The Development of a Decision Analytic Model of Changes in Mean Deviation in People with Glaucoma

The COA Model

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Purpose: To create and validate a statistical model predicting progression of primary open-angle glaucoma (POAG) assessed by loss of visual field as measured in mean deviation (MD) using 3 landmark studies of glaucoma progression and treatment.

Design: A Markov decision analytic model using patient level data described longitudinal MD changes over 7 years.

Participants: Patient-level data from the Collaborative Initial Glaucoma Treatment Study (n = 607), the Ocular Hypertension Treatment Study (OHTS; n = 148; only those who developed POAG in the first 5 years of OHTS) and Advanced Glaucoma Intervention Study (n = 591), the COA model.

Methods: We developed a Markov model with transition matrices stratified by current MD, age, race, and intraocular pressure categories and used a microsimulation approach to estimate change in MD over 7 years. Internal validation compared model prediction for 7 years to actual MD for COA participants. External validation used a cohort of glaucoma patients drawn from university clinical practices.

Main Outcome Measures: Change in visual field as measured in MD in decibels (dB).

Results: Regressing the actual MD against the predicted produced an $R^2$ of 0.68 for the right eye and 0.63 for the left. The model predicted ending MD for right eyes of 65% of participants and for 63% of left eyes within 3 dB of actual results at 7 years. In external validation the model had an $R^2$ of 0.79 in the right eye and 0.77 in the left at 5 years.

Conclusions: The COA model is a validated tool for clinicians, patients, and health policy makers seeking to understand longitudinal changes in MD in people with glaucoma.

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In 2006, the Ocular Hypertension Treatment Study (OHTS) group reported that the lack of understanding of the change in glaucoma severity over time represents an important limitation in evaluation of the cost-effectiveness of glaucoma treatment.1 Although there have been a number of cohort studies of ocular disease, none of these have had been of sufficient size or duration to reliably describe changes in of glaucoma severity.2−5 However, the National Eye Institute has funded 4 large landmark studies of the treatment of glaucoma that might provide insights into changes in disease severity over time: The Collaborative Initial Glaucoma Treatment Study (CIGTS) compared surgical treatment with medical treatment of glaucoma in people newly diagnosed with open-angle glaucoma;6 the OHTS evaluated the efficacy of treatment of ocular hypertension in the prevention of progression to glaucoma in people with normal ocular discs and visual fields;7 the Advanced Glaucoma Treatment Study (AGIS) evaluated surgical methods for the prevention of glaucoma progression in people with advanced glaucoma;8 and the Early Manifest Glaucoma Study evaluated the treatment of people with early disease.9

Using data from clinical trials, it is possible to develop reliable models to describe the progression of chronic diseases, if the investigator uses robust methods of model validation.10 In this report, we describe the development and validation of a statistical model that brings together patient level data from 3 of these landmark clinical trials (CIGTS, OHTS, and AGIS) to develop a mathematical model describing changes in mean deviation in people with glaucoma over a 7-year period.

Methods

We constructed a Markov decision model to estimate changes in MD of a person with open angle glaucoma over 7 years based on...
a patient’s current MD, intraocular pressure (IOP), vertical cup-to-disc ratio (VCD), age, and race. Patient-level data from the CIGTS, OHTS, and AGIS clinical trials were pooled to determine the probability and magnitude of annual change in MD for each year. In the remainder of this paper, we refer to the pooled model as the “COA model” for the 3 trials that contributed data. We assessed the internal validity of the model by comparing the MD predicted by the model at year 7 with that reported for each study participant with similar characteristics. External validity was assessed by comparing the predicted with the actual MD at 5 years for patients drawn from an academically based glaucoma practice.

Acquiring and Preparing the COA Data

Written requests for patient-level data were made to the principal investigators of CIGTS and OHTS and deidentified data were provided after review. Deidentified AGIS study data are in the public domain and were obtained from the AGIS coordinating center. Because the purpose of the COA model was to estimate the change in MD over time, only summary quantitative measures of visual field status from reliable visual fields were requested, along with supporting demographic and clinical information. In addition, because the model is concerned only with the change in MD for patients with glaucoma, we used data only from those OHTS participants who developed primary open-angle glaucoma (POAG) during the study period. This project was approved and monitored by the Washington University Human Research Protection Office.

Building the COA Model

Data from the 3 studies were merged into a single data set and variable names harmonized. To ease the decision analytic modeling process, MD was converted to an integer measure by rounding off at 0.5 dB (e.g., −0.5 to −1.4 dB would be changed to −1.0 dB; −1.5 to −2.4 would be changed to −2.0 dB). This simplification was well within the test–retest reliability of the test; therefore, we have no reason to expect that this led to any clinically relevant bias in prediction of results. Initial descriptive and exploratory analyses were conducted with SAS 9.0 (SAS Inc, Cary NC). Although each study took place at a different period of time, we treated each year as if it occurred at the same time; that is, data for the first year of study after enrollment for each participants was considered “year 1” regardless whether that year occurred in 1989 (AGIS), 1994 (CIGTS), or 1995 (OHTS).

Transition matrices were calculated for each year by combining data from the 3 studies and stratifying by MD, age, race, VCD, and IOP category (see Results for discussion of the categories). The transition probabilities were based on the changes observed in the combined studies for each year. During each year modeled, the simulated participant has the opportunity to transition to a new MD in the following year that is worse (i.e., more negative) by as much as 4 dB, or to improve (become more positive) by as much as 3 dB, or to remain at the current MD. Where the stratification resulted in data too sparse to estimate the transition matrix, we used a “bagging” process to expand and smooth the available data. Bagging is similar to bootstrap sampling in that an iterative process is used to generate modeling estimates by resampling the available data with replacement, then averaging the fitted values. Thus, the available data for model estimation are increased by relying on a data set bounded by the available values and thereby avoiding extrapolation beyond the available data.

The Markov model was constructed using TreeAge Pro 2009 (TreeAge Software, Williamsport, MA). A Markov model is a mathematical representation of an iterative process. In this case, the iterative process describes the annual change in MD seen in people with glaucoma. A schematic of the Markov model is provided in Figure 1. The model estimates the change in MD for a person diagnosed with glaucoma over 7 years. Each simulated “participant” entering the model is assigned an initial MD, age, race characteristic (African American or not African American), IOP, and VCD based on their baseline characteristics (these could be either the characteristics of an actual patient, or characteristics of a hypothetical patient depending on whether an actual person or hypothetical person is being modeled). Each “year” (Markov cycle), the simulated participant proceeds were put through the model by randomly selecting a value from the transition table that determines the change in IOP and MD for that year, first for the right and then the left eye. The distribution of the change in IOP and MD was determined by the frequency of that change seen in the 3 studies for the year (within the categories of age, race, and VCD). After the determination of change in the left eye, the probability of surviving to the next year is determined by assessing the age-specific mortality rate for the participant estimated by the US Census life tables for 2004.

The Markov model was estimated using a microsimulation approach. In a microsimulation, the underlying simulated cohort is estimated with each simulated participant proceeding through the model individually. For example, let us assume that there are 10 000 cohort members who will be “walking” through the model described in Figure 1. At the beginning (root node) on the far left, each participant is assigned an age, race, VCD, IOP, and starting MD. Each participant proceeds individually to the right, and encounters a probability node defining what change in MD she will experience during her first year. She is assigned her progression based on a random draw from a distribution specific to her age, race, current MD, VCD, current IOP and the visit number (i.e., years 1–7). She then proceeds through the model to the next probability node “remembering” this information. This process is repeated for 7 years, and once completed the next “participant” proceeds through the model. The new person would have different baseline characteristics than the first participant, and thus would be likely to face different transition matrices depending on her prognostic factors. As we repeat the process 10 000 times (or some other arbitrary sample size) we develop a distribution that reflects the range and probabilities of change in MD for a cohort of people with the baseline characteristics in question. A similar process might be performed to assess the possible outcomes for a single individual with glaucoma.

Validation of the COA Model

Validation of a model is an essential element of its development, as a model that does not properly represent the underlying epidemiological process will not provide useful information for clinicians or health policy makers. In this project, we conducted internal validation of the model following the example of Kennedy. For those CIGTS, OHTS, and AGIS participants for whom we had 7 years of follow-up data, we compared their actual outcome with that predicted by the COA model for a simulated participant with the same characteristics (i.e., age, race, starting IOP, starting VCD, IOP at each visit, and starting MD). The microsimulation was run with a cohort of 10 000 simulated participants for each combination of characteristics seen among the participants in the CIGTS, OHTS, and AGIS studies. The $R^2$ as calculated by ordinary least-squares regression was used to evaluate the accuracy with which the predicted 7-year MD correlated with the observed. In addition, we assessed the proportion of predicted values that fell outside of a 3-dB range of the observed value at 7 years by following the example of the CIGTS group that found that a 3 dB difference in MD was determined to represent a clinically significant change in MD. We also report the number
of these observations that were overpredicted (i.e., the predicted value was worse than the observed) or underpredicted (i.e., the predicted value was better than the observed).

Internal validation provides insights into the model’s internal consistency of structure; external validation provides insights into the model’s generalizability beyond the data used in its construction. We conducted external validation of the COA model by comparing the predicted outcome at 5 years with that seen in a cohort of participants drawn from the glaucoma practice at Washington University School of Medicine. A report was generated from the medical records system for patients with a diagnosis of POAG for whom there was 5 years of data records within a 7-year period. Each record was reviewed independently by 2 research coordinators trained in review of ophthalmic records. Each coordinator confirmed that the patients met the inclusion criteria and that the patient had at least 5 visual field tests. Demographic data along with MD, IOP, and VCD were recorded for each visit. All data were entered into Excel files and converted to SAS for cleaning and reconciliation. The $R^2$ for the external sample was compared with the $R^2$ for the internal sample using the Fisher $Z$-transformation to determine similarity of fit. A nonsignificant $Z$-statistic was considered to be evidence of similar performance of the model between the external and internal sample, and thus, good external validity.

Results

We have summarized in Table 1 the characteristics of the data provided for participants from the 3 trials. These characteristics did not differ in any manner from that reported in the design or primary papers from these studies. The underlying theory in our model was that the only modifiable risk factor affecting change in MD was IOP and that the only effect of treatment on MD was through this mechanism. We did not consider treatment status of the eye (in CIGTS and AGIS) or participant (in OHTS). Therefore, all eyes of each participant were included in our modeling regardless of treatment status, thus slightly changing the number of participants included in our analyses from that seen in the trials. African Americans represented 45% of the sample. The CIGTS and OHTS participants were of similar age, and AGIS participants were almost a decade older. We expect that this heterogeneity in the training sample used for the model increases the generalizability of the model.

In Table 2, we report the follow-up seen from the participants of the combined trials over time. At 84 months (7 years), 66% of the participants remained in the sample. At 96 months, fewer than half remained. This was owing to a lack of data available for the OHTS cohort after year 7, as well as a drop-off in the follow-up for AGIS and CIGTS after year 7. This was the basis of our rationale to limit our model to change at 7 years.

In Table 3 we detail the changes in MD over 7 years related to baseline factors: age, race, VCD, and IOP. It is clear that older age, African-American race, and a worse VCD at baseline are associated with a larger loss of visual field (as measured by MD) over time. The relationship between baseline IOP and changes over time are not as clear, but because the model considers not only the IOP at baseline, but also IOP at subsequent visits, the influence of baseline IOP did not extend beyond the first year. We estimated...
transition probabilities for each combination of MD (at integer) and time (i.e., years 1–7) stratified by combinations of age (categorized as shown in Table 3), race, and IOP (categorized as shown in Table 3).

The result of internal validation of the COA model is shown in Figure 2. Our internal validation sample consisted of 470 participants from the COA sample (36%) for whom we had complete data over the 7 years. The triangles in the scatter plots represent the combination of the observed and predicted value for a participant whose prediction fell outside of the 3-dB band (161 in the right eye and 172 in the left), 54.6% (88 of 161) were “overpredicted” (i.e., the model predicted that MD at 7 years would be more negative (worse) than participant’s actual reported outcome at 7 years. “Worse” means that the model predicted the patient would have an MD less negative than the actual.

Our external validation sample is detailed in Table 5. Data on 300 participants were extracted from clinic files. Of these, 150 had sufficient longitudinal data to use for validation at 5 years (data for 7 years were too sparse for validation purposes). The characteristics of this subset are not significantly different than the rest of the data extracted from patient records. The predicted outcome for these patients was compared with the actual at 5 years and found to have an R² of 0.79 for the right eye and 0.77 for the left. Comparison of these correlation statistics to those of the internal validation found that the differences were not significant, indicating good external validity of the model.

Table 3. Average Rate of Change in Mean Deviation (MD) Seen in the Right and Left Eyes over 7 Years by Risk Factor

<table>
<thead>
<tr>
<th>Risk Factor/Category</th>
<th>Right Eye (dB)</th>
<th>Left Eye (dB)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age (y)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt; 60</td>
<td>-0.32</td>
<td>-0.60</td>
</tr>
<tr>
<td>≥60 and &lt; 70</td>
<td>-1.86</td>
<td>-1.63</td>
</tr>
<tr>
<td>≥70</td>
<td>-2.24</td>
<td>-2.94</td>
</tr>
<tr>
<td>Race</td>
<td></td>
<td></td>
</tr>
<tr>
<td>African American</td>
<td>-1.49</td>
<td>-1.89</td>
</tr>
<tr>
<td>Non-African American</td>
<td>-1.24</td>
<td>-1.23</td>
</tr>
<tr>
<td>Baseline IOP (mmHg)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>≤17.5</td>
<td>-1.40</td>
<td>-2.13</td>
</tr>
<tr>
<td>&gt;17.5 and ≤20.5</td>
<td>-0.86</td>
<td>-1.45</td>
</tr>
<tr>
<td>&gt;20.5 and ≤23.5</td>
<td>-1.29</td>
<td>-0.54</td>
</tr>
<tr>
<td>&gt;23.5 and ≤26.5</td>
<td>1.40</td>
<td>-1.82</td>
</tr>
<tr>
<td>&gt;26.5</td>
<td>-1.54</td>
<td>-1.87</td>
</tr>
<tr>
<td>Vertical cup:disc ratio</td>
<td></td>
<td></td>
</tr>
<tr>
<td>&lt;0.6</td>
<td>-0.96</td>
<td>-1.52</td>
</tr>
<tr>
<td>≥0.6</td>
<td>-1.49</td>
<td>-1.50</td>
</tr>
</tbody>
</table>

Table 2. The Number of COA Participants Examined at Each Annual Follow-up Visit

<table>
<thead>
<tr>
<th>Visit Number (Months Since Randomization)</th>
<th>Number of Participants</th>
</tr>
</thead>
<tbody>
<tr>
<td>Baseline</td>
<td>1418</td>
</tr>
<tr>
<td>12</td>
<td>1300</td>
</tr>
<tr>
<td>24</td>
<td>1264</td>
</tr>
<tr>
<td>36</td>
<td>1218</td>
</tr>
<tr>
<td>48</td>
<td>1173</td>
</tr>
<tr>
<td>60</td>
<td>1144</td>
</tr>
<tr>
<td>72</td>
<td>1094</td>
</tr>
<tr>
<td>84</td>
<td>929</td>
</tr>
<tr>
<td>96</td>
<td>682</td>
</tr>
</tbody>
</table>

COA = data pooled from the Collaborative Initial Glaucoma Treatment Study, the Ocular Hypertension Treatment Study, and Advanced Glaucoma Intervention Study.

COA = data pooled from the CIGTS, the OHTS, and AGIS.
IOP = intraocular pressure; SD = standard deviation.
*The numbers listed for each study differ from previous published totals due to use of both eyes in the model and requirement of follow-up (see text). For OHTS, only those who achieved a glaucoma endpoint as reported in Kass 2002 are included.†
†Average total change in mean deviation (MD) over 7 years.
Discussion

For this report, we pooled patient-level data from the CIGTS, OHTS, and AGIS studies to create a model of change in MD over time and over most of the range of glaucomatous damage classifications. We extensively validated the model, including comparison with a sample of patients not enrolled in the clinical trials. Our validation process demonstrates that the model provides excellent prediction of the changes in MD of patients with POAG over 7 years. This could be an important tool for people with glaucoma and their physicians who are seeking to understand their disease prognosis, as well as clinical and health policy investigators seeking to model the impact of population-based interventions for glaucoma.

In our 2006 report from the OHTS Group, we demonstrated that treatment of people with ocular hypertension with moderate to high risk of progression (i.e., ≥2% annual risk of developing POAG) was cost effective. However, to make it clear that our results said nothing concerning the treatment of glaucoma to prevent disease progression, we added a key caveat:

Our . . . simulation also makes clear the significant limits to our knowledge of important aspects of the natural history of glaucoma and its impact on quality of life. . . . Among the most influential parameters in the model are the . . . estimate of the progression of POAG. Our results demonstrate that investigation concerning these aspects of the disease would provide important information to support future evaluations of treatment to prevent glaucoma or slow its progression.

Because the OHTS Economic Model lacked a validated estimate for glaucoma progression, it would not provide a basis for conducting cost-effectiveness or cost-benefit of glaucoma treatments such as improved IOP control or neu-

Table 4. Comparison of the Predicted Mean Deviation (MD) at 7 Years by COA Model to the Actual MD for a Person with the Same Age, Race, and Starting MD and Intraocular Pressure at 7 Years

<table>
<thead>
<tr>
<th>Difference between Predicted and Actual at 7 Years (dB)</th>
<th>Right Eye, n (%)</th>
<th>Source of Study Eye</th>
<th>Left Eye, n (%)</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>CIGTS</td>
<td>OHTS</td>
<td>AGIS</td>
<td>Total</td>
</tr>
<tr>
<td>&gt;4</td>
<td>29 (12)</td>
<td>6 (7)</td>
<td>12 (9)</td>
<td>47 (10)</td>
</tr>
<tr>
<td>&gt;3 but &lt;4</td>
<td>26 (10)</td>
<td>8 (9)</td>
<td>7 (5)</td>
<td>41 (9)</td>
</tr>
<tr>
<td>&gt;2 but &lt;3</td>
<td>43 (17)</td>
<td>15 (17)</td>
<td>13 (10)</td>
<td>71 (15)</td>
</tr>
<tr>
<td>&lt;2 but ≥2</td>
<td>107 (43)</td>
<td>44 (51)</td>
<td>62 (46)</td>
<td>213 (45)</td>
</tr>
<tr>
<td>&lt;2 but &gt;3</td>
<td>8 (3)</td>
<td>5 (6)</td>
<td>12 (9)</td>
<td>25 (5)</td>
</tr>
<tr>
<td>&lt;3 but ≥4</td>
<td>9 (4)</td>
<td>2 (2)</td>
<td>5 (4)</td>
<td>16 (3)</td>
</tr>
<tr>
<td>&lt;4</td>
<td>26 (10)</td>
<td>7 (8)</td>
<td>24 (18)</td>
<td>57 (12)</td>
</tr>
<tr>
<td>Total</td>
<td>248</td>
<td>87</td>
<td>135</td>
<td>470</td>
</tr>
</tbody>
</table>

AGIS = Advanced Glaucoma Intervention Study; CIGTS = Collaborative Initial Glaucoma Treatment Study; COA = data pooled from the Collaborative Initial Glaucoma Treatment Study, the Ocular Hypertension Treatment Study, and Advanced Glaucoma Intervention Study; OHTS = Ocular Hypertension Treatment Study.
The ability of clinical investigators to address this weakness of the literature has been limited by the expense and logistic difficulty of assembling the necessary cohort of glaucoma patients to evaluate changes in MD over time. Broman et al20 brought together cross-sectional data from 9 studies totaling 1066 participants, and used a parametric approach to estimate changes in MD relying on age and age-specific incidence to simulate the duration of disease. Using these methods, the investigators found an average annual rate of progression of glaucoma ranging from 1.12 dB per year for people of European ancestry to 1.56 dB per year for people of Chinese ancestry. People of African-American ancestry progressed at a rate of 1.33 dB per year.

The investigators from the Early Manifest Glaucoma Trial (EMGT) found that, among their participants who progressed, there was an average worsening of 1.93 dB over 5 years.21 Broman et al acknowledged the discrepancy between their findings and those of the EMGT and pointed out that this might be because of the difference in treatment and IOP control between the participants in the clinical trials and those enrolled in the cross-sectional, population-based cohort studies.

It is difficult to compare the results of the COA model with those of the natural history arm of EMGT and the work of Broman et al. The EMGT and Broman’s studies are designed as natural history studies and report the result of a population mean. The COA is designed to predict the progression of individual patients given a specific vector of risk factors. It is interesting to note, however, that the average change in MD over 7 years in both OHTS and AGIS was approximately the same seen in EMGT at 5 years (Table 1). We also found that in modeling the change over time for a hypothetical patient with a starting MD of −4 dB the average change in MD was similar to that seen in EMGT (results available by contacting the corresponding author).

Broman et al20 have the additional limitation for assessing the influence of treatment on progression in that it does not assess the influence of IOP control, and assumes a linear relationship across the disease spectrum. The COA model addresses both of these limitations. We predicted ending MD within 3 dB in nearly two-thirds of participants at 7 years. An important strength to our model over the other models mentioned is that we have explicitly incorporated longitudinal changes in IOP as a risk factor. This allows investigators to assess the importance of IOP reduction and stability in slowing patient progression.

Our test of internal validity found that this model has a good fit, as measured by our $R^2$s of 0.68 for the right eye and 0.63 for the left (Figure 2). We searched the literature to determine how the fit of this model compared with other predictive models relying on continuous outcomes such as MD and identified a range of $R^2$ ranging from 0.45 to 0.76.22–25 Other examples we found had a fit considerably lower than these. Although we do not claim that this is an exhaustive review of the predictive literature in medicine, we believe that this indicates that our model well meets most accepted standards of prediction. Furthermore, the fact that the model correctly classifies nearly two-thirds of COA participants indicates that the model has clinical value in providing prognostic information to people with glaucoma, particularly those with early disease.

Our external validation of the COA model is an important strength. It demonstrates that our model is capable of prediction beyond the component studies and thus has good generalizability, an essential quality if the model is to be used by policy makers or physicians. Some might find it surprising that the COA model is more accurate in predicting the outcomes of the external validation sample than it does for the sample on which it was based. We posit 2 reasons for this. First, owing to problems in obtaining a sufficient sample, we were only able to model 5 years of data with our clinic-based sample. It is understandable that the COA model
would be more accurate at 5 years than 7. Second, what we have actually found is that the $R^2$ for the external sample is not different from that of the internal sample. The proper interpretation of this is not that the external sample fits better than the internal one, but that the difference between the samples is because of random variation, not any systemic source. In other words, we found that the fit statistics for each sample lies within the confidence bound of the other.

Some might also question our external validation using the results of a single institution. It is important to note here that the COA sample itself is a national sample (albeit one that incorporates participants in clinical trials); therefore, our external validation consists of a comparison of a local sample to a national one, and we recognize that our claim of generalizability is limited by the use of a sample drawn from a single institution. This limitation is, of course, necessitated by the difficulty of obtaining similar data from other centers or nationally. Therefore, we invite centers that are interested in collaborating with us in conducting further validation of the model using their own data to contact the corresponding author.

It is an important limitation to our knowledge of the epidemiology of glaucoma that there are few validated models describing changes in glaucoma severity. This deficit has important consequences for patients and policy makers. For the newly diagnosed patient and her physician, this makes it difficult to describe prognosis and the benefit of treatment. For the health policy maker, this deficit makes it nearly impossible to properly characterize the value of new medical and surgical treatments. The COA model represents an innovative evidence-based approach to prediction of changes in MD using rigorous statistical methods. In the future, we hope to make this important tool readily available to patients, physicians, and investigators interested in better predicting change in visual function over time.

References


Footnotes and Financial Disclosures

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