampling (500 samples). A false positive was defined by the frequency of assigned neurons being higher in the shuffled dataset compared to in the original dataset. If \( P_{\text{false positive}} \) was lower than 0.05, the frequency of neurons assigned to a given response regressor was considered significantly higher than noise (or chance) levels.

Comparison of response classes and functional groups among visual cortical areas

To examine similarities and disparities in response class distributions across visual areas in wild-type and Frmd7tm mice, we performed a hierarchical clustering analysis. For this, we used the mean proportion of neurons assigned to each of the 33 response types (i.e., regressors) within the simple, translation- and rotation-selective, and binocular-suppressed response classes to create a monocular and binocular motion flow “fingerprint” for each visual area in wild-type and Frmd7tm mice. To create a hierarchical cluster tree, we used the linkage function in MATLAB, and visualized the result in a dendrogram (Figure 5A). For quantifying similarities and disparities across visual areas within wild-type and Frmd7tm mice (Figures 5B and 4C), we computed Pearson’s correlation coefficients using the motion flow fingerprint of each visual area. To quantify the proportions of monocular versus binocular functional groups within each visual cortical area, we computed a binocular optic flow index (BOFI). For this, we determined the proportion of monocular driven (simple MoDS and non-DS, and binocular-suppressed DS and non-DS) and binocular driven (simple BiDS, and FT, BT, CR, and IR) functional groups, and computed the BOFI as:

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with a BOFI of 1 indicating that only binocular driven functional groups are represented, while a BOFI of −1 indicates only monocular driven groups are represented.

Quantification and statistical analysis

To statistically evaluate differences between functional groups or individual response types in wild-type and Frmd7tm mice, we used the two-sided Mann-Whitney U test. To compare the Pearson’s correlation coefficients obtained from two independent samples, i.e., wild-type and Frmd7tm mice, we used the Fischer’s r-to-z transformation and obtained the corresponding two-sided p value. To compare eye movement speeds, or saccade rates, before and during visual stimulation or between wild-type and Frmd7tm mice,
we used two-sided Student’s t tests (paired or unpaired where appropriate). To compare stimulus-related changes in eye movement speeds, or saccade rates, across stimulus conditions, we used a repeated-measure analysis of variance (ANOVA) followed by multiple comparison testing using the Tukey honestly significantly different post hoc test. Center and spread values are reported as mean ± SEM. We used no statistical methods to plan sample sizes but used sample sizes similar to those frequently used in the field\textsuperscript{10,15,17}. Exact \( n \) (i.e., number of animals and neurons) is included in the Result section and Table S1. Data collection and analysis were not performed blind to the conditions of the experiments. No collected data were excluded from analysis. Statistical significance was defined as \( p < 0.05 \), where * \( p < 0.05 \) and ** \( p < 0.01 \). All statistical analyses were carried out in MATLAB.
Supplemental Information

Binocular integration of retinal motion
information underlies optic flow
processing by the cortex

Rune Nguyen Rasmussen, Akihiro Matsumoto, Simon Arvin, and Keisuke Yonehara
Figure S1. Identification of visual cortical areas using intrinsic signal optical imaging. Related to Figure 1. (A) Maps of horizontal (left) and vertical retinotopy (middle), and the corresponding visual field sign map (right) from four example mice. (B) Thresholded visual field sign patches showing the location of primary visual cortex (V1), and the higher visual areas: lateromedial (LM), anterolateral (AL), rostrolateral (RL), anterior (A), anteromedial (AM), and posteromedial (PM). Coordinates indicate posterior (P) and medial (M) directions.
Figure S2. Eye movements in wild-type and *Frdm7tm* mice during visual stimulus protocol. Related to Figure 1. (A) Upper: example traces of the horizontal position of the right eye (T, temporal; N, nasal), recorded in a wild-type mouse (left) and a *Frdm7tm* mouse (right) in response to the monocular and binocular horizontal motion conditions presented at 10 °/s. Middle and lower: trial-averaged horizontal eye speed and saccade rate time courses for the right eye, recorded in wild-type mice (left; n = 9 recordings) and *Frdm7tm* mice (right; n = 9 recordings) in response to the monocular and binocular horizontal motion conditions presented at 10 °/s. Error bars are mean ± SEM. (B) Quantification of mean horizontal eye speed (upper) and mean saccade rate (lower) before and during visual stimulation in wild-type (left) and *Frdm7tm* mice (right). Error bars are mean ± SEM. n.s., not significant, p ≥ 0.05, two-sided paired Student’s *t*-test. (C) Pairwise comparison of stimulus-related changes (pre vs stim) in mean horizontal eye speed (Δ eye speed,
left) and mean saccade rate (Δ saccade rate, right) across visual stimulus conditions in wild-type (upper) and Frmd7tm mice (lower). Heatmap color indicates p-values computed from the repeated measures ANOVA followed by Tukey honestly significantly different post hoc testing; all p-values ≥ 0.05. (D) Quantification of Δ eye speed and Δ saccade rate between wild-type (WT) and Frmd7tm mice across stimulus conditions. Error bars are mean ± SEM. n.s., p ≥ 0.05, two-sided unpaired Student’s t-test.
Figure S3. Regression analysis for classifying individual neurons to discrete response types. Related to Figures 1 and 3. (A) Trial-averaged fluorescence intensity (ΔF/F) time course for example layer 2/3 V1 neuron (#176) from a wild-type mouse in response to the monocular and binocular motion conditions at 10°/s. (B) The tuning curve at the preferred speed was correlated to each of the 256 regressors, yielding a correlation profile. Correlation coefficients were calculated as Pearson’s r. Neuron #176 showed the highest correlation with regressor #56 (r = 0.99) and was thus assigned to this response type. (C) Tuning
profile of neuron #176 and response profile of regressor #56. (D) Regressor profiles and response time courses for V1 neurons assigned to functional groups within the simple, translation- or rotation-selective, and binocular-suppressed response classes. MoDS, monocular DS; BiDS, binocular DS; FT, forward translational; BT, backward translational; CR, contraversive rotational; IR, ipsiversive rotational; N, nasalward; T, temporalward; L, left eye; R, right eye; NDS, non-DS; BiS, binocular suppressed; E, excited by; SP, specific. (E) Distributions of the correlation strength of response time courses for neurons assigned to the same regressor profile in the dataset obtained from V1 of wild-type mice.
Figure S4. Tuning of HVA neurons assigned to functional groups. Related to Figures 1 and 3. (A–C) Regressor profiles and tuning of RL/A (A), AM (B) and PM (C) neurons from wild-type mice assigned to functional groups within simple, translation- or rotation-selective, and binocular-suppressed response classes. MoDS, monocular DS; BiDS, binocular DS; FT, forward translational; BT, backward translational; CR, contraversive rotational; IR, ipsiversive rotational; N, nasalward; T, temporalward; L, left eye; R, right eye; NDS, non-DS; BiS, binocular suppressed; E, excited by; SP, specific.
Figure S5. Obtaining regressor frequency distribution. Related to Figure 3. (A) Distribution of all reliably responsive V1 neurons from wild-type mice (n = 3010 neurons from 4 mice) grouped according to the 256 regressor profiles (white, active; black, inactive) and response class (simple, translation- or rotation-selective, binocular-suppressed, and unclassified). Error bars are mean ± SEM. (B) Distribution from (A) ranked according to regressor frequency. The shaded region depicts the 50 most abundant regressors (as shown in Figure 3A).
Figure S6. Proportional changes for individual response types in wild-type and Frmd7tm mice. Related to Figure 4. (A–D) Proportion of V1 (A), RL/A (B), AM (C), and PM (D) neurons in simple, translation- or rotation-selective, and binocular-suppressed functional response types for wild-type and Frmd7tm mice. Error bars are mean ± SEM. *p < 0.05, **p < 0.01, two-sided Mann-Whitney U test. MoDS, monocular DS; BiDS, binocular DS; FT, forward translational; BT, backward translational; CR, contraversive rotational; IR, ipsiversive rotational; N, nasalward; T, temporalward; L, left eye; R, right eye; NDS, non-DS; BiS, binocular suppressed; E, excited by; SP, specific.
| Area | Total neurons | Consistently responsive | Animals |
|------|---------------|-------------------------|---------|
|      | WT: n         | Frmd7tm: n              |         |
| V1   | 5748          | 5534                    | 4       |
|      | 3010 (52%)    | 2925 (53%)              |         |
| RL/A | 6563          | 6746                    | 5       |
|      | 4165 (63%)    | 3125 (46%)              |         |
| AM   | 6664          | 6919                    | 5       |
|      | 4006 (60%)    | 3375 (49%)              |         |
| PM   | 5419          | 5868                    | 4       |
|      | 3059 (56%)    | 3047 (52%)              |         |

Table S1. Numbers of neurons sampled by visual cortical area and genetics. Related to Figures 1 and 3. Total neurons: total number (n) of neurons recorded in wild-type (WT) and Frmd7tm mice experiments for each visual cortical area. Consistently responsive: number (n) and percent of total of neurons that met the inclusion criteria for responsiveness ($\Delta F/F > 10\%$), reliability ($\delta > 0.5$), and signal-to-noise (SNR > 0.5) and were included for regressor correlation analysis. Animals: number (n) of WT and Frmd7tm mice that data were collected from for each area.