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Retinitis pigmentosa is associated with shifts in the gut microbiome

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The gut microbiome is known to influence the pathogenesis and progression of neurodegenerative diseases. However, there has been relatively little focus upon the implications of the gut microbiome in retinal diseases such as retinitis pigmentosa (RP). Here, we investigated changes in gut microbiome composition linked to RP, by assessing both retinal degeneration and gut microbiome in the rd10 mouse model of RP as compared to control C57BL/6J mice. In rd10 mice, retinal responsiveness to flashlight stimuli and visual acuity were deteriorated with respect to observed in age-matched control mice. This functional decline in dystrophic animals was accompanied by photoreceptor loss, morphologic anomalies in photoreceptor cells and retinal reactive gliosis. Furthermore, 16S rRNA gene amplicon sequencing data showed a microbial gut dysbiosis with differences in alpha and beta diversity at the genera, species and amplicon sequence variants (ASV) levels between dystrophic and control mice. Remarkably, four fairly common ASV in healthy gut microbiome belonging to *Rikenella* spp., *Muribaculaceae* spp., *Prevotellaceae* UCG-001 spp., and *Bacilli* spp. were absent in the gut microbiome of retinal disease mice, while *Bacteroides caecimuris* was significantly enriched in mice with RP. The results indicate that retinal degenerative changes in RP are linked to relevant gut microbiome changes. The findings suggest that microbiome shifting could be considered as potential biomarker and therapeutic target for retinal degenerative diseases.

Neuronal degeneration is an intricate process in which intrinsic and environmental stress can affect vulnerable neurons to promote disease. Mounting evidence highlights the importance of a bidirectional crosstalk between the gastrointestinal bacteria and the central nervous system^{1–3}, and the impact of gut microbiome on brain and behavior is being extensively reported in the literature^{4–8}. The homeostasis of the gut microbiome is critical for maintaining human health, and imbalances in the microbial composition of the gut profoundly influences critical features of host physiology, including the development of metabolic disorders such as diabetes and obesity^{7,9}. Emerging data support the potential role for the gut microbiome in modulating many aspects of the brain function and behavior, with effects on the stress response, mood and anxiety disorders, motor activity, social interaction and memory, among others^{10–15}.

The interplay between the brain and the gut bacteria is mainly mediated by neural and immune networks, with crosstalk interactions between both systems^{3,16}. Thereby, the microbiome–gut–brain signaling system influences key brain processes, including neurogenesis, neurotransmission, neuroinflammation and neuronal degeneration^{17–20}. In this context, experimental data has proved that intestinal microbiome influences brain response to injury^{21–23}, and vice versa²⁴, so that changes in gut microbiome may affect recovery and treatment following brain damage²⁴. Besides, dysbiosis of the human gut microbiome has been associated with neurodegenerative disorders of the central nervous system that include Parkinson's, Alzheimer's and Huntington's disease^{25–29}.

The retina has been historically considered a window to the brain, and anatomically the retina can be regarded as an extension of the central nervous system. The structural and functional features of the retina make this tissue highly vulnerable to stressors, and homeostasis alterations significantly influence the progress of retinal pathologies³⁰. Moreover, the retina reflects some of the pathological alterations of many neurodegenerative diseases and may provide information of brain pathology severity^{31,32}. In this context, a few recent studies have linked gut microbiome changes with some retinal degenerative diseases^{33,34}, including age-related macular degeneration (AMD)^{35–39}, glaucoma^{40–43} and diabetic retinopathy⁴⁴, even though the published results vary depending on the type and stage of the disease and between studies. On the other hand, in a previous study we have demonstrated that invasive infection from gastrointestinal microbiome can induce activation of retinal microglia⁴⁵, the primary resident immune cell of the retina.

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Retinitis pigmentosa (RP) is a heterogeneous group of inherited diseases that cause photoreceptor degeneration, eventually leading to complete blindness³⁰. The death of photoreceptors is accompanied by chronic microglial activation and neuroinflammatory processes^{46–48}, concomitant with an increase of reactive oxygen species^{49–52}. RP disease-causing mutations have been identified in more than 80 different genes⁵³. The rd10 mouse model of RP has a missense mutation in the phosphodiesterase 6b (*Pde6b*) gene^{54,55}, inducing rod photoreceptor degeneration, which leads to secondary cone photoreceptor death^{30,56}. Time courses of photoreceptor cell death and subsequent retinal degeneration in rd10 mice closely resembles the human disease process^{30,57}.

To date, there are no empirical studies in the literature analyzing the gut microbiome composition in retinitis pigmentosa. In this study we analyzed the gut microbiome in control and rd10 mice at postnatal day (P) 32, when dystrophic animals are expected to have suffered from extensive retinal degeneration. We assessed retinal degeneration by functional electroretinography (ERG) and morphological techniques, and we evaluated the gut microbiome by Illumina 16S rRNA gene amplicon sequencing. We have confirmed degenerative changes in neuronal and glial retinal cells and demonstrated alterations in gut microbiome populations of RP animals. These results reinforce the general concept of the interdependence of gut microbiome and the central nervous system homeostasis and suggest that gut microbiome could potentially constitute a therapeutic target for RP and other retinal degenerative diseases.

Results

Degenerative changes in retinitis pigmentosa mice. RP mice showed altered retinal function and morphology. ERG flash responses from rd10 mice were smaller than those obtained in C57BL/6J mice (Fig. 1a). In rd10 mice, maximum amplitudes observed for scotopic a- and b-waves were 12% and 34% (respectively) of the values obtained in C57BL/6J mice (Fig. 1b,c). Also, visual acuity showed visual thresholds significantly smaller in rd10 mice (50% less) than those obtained in control mice (Fig. 1d). On the other hand, the mean thickness of the ONL was smaller in rd10 than in control mice throughout the retina (Fig. 1e). On average, the ONL thickness in rd10 mice was 31% of the values obtained in C57BL/6J mice (18.6 ± 1.6 vs. 60.4 ± 2.0 μm). Cone photoreceptors in control mice showed a normal morphology with visible inner and outer segments and long axons, and normal pedicles (Fig. 1f). Conversely, in rd10 mice cones exhibited a degenerated morphology, with small size cones and an almost absent inner and outer segments (Fig. 1g). In these animals, cone axons were almost lost and pedicles came out from the cell bodies. Besides, rod outer segments of RP mice were shorter and more disorganized than those of control animals (Fig. 1f,g).

Photoreceptor death was associated to reactive gliosis in the retina of rd10 mice. In control mice, Iba1-positive microglial cells were scarce in the outer retina and exhibited morphological features typical of resting microglia (Fig. 1h). By contrast, rd10 mice showed evident changes in Iba1-positive cells, with higher number of positive cells than observed in C57BL/6J retinas, and abundant Iba1-positive cells in the outer nuclear layer (Fig. 1i). Moreover, Iba1-positive cells in rd10 retinas showed a phenotype characteristic of reactive microglia (Fig. 1i). Immunoreactivity for glial fibrillary acidic protein (GFAP) also evidenced a reactive gliosis in rd10 retinas. In C57BL/6J retinas, GFAP immunoreactivity was present only in the inner margin of the retina, corresponding to astrocyte cells (Fig. 1j). By contrast, retinal GFAP immunoreactivity in rd10 was present not only in the inner margin of the retina but also throughout Müller cells (Fig. 1k), which points to the activation of macroglial cells.

General gut microbial composition features. DNA from 8 mice's gut and stool (4 from C57BL/6J mice and 4 from rd10 mice) was extracted and the 16S rRNA marker gene was amplified with PCR using the primers 341F and 805R, and then sequenced with Illumina technology. Reads were quality-filtered, merged (see methods and Supplementary Table S1) and analyzed with QIIME2.2020⁵⁸. The denoiser tool Deblur, available in QIIME2⁵⁸, was used to remove sequencing errors (57–66% sequences per sample, Supplementary Table S2). After quality control processing, as a mean, $\approx 100,000$ final joined reads were obtained for denoising analysis that delivered $\approx 38,000$ reads for taxonomic classification and 16S rRNA gene data analysis (Supplementary Tables S1 and S2). Regarding general taxonomic features (Supplementary Figure S1), in both healthy C57BL/6J mice and diseased rd10 mice, the phyla *Bacteroidota* and *Firmicutes* were predominant in the gut representing 96% of the relative microbial abundance, followed by *Deferribacterota* and *Desulfobacterota* (Fig. 2a). At the species level, 12 were predominant in both mice groups and represented from 94.23 up to 97.91% of the relative abundance per sample (Fig. 2b). *Lactobacillus* spp. was the most abundant specie in both rd10 ($\approx 53\%$) and C57BL/6J ($\approx 38\%$) mice while an uncultured *Muribaculaceae* bacterium was placed the second most abundant specie (Fig. 2b). Other common bacterial species in the gut were also detected, such as *Bacteroides* spp. and *Alistipes* spp.

Altered gut microbiome: differences in alpha and beta diversity. Despite these similarities on general microbial features, apparent alpha and beta diversity differences were found in the microbial gut composition between healthy and diseased mice. First, regarding richness of amplicon sequence variants (ASV), higher number of ASV were found for control mice group ($n=94 \pm 2$) compared to diseased rd10 mice ($n=86 \pm 3$) ($p=0.0017$, Fig. 3). In addition, 49 unique ASV were only found in healthy mice (representing an accumulative relative abundance of 26.7%) whereas 48 were only found in rd10 mice (17.6% of the relative abundance) (see details in Supplementary Table S3). Second, more alpha-diversity was obtained for C57BL/6J healthy control mice when measured with Pielou's Evenness, Shannon's Diversity and Faith's Phylogenetic Diversity indices (Supplementary Figure S1 and Supplementary Table S4). Furthermore, the principal coordinate analysis (PCoA) for the beta-diversity at different taxonomic ranks from family to species (Fig. 4a) and ASV (Fig. 4b) levels showed that C57BL/6J control mice grouped together and separately from rd10 disease mice. Indeed, these beta-diversity differences based on unweighted UniFrac distance were statistically significant (PERMANOVA test, $p=0.03$, Supplementary Table S4) at the ASV level (Fig. 4b) between control and disease mice using Jaccard

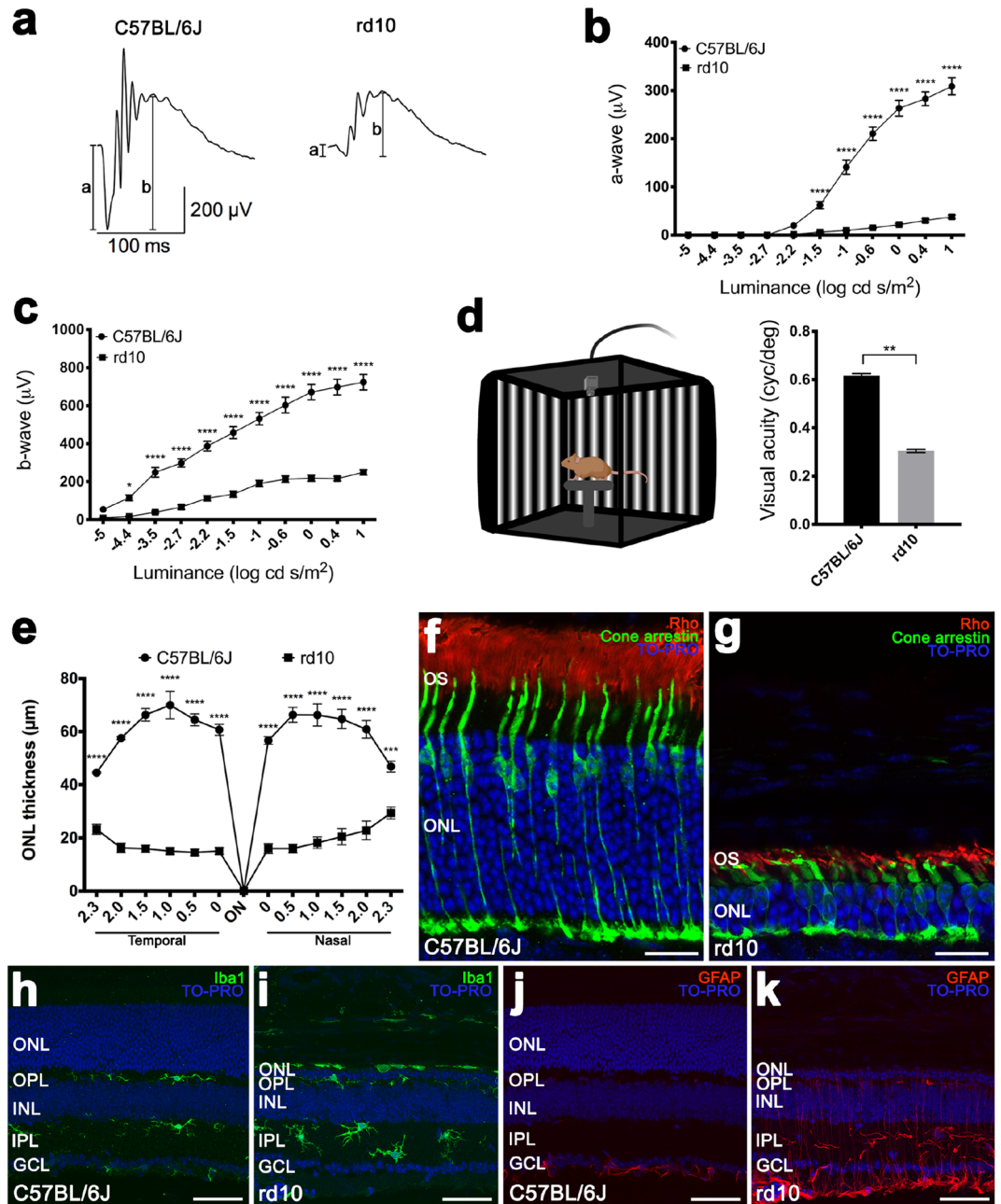


Figure 1. Retinal changes in RP mice. (a) Scotopic ERG responses to 1 log cd s/m² flashes from a normal C57BL/6J (left) and dystrophic rd10 (right) mouse. The amplitudes of both the a- and b-waves are represented. (b, c) Luminance-response curves for the a- (b) and b- (c) waves of C57BL/6J (circles) and rd10 mice (squares). (d) Configuration of the optomotor system (left, image created using BioRender; <https://biorender.com/>) and visual acuity thresholds for C57BL/6J and rd10 mice (right). (e) Mean outer nuclear layer thickness in C57BL/6J (circles) and rd10 (squares) mice, quantified in both the temporal and the nasal side of the retina. (f, g) Retinal sections showing the outer retina of a C57BL/6J (f) and rd10 (g) mouse immunolabeled against cone arrestin (cone cells, in green) and rhodopsin (Rho, rod cells, in red). Nuclei were stained with TO-PRO 3 (in blue). (h–k) Retinal sections from a C57BL/6J (h, j) and rd10 (i, k) mouse, immunolabeled against Iba1 (microglia, in green) or GFAP (activated macroglia, in red). The cell nuclei were stained with TO-PRO 3 (in blue). ANOVA, Bonferroni's test: **p* < 0.05, ***p* < 0.01, ****p* < 0.001, *****p* < 0.0001. ON: optic nerve, OS: outer segments, ONL: outer nuclear layer, OPL: outer plexiform layer, INL: inner nuclear layer, IPL: inner plexiform layer, GCL: ganglion cell layer. Scale bars: (f, g), 20 µm; (h–k), 50 µm.

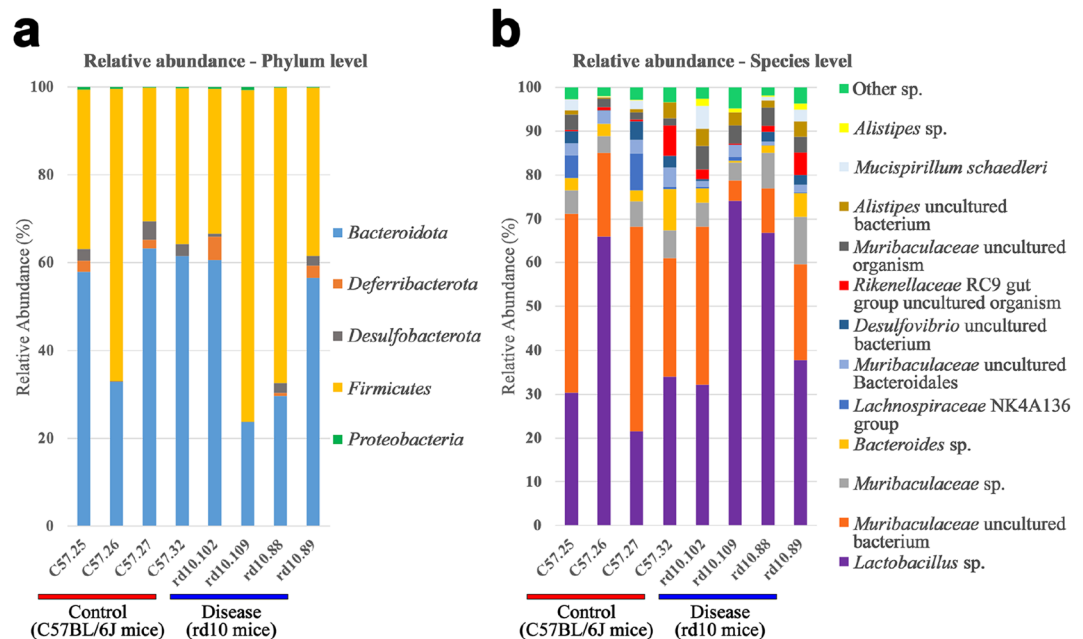


Figure 2. Taxonomic analysis of the mouse gut microbiome. (a) Relative abundance (%) at Phylum level for C57BL/6J and rd10 mice gut microbiota. Dominant phyla were *Bacteroidota* and *Firmicutes* in both groups. (b) Relative abundance (%) of the most abundant species found in C57BL/6J and rd10 samples. Only species which had at least 1% of relative abundance in one of the samples were represented.

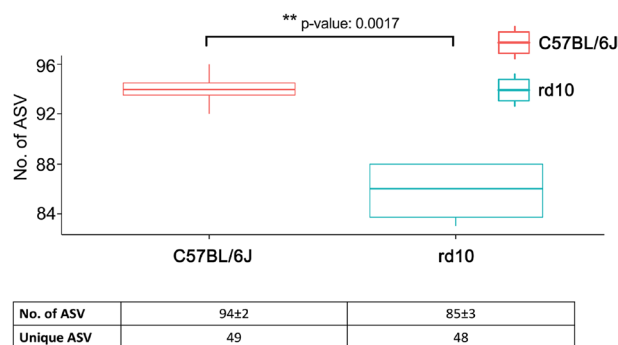


Figure 3. Amplicon sequence variant analysis. Top panel: boxplot comparing the number of amplicon sequence variants (No. of ASV). One-way ANOVA test showed statistically significant differences between the number of ASV in healthy (red) and rd10 (blue) mice. First row of the table contains mean \pm SD of ASV, the second one contains the number of unique ASV (not shared with the other group) for healthy (left) and rd10 mice (right). Graphic performed with the library ggboxplot available in R.

($p=0.03$) and Bray-courtis ($p=0.027$) distance indices (Supplementary Table S4). Remarkably, when analyzed those taxa significantly enriched in control and disease mice with ANCOM⁵⁹, which compares the relative abundance of each taxon with all the remaining features of the same category, data showed that four species (*Rikenella* spp., *Muribaculaceae* spp., *Prevotellaceae* UCG-001 spp., and *Bacilli* spp.) commonly present up to nearly 1% in healthy gut microbiome were absent in rd10 disease mice (Fig. 5 and Supplementary Tables S3 and S5). On the other hand, *Bacteroides caecimuris* was significantly overrepresented in rd10 mice with an average relative abundance of 0.7% (Fig. 5 and Supplementary Tables S3 and S5), while lacking in healthy gut mice. Finally, no difference in microbial composition was found between female versus male mice from both analyzed healthy and disease groups (tested with PERMANOVA, $p>0.05$, Supplementary Table S4).

Discussion

Previous studies have linked gut microbiome changes with retinal degenerative diseases. Here we demonstrate for the first time that degenerative changes in neuronal and glial retinal cells concur with shifts in gut microbiome composition in an animal model of retinitis pigmentosa. The reported deteriorations in retinal responsiveness and in photoreceptor cell number and morphology of rd10 mice agrees with that previously shown for these animals^{51,52}. Also, retinal reactive gliosis observed in the dystrophic animals are consistent with the increases in

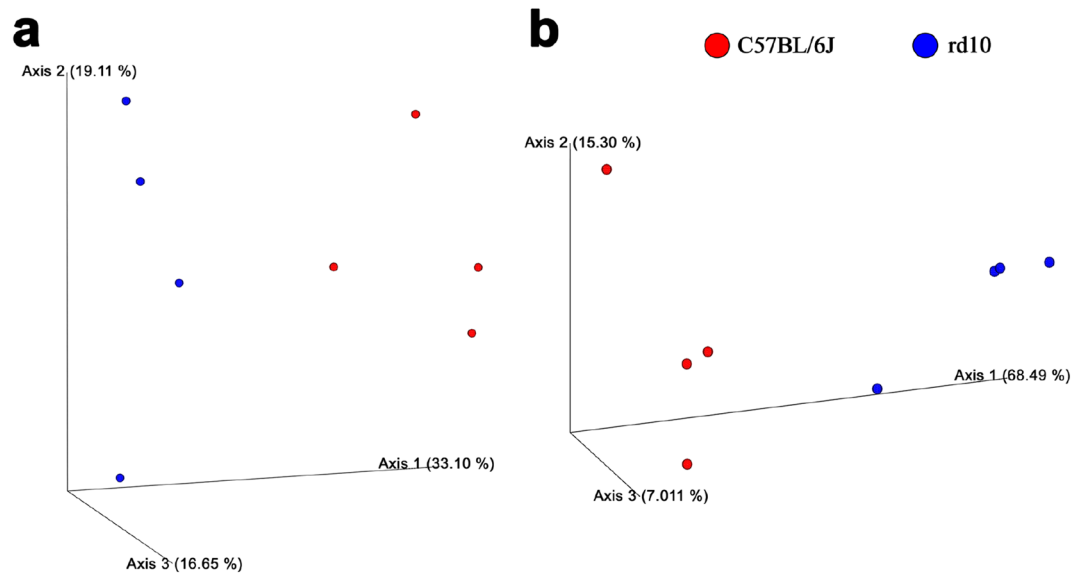


Figure 4. PCoA representation. (a) Principal coordinate analysis (PCoA) at species level, where C57BL/6J (red) and rd10 (blue) groups could be differentiated. (b) PCoA representing unweighted UniFrac distance for C57/6J (red) and rd10 (blue) mice gut were the two groups are separated from each other. The PERMANOVA test performed showed significant differences between the two groups.

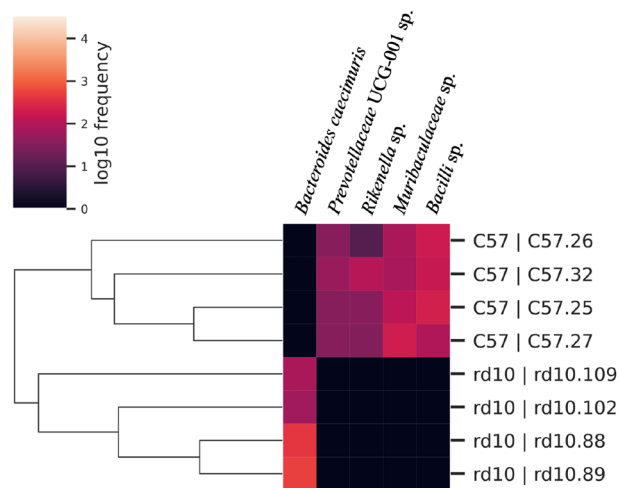


Figure 5. Heatmap that shows species that were identified by ANCOM as more abundant. *Bacteroides caecimuris* was more abundant in rd10 mice, while *Prevotellaceae* UCG-001 spp., *Rikenella* spp., *Muribaculaceae* spp. and *Bacilli* spp. were more frequent in C57BL/6J (C57) compared with the other mice group.

microglial cell numbers and Müller cell reactivity described in previous studies^{51,52}, and point to the activation of pro-inflammatory pathways in these animals. In this context, previous results have demonstrated significant increases of inflammation markers in rd10 mice⁵², and augmented expression of proinflammatory cytokines has been previously reported by us in RP animals⁴⁸. The inflammatory state in retinitis pigmentosa animals persists throughout the life span even after photoreceptor loss⁴⁷, and concurs with significant increase of oxidative stress⁵². In fact, it is assumed that apoptotic cell removal, inflammation and oxidative stress are common features in all retinal neurodegenerative diseases, including age-related macular degeneration, glaucoma, diabetic retinopathy and retinitis pigmentosa³⁰.

Initiation and progression of some prevalent retinal neurodegenerative diseases have also been linked to changes in the homeostasis of gut microbiome^{33,34}. In our results, sequencing analysis of the gut microbiome in dystrophic and control mice showed differences in alpha and beta diversity and interestingly, these differences were statistically supported at the ASV level. In recent reviews on best practices for analyzing microbiomes^{60,61}, ASV methods have been proposed as the reference metric to unveil differences in terms of microbial composition and have demonstrated sensitivity and specificity as good or better than previous methods and better discriminate ecological patterns^{62–65}. Remarkably, there were a large fraction of unique ASV present in only one of the

groups (diseased or healthy), which overall contribution in relative microbial abundance varies between 17.6% (diseased mice) and 26.6% (healthy mice). For instance, four ASV classified as *Rikenella* spp., *Muribaculaceae* spp., *Prevotellaceae* UCG-001 spp., and *Bacilli* spp. were common in healthy gut microbiome but absent in the gut microbiome of retinal disease mice. Oppositely, *B. caecimuris*, normally rare in healthy gut microbiome, was significantly abundant in diseased mice. Thus, data showed a taxonomic partitioning for several ASV in the gut microbiome of diseased and healthy gut microbiomes. Precisely, these striking differences in terms of presence vs. absence of these unique ASVs likely explain our results on ASV microbial composition (Fig. 3) based on unweighted beta-diversity model (i.e. low and high abundant ASV have the same importance)⁶⁶. When analyzing the data based on weighted beta diversity metric, which takes into account the relative abundance of all ASVs, differences were not statistically significant between diseased and healthy mice. This might be explained because with weighted beta diversity model, the relative contribution of most predominant species and ASVs, such as *Lactobacillus* and other abundant species described in Fig. 2b, likely mask the overall contributions of those less abundant species/ASV, which individually have a minor contribution with a relative abundance between 0.06 and 6.85% each one depending on the group, despite there were contrasting differences in absence or presence for several ASV taxa (Fig. 5). Those unique low abundant ASV representing different rare bacterial taxa could be important in the gut's ecosystem since it has been proved that rare or low frequent bacteria have key roles driving ecosystems⁶⁷, for instance, determining the bacterial gut composition in termite after different diet variations⁶⁸.

It has been reported that genera *Rikenella* and *Prevotella* were prevalent in 101 healthy mice gut microbiomes (including the C57BL/6J strain), being identified in 73.3% and 79.2% of the analyzed samples, thus being considered part of the healthy core of mice gut⁶⁹. In addition, bacteria belonging to the family *Muribaculaceae* are related to colonic inner mucus layer formation and barrier function⁷⁰, and the abundance of *Muribaculaceae* correlates with increased production of short-chain fatty acids and enhanced longevity in mice⁷¹. Besides, it has been demonstrated that relative abundance of *Muribaculaceae* negatively correlates with inflammatory mediators^{70,72}, and that fecal short-chain fatty acids concentrations are significantly reduced in Parkinson disease patients compared to controls⁷³. On the other hand, the abundance of *Prevotellaceae* has been reported to be reduced in feces of patients with neurological and psychiatric disorders¹⁶, including multiple sclerosis⁷⁴, Parkinson disease^{25,73} or major depressive disorder⁷⁵. Furthermore, previous studies have proved that the presence in gut microbiome of *Bacilli* spp., as *Lactobacillus*, can contribute to the production of short-chain fatty acids and collaborate in the maintenance of immune cells and the production anti-inflammatory response^{76,77}. Therefore, we can infer that the decline in the population density of these bacterial species may be related to the inflammatory and degenerative processes in RP mice.

Several different mechanisms have been proposed to explain how changes in the gut microbiome are linked to ocular diseases³⁴. Microbial imbalance can result in disruptions of the intestinal permeability and the blood-retinal barrier⁴³, thus allowing bacteria and their products to induce ocular cells to an inflammatory state^{34,35}. Moreover, it has been hypothesized that gut dysbiosis may be a cause of increased levels of oxidative stress in the central nervous system⁷⁸. But also vice versa, central nervous system injuries may cause changes in the gut environment, and trigger alterations of gut microbiome⁷⁹. In this context, it has been demonstrated that brain injury may induce changes in the gut microbiome composition via altered autonomic balance²⁴. All these hypotheses are in concordance with the context of neuroinflammation, oxidative stress and cell death observed in RP mice. The link between gut microbiome composition and retinal health suggests that different stages of retinal degeneration might correspond to different gut microbiome changes. In this context, it has been demonstrated that human pharyngeal microbiome varies depending on the stage of the disease in age-related macular degeneration⁸⁰. On the other hand, in our opinion, the restoration of the gut microbiome could prevent or reverse retinal degeneration. Previous studies have demonstrated that modification of the gut microbiota by microbiota transplantation⁸¹, or by changing the diet³⁶, can attenuate the development of age-related macular degeneration, and that restructuring of the gut microbiome by intermittent fasting prevents retinopathy in diabetic mice⁸².

Conclusions

Our results confirm previously described alterations in the morphology and function of the rd10 mouse, an animal model of retinitis pigmentosa, and demonstrate for the first time that retinal degenerative changes in neuronal and glial cells occurring in retinitis pigmentosa are concomitant with relevant gut microbiome changes. The findings could be extrapolated to patients suffering from retinitis pigmentosa or other ocular degenerative diseases and suggest that microbiome shifting could be considered as potential biomarker and therapeutic target for human retinal degenerative diseases. We realize that our results are preliminary and hope that it will lead and trigger further studies to elucidate the specificity of the interactions between the gut microbiome and retinitis pigmentosa or other retinal diseases. Continued investigations of the gut-retina axis could reveal unknown aspects of retinal diseases and potentially identify new relevant targets for therapeutic strategies.

Methods

Animals. Mice homozygous for the rd10 mutation (B6.CXBI-Pde6brd10/J) (n = 8) and wild-type C57BL/6J mice (Harlan Laboratories, Barcelona, Spain) (n = 8), half male, half female, were used in the study. Animals were maintained in cages under controlled temperature (23 ± 1 °C), humidity (60%) and photoperiod (12 h light/12 h dark, 50 lx). Water and food were provided ad libitum. At the end of the study, animals were humanely sacrificed by a lethal dose of sodium pentobarbital. The study has been approved by the Ethics Committee of the University of Alicante (UA-2018-07-06). All procedures were performed in conformity with current guidelines and regulations on the use of laboratory animals (European Directive 2010/63/EU, NIH, ARVO and ARRIVE) in an effort to reduce the number of animals used and limit unnecessary animal suffering.

Electroretinographic records. In the morning of postnatal day 32, scotopic ERG responses were recorded bilaterally following previously reported methodology⁵². After overnight dark adaptation, animals were anesthetized under dim red light by intraperitoneal administration of 100 mg/kg of ketamine (Imalgene, Merial Laboratorios S.A., Barcelona, Spain) and 4 mg/kg of xylazine (Xilagesic 2%, Laboratorios Calier, Barcelona, Spain), pupils were dilated with tropicamide 1% (Alcon Cusí, Barcelona, Spain), and the eyes were instilled with 0.2% polyacrylic acid carbomer (Novartis, Barcelona) to reduce dehydration and improve electrical connectivity with the recording electrodes (DTL fiber; Sauquoit Industries, Scranton, PA, USA). A reference needle electrode was placed in the head, under the scalp, and a ground electrode was placed in the mouth. During the recordings, into a Faraday cage, stable body temperature (37 ± 0.3 °C) and absolute darkness was maintained. Light stimuli (10-ms duration) were presented for at 11 logarithmically increasing luminance (from -5 to 1 log cd s/m²) by a Ganzfeld led stimulator. The responses to 3 to 10 consecutive stimuli were averaged for each light intensity. The spacing between flashes was 10 s for dim flashes (-5 to -0.8 log cd s/m²) and 20 s for bright flashes (0 to 1 log cd s/m²). A data acquisition board (DAM50; World Precision Instruments, Aston, UK) was used to amplify and band-pass filter the signal (1–1000 Hz, without notch filtering). Stimuli administration and data acquisition (4 kHz) were accomplished using PowerLab-AD system (AD Instruments, Oxfordshire, UK).

Optomotor test. Visual acuity (VA) was assessed in C57BL/6J and rd10 mice, by evaluating optomotor responses in the Argos system (Instead, Elche, Spain). As described previously⁵², spatial frequency thresholds were obtained by analyzing the response of the animals to vertically oriented drifting gratings (Fig. 1d). The initial spatial frequency tested was 0.088 cyc/deg and the temporal frequency was 0.8 Hz.

Tissue and stool collection. After ERG recording, animals were sacrificed, and tissue samples were collected. For microbial analysis, colon and ileum segments were removed and stored at -80 °C after quick immersion in liquid nitrogen. For morphological analysis of the retinas, the eyes were enucleated after the placement of a suture to mark the dorsal margin of the limbus. The eyes were then fixed with 4% (w/v) paraformaldehyde for 1 h at room temperature, washed with 0.1 M phosphate buffer (PB, pH 7.4) and cryoprotected through a series of increasing concentrations of sucrose (15, 20 and 30% (w/v)). Following, the cornea, lens and vitreous body were gently removed, the eyecups were embedded in Tissue-Tek OCT (Sakura Finetek, Zoeterwoude, Netherlands), frozen with liquid nitrogen and cut with a cryostat (CM 1900, Leica Microsystems, Wetzlar, Germany). Sections of thickness 16 µm were mounted on glass slides (Superfrost Plus; Menzel GmbH and Co. KG, Braunschweig, Germany) and stored at -20 °C.

DNA extraction. For the microbiome study, 8 tissue and stool samples were used. Half of them were rd10 and the other half were C57BL/6J, also there were 2 males and 2 females in each group. DNA was extracted from the samples using DNeasy PowerSoil Pro (QIAGEN, Germany) according to the manufacturer's protocol, including an extra sample incubation with CD2 at 4 °C during 5 min before being centrifuged. All centrifugations were carried at 15,100 G, minus the one used for removing the residual solution C5, centrifuged at 16,100 G.

PCR and sequencing of 16S rRNA gene amplicons. DNA from fecal and colon samples was subjected to amplification of polymerase chain reaction (PCR) using Pro341F (5'-TCGTCGGCAGCGTCAGATGTGTAT AAGAGACAGCCTACGGGNBGCASCAG3') and Pro805R (5'-GTCTCGTGGGCTCGGAGATGTGTATAA GAGACAGGACTACNVGGGTATCTAATC-3'), targeting the V3-V4 region of 16S rRNA gene. The PCR conditions were: 94 °C for 3 min, 25 cycles of 94 °C for 45 s, 51 °C for 1 min and 72 °C for 10 min. This was followed by 72 °C for 10 min. PCR amplicons were cleaned and indexed as indicated in the Illumina's MiSeq 16S Sequencing Library Protocol and sequenced with Miseq (2 × 300 pb). Sequencing was performed at the Genomics Center (FISABIO, Valencia, Spain).

Microbiome analysis. The sequenced data was quality filtered using prinseq-lite⁸³, eliminating 0.89% of the reads, with the following parameters min_length: 50, trim_qual_right: 30, trim_qual_type: mean, trim_qual_window: 20 and then joined with FLASH⁸⁴, using default parameters producing 814,069 amplicons (Supplementary Table S1). The primers were removed with cutadapt⁸⁵, and the cleaned merged reads were analyzed with QIIME2.2020⁸⁶. Low quality reads were eliminated with quality-filter q-score, eliminating ≈54 merged reads/ sample. Deblur was used to trim the sequences at position 417 to remove low quality regions⁸⁶.

Diversity was studied using the QIIME2 plugin q2-diversity for C57BL/6J-rd10 mice and male-female mice⁸⁸. Specifically, alpha-diversity was evaluated with Pielou's Evenness, Shannon's Diversity index and Faith's Phylogenetic Diversity index and compared with the no-parametric Kruskal-Wallis test. Beta-diversity was studied using PERMANOVA with the Bray-Courts distance, Jaccard distance and weighted UniFrac and unweighted UniFrac distances. PCoAs (-p-metric seculidean) were performed for representing beta-diversity and for all the taxonomic levels, that were previously collapsed. Taxonomy was assigned with the already pre-formatted SILVA 138 database (reproducible sequence taxonomy reference database management for the masses)⁸⁷. The comparison between taxa's relative abundance to find differentially abundant features was performed with ANCOM⁵⁹. Accumulative relative abundance for unique ASV (present in C57BL/6J or rd10 mice, never both) was calculated adding the relative abundance of each unique ASV.

Immunohistochemistry. Immunohistochemical assessment of the retinas was achieved following previously reported methodology⁵². Briefly, retinal sections were thawed at room temperature, washed 3 times with PB and incubated for 1 h in 0.1 M PB with 10% (v/v) normal donkey serum and 0.5% Triton X-100. After

that, sections were immunolabeled overnight at 4 °C under agitation using combinations of primary antibodies at different dilutions in 0.1 M PB with 0.5% Triton X-100: mouse monoclonal anti-rhodopsin (MAB5356, Merk Millipore, Darmstadt, Germany, 1:100), rabbit polyclonal anti-cone arrestin (AB15282, Merk Millipore, 1:200), rabbit polyclonal anti-ionized calcium-binding adapter molecule 1 (Iba1) (019-19741, Wako Chemicals, Richmond, VA, USA, 1:1000) and mouse monoclonal anti-glial fibrillary acidic protein (GFAP) (G3893, Sigma-Aldrich, Steinheim, Germany, 1:500). For objective comparison, rd10 and C57BL/6J retinas were processed in parallel. The slides were washed and then incubated with a mixture of corresponding secondary antibodies at a dilution of 1:100 in PB with 0.5% Triton X-100: AlexaFluor 488-anti-rabbit and AlexaFluor 555-anti-mouse (Invitrogen, Carlsbad, CA, USA). When corresponded, the nuclei marker TO-PRO 3-iodide (Invitrogen) was added at a dilution of 1:1000. Images were acquired on a Leica TCS SP8 confocal laser-scanning microscope (Leica Microsystems, Wetzlar, Germany).

Measurement of retina outer nuclear layer thickness. In order to assess photoreceptor death in retinal degenerative conditions, the thickness of the outer nuclear layer (ONL) was quantified in at least two non-consecutive sections per retina stained with hematoxylin. Retinal sections included the optic nerve and the temporal and nasal *ora serrata*. As the progression of the degeneration is not uniform throughout the retina, the quantification was performed every 0.5 mm, at distances of 0, 0.5, 1.0, 1.5, 2.0 and 2.3 mm from the optic nerve toward the periphery.

Statistical analysis. A one-way ANOVA was performed to assess the effects of genotype (rd10 vs. C57BL/6J) on ERG amplitude and ONL thickness, using the IBM SPSS statistics 24 software package (SPSS Inc, Chicago, IL, USA). Post hoc pairwise comparisons were done with the Bonferroni's test. To assess the effects of genotype on visual acuity, a Mann–Whitney U test was applied. Diversity parameters were statistically evaluated using different QIIME2 tools (<https://qiime2.org/>): the nonparametric Kruskal–Wallis test was used to compare alpha-diversity whereas beta-diversity was studied using PERMANOVA. The comparison between taxa's relative abundance was performed with ANCOM⁵⁹, which found features that were more abundant in a group as compared with the other. One-way ANOVA was applied to study abundance differences between different taxon levels and ASV numbers using the R statistical software (4.0.2)⁸⁸. A p value of less than 0.05 was considered to be statistically significant. All data were plotted as the average ± standard error of the mean.

Ethics declarations. All procedures were performed in conformity with current guidelines and regulations on the use of laboratory animals (European Directive 2010/63/EU, NIH and ARVO) in an effort to reduce the number of animals used and limit unnecessary animal suffering.

Approval for animal experiments. This study was approved by the Ethics Committee of the University of Alicante (UA-2018–07-06).

Data availability

The 16 s rRNA raw sequences generated during the current study were deposited at Sequence Read Archive (SRA) database which belongs to the National Center for Biotechnology Information. Bioproject number: PRJNA675447. Biosamples ID for C57BL/6J mice: SAMN16708365 (mouse 25), SAMN16708366 (mouse 26), SAMN16708367 (mouse 27) and SAMN16708368 (mouse 32). Biosamples ID for rd10 mice: SAMN16708371 (mouse 88), SAMN16708372 (mouse 99), SAMN16708369 (mice 102) and SAMN16708370 (mouse109).

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Author contributions

P.L., M.M.G. and N.C. initiated and led the study. O.K. and M.L.G. collected the data. The analysis was performed by O.K. and L.M.C.. P.L., L.M.C. and M.M.G. wrote the manuscript. All authors read and approved the final manuscript.

Competing interests

The authors declare no competing interests.

Additional information

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