Can artificial intelligence accelerate the diagnosis of inherited retinal diseases? Protocol for a data-only retrospective cohort study (Eye2Gene)

Quang Nguyen, William Woof, Nathaniel Kabiri, Sagnik Sen, Malena Daich Varela, Thales Antonio Cabral De Guimaraes, Mital Shah, Dayyanah Sumodhee, Ismail Moghul, Saoud Al-Khuzaeii, Yichen Liu, Catherine Hollyhead, Bhavna Tailor, Loy Lobo, Carl Veal, Stephen Archer, Jennifer Furman, Gavin Arno, Manuel Gomes, Kaoru Fujinami, Savita Madhusudhan, Omar A Mahroo, Andrew R Webster, Konstantinos Balaskas, Susan M Downes, Michel Michaelides, Nikolas Pontikos, Eye2Gene Patient Advisory Group

ABSTRACT

Introduction Inherited retinal diseases (IRD) are a leading cause of visual impairment and blindness in the working age population. Mutations in over 300 genes have been found to be associated with IRDs and identifying the affected gene in patients by molecular genetic testing is the first step towards effective care and patient management. However, genetic diagnosis is currently slow, expensive and not widely accessible. The aim of the current project is to address the evidence gap in IRD diagnosis with an AI algorithm, Eye2Gene, to accelerate and democratise the IRD diagnosis service.

Methods and analysis The data-only retrospective cohort study involves a target sample size of 10,000 participants, which has been derived based on the number of participants with IRD at three leading UK eye hospitals: Moorfields Eye Hospital (MEH), Oxford University Hospital (OUH) and Liverpool University Hospital (LUH), as well as a Japanese hospital, the Tokyo Medical Centre (TMC). Eye2Gene aims to predict causative genes from retinal images of patients with a diagnosis of IRD. For this purpose, 36 most common causative IRD genes have been selected to develop a training dataset for the software to have enough examples for training and validation for detection of each gene. The Eye2Gene algorithm is composed of multiple deep convolutional neural networks, which will be trained on MEH IRD datasets, and externally validated on OUH, LUH and TMC.

Ethics and dissemination This research was approved by the IRB and the UK Health Research Authority (Research Ethics Committee reference 22/WA/0049) 'Eye2Gene: accelerating the diagnosis of IRDs’ Integrated Research Application System (IRAS) project ID: 242050. All research adhered to the tenets of the Declaration of Helsinki. Findings will be reported in an open-access format.

INTRODUCTION

The retina is the light-sensitive tissue at the back of our eyes, which transforms light into electrical signals to the brain and is responsible for vision. The inherited retinal diseases (IRDs) are a group of diseases resulting from variation in proteins involved in retinal function. They represent the most common cause of blindness in young people in the UK and a leading cause of severe visual impairment and/or blindness in the working age population. IRDs affect more than 2 million people globally and over 1 in 3000 people in the UK.

The age of disease onset varies with different IRDs, and patients usually have a progressive deterioration of their peripheral or central vision over several decades. Hence, it is important to identify an IRD at an early stage, so that patients can undergo proper characterisation of the disease accurately.
Treatments are emerging for some IRDs, but most are gene-specific, requiring identification of the precise causative genetic mutation.\textsuperscript{5,6} 

Mutations in over 300 genes are associated with IRDs.\textsuperscript{7,8} Identifying the causative gene is the first step towards diagnosis, prognosis and treatment. Currently, IRDs are usually detected first by community opticians and referred to ophthalmology departments for retinal imaging and diagnosis with a subsequent referral to specialist centres, such as Moorfields Eye Hospital (MEH), for further imaging and a genetic test.

However, due to limitations in the availability of IRD clinical expertise, detection and diagnostic rates remain poor, with most individuals having to wait for an average of 5.6 years in the UK for a diagnosis.\textsuperscript{9} In addition, the diagnosis can cost the establishment £10,000 to obtain a final diagnosis for the patients and their families, starting from primary referral, to tertiary care, testing, investigation and genetic counselling.\textsuperscript{1,10} Hence, insufficient data on understanding of the disease prevalence and detection has contributed to insufficient funding available for testing of IRDs and associated counselling for patients and families. This delays development of possible treatment pathways and assistance with sight loss. As a result, a significant proportion of patients remain undiagnosed (figure 1).

The proposal here is to prepare images of historical IRD participant retinal scans (datasets) from eye hospitals located in the UK and in Japan:
\begin{itemize}
  \item MEH.
  \item Oxford University Hospital (OUH).
  \item Liverpool University Hospital (LUH).
  \item Tokyo Medical Centre (TMC).
\end{itemize}

Retinal scan datasets will be used to benchmark, train and test Eye2Gene, a deep-learning algorithm designed to detect and diagnose IRDs from a participant’s retinal scan (figure 2).

Aims and objectives

The aim of Eye2Gene is to provide detection and assist in diagnosis of IRDs through non-specialist centres within months instead of years. Eye2Gene does not aim to replace molecular diagnosis obtained through genetic testing, but it serves to narrow down the possibilities of genetic diagnosis based on imaging features, so that an early decision regarding patient care can be taken, and further testing offered after careful discussion with all stakeholders. It will also act as a tool for dissemination of expert IRD knowledge locally across the National Health Service (NHS).

By increasing the diagnostic rate for IRDs at a decreased cost, and by offering equitable access to a genetic diagnosis, the anticipated impacts for participants are:
\begin{itemize}
  \item Improved health outcomes.
  \item Earlier clinical diagnosis.
  \item Personalised treatment plans (emerging treatments or clinical trials).
  \item Better understanding of the condition, its prognosis and its heritability for family planning.
\end{itemize}

For the NHS:
\begin{itemize}
  \item Improved operational efficiency both for the prescription and interpretation of genetic tests.
  \item Increased genetic diagnostic rate at eye hospital.
  \item Reduced economic burden by not needing to test large gene panels or whole genome in every case.
\end{itemize}

The broad aim is to address the evidence gap in IRD diagnosis with an artificial intelligence (AI) algorithm, Eye2Gene, to accelerate and increase availability of a specialist IRD diagnostic service at point of care.

Our primary objectives are training and further validation of Eye2Gene on independent datasets from three external sites: OUH, LUH and TMC, which include:
\begin{itemize}
  \item To refine and improve our model, particularly with respect to rarer genes.
  \item To provide explainability by identifying segmented IRD-specific features in classified images.
  \item To investigate and develop saliency maps for our networks.
  \item To validate Eye2Gene on external datasets to ensure it performs consistently well in different contexts (ie, that the model is generalisable).
\end{itemize}

Our secondary objectives are:
\begin{itemize}
  \item To provide explainability by accurately identifying specific abnormalities (IRD-specific features) in retinal scans.
  \item To lay the groundwork for development of Eye2Gene into a medical device.
\end{itemize}
METHODS AND ANALYSIS

Work plan and timelines for delivery

Eye2Gene project will be divided into eight work packages (WP) (illustrated in figure 3).

WP1: development of classification algorithm

This will involve developing a Convolutional Neural Network (CNN) model that can generalise to the $N$ most common IRD genes at Moorfields and provide a top-five accuracy of at least 88%. In particular, we will focus on achieving high per-gene accuracy for the rarer genes (which the current iteration of the model currently underpredicts). In addition, part of this milestone will be to establish the value of $N$ (the number of genes covered by the model), which we will pick based on all the data available across the four sites. We will assume $N$ to be at least 10 for now, as this covers 70% of IRD cases and will be represented in the datasets of the four centres.

WP2: development of segmentation algorithm

This will involve the manually curated and segmented dataset provided by the Moorfields Reading Centre IRD Segmentation Team. The team will consist of graders and software developers under the lead of Dr Balaskas at the Moorfields Reading Centre. These segmented IRD datasets will be useful for the training of multiple AI algorithms including Eye2Gene. These will include a total of 14 retinochoroidal features detectable by spectral domain optical coherence tomography (SD-OCT), infrared reflectance (IR) or short-wavelength fundus autofluorescence (FAF) or both, as well as their location, shape and distribution. A segmentation algorithm based on U-Net\(^1\) will be developed using this dataset.

WP3: development of explainable AI algorithm

By combining the output of the classification algorithm (WP1) with the segmentation/classification algorithms (WP2), we will build an explainable AI algorithm that combines accuracy (WP1) and explainability (WP2). The final output of these models will be combined in a multinomial logistic regression with additional optional inputs such as age, sex, ethnicity and mode of inheritance, to enhance predictive power. We will also be continuing to investigate and improve saliency maps for our models, and other explainability measures such as model confidence scores.

WP4: phenotype-driven genetic variant prioritisation

Deriving gene score based from the Eye2Gene classification gene probability from WP3. Also segmented IRD features may be translated to Human Phenotype Ontology (HPO) terms in order to support HPO-base phenotype prioritisation using approaches such as Exomiser.\(^2\) We will assess the utility of Eye2Gene for phenotype-driven variant prioritisation to help solve cases with multiple candidate variants. This will fulfil the ACMG annotation guidelines PP4 criteria, namely that the patient’s phenotype or family history is highly specific for a disease with a single genetic aetiology.\(^3\)

Figure 2  Eye2Gene supports the three main retinal imaging modalities: (A) Infrared (IR), (B) Autofluorescence (FAF) (C) Spectral domain optical coherence tomography (SD-OCT). FAF, fundus autofluorescence.
WP5: health economic evaluation
Health economic evaluation comparing the current treatment process to that of Eye2Gene will be conducted. The evaluation will consider two treatment pathways (standard care and the use of Eye2Gene), and will model resource use and cost, including the cost of validation, the cost of genetic tests, the time to find the genetics diagnosis (staff time) and the estimated cost of misdiagnosis, as well as the outcomes of standard and early diagnosis.

WP6: Eye2Gene medical software
Once we have completed the prototype as part of WP3, a software consultancy company (Phenopolis) will, under the oversight of regulatory consultants and the UCL Translational Research Office, develop Eye2Gene as medical device software following a QMS approach. In the first instance, the software will be developed to be hosted on a server that will likely be cloud based.

WP7: patient and public involvement
Patient advisory group (PAG) will feed into the decision making and the dissemination of results. The PAG will meet three times a year (January, May and September), each meeting will be 90 min and feed directly into the input of Eye2Gene. During this process, any risks raised by participants will be added to the risk register for the QMS.

WP8: human factors
User experience, usability and accessibility research will underpin the development of Eye2Gene. Following completion of WP3, we will have a working version of Eye2Gene to explore human factors around user expectations and experience.

Study design and population
This is an investigation aiming to develop an AI software as a medical device. It is a data-only retrospective cohort study.
study that will use images (retinal scans), associated scan-specific (such as laterality, scan date and modality) and participant-specific (such as molecular diagnosis, mode of inheritance, age and ethnicity) labels.

The study population includes data from participants that have received an IRD diagnosis, which has been molecularly confirmed via means of genetic test and have had retinal scans acquired using the Spectralis from Heidelberg Engineering (Dossenheim, Germany) with one of the following imaging modalities: IR, SD-OCT and FAF.

The study population at MEH has been derived by querying the OpenEyes Electronic Health Record (EHR) for IRD participants with a known genetic diagnosis and joining it up to the imaging databases of retinal scans (Heidelberg Medical Image Database) on hospital numbers. This enabled inclusion of all participants at MEH with an IRD who have both a genetic diagnosis and retinal scans available.

The study populations at OUH, LUH and TMC have been estimated based on information provided by the respective Principal Investigators, Prof Downes, Dr Madhusudhan and Prof Fujinami. This information has also been obtained by querying their local EHR databases and joining the dataset from the imaging database by hospital number.

Derivation of sample size

The target sample size of 10000 participants has been derived based on the number of participants with IRD at the three UK eye hospitals participating in this study (MEH, OUH and LUH), as well as a Japanese hospital, the TMC. Given the rare nature of IRDs and that the study works on retrospectively collected anonymised data, we are targeting the largest datasets available in the UK.

The 36 most common genes have been selected as these should have enough training examples to ensure at least 10 example images for each fold, when split into 5 folds.
such as genetic diagnosis and phenotype, may need to
information including demographics and clinical information
various spreadsheets containing participant-
but instead located in study spreadsheets. Therefore,
tation is not always in the EHR in a research-
controlled by the site PI. First, patient genetic informa-
tion as described in the Inclusion criteria section below and
following export, the images will be quality controlled
at UCL will not have access to the original hospital IDs.
The study team working on developing the AI algorithm
original hospital number at each of the respective sites.
keeping the link between each study ID and
available. Data will be pseudonymised by the respec-
tive clinical teams, assigning a unique study ID to each
participant, and keeping the link between each study ID and
original hospital number at each of the respective sites.
The study team working on developing the AI algorithm
at UCL will not have access to the original hospital IDs.
Failing export, the images will be quality controlled
as described in the Inclusion criteria section below and
uploaded, for each patient, to the Moorfields Reading
Centre data-sharing platform (grading.readingcentre.
org). These scans are then converted to an open format
so they can be processed by the AI or annotated as part of
WP2.

**Data acquisition**

Participants will be identified by the care team of the
respective site PIs by searching their medical records
for patients who have received a molecularly confirmed
 genetic diagnosis for IRD. Data from MEH will be
obtained by searching the EHR (OpenEyes) for partici-
pants with genetic reports entered in the EHR. The
hospital numbers for these participants will be extracted
and cross-referenced with the hospital numbers extracted
from the imaging database, as shown in figure 4. A similar
approach will be undertaken at LUH, OUH and TMC to
link the imaging data to the genetic reports and other
associated metadata (age, mode of inheritance and
ethnicity) using the respective medical records in those
sites and collating information from various spreadsheet,
as needed.

Participants’ data extracted from medical records and
imaging databases at each site will be used to produce a
list of images labelled with genes and metadata, where
available. Data will be pseudonymised by the respective
clinical teams, assigning a unique study ID to each
patient, and keeping the link between each study ID and
original hospital number at each of the respective sites.
The study team working on developing the AI algorithm
at UCL will not have access to the original hospital IDs.
Failing export, the images will be quality controlled
as described in the Inclusion criteria section below and
uploaded, for each patient, to the Moorfields Reading
Centre data-transfer portal secure database (grading.
readingcentre.org).

Note that data collection for this study at each site is
often an involved process since the data requires prepa-
ration, which needs to be overseen and carefully quality
controlled by the site PI. First, patient genetic informa-
tion is not always in the EHR in a research-ready format
but instead located in study spreadsheets. Therefore,
various spreadsheets containing participant-level infor-
information including demographics and clinical information
such as genetic diagnosis and phenotype, may need to
be collated. Once the participant information has been
prepared, their corresponding scans need to be extracted
from the Heyex medical imaging database. Since most
sites lack a Vendor Neutral Archive (unfortunately these
are still rare in ophthalmology), this process requires
cross-referencing of scans to participant, extracting them
from the Heyex database as E2E files one patient at a
time, and uploading them to the Moorfields Reading
Centre data-sharing platform (grading.readingcentre.
org). These scans are then converted to an open format
so they can be processed by the AI or annotated as part of
WP2.

**Consent/consent exemptions**

The project is limited to the use of previously collected,
on-identifiable information. As only anonymised clinical
data will be made available to the research team and no
study procedures will be carried out as part of this study,
informed consent will not be separately sought from
participants. However, consent will be obtained from the
participants of the human factors research conducted as
part of WP8 to gather user feedback on Eye2Gene.

**Inclusion criteria**

There will be no age restrictions for participants contrib-
uting data to train Eye2Gene, however, it is anticipated
that most will be over the age of 18. The inclusion criteria
require participants to have both a confirmed IRD genetic
diagnosis available that conforms to criteria (A) below
and retinal imaging scan data available that conforms to
criteria (B) below.

**Criteria for IRD genetic diagnosis**

> An IRD genetic diagnosis consists of the identification of
the IRD gene thought to be associated with the IRD
condition of the participant.

> An IRD genetic diagnosis will often include the
specific genetic variations which are thought to cause
the disease.

> The IRD genetic diagnosis may have been conducted
via a clinical NHS genetic testing service or through a
research study.

> Both sources will be included in this study.

**Criteria for retinal imaging scan data**

Retina imaging scans will have been acquired with
a medical imaging device (such as the Spectralis,

<table>
<thead>
<tr>
<th>Quality of the FAf scans</th>
<th>Quality of the IR scans</th>
<th>Quality of the OCT scans</th>
</tr>
</thead>
<tbody>
<tr>
<td>BRISQUE score &lt;120</td>
<td>BRISQUE score &lt;80</td>
<td>BRISQUE score &lt;150</td>
</tr>
<tr>
<td>Median intensity &gt;0.05</td>
<td>Median intensity &gt;0.1</td>
<td>Max intensity &lt;1</td>
</tr>
<tr>
<td>‘Noise level metric’ (ie, sum of square differences compared with blurred image via 5×5 box filter) &lt;2200</td>
<td>OR mean intensity &lt;0.2</td>
<td></td>
</tr>
</tbody>
</table>

BRISQUE, Blind Referenceless Image Spatial Quality Evaluator; FAF, fundus autofluorescence; IR, infrared reflectance; MEH, Moorfields Eye Hospital; OCT, optical coherence tomography.
Heidelberg) fixated on the macula and may belong to one of the following three categories:

- FAF.
- IR.
- SD-OCT.

Image quality will be an important factor to consider. In order to assess image quality objectively, image quality scores such as the Blind Referenceless Image Spatial Quality Evaluator\(^\text{15}\) image quality score will be applied. The criteria currently applied for image quality at MEH have been summarised in Table 1. These scan quality thresholds will be reviewed and potentially adjusted depending on the data quality available across sites.

**Exclusion criteria**

Participants that do not have a confirmed IRD genetic diagnosis or no retinal imaging data available. No other exclusion criteria apply.

**Time period of data collection and follow-up**

The data collection will happen in the first 2 years of the study (January 2022–January 2024) to obtain retrospective observational data from all four sites. There will be no follow-up as all data are collected retrospectively for participants that have already received a genetic diagnosis. Following lead-in times, including ethics approvals, contractual procedures and data sharing agreements, data collection from UK sites (MEH, OUH and LUH) started in June 2022 and is likely to finish towards December 2023. Due to the additional challenges surrounding international data sharing and transfer arrangements, data collection at the Japanese site (TMC) was delayed to December 2022 and consequently, is likely to complete in January 2024. For the reasons explained above in ‘Data Acquisition’, the data collection, although retrospective, is a lengthy process which should finish by the end of 2023. In addition, as part of WP2, there is also an additional manual process undertaken of manually grading scans which will likely continue in the background for the entire duration of the project.

**Description of collected data**

Along with the gene diagnosis and the retinal scans, the following information will be collected where available:

- Site: MEH, OUH, LUH or TMC
- Scan metadata:  
  - Laterality.
  - Scan modality.
  - Date scan was acquired.
- Participant demographic data:  
  - Age when scan was acquired.
  - Biological sex.
  - Ethnicity.
- Clinical information pertinent to disease:  
  - Mode of inheritance.
  - Age of onset.

All data will be consistently coded across sites and pseudonymised. A unique study ID will be assigned to each participant and the link between the study IDs and original hospital number identifiers will be kept at each of the respective sites and not shared with the research team.

**Deep learning protocols**

A CNN\(^\text{26}\) will be used to classify the images. It will be trained on retinal images from patients with IRD labelled with the causative gene. The aim will be to input a previously unseen retinal image and output a prediction of the causative gene (WP1) (see online supplemental figure 1).

Next, a subset of scans will be manually annotated, as part of WP2. This data will be used to train a U-Net,\(^\text{11}\) a commonly used neural network architecture for image segmentation tasks (online supplemental figure 2). These will include a total of 14 retinochoroidal features detectable by SD-OCT or FAF or both, as well as their location, shape, and distribution.

Specifically, on SD-OCT, we will segment eight features:

- Drusen.
- Subretinal fluid.
- Intraretinal fluid (cysts).
- Subretinal hyper-reflective material.
- Ellipsoid zone loss.
- Retinal pigment epithelium loss.
- Choroidal hypertransmission.
- Foveal hypoplasia.

On FAF/IR we will segment six features:

- Hypo/IR we will segment six features:
- Drusen
- Flecks
- Peripapillary sparing
- Vessel attenuation.

- Foveal hyp Autofluorescence loss.

On the other hand, development of Deep Neural Network will also involve the manually curated and segmented dataset provided by the Moorfields Reading Centre IRD Segmentation Team. The team will consist of four graders and two software developers under the lead of an IRD expert at Moorfields and the director of the Moorfields Reading Centre.

**Statistical methods and performance evaluation**

We aim to develop a model that can generalise to the N most common IRD genes at Moorfields and provide a top-five accuracy of at least 95%. In particular, we will focus on per-gene accuracy for the rarer genes (which the current iteration of the model currently underpredicts). Gene-specific or phenotype-specific segmentation features will be delineated in the Moorfields dataset and will be internally validated by the clinical team using the Dice similarity coefficient score\(^\text{17}\)

\[
DSC = \frac{2 \times |A \cap B|}{|A| + |B|}
\]

where A and B are the regions defined by the two annotated features, to assess overlap with manual segmentation. Images with Dice score over 0.8 will be selected for training and validation.
The final output of classification and segmentation models will be combined in a multinomial logistic regression with additional optional inputs such as age, sex, ethnicity and mode of inheritance, to enhance predictive power (online supplemental figure 3). We will also be continuing to investigate saliency maps for our models, and other explainability measures such as model confidence scores.

The algorithm will be externally validated on the multisite data. This might require further calibration of parameter weights for age, sex, ethnicity and mode of inheritance, per site. We will use top-one and top-five classification accuracy and mean per-gene defined as area under the receiver operator curve score as the metrics for evaluation. We will also review the interpretation of the output (segmentation and saliency maps) qualitatively as part of Humans Factors (WP8). The data are stored at https://grading.readingcentre.org and can be made available for viewing at the discretion of the chief investigator NP. The code can be found at https://github.com/Eye2Gene.

**Measures to avoid bias**

Site, age, gender, ethnicity and mode of inheritance are all potential sources of bias. These will be fitted as extra covariables into the classification algorithm using a multinomial regression or another equivalent statistical method to avoid confounding. Since only retrospective data will be used in a first instance, there will be no ascertainment bias as the data will represent routine IRD department activity at the respective hospitals.

A known source of bias in the data which cannot be corrected is the inherent imbalance in the data: more common diagnosis versus rarer diagnosis. The implications of this is that Eye2Gene will tend to overpredict the common classes and underpredict the rare classes.

**Patient and public involvement**

Participants will be engaged for service mapping; knowledge gathering; acceptability testing and questionnaire design. A number of activities planned for the PAG and charity partners include:

► Interviews and focus groups.
► Survey and report designs.
► Data monitoring advice.
► Dissemination of research.

We will attend and present at participant days organised by charity partners, including Stargardt’s Connected and RetinaUK. During these conferences, Eye2Gene will be presented to participants who will be invited to share their views on the project. In the third year, we will liaise with the Moorfields patient and public involvement (PPI) team to organise an Eye2Gene participant day. Participants will be invited to discuss their diagnostic experience and view a presentation of Eye2Gene.

As the project advances, participants will be given the opportunity to participate in one-to-one interviews and focus groups, aiming at an in-depth investigation of participant experience and feedback on Eye2Gene (WP8). During these interviews, participant experience surrounding the genetic diagnosis process will be further explored:

► Aspects of health psychological support and health anxiety.
► Participant expectations about AI.
► Education and genetic counselling.
► Areas for improvement.

Exploration of participant preferences (WP5) in year 2 (2023) will be formally quantified using a discrete choice experiment to investigate participants’ preference in their diagnostic journey will help inform future research and implementation.

The chief investigator, NP, and coinvestigators, supported by a health psychologist collaborator, DS, with experience in qualitative research and genetic counselors will gather patient feedback at various events. These include focus groups held once a year (September), a participant day in the third year (June 2024) and any others organised by the Moorfields PPI teams.

During these events, participant experiences surrounding the genetic diagnosis process will be further explored. In particular, aspects of health psychological support, doctor–participant communication when discussing diagnosis and health anxiety, and participant education and genetic counselling. We will seek to understand how the current genetic diagnosis service experience can be improved with Eye2Gene. Preliminary results of PPI engagement can be found in online supplemental appendix A.

**ETHICS AND DISSEMINATION**

This research was approved by the IRB and the UK Health Research Authority (Research Ethics Committee (REC) reference 22/WA/0049) ‘Eye2Gene: accelerating the diagnosis of inherited retinal diseases’ Integrated Research Application System (IRAS) (project ID: 242050). All research adhered to the tenets of the Declaration of Helsinki. Findings will be reported in an open-access format.

Our PPI group will be involved in dissemination of the results. In addition, our IRD participants that are signed up to our mailing lists will be informed of our progress.

**Author affiliations**

1. UCL Institute of Health Informatics, University College London, London, UK
2. UCL Institute of Ophthalmology, University College London, London, UK
3. Moorfields Eye Hospital NHS Foundation Trust, London, UK
4. Oxford Eye Hospital, Oxford, UK
5. ICL Cancer Institute, University College London, London, UK
7. UCL Translational Research Office, University College London, London, UK
8. UCL Department for Applied Health Research, University College London, London, UK
9. National Institute of Sensory Organs, National Hospital Organization Tokyo Medical Center, Kankaku Cente, Meguro-ku, Tokyo, Japan
10. Royal Liverpool and Broadgreen University Hospitals NHS Trust, Liverpool, UK
Twitter Quang Nguyen @quangngn, William Woof @william_woof, Malena Daich Varella @MalenaDaich, Thales Antonio Cabral De Guimarães @taquiciguimaraes, Dayyanah Sumodhee @DayyanahSumodhee, Ismail Moghul @ismail_moghul, Gavin Arno @Gavin_Arno, Manuel Gomes @ManuelGomes, Kauru Fujinami @KauruFujinami, Comativa Madhusudhan @ComativaMadhu, Konstantinos Balaskas @konbalaskas, Michel Michaeliades @MichelMichaeliade and Nikolas Pontikos @npontikos

Acknowledgements As NIHR funded this study, publications will be published according to their guidelines. Moorfields Eye Charity has seed-funded part of this work, hence will be acknowledged in publications. Furthermore, the Archer family has also made a donation to the Eye2Gene project previously, hence will also be acknowledged.


Contributors The chief investigator NP conceptualised the Eye2Gene study, designed and developed the protocol. NP, GN, YL and WW contributing to deep learning protocols. SS, TACDG, MDV, SA-K, KF, OAM, MM, ARW, KB, SD and SM contributed to reviewing the clinical background. QA contributed to writing the phenotype-driven interpretation work package. MG contributed to writing the health economics work package. IM and MS contributed to writing the software as a medical device work package. DS, NK and WW and the Eye2Gene Patient Advisory Group (CH, BT, LL, CV and SA) contributed to writing the Patient and Public Involvement work package. DS contributed to writing the human factors work package. JF and SA contributed to writing the project management work package. NP contributed to all work packages. All authors contributed to critically reviewing and formatting the protocol. All authors approved the final version of the manuscript for submission for publication.

Funding The work was funded by a National Institute for Health Research (NIHR) AI Award (AI_AWARD02488), by Moorfields Eye Charity Career Development Award (R190031A) and by the Biomedical Research Centre (BRC) at Moorfields Eye Hospital NHS Foundation Trust and UCL Institute of Ophthalmology.

Competing interests Neither the chief investigator (NP) nor any of the coinvestigators have any direct personal involvement (financial, shareholder or personal) with the NIH that may give rise to a conflict of interest. As has been disclosed to the study sponsor (UCL) and the funder (NIHR). NP and IM are shareholders at Phenopolis, which is one of the two subcontractors mentioned above (NP6). Phenopolis will provide professional software development services for Eye2Gene.

Patient and public involvement Patients and/or the public were involved in the design, or conduct, or reporting, or dissemination plans of this research. Refer to the Methods section for further details.

Patient consent for publication Not applicable.

Provenance and peer review Not commissioned; peer reviewed for ethical and funding approval prior to submission.

Supplemental material This content has been supplied by the author(s). It has not been vetted by BMJ Publishing Group Limited (BMJ) and may not have been peer-reviewed. Any opinions or recommendations discussed are solely those of the author(s) and are not endorsed by BMJ. BMJ disclaims all liability and/or omissions arising from translation and adaptation or otherwise.

Open access This is an open access article distributed in accordance with the Creative Commons Attribution 4.0 Unported (CC BY 4.0) license, which permits others to copy, redistribute, remix and build upon this work for any purpose, provided the original work is properly cited, a link to the licence is given, and indication of whether changes were made. See: https://creativecommons.org/licenses/by/4.0/.

Author note For the latest information about the Eye2Gene project please see the project website (www.eye2gene.com)

ORCID iDs Quang Nguyen http://orcid.org/0000-0003-2665-3125
William Woof http://orcid.org/0000-0001-7555-326X

Nathaniel Kabiri http://orcid.org/0000-0003-3684-0020
Malena Daich Varella http://orcid.org/0000-0003-4960-4510
Thales Antonio Cabral De Guimarães http://orcid.org/0000-0002-7936-6851
Mital Shah http://orcid.org/0000-0001-7496-6121
Dayyanah Sumodhee http://orcid.org/0000-0001-7274-6088
Ismail Moghul http://orcid.org/0000-0003-3653-2327
Saoud Al-Khuzaei http://orcid.org/0000-0002-2779-3202
Yichen Liu http://orcid.org/0000-0002-9563-9151
Gavin Arno http://orcid.org/0000-0001-7086-7188
Manuel Gomes http://orcid.org/0000-0002-1420-1232
Kauru Fujinami http://orcid.org/0000-0003-4248-0033
Savitah Madhusudhan http://orcid.org/0000-0002-9203-9029
Omar AMahroo http://orcid.org/0000-0003-1254-0832
Andrew R Webster http://orcid.org/0000-0001-6915-9560
Konstantinos Balaskas http://orcid.org/0000-0002-7690-6277
Susan M Downes http://orcid.org/0000-0001-7373-2665
Michel Michaeliades http://orcid.org/0000-0002-1552-7046
Nikolas Pontikos http://orcid.org/0000-0003-1782-4711

REFERENCES