The Cis-regulatory Logic of the Mammalian Photoreceptor Transcriptional Network

Timothy H.-C. Hsiau1, Claudiu Diaconu1, Connie A. Myers1, Jongwoo Lee1, Constance L. Cepko2,*, Joseph C. Corbo*1

1 Department of Pathology and Immunology, Washington University School of Medicine, St. Louis, Missouri, United States of America, 2 Howard Hughes Medical Institute, Department of Genetics, Harvard Medical School, Boston, Massachusetts, United States of America

The photoreceptor cells of the retina are subject to a greater number of genetic diseases than any other cell type in the human body. The majority of more than 120 cloned human blindness genes are highly expressed in photoreceptors. In order to establish an integrative framework in which to understand these diseases, we have undertaken an experimental and computational analysis of the network controlled by the mammalian photoreceptor transcription factors, Crx, Nrl, and Nr2e3. Using microarray and in situ hybridization datasets we have produced a model of this network which contains over 600 genes, including numerous retinal disease loci as well as previously uncharacterized photoreceptor transcription factors. To elucidate the connectivity of this network, we devised a computational algorithm to identify the photoreceptor-specific cis-regulatory elements (CREs) mediating the interactions between these transcription factors and their target genes. In vivo validation of our computational predictions resulted in the discovery of 19 novel photoreceptor-specific CREs near retinal disease genes. Examination of these CREs permitted the definition of a simple cis-regulatory grammar rule associated with high-level expression. To test the generality of this rule, we used an expanded form of it as a selection filter to evolve photoreceptor CREs from random DNA sequences in silico. When fused to fluorescent reporters, these evolved CREs drove strong, photoreceptor-specific expression in vivo. This study represents the first systematic identification and in vivo validation of CREs in a mammalian neuronal cell type and lays the groundwork for a systems biology of photoreceptor transcriptional regulation.

INTRODUCTION

Transcriptional regulatory networks (TRNs) lie at the center of organismal development and physiology [1,2]. Transcription factors (TFs) within these networks control the spatiotemporal expression of their target genes by binding to CREs, short (~300–600 bp) stretches of genomic DNA which can lie upstream, downstream, or within the introns of the genes they control. Significant progress has been made in the computational identification of putative CREs in a variety of species [3–7]. One recent study demonstrated the effectiveness of using deep phylogenetic conservation of non-coding DNA to identify developmentally active CREs in the mouse [4]. However, given the importance of cis-regulatory change in evolution [8] and the relatively limited number of deeply conserved non-coding regions in the mouse (~3000 were identified in [4]), it is likely that there are many tissue-specific CREs which do not fall within such regions.

Despite advances in the design of computational algorithms to identify CREs in mammalian genomes, the development of cheap, high-throughput assay systems for validating these computational predictions in vivo has lagged far behind. Most studies of mammalian cis-regulation to date have relied on mouse transgenesis as a means of assaying the enhancer function of CREs [4]. This technique is time-consuming, costly and subject to insertion site effects. On the other hand, rapid assays for mammalian CRE function have been developed in tissue culture systems, but it is not clear how such results translate into the in vivo behavior of CREs. We aim to demonstrate in this paper that rapid, inexpensive, high throughput analysis of mammalian CREs can be achieved by exploiting electroporation to introduce CRE-reporter fusion constructs either into living tissue in vivo or in ex vivo explant culture. This approach retains many of the desirable features of in vivo transgenic approaches to CRE analysis but is much more rapid and inexpensive.

Photoreceptor cells are sensory neurons that elaborate a highly specialized, membrane-rich organelle, the outer segment, which is exquisitely sensitive to light. These cells are particularly susceptible to degeneration. There are currently over 180 mapped disease loci which cause blindness in humans (http://www.sph.uth.tmc.edu/RetNet/). Of these, more than 120 have been cloned, and the majority of these genes have been shown to be specifically expressed, or highly enriched, in photoreceptors [9]. Unfortunately, there is currently no systems-level understanding of how transcriptional regulation of these disease genes is globally coordinated.

We aim to provide such understanding via analysis of the mouse photoreceptor transcriptional network. Numerous prior studies have demonstrated a central role in this network for the TFs Crx, Nrl, and Nr2e3 [10–15]. Crx is expressed in both rods and cones and activates numerous genes in both [9–11,16]. Nrl and Nr2e3, in contrast, are rod-specific and are required for activation of rod genes and repression of cone genes [13,15,17–20]. Nrl appears to be a molecular switch between cone and rod cell fate: if a photoreceptor precursor expresses Nrl it becomes a rod, otherwise it becomes a cone [21]. All three genes have been implicated in a variety of binding diseases in humans [14,22,23].


* To whom correspondence should be addressed. E-mail: cepko@genetics.harvard.edu (CC); jcorbo@pathology.wustl.edu (JC)
Previous studies of mice with mutations in these TFs identified a range of potential target genes [9,18,20,24,25].

Here, we present a more complete analysis of the genes affected by these mutations in order to define the nodes of the photoreceptor TRN. To understand how gene expression in this network is orchestrated, we identified and characterized many of the CREs linking these nodes via a combination of computational prediction and in vivo validation using electroporation of CRE-reporter fusion constructs. This analysis resulted in the identification of a cis-regulatory motif associated with high-level expression in photoreceptors. To test the importance of this motif for photoreceptor expression, a selection filter based on this motif was used to evolve photoreceptor-specific CREs in silico, and their functional activity was then demonstrated in photoreceptors. This study demonstrates the feasibility of a high throughput, in vivo, non-genomic approach to mammalian CRE analysis which can easily be applied to a wide range of different tissues.

RESULTS

The transcriptional network controlled by Crx, Nrl, and Nr2e3

In order to elucidate the global architecture of transcriptional regulation in mouse photoreceptors, analyses of genes expressed in Crx-/- retinas at P21 were carried out on Affymetrix microarrays. These data were integrated with those of previous studies of Nrl-/- and Nr2e3-/- retinas [20,25]. Using stringent criteria to define up- and downregulation, a total of 628 genes were identified as dysregulated in at least one of the three mutants (Fig. 1A; Tables S1, S2, S3, S4, S5 and S6). 179 genes were downregulated in Crx-/- (compared to 140 in Nrl-/- and 12 in Nr2e3-/-) whereas 93 genes were upregulated (compared to 297 in Nrl-/- and 55 in Nr2e3-/-). Our results accord well with two previous gene expression studies of the Crx mutant using cDNA microarrays and SAGE [9,16]. The dysregulated genes comprise many known photoreceptor genes including numerous components of both rod and cone phototransduction cascades.

A remarkable degree of overlap between genes downregulated in Crx-/- and Nrl-/- was discovered. 51% (72/140) of Nrl-downregulated genes were also downregulated in Crx-/- using stringent criteria (Fig. 1A). These results suggest that many photoreceptor genes are co-regulated by Crx and Nrl and are consistent with previous studies of individual photoreceptor genes showing co-regulation by these two TFs [26,27]. To analyze the effects of removing Crx, Nrl, and Nr2e3 activity from the retina simultaneously, Crx-/-;Nrl-/-;Nr2e3-/- double mutant mice were generated and their retinas were subjected to microarray analysis at P21. Since Nr2e3 is not expressed in Nrl-/- [13], these mice are effectively Crx-/-;Nrl-/-;Nr2e3-/-. We found an even greater number of downregulated genes in the double mutant (298) than in either of the single mutants using similar criteria (Table S2). This finding confirms the status of these TFs as major hubs within the photoreceptor TRN.

In order to validate the microarray results, we examined an expression profile using probes against 76 photoreceptor genes (~10% of the genes in the proposed network) on four mutant (Nr2e3-/-, Crx-/-, Nrl-/-, and Crx-/-;Nrl-/-) and two wild-type (C57BL/6 and 129S6/SvEv) backgrounds. The results for Nrl-/- and a subset of the results for Nr2e3-/- were reported previously [20,25] and are reproduced here for comparison. The ISH results directly reflected the changes seen by microarray for nearly all of the genes and genotypes examined. An exception was that five of the probes examined showed a greater degree of downregulation in Crx-/- than was predicted by microarray (see Table S1). This discrepancy may relate to the fact that the ISH was performed on retinas from a somewhat later timepoint (4–5 weeks postnatal) than when the microarray was performed.

Several examples of ISHs for genes residing at points of intersection in the Venn diagram in Fig. 1A are shown in Fig. 1C–F (the full intersection dataset is given in Table S7). Rho, which is typical of genes co-regulated by Crx and Nrl, shows marked downregulation in both of these mutants as well as in Crx-/-;Nrl-/- (Fig. 1C). Many cone genes such as Ombd3 were upregulated in both Nrl-/- and Crx-/- (Fig. 1D). In contrast, some cone genes such as Opa1 were upregulated in Nrl-/- and downregulated in Crx-/- (Fig. 1E). The reason for this difference is not known. A minority of photoreceptor-specific genes such as Pde6h were not changed in any of the mutants (Fig. 1F), suggesting the possibility of regulation by other TFs. ISH results for all 76 genes are presented in Table S1 and Fig. S1. Overall, these results show that the microarray data are a powerful predictor of photoreceptor gene expression patterns.

Downstream transcription factors and retinal disease genes in the photoreceptor network

Given the importance of TFs as regulatory nodes within TRNs we subjected all 628 genes in our network to Gene Ontology (GO) analysis to identify those downstream TFs which were dysregulated in Crx-/-, Nrl-/- or Nr2e3-/- retinas. Out of 27 genes with the GO classification “transcription factor activity”, we identified 17 sequence-specific TFs (Table S8). Aside from Crx, Nrl, and Nr2e3, this set included two known photoreceptor TFs: Otx2 and Rtrg. Otx2 was previously shown to be required for photoreceptor formation in the mouse and to activate Crx transcription [29]. Derepression of Otx2 in both Crx-/- and Crx-/-;Nrl-/- suggests that it is under negative feedback regulation by Crx (Fig. 1F). Rtrg is required for regulation of S-cone opsin in mice [29] and is also upregulated in Nrl-/-. Neurod1 mutant photoreceptors undergo degeneration suggesting a role for this TF in photoreceptor gene regulation [30]. Although not in the GO list, we also found Neurod1 to be upregulated in Nrl-/- (Table S1). Derepression of Neurod1 and Rtrg in Nrl-/- suggests that both genes are cone-enriched in the adult. Surprisingly, Nrl itself was downregulated in Crx-/- both by microarray and ISH (Fig. 1 and Table S1), suggesting a requirement for Crx in maintenance of Nrl expression.

Aside from Nrl, Nr2e3 is the only known rod-specific TF. Yet, the microarray data suggest that Esrh and Myf2c are also rod-enriched (Fig. 1G, K). Esrh is an orphan nuclear receptor required for placental development and maintenance of “stemness” in embryonic stem cells and germ cells [31-33]. Although previously shown to be photoreceptor-enriched [34], complete loss of expression in Nrl-/- strongly suggests that Esrh is rod-specific in the adult. Myf2c is required for heart development [35]. It was markedly downregulated in both Crx-/- and Nrl-/- by microarray suggesting that it too might be rod-enriched in the adult. Pdan1, also known as Blimp-1, is a well known regulator of B-cell and germ cell differentiation [36-38]. We found that Pdan1 is expressed at high levels in developing photoreceptors in a pattern indistinguishable from Crx itself (Fig. 1H, I). In contrast to Crx, Pdan1 expression is undetectable in the adult wild-type retina. Intriguingly, homeodomain-containing TFs on either side of the Crx locus were also dysregulated: Otx2 (upregulated in Crx-/-) and Cxox1 (upregulated in Nrl-/-). The latter result suggests that Cxox1, previously reported as a possible antisense regulator of Crx [39], is cone-enriched. Overall, these findings implicate a range of TFs in regulation of photoreceptor gene expression downstream of Crx and Nrl.
Given the remarkable genetic heterogeneity of human retinal disease, we wished to determine the extent to which the causative genes are regulated by Crx, Nrl, and Nr2e3. A set of 58 mouse orthologs of human retinal disease genes from the Retnet database were manually curated (Fig. 2). This set is meant to encompass all disease genes with known or probable photoreceptor-enriched patterns of expression based on analysis of the literature. 89% (50/56) of those disease genes for which microarray data were available.
Figure 2. Retinal disease genes in the photoreceptor network. This figure is a manually curated list of all retinal disease genes which are known or are likely to have a photoreceptor-enriched expression pattern. Those gene symbols which differ between mouse and human are marked with a single asterisk, and the human equivalents are given here: $\text{Cngb1} = \text{CNGB1}; \text{Gpr98} = \text{USH2C}; \text{Grk1} = \text{RHOK}; \text{Gucy2e} = \text{GUCY2D}; \text{Prom1} = \text{PROML1}; \text{Rgs9bp} = \text{R9AP}; \text{Rph1} = \text{RPI}; \text{Ush1g} = \text{SANS}$. 'Known CRE' indicates the location of published photoreceptor CREs which have experimental support (references are given in MATERIALS AND METHODS). Annotations in this column such as '6.5 Kb 5'' means the tested CRE contained the first 6.5 Kb upstream of the TSS. Known CREs predicted by our algorithm are highlighted in light green. 'Predicted CRE(s)' lists candidate CREs predicted by our algorithm which lie within +/- 15 Kb of the TSS. In some cases, two locations are given which represent that CRE prediction closest to the TSS and that with the highest score. Only one location is given when they are the same. Experimentally tested CREs are highlighted in red. The CREs tested for Crx and Elo4 lie more than 15 Kb downstream of the TSS. The CRE tested for Tulp1 scores below the cutoff threshold of 200. This CRE was selected for testing during the early phases of this project using an earlier version of PhastCons with which it scored above threshold. A third predicted CRE location is given for Opn1mwi which corresponds to the locus control region. 'CRE strength' indicates the estimated strength of the indicated CRE as tested by electroporation (see Fig. S2). The last four columns of the table show the wild-type-to-mutant ratios of the averaged microarray scores for the given gene. Dark green = downregulated under high stringency (as described in MATERIALS AND METHODS); light green = downregulated under low stringency; red = upregulated under high stringency; orange = upregulated under low stringency. A dash indicates that the gene was not significantly altered in the given mutant. 'NA' indicates that this gene is not represented on the Affymetrix Mouse 430 2.0 microarray.
were dysregulated to some extent in at least one of the four mutant backgrounds. This finding suggests that the majority of known human retinal disease genes expressed in photoreceptors is under the transcriptional control of Crx, Nrl, and/or Nrl2e3.

Computational and experimental analysis of photoreceptor cis-regulatory elements

These microarray and ISH studies identified many of the important nodes within the photoreceptor TRN. Next we sought to elucidate the connectivity between these nodes by testing the hypothesis that the majority of Crx, Nrl, and Nrl2e3 target genes are direct transcriptional targets (i.e., these TFs bind directly to the CREs of the genes they regulate). Accordingly, a computational algorithm was developed to identify putative CREs around genes within the network. The algorithm analyzes the genomic region 15 Kb upstream and downstream of the transcriptional start site (TSS) of a gene, assigning a score to successive 500 bp stretches of genomic DNA based on the number, affinity, and clustering of phylogenetically conserved Crx, Nrl, and Nr2e3 binding sites. Those 500 bp blocks with a score above a threshold value of 200 are considered likely photoreceptor-specific CREs. This algorithm was used to make CRE predictions around all genes in the photoreceptor network (Table S1).

Fig. 3A depicts the output of this algorithm for the Rhodopsin (Rho) locus. Note that a strong photoreceptor-specific CRE is predicted within the first 500 bp upstream of the TSS. This peak corresponds to a previously well characterized photoreceptor-specific CRE [40,41]. Fig. 3C depicts a retina harvested two weeks after being electroporated in vivo at P0 with a construct containing the proximal 2.2 Kb upstream of the bovine Rho gene fused to DsRed. This promoter, which contains the regulatory peak predicted by the algorithm, drives very strong, photoreceptor-specific expression, as shown previously [41]. A literature search resulted in the identification of 14 genes (out of 58 in the disease gene list) for which corresponding CREs had been previously reported (Fig. 2). The algorithm predicts a potential regulatory peak corresponding to these validated CREs in 79% (11/14) of the cases (Fig. 2). In addition to the Rho regulatory peak, the successful predictions include a strong regulatory peak ∼3 Kb upstream of the TSS of Opx1mla, the location of the well characterized locus control region which drives expression of the X-linked cone opsins in humans [42]. Furthermore, a prior computational and experimental analysis of the photoreceptor TRN validated three novel CREs in vitro (Abcd4, Gucy2e, and Rp1lh) [43], all of which were predicted by the algorithm. These findings suggest that the algorithm has a high sensitivity for identifying active photoreceptor-specific CREs around known photoreceptor-enriched genes.

In order to test how reliably the algorithm can predict novel functional photoreceptor CREs, we assayed the activity of predicted regulatory peaks around 26 retinal disease genes that had not been previously studied (Fig. 2). Short PCR products (range: 0.35 to 1.7 Kb) containing the individual predicted peaks were tested for enhancer activity by electroporation as CRE-reporter fusions into retina (Table S9). 73% (19/26) of the computationally predicted CREs tested drove detectable expression in photoreceptors (Fig. S2; Table S1). Novel CREs residing upstream, downstream, and within introns of the genes they control were identified (Fig. 4). For example, Fig. 4A depicts strong expression driven by a 1st intronic CRE from the Rgs9 gene. When a portion of this same intron not corresponding to a computationally predicted CRE was tested by electroporation, it failed to drive any expression in the retina (Fig. 4B), thus demonstrating the specificity of the prediction. The algorithm predicts a regulatory peak immediately upstream of the TSS of Crx and Nrl, both regions shown previously to harbor functional CREs [24,44] (Fig. 4F, G). Furthermore, the algorithm predicts additional peaks downstream of Crx and within the 3rd intron and 3′ UTR of Nrl. In order to test whether these peaks might also contribute to regulation of these genes, we assayed the peak downstream of Crx, which drove very strong expression in photoreceptors (Fig. 4F). The fact that active CREs around the Crx locus contain clustered Crx sites suggests that they may represent autoregulatory elements, as noted previously [44]. Alternatively, since Otx2 has a nearly identical binding preference to that of Crx, some of these sites may actually be bound by Otx2 in vivo. We also validated a very strong CRE ∼3 Kb upstream of the TSS of Nrl2e3 (Fig. 4H). This CRE is the strongest we have characterized to date. Our analysis has more than doubled the number of known, experimentally validated photoreceptor-specific CREs.

In order to classify the novel CREs according to expression strength, flatmount images of the retinas electroporated with the novel CRE-reporter constructs were photographed at a defined exposure time. The images were compared with each other and with retinas electroporated with two previously characterized CREs, mouse Rho and Nrl [24,40]. The novel CREs were categorized into five expression groups (Fig. S2): very strong (4 CREs), strong (6), medium (4), weak (5) and absent (7). Over half of the CREs which showed activity were classified as either ‘very strong’ or ‘strong’. By comparison, Rho-CRE was classified as ‘very strong’ and Nrl-CRE as ‘strong’. Next we examined the CREs within the five expression classes to determine whether any rules could be derived to explain their differences in expression level. A very simple cis-regulatory motif associated with high level expression in photoreceptors was found. The motif consists of a single Nrl site less than 40 bp from a Crx site, with individual and combined affinity scores above defined cutoffs (Fig. S3). When we assessed all 26 tested photoreceptor CREs for the presence of this motif, we found that it was present in the following percentage of CREs: 100% (4/4) of ‘very strong’ CREs, 17% (1/6) of ‘strong’ CREs, 25% (1/4) ‘medium’ CREs, and 0% of ‘weak’ (0/5) and ‘absent’ (0/7) CREs (Fig. S3). This motif is also present within the first 100 bp of the Rho promoter. These findings suggest that synergistic activation of photoreceptor gene transcription by Crx and Nrl as first reported for Rho-CRE [26], is a common principle of photoreceptor CRE design, at least for CREs which drive high level expression in rods. The absence of this motif from many photoreceptor CREs, some with ‘strong’ expression, demonstrates that clusters of Crx sites (perhaps in combination with sites for other unidentified TFs) can also support strong photoreceptor-specific expression.

If the novel photoreceptor CREs identified here represent the principal regulatory regions controlling expression of the genes in question, then one might expect there to be a correlation between the expression strength of a given CRE and the corresponding gene’s transcript levels in vivo. In order to test this idea, we compared the expression strength classification of 21 photoreceptor CREs (19 novel CREs along with Rho-CRE and Nrl-CRE as controls) with their mRNA levels (using the average array value of the gene in question as a surrogate for mRNA levels). We found that the average wild-type (C57BL/6) array values for genes in each of the four expression strength categories were as follow: ‘very strong’ = 19,607; ‘strong’ = 15,308; ‘medium’ = 11,016; and ‘weak’ = 7,927. This result demonstrates that the average transcript levels of genes within a given CRE expression category correlate with the strength of that category. Despite this trend, the genes within individual expression strength categories show a wide range of array values. For example, Rgs9-CRE was rated as...
‘strong’, but the array value of the corresponding gene was quite low (i.e., <300). There are a couple possible explanations for this type of discrepancy. First, although array values roughly correlate with endogenous transcript levels, they are not a strictly quantitative measure of mRNA levels. Non-representative array values could therefore result in a lack of correlation between CRE strength and endogenous transcript levels. In addition, for a number of genes (e.g., Crx) there appears to be more than one functional CRE. In such cases, the endogenous transcript levels would reflect the concerted activity of all associated CREs and so would not be expected to correlate perfectly with the activity of any one CRE tested in isolation. Despite a number of individual exceptions, the overall correlation between CRE strength and transcript levels suggests that we have successfully identified the principal regulatory regions of many of the genes under consideration.

Figure 3. Computational algorithm to identify photoreceptor cis-regulatory elements. A, Output of computational algorithm for 12 Kb of genomic DNA around the Rho locus. ‘PhastCons’ indicates the degree of phylogenetic conservation of a given stretch of DNA as determined by the PhastCons algorithm. ‘Conserved Crx sites’ indicates the occurrence of a phylogenetically conserved Crx site. The height of the bar representing a site indicates the ‘affinity’ of the site as measured as the percentile rank of that site’s log odds score relative to the consensus site for that transcription factor = 1 (i.e., 100%ile). ‘CRE score’ indicates the likelihood that a given 500 bp stretch (on which the peak is centered) harbors a photoreceptor CRE (details about scoring are given in MATERIALS AND METHODS). Color code: red if score ≥500, dark blue if 500>score ≥200, light blue if score <200. B, Pattern of phylogenetic conservation around Rho CRE as displayed by UCSC genome browser. C, Sections of mouse retina electroporated in vivo at P0 with the indicated constructs and harvested at P14. The Rhodopsin promoter used here derives from cow. The white arrows highlight CAG-eGFP-positive cells in the INL in which DsRed is not expressed. IS = inner segment.

doi:10.1371/journal.pone.0000643.g003
**Figure 4. Novel photoreceptor-specific cis-regulatory elements.**

A. The top panel shows the output of our computational algorithm for the first intron of Rgs9. The red bar indicates the region of genomic DNA assayed for CRE activity. The bottom panels show flatmount and cross-sectional images of a mouse retina electroporated at P0 with the indicated constructs, cultured as an explant, and harvested at P8. The CAG-eGFP is a ‘loading’ control expressed in all retinal cell types. Rgs9-CRE consists of the Rgs9 regulatory peak (indicated by the red bar in top panel) fused to DsRed. The cross-sectional image is a composite of images taken in the green and red channels. Size bar = 500 μm for flatmount images and 100 μm for cross-sections.

B. A control construct containing the indicated region of DNA from the first intron of Rgs9 which does not correspond to a predicted regulatory peak.

C–E, Additional examples of novel photoreceptor-specific CREs. Note that some annotations of the Rgs9bp locus suggest that the indicated regulatory peak falls within the 3’ UTR of this gene.

F–H, CREs around the Crx, Nrl, and Nr2e3 loci. The regulatory peak upstream of the TSS in F indicates a previously characterized Crx CRE [44]. The Crx-CRE depicted in the bottom panels corresponds to the downstream regulatory peak identified in the present study (indicated by the red bar). The Nrl-CRE shown in G was previously characterized [24] and is included here as a control.

doi:10.1371/journal.pone.0000643.g004
**In silico evolution of functional photoreceptor cis-regulatory elements**

Comparative analysis of the CREs identified in this study demonstrated a correlation between a closely linked pair of Crx and Nrl sites and strong expression in photoreceptors. In order to test the generality of this association, we created a genetic algorithm to evolve photoreceptor CREs from random DNA sequences in silico using a selection filter based on clustering and affinity of Crx and Nrl binding sites (see MATERIALS AND METHODS for details). The purpose of this experiment was to produce sequences that retain the essential features of photoreceptor-specific CREs (i.e., closely clustered Crx and Nrl binding sites), while randomizing all intervening sequences. We found that in less than 100 generations it was possible to evolve CREs with features highly reminiscent of naturally occurring photoreceptor-specific CREs, including some with a strong resemblance to the well characterized Rho-CRE (compare Fig. 5A with B-D; Table S10). In order to test whether these in silico evolved CREs could drive photoreceptor-specific transcription in vivo, we selected three ‘organisms’ from three independent evolutionary runs which had a distribution of Crx and Nrl binding sites resembling that of Rho-CRE and which contained the cis-regulatory motif described above. None of these evolved sequences had significant linear sequence homology with each other or with Rho-CRE. We then synthesized these three 400 bp sequences and cloned them upstream of a minimal basal promoter driving DsRed. This basal promoter alone does not drive any expression in photoreceptors (Fig. S2). When electroporated into explanted P0 retinas, these synthetic CREs drove photoreceptor-specific expression after several days in culture. All three synthetic CREs were scored as ‘strong’ with syn1-G70 and syn2-G65 being somewhat stronger than syn3-G55. These results strongly suggest that closely clustered Crx and Nrl sites are a critical determinant of strong photoreceptor-specific expression and that there is significant flexibility in the architecture of photoreceptor-specific CREs.

We wished to establish with certainty that clustering of Crx and Nrl sites alone is sufficient to support high-level photoreceptor-specific expression in these synthetic constructs. However, the basal promoter used in these experiments contains the proximal 36 bp upstream of the ‘TATA’ box from the bovine Rho locus. This stretch of DNA includes the so-called ‘Ret4’ element which consists of a single low-affinity Crx site as well as a binding site for an unidentified ubiquitous factor [45]. To rule out the possibility that this ubiquitous factor is playing a significant role in driving expression in these synthetic constructs, we repeated these experiments using a basal promoter carrying a mutation known to eliminate binding of this ubiquitous factor. The expression level of two constructs (Syn2-G65 and Syn3-G500) were basically unchanged and the levels of two others (Syn1-G70 and Syn3-G55)
were only modestly reduced (data not shown). These findings suggest that clusters of Crx and Nrl sites alone are sufficient to drive high-level expression in photoreceptors in this assay system.

In order to determine whether further evolution could be used to generate CREs with progressively stronger expression, CRE-reporter constructs corresponding to the fittest organism from the 150th, 300th, and 500th generations of run number 3 (the same run from which syn3-G55 was derived) were synthesized. The 500th generation construct drove very strong expression in photoreceptors, comparable to some of the strongest naturally occurring CREs we examined (Fig. S4D). In addition, we found that expression was detectable at progressively earlier timepoints the more highly evolved the element (See Fig. 5E; F; Fig. S4). However, there was not a clear linear progression in the ultimate strength of expression. In fact, the CRE from generation 150 was significantly weaker than that from generation 55 (Fig. S4). The reason for this decrease is unclear but could relate to the acquisition of a binding site for an unidentified repressor.

**DISCUSSION**

We have employed experimental and computational analyses of Crx, Nrl, and Nrl23 mutant retinas to define a comprehensive model of the mouse photoreceptor TRN which contains over 600 genes. This model will serve as a template for understanding retinal diseases at a systems level. Using computational prediction and in vivo validation we were able to identify 19 new photoreceptor-specific CREs, thus doubling the number of known photoreceptor-specific CREs. Given recent progress in using viral vectors for intraretinal gene therapy in animal models of blindness [46,47], these novel CREs could prove useful as gene-specific drivers for rescue constructs in human gene therapy. This study also demonstrates the usefulness of electroporation into living tissue as a means of assaying mammalian CREs in a rapid manner which circumvents costly mouse transgenesis. We believe that this approach can be extended to other cell types and organs and can easily be made quantitative.

Our microarray and in situ hybridization results showed that many photoreceptor genes appear to be co-regulated by Crx and Nrl, a finding which is corroborated by prior analyses of individual photoreceptor-specific CREs [26,27]. At the same time, we also showed that maintenance of normal Nrl transcript levels depends on Crx. This latter result suggests that the similarities between the sets of target genes regulated by Crx and Nrl could, in part, be attributable to the decrease in Nrl expression in Crx. Although we cannot completely rule out this possibility, multiple lines of evidence suggest that, for many co-regulated genes, Crx’s transcriptional control is independent of Nrl. First, several cone-enriched genes (e.g., Opn1sw, Ptk6, and Arr3) are downregulated in Crx and upregulated in Nrl (see Table S1). This pattern is the opposite of what one would expect if Crx’s regulation of these genes was exclusively mediated via maintenance of Nrl transcription. Secondly, we showed that Crx binding sites are abundant within many functional CREs around co-regulated photoreceptor genes. Again, this finding would not be expected if Crx’s control of these genes was mediated solely via downstream activation of Nrl. Despite these points, the extent to which the transcriptional changes seen in Crx are due to a decrease in the levels of Nrl is an important problem for future studies.

Using synthetic CREs, we found that clusters of Crx and Nrl binding sites alone are sufficient to drive strong photoreceptor-specific expression in the mouse retina. This finding was somewhat surprising given the proposed role of other photoreceptor TFs in the regulation of rod photoreceptor gene expression. For example, recent studies have shown that mice carrying mutations in the nuclear receptor TF, Nrl23, show a delay in the onset of Rho expression [20] and modest decreases in the levels of expression of several other rod genes [19]. In addition, Neurod1 is thought to regulate gene expression in rods since mice carrying mutations in this TF show rod degeneration [30]. Furthermore, we demonstrated in this study that there is a small subset of photoreceptor genes whose expression is unchanged in Crx or Nrl (see Fig. S1), suggesting that there must be alternative activators of rod gene expression. In general, however, the effects of Nrl23 and Neurod1 mutations on rod gene expression are modest [18-20] (and data not shown). We therefore propose that Crx and Nrl are the primary activators of gene expression for the majority of rod genes in the mouse, and that other TFs expressed in rods such as Nrl23 and Neurod1 are involved in fine-tuning expression.

In analyzing the novel photoreceptor CREs identified in the present study, we were able to define a simple cis-regulatory motif associated with high-level expression in photoreceptors which consisted of a pair of Crx and Nrl binding sites. This motif is reminiscent of a cis-regulatory motif recently identified in the regulatory regions of Drosophila immune response genes which consists of a heterotypic pair of REL and GATA binding sites with a defined orientation relative to each other [48]. The mammalian eye offers other intriguing examples of cell types in which two different TFs play the predominant role in regulation of gene expression. For instance, Otx2 and MITF are known to be the principal TFs involved in regulation of retinal pigment epithelium (RPE) gene expression and are thought to act at the same hierarchical level in the RPE gene network [49]. In addition, gene expression in the mammalian lens is known to depend crucially on binding sites for two types of TF, Maf and Sox [50]. In this case, multiple members of the Sox family are known to participate in different stages of lens differentiation [51], but they all share a very similar binding site. These findings suggest that such two-site motifs may be a common feature of metazoan cis-regulatory regions. We propose that despite the apparent diversity of TFs expressed in a given cell type (e.g., Fig. 1G) in many cases only two or three TFs may play the major role in control of gene expression. The host of other TFs in that cell type may either be involved in fine-tuning gene expression or in regulating the expression of a selected subset of genes.

Another example of this general principle is the control of cone photoreceptor gene expression in the mouse. Many cone genes are activated by Crx and many are repressed (in rods) by Nrl and Nrl23 [12,13,18-20,25]. These three TFs therefore appear to be the primary regulators of cone gene expression. However, several other cone TFs have been shown to regulate a small subset of cone genes. For example, Thrb2 is known to be required for activation of Opn1sw and dorsal repression of Opn1sw in the mouse [52]. The TF, Rxrγ, appears to have an even more specialized role in cone photoreceptors in that it is involved in the dorsal repression of Opn1sw, but does not appear to regulate Opn1sw [29]. Lastly, a recent paper showed that the orphan nuclear receptor TF, RORBeta, is required for the activation of Opn1sw [53]. Although it is possible that these three TFs have other transcriptional targets in cone photoreceptors, it may also be the case that their transcriptional regulatory role is restricted to only a subset of genes in this cell type.

A recent study identified functional CREs in the human genome by searching for contiguous blocks of extreme phylogenetic conservation and then testing them in transgenic mice [4]. We compared the location of the 19 novel photoreceptor CREs identified here with a set of over 3000 elements conserved between human and Fugu identified in that study. We found that none of the 19 CREs were included in that set (data not shown). This result suggests that approaches which rely solely on identifying extended
blocks of extreme phylogenetic conservation may fail to detect many cell-type specific CREs. The algorithm used in the present study only requires that the identified binding site itself be phylogenetically conserved. This approach was employed because examination of known photoreceptor CREs demonstrated that, in many cases, blocks of conservation were found to encompass only a single binding site (e.g., the rightmost Crx site in Fig. 3B).

Given the reliance of the CRE-finding algorithm used in the present study on clustering of phylogenetically conserved Crx binding sites which contain the canonical ‘TAAT’ sequence bound by many homeodomain TFs, putative regulatory peaks predicted around non-photoreceptor genes must be interpreted with caution. A preliminary analysis of the genomic regions around 20,000 mouse genes using this algorithm detected a marked enrichment of predicted regulatory peaks around TFs and other developmental regulators (data not shown). Nearly half of the 100 top-ranked genes in this analysis were homeodomain TFs, most of which are unlikely to be involved in photoreceptor development and therefore represent false positives. We believe there are two possible explanations for this result. First, the scoring algorithm is not normalized according to the extent of regional phylogenetic conservation of non-coding DNA. Since developmental regulators including TFs are known to exhibit a greater degree of phylogenetic conservation in their surrounding non-coding regions [54], the algorithm would be expected to predict more regulatory peaks in such regions by chance alone. Secondly, the abundance of homeodomain TFs among the top-ranked genes suggests that the algorithm may be identifying bona fide CREs containing clusters of binding sites for homeodomain TFs other than Crx. Given these caveats, the algorithm presented here identifies functional CREs with considerable reliability when applied to the genomic region around photoreceptor-enriched genes.

MATERIALS AND METHODS

Oligonucleotide microarray analysis

Microarray analysis of P21 Crx-/- and Crx-/-;Nil-/- retinas was performed on Affymetrix mouse genome 430 2.0 GeneChip arrays (Affymetrix, Santa Clara, California, United States). The full microarray datasets for the Crx-/- and Crx-/-;Nil-/- analyses are given in Tables S3 and S6. A comparison of these microarray datasets alongside the results of our prior microarray studies of Nr2e3-/- and Nil-/- are presented in Table S4. A total of three microarray hybridizations were performed for each mutant. Since the Crx mutant allele is on a 129 background, we also performed three control microarray hybridizations with retinas derived from 129S6/SvEv mice at P21. For each microarray, RNA was prepared from 4-6 freshly dissected retinas derived from 2-3 animals at P21. Probes were synthesized starting with 10 μg of total RNA for each sample according to manufacturer’s instructions (Affymetrix). Hybridization, washing, and scanning of the microarrays were all performed at the Bauer Center for Genomics Research at Harvard University according to manufacturer’s instructions (Affymetrix).

Initial data analysis was carried out using the GeneChip Operating System (GCOS) software from Affymetrix. Pairwise comparisons were made between individual mutant microarray results and controls. For our ‘low stringency’ analysis, up- and downregulated genes were determined by using Wilcoxon’s signed rank test to compare mutant and control microarrays in a pairwise fashion with a p-value = 0.002. Only those array features which were significantly ‘increased’ or ‘decreased’ in three out of three microarray comparisons were considered to have met our criteria for inclusion in the low stringency datasets.

To create a ‘high stringency’ dataset, we used the same approach used previously for the Nil mutant [25]. First, we separated the features in the low stringency datasets into three expression level categories: (for analysis of downregulated features) (1) features whose average microarray score in the wild-type retina (score129) is ≥5000; (2) 1000≤score129<5000; and (3) 100≤score129<1000. For analysis of upregulated features, the same categories were created using the average microarray score for the mutant. Next, we sorted each of these three sub-lists according to the ratio of their average wild-type to average mutant scores and retained only those features that passed the following thresholds for the three expression categories: category (1) at least a 2-fold change; category (2) at least a three-fold change and category (3) at least a 5-fold change. In this manner, we applied progressively more stringent fold-change requirements for the features showing lower levels of expression. Next, in order to create a non-redundant list of up- and downregulated genes (since our initial feature lists contained multiple Affymetrix features which correspond to a single gene) we used the DAVID functional annotation tool (http://david.abcc.ncifcrf.gov/) to annotate the low and high stringency lists. We then wrote a Perl script to sort and format the DAVID outputs (script available on request). Data from our previously published microarray study of the Nr2e3 mutant [20] were processed in the same way to generate high and low stringency datasets. High stringency datasets for the Nil mutant were reported previously [25], and low stringency datasets for the Nil mutant were obtained as described above.

Since the Crx and Nil mutant alleles are on different backgrounds (129 and B6, respectively), the Crx;Nil double mutant created in the present study was on a mixed 129 X B6 background. Since microarray data from a mixed wild-type background were not available as a control, it was necessary to use the pure 129 and B6 wild-type datasets as controls for the Crx;Nil double mutant. Accordingly, pairwise comparisons were made between individual Crx-/-;Nil-/- microarray results and both 129 and B6 controls using the GeneChip Operating System (GCOS) software from Affymetrix (i.e., a total of six pairwise comparisons). For our ‘low stringency’ analysis, up- and downregulated genes were determined as described above for the single mutants with the exception that we required that a given feature be up- or downregulated in all six pairwise comparisons in order to be included in our low stringency dataset. High stringency datasets for Crx-/-;Nil-/- were then obtained in the same manner as described for the single mutants.

RNA in situ hybridization

ISH on tissue sections was performed essentially as described previously [55]. Most of the ISH images of Nil-/- and B6 were published previously [25] with the exception of the images for Eorb, Otx2 and Neurod1. ISH images of Nr2e3-/- using a subset of the probes employed in the present study were also published previously [20] but on sections from earlier developmental timepoints (P14 and P28). Briefly, retinas from 6-9 week old Nr2e3-/-, Nil-/-, Crx-/-;Nil-/-, C57BL/6 and 129S6/SvEv mice and from 4-5 week old Crx-/- mice were harvested in PBS and immediately put in 4% paraformaldehyde fixative at 4°C overnight. The earlier timepoint was used for Crx-/- because of more rapid onset of degeneration in this mutant. Tissue was equilibrated in 30% sucrose/PBS and then embedded in OCT (Tissue Tek). Sectioned tissue from all four mutants and both wild-type controls were placed on a single glass slide which was then hybridized with RNA riboprobes synthesized from PCR products derived from the templates described in our prior study [25] with the exception of probes for Eorb, Otx2 and Neurod1. The latter probes were derived...
from the following templates: BC044858 (Earr), AB836617 (Otx2), and an in-house clone corresponding to nucleotides 966-1045 of NM_101094 (Neurod).

Photoreceptor cis-regulatory element scoring algorithm

Position frequency matrices (PFMs) were constructed for Crx, Nrl, and Nr2e3 based on previous literature [10,18,56]. Given that all members of the maf subfamily of leucine zipper TFs (which includes Nrl) have very similar DNA binding preferences [57], a well characterized PFM for v-maf [56] was used as the ‘Nrl’ PFM. Nrl and other maf subfamily members can bind to 13-bp and 14-bp variants of their binding site which differ by one nucleotide in the middle of the motif. In preliminary computational analyses we found very few instances of the 14-bp motif around photoreceptor genes (data not shown) and we therefore only used the 13-bp motif in all subsequent analyses. PFMs were converted to position weight matrices (PWMs) using the following formula:

\[
\log_2 \left( \frac{n(b,i) + \frac{N}{2}}{\sum N + \sqrt{N}} \right) \frac{p(b)}{p(\bar{b})} \]

where \(n(b,i)\) is the number of occurrences of base \(b\) at position \(i\), \(N\) is the number of sequences used to create the PFM, and \(p(b)\) is the background genomic frequency of base \(b\) (given a GC content of 0.43 in the mouse). Site scoring was then performed on the NCBI Mouse Build 36 using the TFBS Perl module [58]. PhastCons release mm8 was used to estimate the degree of conservation for each binding site [59]. Average PhastCons scores for each binding site were calculated and then the binding site scores were weighted according to the formula:

\[
w(i) = 0.8 \times r_{LOS} + 0.2 \times \mu_{ave}
\]

Where \(w(i)\) is the weighted binding site score, \(r_{LOS}\) is the raw log odds score for the site and \(\mu_{ave}\) is the average PhastCons score for the site. This weighting ensures that binding sites that lie within phylogenetically conserved stretches of the genome contribute a higher score to the overall score of that 500 bp window. CREs typically consist of closely linked clusters of binding sites for multiple TFs [1,2]. Accordingly, closely clustered phylogenetically conserved binding sites were given more weight than sites that were less closely clustered. Our scoring scheme computed module scores for 500 bp windows according to the following formula:

\[
S = \sum \left(1 + n_{75}(i) + 2 \times m_{75}(i) \right) \times w(i) \times f(w(i))
\]

where \(S\) is the score for the 500 bp window. The summation is done for all binding sites in the window with \(i\) being the iteration index, \(n_{75}\) is the number of homotypic neighbors (i.e., Crx-Crx, Nrl-Nrl, or Nr2e3-Nr2e3 pairs) within \(\pm 75\) bp, \(m_{75}\) is the number of heterotypic neighbors (i.e., Crx-Nrl pairs etc.) within \(\pm 75\) bp, \(w(i)\) is the weighted binding score, and \(f(w(i))\) is defined to be 4 if \(w(i)\) is greater than 90% of the maximum log odds score for that particular PWM. This scoring scheme gives greater weight to high affinity sites (i.e., sites whose log odds scores are in the top 10% of all sites for that TF) and closely clustered (i.e., within 75 bp of each other) heterotypic pairs of sites. Once a given 500 bp window is scored, the window is shifted by 100 bp and the process is repeated. CRE predictions were made for a \(\pm 15\) Kb interval around genes in our dataset. Gene structure information was accessed using the Ensembl Perl modules (version 40) [60]. The BioPerl Graphics package [61] was used to visualize predictions superimposed on gene structures. The script for this algorithm is available from the authors on request.

Literature analysis of previously characterized photoreceptor cis-regulatory elements

Through a literature search we identified a total of 14 retinal disease genes for which mammalian CREs had been previously experimentally analyzed: Abca4, Crx, Gnat1, Gnat2, Gek1, Gacy2e, Nrl, Opn1mu, Opn1sw, Pde6a, Pde6b, Rho, Rpl1h, and Sag [24,27,41-44,62-67].

Construction of cis-regulatory element fluorescent reporters

PCR primers were designed around computationally predicted CREs using Primer3 [68]. The sequences of all PCR primers used (along with restriction sites added for purposes of cloning) are given in Table S9. Two different starting vectors were used for the construction of the CRE-DsRed reporters: Rho-basal and no-basal. The Rho-basal vector was used for all CRE that resided at some distance from the gene’s endogenous basal promoter region. The no-basal vector was used for those CRE that occurred immediately upstream of the TSS. In the latter case, the gene’s own endogenous basal promoter could be cloned directly upstream of the DsRed coding sequence. The Rho-basal vector was created by replacing the 2.2 Kb bovine Rho promoter region in pRho-2.2K-DsRed [69] with a minimal basal promoter which contains nucleotides -36 to +79 around the ‘TATA’ box of bovine Rho. This Rho minimal basal promoter includes the so-called ‘Ret’ element which contains a single low affinity Crx site (log odds score \(\geq 5.1\)) and a binding site for an unidentified ubiquitous factor [45]. The following PCR primers were used to create this minimal basal promoter: 5‘- taccgcTTCGATGatctctaggtacggaattcagcgagaggcttag and 5‘- tccgagCCCGGGGatgcttct. The 5‘ primer includes an Xhol site for cloning (in upper case) and the following additional enzyme sites for subsequent subcloning of CREs into the Rho-basal vector (from 5‘ to 3‘): EcoRV, XhoI, Kpnl, and EcoRI. The downstream primer contains a Smal site (in upper case) for cloning. This Rho basal promoter PCR product was digested with Xhol and Smal and ligated into the pRho-2.2K-DsRed vector that had been previously digested with Smal and Smal, treated with calf intestinal alkaline phosphatase (AP), and separated from the 2.2 Kb bovine Rho promoter fragment by gel electrophoresis. The Rho-basal vector was the product of this ligation. The no-basal vector was created by digesting pRho-2.2K-DsRed with SacI, treating the cut vector with AP, and ligating in a short linker created by kinasing and annealing the following two oligonucleotides: 5‘- tcgagcttcgagtcgacgag and 5‘- tcggagcttcgagtcgag. The CAG-eGFP vector used as a loading control for all electroporations was described previously [referring to in that paper as pCAG-GFP] [69].

A total of six 400 bp synthetic CREs (Syn1-G70, Syn2-G65, Syn3-G55, Syn4-G150, Syn3-G300, and Syn3-G500) were synthesized (Integrated DNA Technologies Inc.) with appropriate restriction enzyme sites added to the ends for purposes of subcloning into Rho-basal.

Retinal electroporation and explant culture

In vivo and in vitro electroporation and explant culture were performed as described previously [69] with the following
The expression strength of the novel CREs were categorized in the following manner. Images of the best electroporated retina for each CRE-reporter construct (defined as that retina with the greatest area and intensity of green fluorescence of the retinas examined for a given construct) were captured in both green and red channels at the same exposure time (1/300 second) for all constructs. These images were then compared side-by-side with images for all other constructs as well as images of retinas electroporated with Rho-CRE and Nr2e3-CRE controls. 'very strong' expression was defined as red fluorescence (i.e., expression levels of DiRed driven by the CRE in question) that showed extensive pixel saturation (i.e., yellow areas on the image) at 1/300 second exposure. 'strong' expression was defined as widespread robust red fluorescence without significant saturation. 'medium' expression was defined as red fluorescence that was visible at 1/300 second exposure but that was considerably weaker or more patchy than that seen in 'strong' constructs. 'weak' expression was defined as any definite expression in photoreceptors that was not easily visible in a 1/300 second exposure. 'absent' expression was defined as absolutely no detectable expression in the retina.

Algorithm for in silico evolution of photoreceptor cis-regulatory elements

A genetic algorithm was developed in Matlab to evolve photoreceptor-specific CREs in silico. The original inspiration for this algorithm came from a paper by Thomas Schneider [70]. The algorithm consists of four distinct steps: initiation, mutation, selection, and reproduction. In the initiation step, random 400 bp DNA sequences were generated in accordance with the GC content of the mouse genome (0.43), and a single such sequence lacking Crx and Nr2e3 sites (i.e., containing no Crx or Nr2e3 sites with log odds scores greater than 4 and 6, respectively) was selected. This sequence was then assigned to 100 ‘organisms’, each consisting of a single 400 bp genome. In the mutation step, random point mutations were assigned to individual organisms according to a Poisson distribution such that, on average, each organism received one point mutation per generation.

In the selection step, the ‘fitness’ of each organism was determined by assigning the affinity of any Crx or Nr2e3 sites in the organism’s genome (scored using the PWMs and TFBS Perl module described above), the spacing between pairs of sites, whether the pairs of sites were homotypic (i.e., Crx-Crx or Nr2e3-Nr2e3) or heterotypic (i.e., Crx-Nr2e3) and the distance of sites from the TSS. Specifically, the fitness score, S, for each 400 bp organism was calculated using the following formula:

\[ S = \sum_{i=1}^{n} h_{TSS}^{y} \left( 1 + \frac{1}{S_{100}} \sum_{j=1}^{n} f(i)_{pair} \cdot g(i)_{dist} \right) \]

where \( h_{TSS}^{y} \) is the raw log odds score of the site being scored; \( y \) is a normalizing factor to even the contribution made by Crx and Nr2e3 sites, where \( y = 3 \) for Crx sites and 2.5 for Nr2e3 sites (the raw log odd score for a high affinity Nr2e3 site is greater than that for a high affinity Crx site since the Nr2e3 binding site is almost twice as long); \( i \) represents all other Crx or Nr2e3 binding sites within the 400 bp genome; \( f(i)_{pair} \) is a weighting term which equals 2 for homotypic neighboring sites (i.e., Crx neighbors if site being scored is Crx; Nr2e3 neighbors if site being scored is Nr2e3) and 8 for heterotypic neighboring sites (i.e., greater weight is attributed to Crx-Nr2e3 pairs to reflect the cis-regulatory motif defined in this study); \( g(i)_{dist} \) is a second weighting term which reflects the importance of site clustering; \( g(i)_{dist} = 0 \) if \( i > 100 \) bp from the site being scored; \( g(i)_{dist} = 1 \) if \( 40 \) bp \( < i \leq 100 \) bp; and \( g(i)_{dist} = 2 \) if \( i \leq 40 \) bp (this weighting term also reflects the important contribution of sites within 40 bp of each other as demonstrated in our cis-regulatory motif); finally, \( h_{TSS} \) is a weighting term which increases from 1 to 4 the closer the location of the site being scored is to the TSS; several of the most strongly expressing endogenous CREs examined (e.g., Rho, Kcnv2, and Nr2e3) have key TF binding sites less than 200 bp from the TSS. For purposes of scoring, Crx and Nr2e3 sites with log odds scores less than 6 and 9, respectively, were ignored.

After assignment of a fitness score, the organisms are ranked from highest to lowest fitness, the bottom 50 organisms are culled, and the top 50 are permitted to differentially reproduce to restore the original population of 100. In this reproduction step, the top ten scoring organisms each contribute three identical offspring to the next generation, the 21st to 40th highest scoring organisms contribute two identical offspring to the next generation, and the 41st to 50th highest scoring organisms each contribute one offspring to the next generation. The new population of 100 organisms reenters the algorithm at the mutation step and the cycle is repeated. The script for this algorithm is available from the authors on request.

The full datasets for the three evolutionary runs used in this study are given in Table S10.

Graphing of cis-regulatory element architecture

A Matlab graphic user interface (GUI) was developed to visualize the spatial arrangement of Crx and Nr2e3 sites on a segment of DNA. For color-coding of Crx and Nr2e3 binding sites the following log odds score thresholds were used: 5 \( \leq \text{Crx LOS} < 7 \) (yellow); 7 \( \leq \text{Crx LOS} < 9 \) (orange); 9 \( \leq \text{Crx LOS} < 12 \) (light blue); 12 \( \leq \text{Nr2e3 LOS} < 15 \) (blue); 15 \( \leq \text{Nr2e3 LOS} \) (dark blue). The term ‘affinity’ used in this paper to describe individual Crx and Nr2e3 sites is not actually the true affinity of the binding sites in question but...
rather a measure of closeness of a given binding site’s sequence to that of the consensus site for that TF as derived from the PFM.

**SUPPORTING INFORMATION**

**Figure S1** In situ hybridizations for genes not in Table S1. ISH images on four mutant and two wild-type backgrounds for nine photoreceptor-enriched genes which were not dysregulated in Crx<sup>-/-</sup>, Nrl<sup>-/-</sup>, or Nr2e3<sup>-/-</sup> and therefore not included in Table S1. Size bar = 100 μm.

**Figure S2** Expression strength categorization of novel photoreceptor cis-regulatory elements. Images of retinas electroporated with the indicated constructs at P0, cultured as explants, and harvested at P8. All 26 predicted CREs tested in the present study are shown in flatmount along with previously characterized mouse Rho-CRE and Nrl-CRE as controls. Also shown are retinas electroporated with the ‘vector only’ controls: ‘Rho-basal’ and ‘No-basal’ (described in METHODS). For comparison purposes, all images were taken at the same exposure time (1/300 second) except Nr2e3-CRE (1/600 second) and all CREs categorized as ‘absent’ (1/150 second). The ‘vector only’ controls were also taken at (1/150 second). Images taken with longer exposures to highlight expression of ‘weak’ CREs are given in Table S1. Some of the retinas electroporated with constructs with ‘absent’ expression show faint, diffuse red fluorescence which represents autofluorescence in the lens. Size bar = 100 μm.

**Figure S3** cis-regulatory motif associated with high-level expression in photoreceptors. A-D. Images showing the distribution of Crx and Nrl sites (as described in Legend at the top of the figure) in four ‘very strong’ CREs. All four CREs show a simple cis-regulatory motif (summarized in E) which is boxed. E. Summary of novel cis-regulatory motif associated with high-level expression in photoreceptors. ‘Crx<sup>LOSI</sup>’ indicates the log odds score for the Crx site, a measure of its closeness to consensus. In addition to minimal log odds scores for the individual Crx and Nrl sites, this grammar rule requires that their combined log odds scores be ≥ 16.0 as indicated and that the two sites be <40 bp apart. F. Summary of percentage of novel CREs from each expression strength category which contain the cis-regulatory motif described in E.

**Figure S4** Expression of synthetic constructs from evolutionary run 3. A-D. Images of retinas electroporated with the indicated constructs at P0, cultured as explants, and harvested at P8. Syn3-G55 is shown in Fig. 5 and is reproduced here for comparison. All four constructs derive from the ‘organisms’ indicated in Fig. 5E. The expression strengths of these four constructs over time are summarized in Fig. 5E. Details about ‘CRE architecture’ notation are given in Fig. 5. Size bar = 500 μm for flatmount images and 100 μm for cross-sections.

**Table S2** Genes dysregulated in Nr2e3<sup>-/-</sup>, Crx<sup>-/-</sup>, Nrl<sup>-/-</sup>, and Crx<sup>-/-</sup>; Nrl<sup>-/-</sup> under high stringency criteria. This list contains eight sub-lists of genes which met high stringency criteria for up- or downregulation in Nr2e3<sup>-/-</sup>, Crx<sup>-/-</sup>, Nrl<sup>-/-</sup>, and Crx<sup>-/-</sup>; Nrl<sup>-/-</sup> retinas (described in METHODS). The data for Nrl<sup>-/-</sup> were published previously [25] and are reproduced here for convenience. The microarray datasets for Nr2e3<sup>-/-</sup> were also published previously [20], but this analysis is new. Each sub-list is subdivided into three parts by color coding (taking Nrl-downregulated genes as an example): light green: genes whose average raw microarray score in the given mutant retina (score<sub>Nrl</sub>) is ≥5000; brown: 1000≤score<sub>Nrl</sub><5000; and pink: 100≤score<sub>Nrl</sub><1000. "Nrl_ave" and "B6_ave" indicate the average scores for Nrl mutant and wild-type (C57BL/6) obtained for the indicated feature from three separate microarray experiments. Within each of the three sub-lists, genes are listed in order of their degree of change in the Nrl mutant retina (relative to wild-type) as indicated by the ratio of their average wt::Nrl ratios ("B6_ave::Nrl_ave"). The columns labeled "Additional Affy_IDs" contain additional Affymetrix image of our computational prediction in the region 15 Kb on either side of the gene’s TSS. Black bars within this image highlight the following regulatory peaks (if any are present): the peak closest to the TSS and the peak with the highest CRE score within this 30 Kb window. For those genes whose CRE predictions were tested experimentally, a red bar in this image indicates the location of the CRE that was tested. For Crx and Elsl4 the tested CRE is outside of this 30 Kb window and is therefore indicated in the ‘Large image’. ‘Links’ contains links to the UCSC genome browser, ENSEMBL, and NCBI database entries for the indicated gene, if available. ‘Max score (threshold of 200)’ contains the score of the highest predicted regulatory peak within the 30 Kb window around the gene’s TSS. If this window does not contain any predicted peaks ≥ 200 (our cutoff threshold) no value is given (indicated by a dash). The next four columns of the table (‘Nr2e3-/-’ etc.) show the wild-type-to-mutant ratios of the averaged microarray scores for the given gene. For those genes to which more than one Affy tag correspond, the data for the first one listed under ‘Affy Mouse 430 2.0’ is given. Dark green = downregulated under high stringency (as described in METHODS); light green = downregulated under low stringency; red = upregulated under high stringency; orange = upregulated under low stringency. A dash indicates that the gene was not significantly altered in the given mutant. ‘NA’ indicates that the gene is not represented on the Affymetrix Mouse 430 2.0 microarray. In situ hybridization’ links to ISH images of the gene, if available. Alterations in expression level ≤ 2-fold on microarray are generally not detectable by ISH. ‘Electroporation’ links to images of retinas electroporated with the CRE indicated by the red bar in ‘Closeup image TSS +/- 15 Kb’, if available. All images were taken at the same exposure time (1/300 second) unless indicated in the image. ‘Averaged Affy scores’ includes the average score for the first Affy tag listed under ‘Affy Mouse 430 2.0’ from three independent microarray experiments in the given mutant or wild-type background. Each mutant is shown in association with the score for its respective wild-type background. The score for the ‘B6:d129’ control for ‘Crx<sup>-/-</sup>; Nrl<sup>-/-</sup>’ is the average of the averaged B6 and 129 scores (as described in more detail in METHODS). ‘Locus’ indicates the chromosomal locus of the gene. ‘Affy Mouse 430 2.0’ contains all Affy tags corresponding to the gene in question which were dysregulated under high stringency criteria in any of the single mutants. An interactive version of this table is available at: http://www.plosone.org/supplinfo/pone.0000643/.
microarray feature identifiers which correspond to the gene in question and which were also met our stringent criteria for upregulation in the Nrl mutant.

Table S3 Genes dysregulated in Nr2e3+/−, Crx−/−, Nrl−/−, and Crx−/−; Nrl−/− under low stringency criteria. This table contains eight sub-lists of genes which met low stringency criteria for up- or downregulation in Nr2e3+/−, Crx−/−, Nrl−/−, and Crx−/−; Nrl−/− retinas (described in METHODS). The data for Nrl−/− and Nr2e3−/− were published previously [20,25], but this analysis is new and they are therefore included here for completeness. Each row corresponds to a single gene which is dysregulated in the given background. ‘Affy_ID’ is an Affymetrix feature identifier representative of the given gene. Taking ‘Nr2e3-downregulated genes as an example, ‘Nr2e3_ave’ and ‘B6_ave’ columns show average Affymetrix scores for the feature in the first column from three independent microarray experiments. ‘B6_ave/Nr2e3_ave’ is the ratio of the B6 to Nr2e3−/− ratios. Within each sub-list, genes are listed in order of their degree of change in the mutant retina (relative to wild-type) as indicated by the ratio of their average wild-type-to-mutant ratios. ‘Additional Affy_ID’s’ show additional Affymetrix tags which correspond to the gene question in those cases where there is more than one.

Table S4 Complete averaged microarray datasets for Nr2e3+/−, Crx−/−, Nrl−/−, and Crx−/−; Nrl−/−. This table contains average scores for all four mutants compared to their respective wild-types for all 45,037 Affymetrix feature identifiers. The data for Nrl−/− and Nr2e3−/− were presented previously [20,25] and are reproduced here for comparison. The descriptors for the columns are as described in Table S3.

Table S5 Complete microarray datasets for wild-type vs. Crx+/− comparison at P21. This table contains the full microarray datasets from six microarray hybridizations (three biological replicates each from wild-type (129S6/SvEv) and Crx−/− retinas at P21). The data from each of the individual microarray experiments are labeled with the following prefixes: ‘Crx-mut_#1’ etc. and ‘129-wt-P21-#1’ etc.

Table S6 Complete microarray datasets for wild-type vs. Crx+/−; Nrl−/− comparison at P21. This table contains the full microarray datasets from nine microarray hybridizations (three biological replicates each from wild-type (C57BL/6), wild-type (129S6/SvEv) and Crx−/−, Nrl−/− retinas at P21). Since Crx−/−; Nrl−/− is on a mixed 129 X B6 background, we present comparisons between both of the individual wild-type mutant backgrounds (as data from the mixed control background were not available). The data from each of the individual microarray experiments are labeled with the following prefixes: ‘Crx-mut_#1’ etc., ‘B6-wt-P21-#1’ etc. and ‘129-wt-P21-#1’ etc.

REFERENCES


Involvement of Sox1, 2 and 3 in the early and subsequent molecular events of lens induction. Development 125: 2521–2532.


Cone Genes in a Mouse Model of Enhanced S-Cone Syndrome. PLoS Genet 1: e643.


2897–2904.

Morphological, Molecular, and Electrophysiological Features of the Photoreceptor Gene Network


Bmp1 is required for the formation of immunoglobulin-secreting plasma cells and pre-plasma memory B cells.


Maf nuclear oncoprotein recognizes specific sequence motifs from mammalian gene expression data. BMC Bioinformatics 7: 292.


Identifying conserved transcription-factor binding sites in multiple alignments using a binding site-specific evolutionary model. Genome Biol 5: R98.


rational folding determines the divergence in binding specificities between Maf and other bZIP proteins. EMBO J 20: 828–840.


