

National Society of
Genetic
Counselors



36th Annual Conference

September 13-16, 2017

Greater Columbus
Convention Center
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COI Disclosure

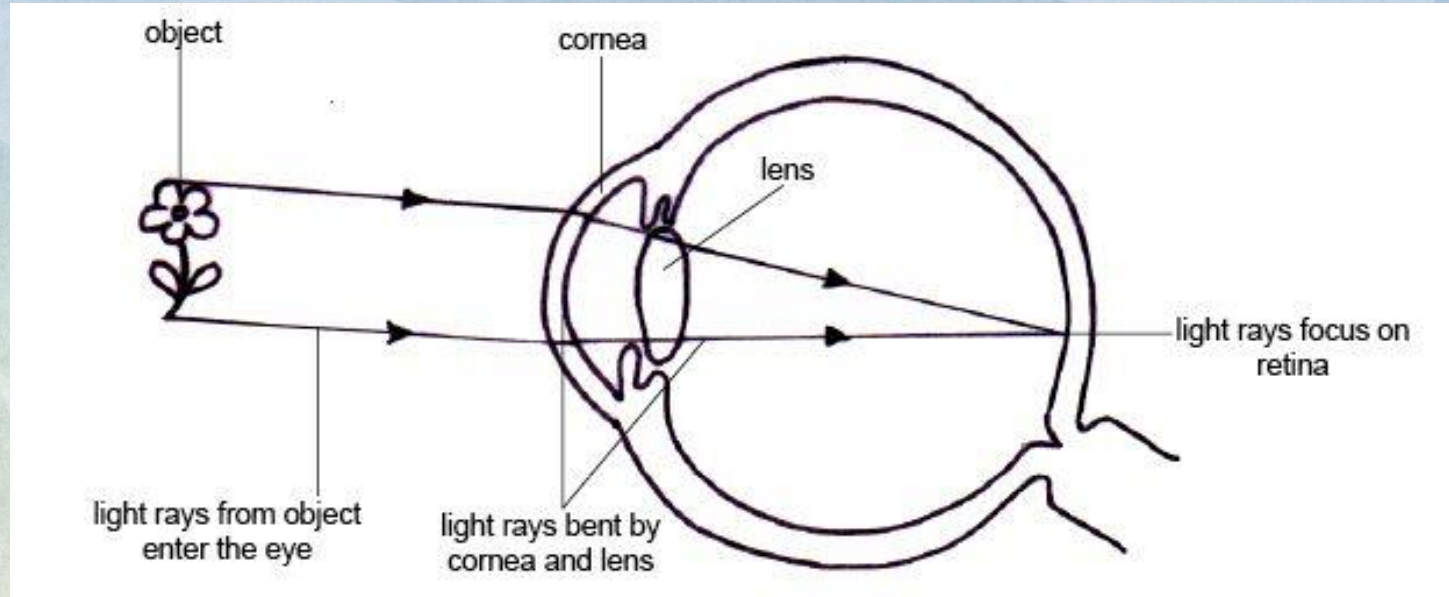
No COI to disclose



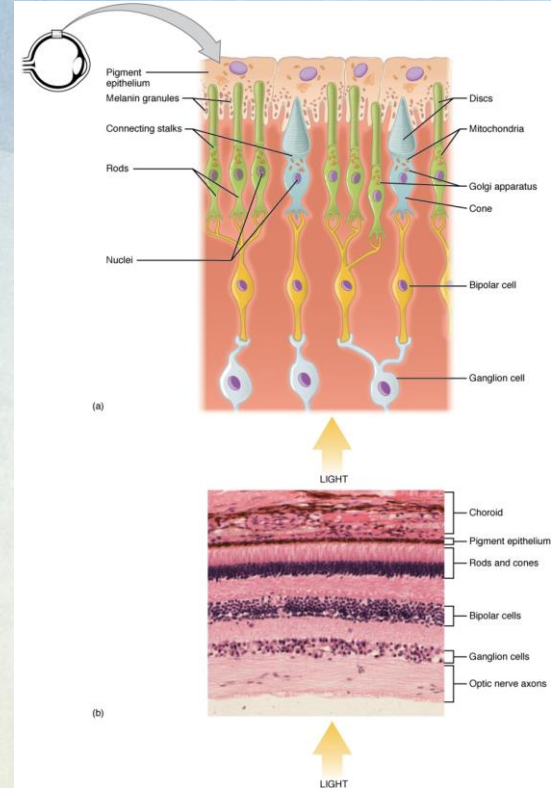
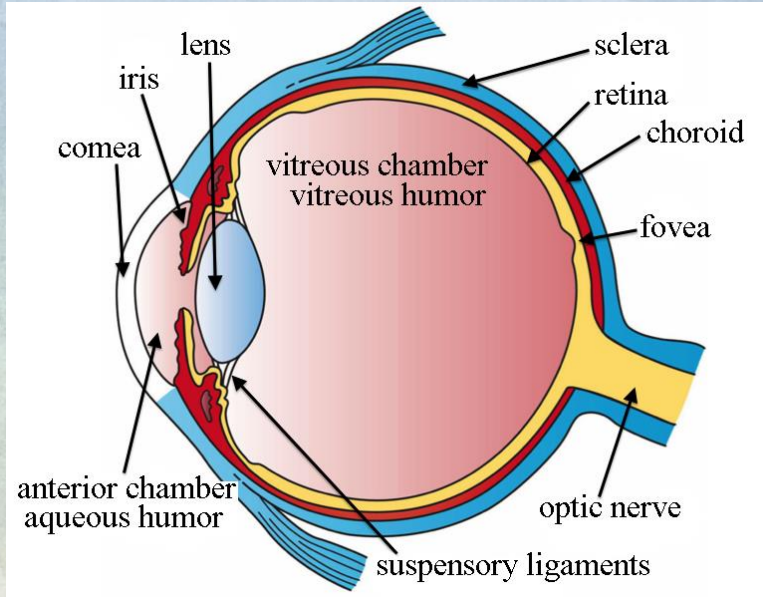
Inheritance Pattern Prediction: An Ophthalmic Model for Digital Pedigree Feature Extraction and Machine Learning

Dana Schlegel, MS, MPH, CGC; Edmond Cunningham; Xinghai Zhang; Yaman Abdulhak; Andrew DeOrio, PhD; K. Thiran Jayasundera, MD

The eye: a brief overview



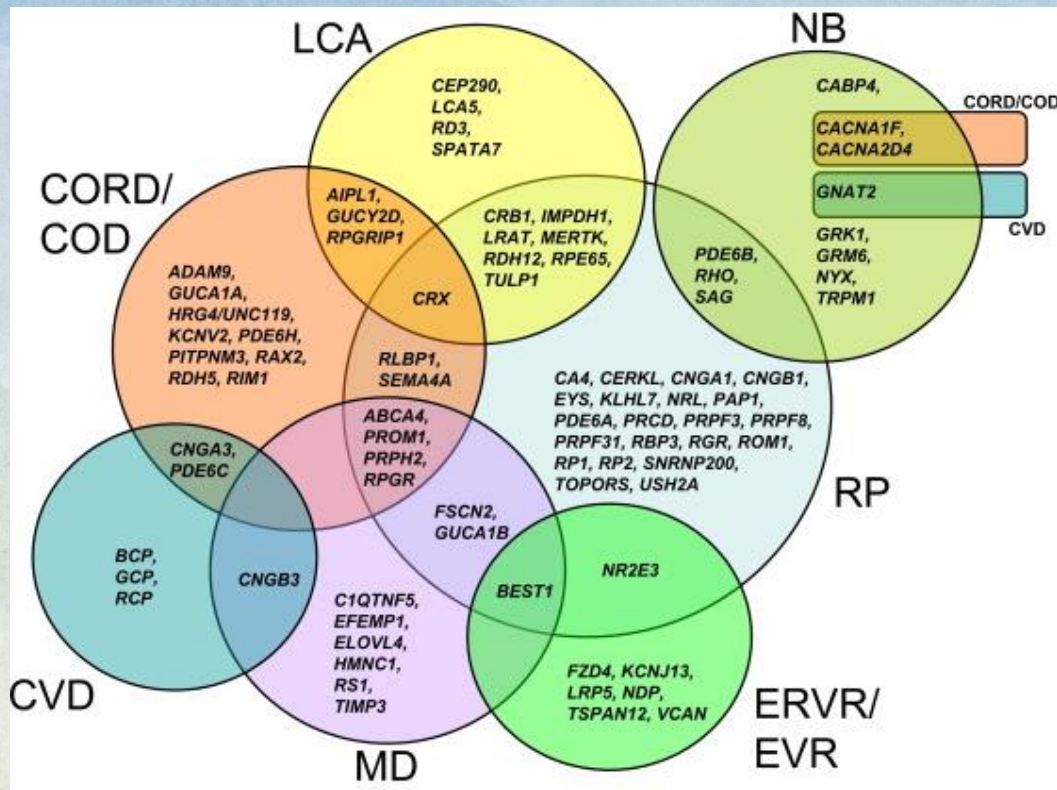
Slightly more detail



Retinal dystrophies

- Inherited retinal degenerative diseases
 - Due to reduced or deteriorating function of cells of retina (ex. photoreceptors, retinal pigment epithelium)
 - Usually progressive, sometimes stationary
- Wide range of conditions
 - Retinitis Pigmentosa, Stargardt, Cone-rod dystrophy, Cone dystrophy, Choroideremia, Leber Congenital Amaurosis, Usher, Bardet-Biedl syndrome...
- Genetically complicated/diverse
 - Clinical heterogeneity, genetic heterogeneity, variable expressivity, incomplete penetrance, some genes with multiple patterns of inheritance

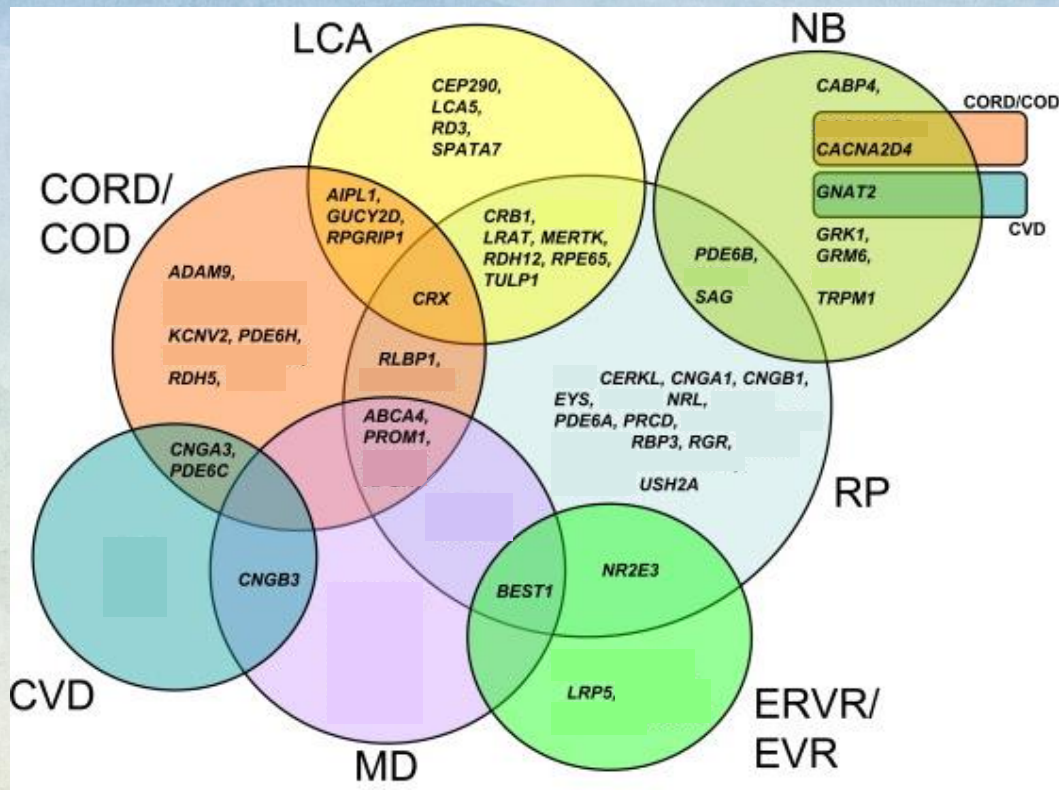
Retinal Dystrophies



Berger W et al, 2010.

Retinal Dystrophies

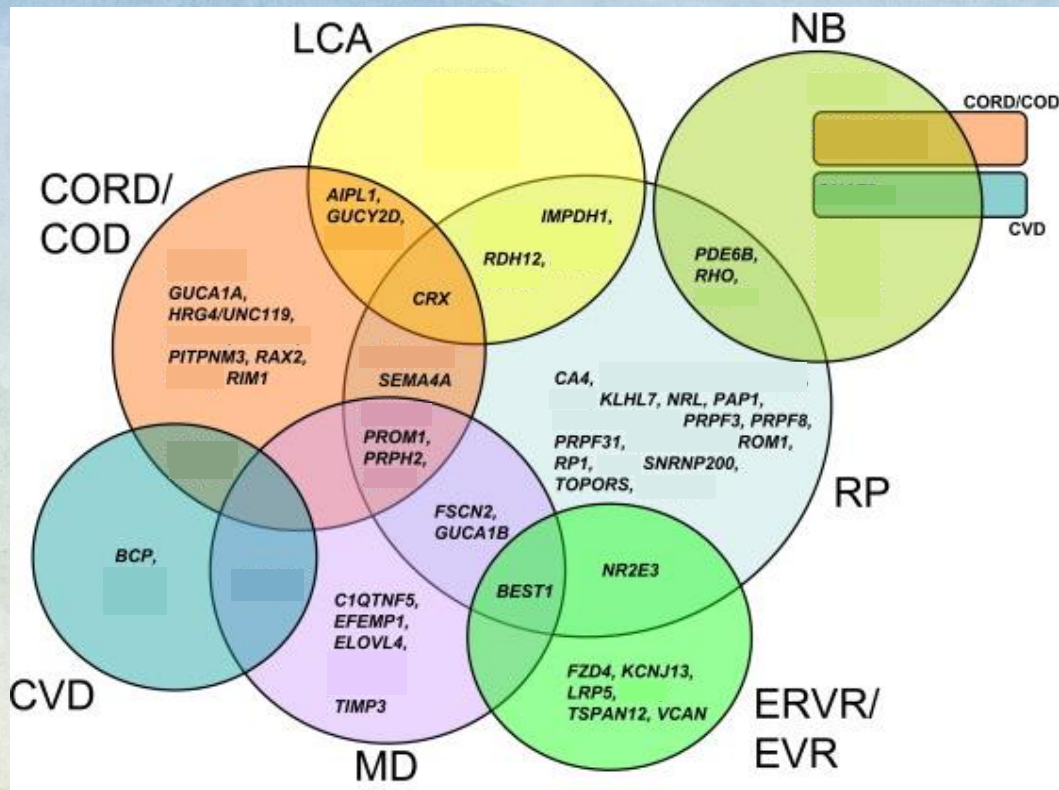
Autosomal
Recessive (AR)



Adapted from Berger W et al, 2010.

Retinal Dystrophies

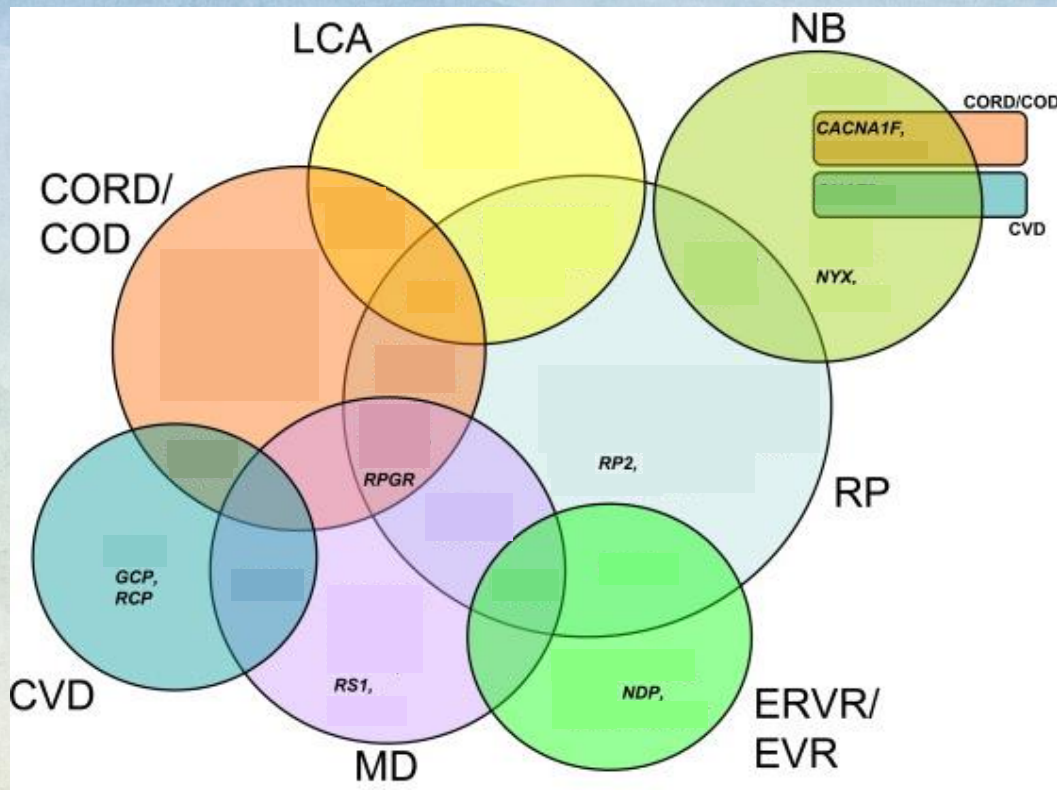
Autosomal
Dominant (AD)



Adapted from Berger W et al, 2010.

Retinal Dystrophies

X-linked (XL)



Adapted from Berger W et al, 2010.

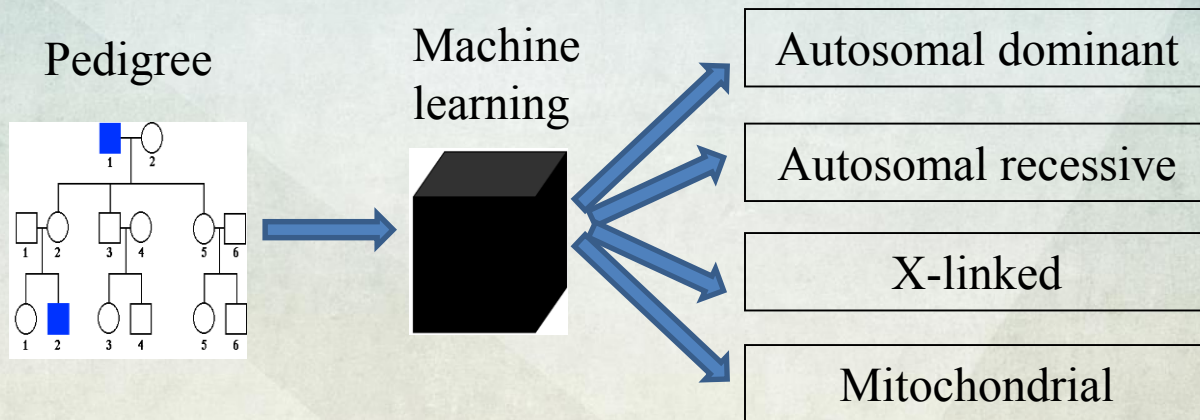
Inheritance Pattern Prediction

- May inform likely diagnosis
- Can guide appropriate genetic testing
- Allows calculation of likely risks to relatives
- Required component of data collection for some retinal dystrophy studies

As far as we are aware, there is no current algorithm to predict pattern of inheritance for a given patient, and not all retinal dystrophy clinics have genetics services

Aim

- Create a machine learning algorithm whose input is patient family history information and whose output is likely pattern of inheritance
- Used retrospective chart review on patients with genetically-proven retinal dystrophies



Data collection

- Kellogg Eye Center retinal dystrophy patients with genetic diagnosis
- Family history obtained by genetic counselors (and, in rare cases, retinal dystrophy specialists) as a part of routine patient care
- Information collected by engineering and medical students trained by genetic counselors and retinal dystrophy specialists
- Pedigrees converted into digital computer-readable form

Data collection methodology

- Students trained in predicting pattern of inheritance based on interpretation of pedigree appearance evaluated likely pattern of inheritance for each patient (277 patients)
- Answers to 12 questions about family history were collected from each patient's pedigree and analyzed with machine learning (100 patients)
- Answers to the same 12 questions were collected through computer feature extraction of a digitized pedigree and analyzed with machine learning (90 patients)
 - Included tolerance for user input error

(Overlap of 70 patients between the three cohorts)

Family history features

| | Question | Possible Answers |
|----|--|--|
| 1 | Is more than one generation affected? | Yes/No |
| 2 | Do any affected males have affected sons? | Yes/No |
| 3 | Do any affected males have affected daughters? | Yes/No |
| 4 | Are there any unaffected individuals who are "skipped"? (Their parents or siblings or grandparents are affected and children or grandchildren are affected, but they themselves are unaffected.) | 1. No 2. Yes - females only are skipped 3. Yes - at least some males are skipped |
| 5 | Are any siblings of the patient affected? | 1. No 2. Yes, and no other relatives are affected 3. Yes, and other relatives are also affected |
| 6 | Are any cousins of the patient affected? | 1. No 2. Yes - maternal cousins only 3. Yes - paternal cousins only 4. Yes - maternal and paternal cousins |
| 7 | Are both males and females affected? | 1. Yes 2. No - only males 3. No - only females |
| 8 | Is onset of disease < or = 20yrs in males? | Yes/No |
| 9 | Do any females have asymmetric disease? | Yes/No |
| 10 | In general, do females have less severe or later onset of disease? | Yes/No |
| 11 | Is there more than one retinal diagnosis in the family? (ex. Stargardt and Pattern Dystrophy) | Yes/No |
| 12 | Is consanguinity present? | Yes/No |

Family history features

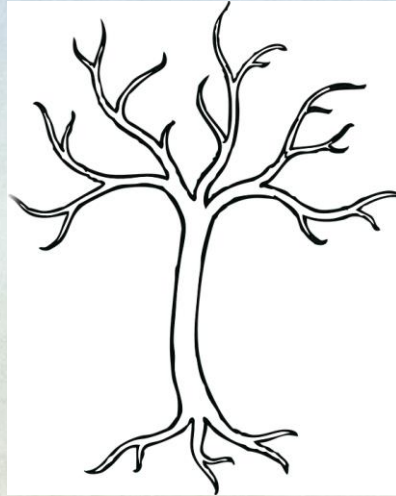
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Machine learning methodology

Gradient-Boosted Tree



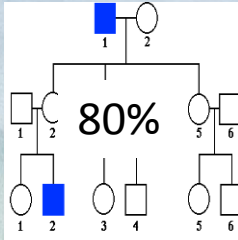
Machine learns appropriate weight for each branch

Decision tree

Machine learning methodology

80/20 training/testing split

Training

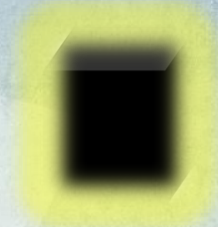


Inheritance pattern

Machine learning



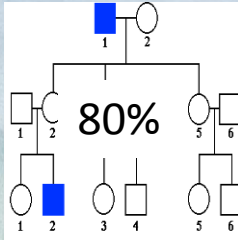
Classifier



Machine learning methodology

80/20 training/testing split

Training

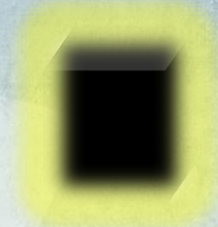


Inheritance pattern

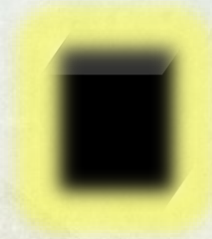
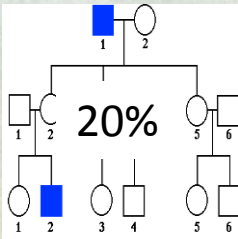
Machine learning



Classifier



Testing



Predicted pattern of inheritance

Results

| Method | Accuracy | Standard Deviation |
|--|----------|--------------------|
| Human-predicted | 84% | -- |
| Machine learning with human-entered answers | 78% | 7.5% |
| Machine learning with computer-extracted answers | 76% | 9.8% |

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| | | | | |
|------------|----|------|------|------|
| True Label | AD | 0.8 | 0.17 | 0.03 |
| | AR | 0.06 | 0.9 | 0.03 |
| | XL | 0.05 | 0.24 | 0.71 |
| | | AD | AR | XL |



| | | | | |
|------------|----|------|------|-----|
| True Label | AD | 0.67 | 0.33 | 0.0 |
| | AR | 0.11 | 0.89 | 0.0 |
| | XL | 0.0 | 0.5 | 0.5 |
| | | AD | AR | XL |



| | | | | |
|------------|----|-----|------|------|
| True Label | AD | 0.4 | 0.6 | 0.0 |
| | AR | 0.0 | 0.88 | 0.12 |
| | XL | 0.0 | 0.0 | 1.0 |
| | | AD | AR | XL |



Challenges

- Small dataset
 - Limited to patients with definitive genetic diagnosis
- Machine learning, but human-written questions
 - Our assumptions about the most important questions to ask may not always be correct
 - Is it better to ask more questions or fewer?
- Machines can make mistakes, too
 - Attributing importance to unimportant features (worse with small dataset)
- Perfect prediction is impossible
 - Ex. Isolated cases

Future Directions

- Collect more data from other institutions
 - Machine learning relies on large datasets for sufficient training
- As data collection increases, adjust questions that are informative/non-informative
 - Our expectations about what questions would be most useful might not have been correct
- Use machine learning directly on pedigree, without answering questions
 - Use statistical analysis (Bayesian inference, hidden Markov models) to supplement or substitute for machine learning methodology



- University of Michigan Kellogg Eye Center
 - Thiran Jayasundera, MD
 - Kari Branham, MS, CGC
 - Naheed Khan, PhD
 - Abigail Fahim, MD, PhD
 - John Heckenlively, MD
 - Eman Al-Sharif
- eyeGENE research project
- University of Michigan Computer Science & Engineering Department
 - Andrew DeOrio, PhD
 - **Edmond Cunningham**
 - Xinghai Zhang
 - Yaman Abdulhak
- Funding
 - University of Michigan Multidisciplinary Program (MDP)
 - Jayasundera startup grant