AIM (Angular Indication Measurement)- Visual Acuity: An adaptive, self-administered, and generalizable vision assessment method used to measure visual acuity

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Abstract

This proof-of-concept study introduces Angular Indication Measurement and applies it to VA (AIM-VA). First, we compared the ability of AIM-VA and ETDRS to detect defocus and astigmatic blur in 22 normally-sighted adults. Spherical and cylindrical lenses (±0.00D, +0.25D, +0.50D, +0.75D, +1.00D, +2.00D and +0.50D, +1.00D, +2.00D each at 0°, 90°, 135°, respectively) in the dominant eye induced blur. Second, we compared repeatability over two tests of AIM-VA and ETDRS. A 2-way-ANOVA showed a main effect for defocus- blur and test with no interaction. A 3-way-ANOVA for the astigmatism experiment revealed main effects for test type, blur, and direction and with no interactions. Planned multiple comparisons showed AIM had greater astigmatic-induced VA loss than ETDRS. Bland-Altman plots showed small bias and no systematic learning effect for either test type and improved repeatability with >2 adaptive steps for AIM-VA. AIM-VA's ability to detect defocus was comparable with that of an ETDRS letter chart and showed greater sensitivity to astigmatic blur, and AIM-VA's repeatability is comparable with ETDRS when using 2 or more adaptive steps. AIM's self-administered orientation judgment approach is generalizable to interrogate other visual functions, e.g., contrast, color, motion, stereo-vision.

Keywords: Visual acuity, psychophysical methods, response-adaptive, self-administered, astigmatism

1. Introduction

Visual function assessment is an essential task for eye care clinicians. Visual acuity (VA) is arguably the key endpoint in current clinical practice to assess central vision as it is sensitive to changes of the refraction of the eye (1), but also changes of central retinal (2) and ocular media structures (3), changes along the visual pathway (4), and atypical visual-cortical structures (5). Since the first known reports for VA tests by Tobias Mayer 1757 (6) and Heinrich Küchler in 1843 (7), various techniques and tasks have evolved to measure VA; however, all of those tests examine the observer's ability to discern different structural details from each other by subjective report, for recent review see (6).

One of the major challenges for eyecare professionals worldwide is the accurate correction of refractive errors, where estimates suggest that approximately 2.6 billion people are affected by myopia and 1.8 billion people by presbyopia (8). Defocus blur, i.e., isotopically distributed blur, may reduce VA in cases of myopia and higher hypermetropia and can be corrected with spherical lenses. Astigmatism, i.e. anisotopic blur with two focal lines of different focal lengths, causes uneven defocus and may generate systematic distortions of an image on the retina.

In practice, VA assessment typically requires a participant to identify letter or number optotypes in 4 alternative forced choice (AFC) letter charts such as HOTV (9), 10AFC letter charts such as Early Treatment of Diabetic Retinopathy Study (ETDRS) (10, 11) or 9AFC letter charts such as Snellen (12). Observers who are non-literate or unfamiliar with the Roman alphabet may be asked to identify the orientation of the gap of tumbling E (13) or Landolt C (14) optotypes in 4 or 8 AFC tasks; to match child-friendly optotypes such as Lea symbols (15) or Auckland Optotypes (16) in 10AFC tasks; or to report the offset of misaligned line stimuli in 2AFC tasks (17). For participants such as babies who are unable to generate a verbal or manual report, preferential looking behavior can be converted to a 2AFC task (18).

Typically, optotypes are presented in printed or computer-generated charts in which the layout, size and spacing of optotypes is organized with monotonically increasing difficulty. The number and spacing of letters can vary (12) or may be logarithmic (19)

following the Weber-Fechner law (20). Various rules have been developed to score

the correct and incorrect responses of the participant, including letter scoring, in

which VA is related to the total number of letters correctly identified; line scoring, in

which VA is defined by the optotype size where a specified proportion of letters are

correctly reported; and stopping criteria, all of which affect the estimate of VA and

repeatability (21–23).

The transition to computer-administered tests avoids memorization artifacts of

printed charts and allows the implementation of adaptive psychophysical testing

principles (24, 25) and automated scoring eliminating inter-examiner differences (24,

26). More recently, numerous web-browser- and app-based VA tests have entered

the field to monitor VA remotely (25, 27, 28).

Current tests do not capture an individual's performance change with changing

visibility levels for targets of differing size but rather only report the threshold, or

acuity at a single point estimated from the psychometric function. The range between

reliably detecting an optotype or its orientation and plateauing in performance is

currently not captured, however parameters such as the slope of the psychometric

function may be informative as well as they can be affected by blur (29). Also, an

individual's baseline error rate (i.e., noise or uncertainty for supra-threshold letters

that are correctly identified) is not captured using chart-based VA tests and might be

systematically disrupted due to factors such as visual distortions.

To address the shortcomings of current visual acuity tests described above (namely

optotype structure, familiarity and # alternatives; behavioral task; and data scoring

and fitting), we introduce the Angular Indication Measurement (AIM) method, which

overcomes these issues and apply it to measure VA. AIM is computer-based,

generalizable, self-administered, and response-adaptive. AIM can be applied to

measure multiple visual function endpoints by adapting different stimulus and

response formats to an orientation judgment. In each case, stimuli are presented at a

range of personalized stimulus levels and random orientations (acuity, color,

contrast, stereoscopic disparity etc.) or directions (motion). The observer's task is to

report the perceived orientation or direction of the stimulus. The difference between

the actual and reported stimulus orientation/direction provides a continuous error

score as a function of stimulus intensity (30). For VA, observers report the orientation

of randomly-oriented C-like stimuli of varying sizes, which requires no literacy or

memory for optotypes by the participant and automated data analysis for the

examiner.

The current study has three objectives. First, we introduce AIM for Visual Acuity

assessment (AIM-VA) and describe its features and outcome measures. Second, we

report AIM-VA's ability to detect defocus and astigmatic blur induced with ophthalmic

lenses in normally-sighted adults and compare it with results of a conventional

ETDRS letter chart (Experiment 1). Third, we investigate the repeatability of AIM-VA

and compare it with ETDRS results (Experiment 2).

2. Methods

Participants were recruited from the translational vision laboratory and from the

undergraduate population at Northeastern University, Boston. Undergraduate

students received course credit towards the completion of their Introductory

Psychology course in exchange for their participation. Written and verbal information

about the project were provided in advance to the participants and they gave

informed consent via online form before taking part. The Institutional Review Board

of Northeastern University approved the study and was in line with the ethical

principles of the Helsinki declaration of 1975.

2.1 Participants

Pilot experiment: Eleven adults from Northeastern University's student cohort without

a known (self-reported) ocular opacity, accommodative or extreme refractive

disorder/disease took part.

Experiment 1: Twenty-three adults from Northeastern University's student cohort and

members of the Translational Vision Science lab (including JS, MF, JBS, JH) took

part in the study. One participant was excluded due to erroneously duplicated data.

Of the remaining twenty-two participants (age range 18-35 years), eight wore no

optical correction, fourteen wore correction. An objective refraction was performed
using autorefractometer (AR(K)7610 Auto Refracto-Keratometer) and prescription of
glasses if worn, were measured using an auto-focimeter (Xinyuan, Auto Lensmeter).
Although no data on subjective, optometrist-performed refraction was available, we
assume that participants who wore corrections were best-corrected.
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Experiment 2: Participants from experiment 1 participated also in the second
experiment.
Control experiment: An additional control experiment was performed to determine
the effect of the number of adaptive steps on repeatability, twelve participants were
recruited, including a subset of seven participants from Experiment 2 and five
additional participants (three with and two without correction).
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178   **2.2 Equipment**
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Stimuli were presented on a HP (USA) All-in-One computer and monitor system
(Model: 32-a1027c Rfrbd PC) with a framerate of 60 Hz with a pixel resolution of
[3840 / 2160], that subtended 10° at a 4m viewing distance with a resolution of 384
pixels per degree. Stimuli were generated using Matlab (MathWorks, R2021a)
software in combination with *The Psychophysics Toolbox* (31, 32). An HP Bluetooth
mouse was used as indication device. The monitor was gamma-corrected using
datacolor Spyder (Switzerland) photometer. We used a retro-luminant ETRDS chart
(Lighthouse, USA, second edition) and measured the background luminance with a
LS-110 (Konica Minolta Sensing Americas) photometer.
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190   **2.3 AIM principal**
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AIM is a generalizable psychophysical paradigm in which one or more stimuli is
presented at a random orientation, and the observer's task is to indicate on a
surrounding response ring the perceived orientation of the target, using a mouse or
touch response. The difference between the true and reported orientation provides a
continuous error score for each stimulus test level/intensity. Across different tests,
stimulus intensity can refer to contrasts and spatial frequencies (33), cone-isolated

contrasts and color differences (34), stereoscopic disparities (30) etc. In the current

study, we varied optotype size to estimate VA. AIM combines a number of features

that address current problems of other tests and also provides additional information

about an observer's performance, namely a) generalizable orientation judgment

method, b) inclusive method as AIM is easy to perform due to the simple task nor

does it require knowledge of letters or numbers, c) self-administered, thus reducing

effort on part of the investigator, d) response-adaptive, thus narrowing in on

individual threshold, e) continuous data, enabling personalized modeling of visual

performance, f) active engagement of the user due to the search aspect of the task,

g) instant analysis due to computerized approach, h) digital, thus deployable on a

wide range of computerized screens, e.g. phones, tablets, laptops, computer

monitors.

We used C optotypes as they have the advantage that the physical and spectral

properties of optotypes do not vary within a chart except by the gap orientation and

do not vary in meaning, unlike object optotypes such as letters, hence the acuity

determined with such stimuli is a more unbiased estimate of an observer's VA.

Following International Council of Ophthalmology (30) recommendations, Cs were

defined as interrupted circles whose stroke width and gap width are one-fifth of its

outer diameter. The only difference of our C stimulus and a Landolt C is that the

inner diameter had a smaller width than the outer part as an arc was used to

construct the gap (Figure 1B).

We presented a range of C optotypes with logarithmically spaced sizes. This feature,

unlike single trial AFC methods, ensures that an observer always views stimuli that

range from likely to unlikely to be seen and thereby ensuring for the inclusion of a

clear exemplar, which may be particularly important for observers with visual or

cognitive impairment. C optotypes were presented on a series of 2 charts, each

comprising a 4x4 grid of 1.3° cells surrounded by a 0.1° thick indication ring. The

background luminance and stimulus luminance were adjusted to match that of the

ETDRS chart. The pixel size was 0.0026° (384 pixel/1°). The C gap width was kept

at 20% of the line width and was constructed as an arc. The inter-cell gap distance

was 0.1°. The surrounding rings had a grey hue and changed color to green to

visualize an indication was made as well as the appearance of a black arc that
indicated the reported gap orientation (Figure 1 A).

The stimulus sizes varied from difficult to easy, with log-step sizes starting from -0.30
logMAR to 0.80 logMAR on the first chart, following the design principle of the
ETDRS chart (19). The range of sizes on the second and subsequent charts were in
equal log steps between the latest threshold size estimate ±2σ from the fit to the
results of previous charts. The orientation of the optotype in each cell was randomly
drawn from a uniform distribution and the different optotype sizes were randomly
assigned to each cell. Participants indicated the perceived orientation of all 16 stimuli
on one chart by clicking on each ring (Figure 1 A), guessing if unsure. Participants
could repeatedly correct their choice by clicking again on any response ring and
proceeded to the next chart by clicking on a 'Next' button that became available only
once responses had been made to all 16 stimuli. The response error is calculated as
the difference between target orientation and indicated orientation in degrees (see
Figure 1 B).

The response errors for each optotype size were converted to 8AFC responses,
where absolute orientation errors less than 22.5° was scored as correct response,
and errors greater than 22.5° were scored as incorrect responses. A cumulative
gaussian psychometric function was fit to the 8AFC data, from which VA threshold
±2σ was used to select the optotype sizes for the next chart.

At the end of the final chart, a cumulative gaussian function was fit to the error report
data as a function of optotype size from all charts, defined as:
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$$\tau_\theta(s) = \left(\theta_i + \left(\theta_{guess} - \theta_i\right)\right)\left(0.5 - 0.5 * erf\left(\frac{s-\tau}{\sqrt{2}\gamma}\right)\right) \qquad (1)$$

Where θ_{guess} is orientation error for a guess response (90° for a stimulus with 360°
range e.g., C; or 45° for a stimulus with 180° range e.g., grating-type stimuli), *s* is
signal intensity (here stimulus size), *τ* is a sensitivity threshold, θ_i is internal angular
uncertainty (*noise* – the angular report error for the most visible stimulus), and γ is
the slope of the function.
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Figure 1 B) shows the different parameters in a schematic diagram of the

psychometric fit after an AIM experiment. AIM's threshold, here AIM-VA, is halfway

between random responses and plateauing of performance, i.e., *noise.* As an

existing test may have a fixed threshold criterion, e.g., ETDRS chart min. 60%

correct, we can calculate an ETDRS-equivalent VA. To do so, we constrained the

free parameters of AIM's fit for noise and slope so that the number of free

parameters was comparable to the ETDRS letter chart. On the other hand, AIM-VA

can be additionally calculated with slope and noise as additional free parameters.

This difference may cause slight variations in VA performance and also enable

personalized models of psychometric fits (See Supplementary materials, Figure S1

for more details on fully versus semi-constrained fits). Specifically, the slope of a

psychometric function reveals the change of the detection probability across stimulus

levels, and is not scored for ETDRS although it can be used to report the logMAR

range from guessing to perfect performance (29). AIM provides a similar range,

which we define as the logMAR range between VA and the 95% point on the

cumulative Gaussian function (5% above an observers min. angular indication error,

Figure 1B). We refer to the logMAR distance between VA and the 95% angular error

point as *range of stimulus detectability improvement* (ROSDI), which describes the

range of optotype sizes over which visual function improves that may provide useful

information concerning vision changes in different eye diseases. As threshold, slope,

and noise interact, ROSDI is a single endpoint that incorporates all these

parameters.

Figure 1C shows example data for one observer where the x axis shows optotype

size in logMAR units, the y axis shows absolute angular error in degrees, blue circles

show the orientation reports of an observer, the red line shows the best fitting

cumulative gaussian function, green dashed lines show 95% confidence intervals at

each optotype size.

Figure 1: Schematic of AIM principal. A) In series of charts, each containing a 4*4

grid of cells, stimuli that range from easy to hard-to-detect are presented while being

scaled logarithmically and randomized in regards of size and gap orientation. Each

cell consists of a stimulus surrounded by a response ring. Participants reported the

orientation of the C optotype in each cell by clicking on the corresponding location on

the response ring. Their response is indicated by an arc on the ring, and participants

are able to adjust their response by clicking again on the response ring. After a

response has been made for all cells, a 'Next' button is presented, which participants

can click when they are satisfied with their responses to the chart. The stimulus sizes

on the next chart are updated, based on the responses to stimuli in previous charts

and again logarithmically scaled in size. B) Schematic of AIM scoring and outcome

parameters. Y axis represent the angular indication error, the x axis the stimulus

size. C) Representative results and fit after three charts including psychometric

function (red line), 95% confidence intervals (dashed green lines), and single

responses (blue circles).

Orientation errors resulting from an observer's responses should be scattered

randomly in either clockwise or anti-clockwise orientation relative to the target

orientation and increase in magnitude with decreasing stimulus size. However, if

optical distortion is present due to astigmatism, a systematic bias in error reports

may occur that is expected to follow a sinewave pattern. If an observer is corrected

for astigmatism, then distortions should be minimized unless neuro-ophthalmic

distortion is present e.g., due to amblyopia (35). AIM can quantify orientation report

error bias. The orientation bias was calculated as a function of error direction

(clockwise or anti-clockwise relative to target orientation), target orientation and large

to small stimulus sizes using a two-parameter sinewave function to calculate phase

(bias) and its amplitude.

$C = A * sin(\varphi + \theta_{error})$ (2)

Where θ_{error} is the observer's orientation error for a target gap orientation, φ is the

phase (bias), and A is amplitude of the function. Also, we calculated the R^2 to

evaluate the goodness of fit.

2.4 Procedure

To ensure comparable luminance levels, we measured at 1m the mean letter and

background luminances for ETDRS with a LS 110 photometer, which were 5.2cd/m^2

±1.3 and 190.6 cd/m^2 ±1.6, respectively. The minimum and maximum luminance

values of the monitor were 0.3 cd/m^2 and 250 cd/m^2, respectively and gamma was

2.0. The LUT values for the background was adjusted to 191 to match AIM-VA to

ETDRS.

Pilot experiment-refractive error methods: We fitted constrained psychometric

functions (threshold as a free parameter for each fit, with slope and error shared

across all fits for each observer) and calculated ETDRS-equivalent VA for AIM. The

test was performed in 4m viewing distance at constant lighting conditions. Two

charts, including one adaptive step, were deployed to measure the VA. One-D

(Diopter)-step ±spherical lenses were placed in a trial-frame until the largest C was

unrecognizable. We analyzed data ranging between -10.00D and +3.00D. Matlab's

anova1 and *multcompare* (Tukey-Kramer test) functions were used for ANOVA

analysis and planned comparisons, respectively.

First experiment: We used spherical (i.e., 0.00D, +0.25D, +0.50D, +0.75D, +1.00D,

+2.00D) followed by cylindrical blur lenses (+0.50D, +1.00D, +2.00D, each in 0º, 90º,

135º) at a 4m viewing distance (thus +0.25D remaining accommodation). For each

part of the experiment, lenses were randomized within and between observers and

placed in front of the dominant eye in a conventional trial frame while the opposite

eye was occluded with a dark occluding lens. Matlab's *anova1*, *anovan,* and

multcompare functions were used for ANOVA analysis and planned, multi-

comparisons, respectively, for thresholds, noise, slope, ROSDI, and the phase and

amplitude (we converted all values to positive numbers) values of the orientation

report error analysis. All other aspects were the same as in the pilot experiment. For

ETDRS, VA is defined at the line at which 3/5 (60% correct) letters are correctly

identified, which corresponds to an angular error of $\pm72°$ for AIM-VA. ETDRS has a

single parameter estimate (threshold -VA), to equate the number of free parameters

we shared the slope and noise level parameters across all data for each subject and

allowed threshold to vary in order to equate the number of free parameters for AIM-

VA and ETDRS. The observer's task during the ETDRS test was to read the letters

until the line that was read with more than 2 incorrect answers. Each correct letter

was count 0.02logMAR within the last line, e.g., if the performance at the last line

was at 0.00logMAR and 4/5 correct answers were given, this would result in

+0.02logMAR VA.

Second experiment: We investigated the repeatability of AIM-VA and compared it to

ETDRS test-retest. The dominant eye was tested, and each test was repeated twice

on the same day by the same experimenter where the order of tests was randomized

using two charts for AIM.

Control experiment: To test whether more adaptive steps for AIM-VA would have an

impact on repeatability, we performed a control experiment with a subset of

participants who were tested with four adaptive steps (5 charts). Bland Altman plots

were used in combination with a linear regression model.

3. Results

3.1 Experiment 1- Testing AIM's ability to detect induced defocus and

astigmatic blur

We initially performed a pilot experiment with negative and positive spherical 1.00D

lenses in young (mean age 21), normally sighted observers. A one-way ANOVA (-

10.00D to +3.00D) revealed a significant effect of defocus on VA [$F(13,140)=14.0$,

$p<0.001$, $\eta_p^2 = 0.57$]. All positive spherical lenses caused a defocus-based decline in

VA with a significant effect using +2.00D (planned comparisons). Negative lenses

caused an accommodation-based flat function of VA until approximately -5.00D, and

then VA decreased significantly for-10.00D. No significant effect of defocus on

psychometric function slopes [$F(13,140)=0.89$, $p=0.57$, $\eta_p^2 = 0.08$] and ROSDI

[$F(13,140)=0.89$, $p=0.57$, $\eta_p^2 = 0.08$] was found but there was a significant increase

in orientation error noise [$F(13,140)=3.86$, $p<0.001$, $\eta_p^2 = 0.26$] due to induced

refractive error. A planned comparison showed this was due to positive blur (see

supplementary materials, Figure S2 & S3).

We then investigated whether AIM would be able to detect defocus blur smaller than

1 diopter and compared results with that of an ETDRS letter chart.

For the group without correction worn, the median spherical (sph), negative

cylindrical (cyl) values in diopters, axis (x) in degrees, spherical equivalent (SE) in

diopters, and their respective minimum to maximum values were:

OD-sph: +0.13 D [-0.25, +1.00] / cyl: -0.75 D [-0.75, ±0.00] x 94° [89, 164] | SE: -0.19 D [-0.75, +0.63]

OS-sph: ±0.00 D [-0.25, +1.75] / cyl: -0.38 D [-1.25, ±0.00] x 65° [9, 179] | SE: ±0.00 D [-0.75, +0.63].

Figure 2: Results for induced spherical defocus results for AIM (left) and ETDRS

(right). Shown are interquartile range (boxes), medians and means (horizonal lines

and squares within each box, respectively), single outcomes (dots) jittered by a

kernel density estimate horizontally, whiskers for extreme values, and red + for

outliers. Depicted are VA results (y axis) for each effective blur level at 4m distance

(i.e., lens power minus 0.25D accommodation), using plano lens, +0.25D, +0.50D,

*+0.75D, +1.00D, +2.00D spherical lenses. *Indicated p<0.05; *** p<0.001 of planned*

comparisons between 0.00D and blur conditions.

The effect of spherical blur induced by using positive spherical lenses is shown in

Figure 2. A two-way ANOVA between test types and blur levels revealed a

significant effect of test type [$F(1,252)=24.0, p<0.001, \eta_p^2 = 0.09$] and spherical blur

[$F(5,252)=43.7, p<0.001, \eta_p^2 = 0.46$], but no interaction between these independent

variables [$F(5,252)=1.4, p>0.05, \eta_p^2 = 0.03$]. Planned multiple comparisons between

the +0.00D and all other blur conditions within each test type showed an elevated,

i.e., poorer, but non-significant acuity for +0.25D and +0.50D and a significant

reduction in acuity with +0.75D but with a stronger effect for AIM [$p<0.001$] compared

to ETDRS [$p<0.05$] and +1.75D blur for both AIM and ETDRS [$p<0.001$].

We also tested pairwise differences between test types for each blur condition using

planned multiple comparison analysis and found that all AIM VA means were

nominal higher but only the +1.00D lens (effective +0.75D blur) was significantly

different [$p < 0.05$]. The mean differences (AIM-ETDRS) were all positive

(supplementary materials, Table 1).

We also investigated whether spherical blur affects the other parameters and found

that the minimum orientation error (noise) was significantly increased

[$F_{(4.0)} = 131, p < 0.01, \eta_p^2 = 0.12$] with blur whereas the slope was unchanged

[$F_{(0.4)} = 131, p > 0.05, \eta_p^2 = 0.01$]. ROSDI (see Figure 1B) includes the distance

between 5% above an individual's minimum angular error (noise) to the threshold

and depends also on the slope of the psychometric function. Therefore, it contains all

three parameters, which may interact with each other. We investigated the effect of

spherical blur and found no significant change of ROSDI [$F_{(5,131)} = 0.4, p > 0.05, \eta_p^2 = $

0.01], (supplementary materials, Figure S4, S5, S6).

The median duration for AIM across blur conditions using two charts was 59 ±25

(standard deviation) sec. We did not collect data for time across blur steps for

ETDRS and found for AIM no significant effect of blur on experimental duration

[$F_{(1,131)} = 0.9, p > 0.05, \eta_p^2 = 0.03$] (supplementary materials, Figure S7).

Next, we investigated the effect of astigmatic blur.

Figure 3: Results for AIM (left) and ETDRS (right) VAs using cylindrical lenses.

Depicted are VA results (y axis) for each blur lens (x-axis), i.e., +0.50D (blue),

+1.00D (green), and +2.00D (red) cylindrical powers in combination with 0°, 90°,

135° induced orientations.

A 3-way ANOVA was performed and there was a significant effect of test type

[$F_{(1,382)}$=63.7,$p<0.001$, η_p^2= 0.14], blur level [$F_{(2,382)}$=72.0,$p<0.001$, η_p^2= 0.27],

and cylindrical axis orientation [$F_{(2,382)}$=8.4,$p<0.001$, η_p^2= 0.04] but no interaction

effects. Planned multiple comparison analysis between test types revealed that

many AIM VAs, including all 90° conditions, were significantly lower (numerically

higher) compared ETDRS and all mean differences (AIM-EDTRS) were positive

(supplementary material, Table 2).

A two-way ANOVA showed no significant effect of blur level [$F_{(2,189)}$ =1.9,$p>0.05$,

η_p^2= 0.02], cylindrical axis orientation [$F_{(2,189)}$=0.7,$p>0.05$, η_p^2= 0.007] nor an

interaction [$F_{(4,189)}$=0.3,$p>0.05$, η_p^2= 0.006]. (supplementary materials, Figure S8).

Noise showed a significant effect of blur level [$F_{(2,189)}$=11.3,$p<0.001$, η_p^2= 0.10]

and cylindrical axis orientation [$F_{(2,189)}$=4.1,$p<0.05$, η_p^2= 0.04], but no interaction

[$F_{(4,189)}$=0.6,$p>0.05$, η_p^2= 0.02](supplementary materials, Figure S9). No significant

effects of ROSDI for cylindrical blur condition [$F_{(2,189)}$=1.9, $p>0.05$, η_p^2= 0.02], axis

orientation [$F_{(2,189)}$=0.7, $p>0.05$, η_p^2= 0.007], or interaction between both

[F(4,189)=0.2, p>0.05, η_p^2= 0.006] were found (supplementary materials, Figure
S10).

The median duration for administering AIM using two charts and using cylindrical
blur was 46.8 sec ±17.7 (supplementary materials, Figure S11).

3.2 Analysis of orientation report error

To confirm that the model is insensitive to defocus blur, we analyzed whether
spherical lens induced blur would affect the two parameters of the sinewave model
using one-way ANOVAs and found that neither phase [F(5,114)=0.7,p>0.05, η_p^2=
0.03] nor amplitude [F(5,114)=1.8,p>0.05, η_p^2= 0.07] were significantly affected
(supplementary materials, Figures S12, S13).

We then tested whether cylindrical lenses would systematically change the
orientation report error, using a sinewave model that accounts for optical distortion of
perceptual to actual location of the gap (Example Figure 4). First, we analyzed the
phase shift using two-way ANOVAs. We found no significant effect of blur
[F(2,171)=0.2,p>0.05, η_p^2= 0.003], no effect of axis orientation [F(2,171)=0.5,p>0.05,
η_p^2= 0.006], nor an interaction [F(4,171)=0.4, p>0.05, η_p^2= 0.009] for the phase
parameter. The amplitude of the model was significantly affected by cylindrical blur
[F(2,171)=9.2,p<0.001, η_p^2= 0.08] i.e. higher blur caused larger amplitude, but not by
axis orientation [F(2,171)=0.9,p>0.05, η_p^2= 0.006], nor was an interaction found
[F(4,171)=1.4,p>0.05, η_p^2= 0.02] (supplementary materials, Figures S14, S15).

Figure 4: Example for an individual's results applying the oblique cylindrical +0.50 D
(top), +1.00D (middle), and +2.00D (bottom) blur. The orientation of each target ring
was randomized. The y axis shows orientation report error, the x axis shows the
target orientation. The orange line represents the best fitting sinewave function, the
blue dashed lines show the 95% confidence intervals, single indications are shown
as blue circles. The phase, amplitude, and goodness-of-fit R^2 are depicted above.
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493 **3.3 Experiment 2- Repeatability of AIM-VA**
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Figure 5: Bland-Altman plots for repeatability comparing AIM-VA with an initial chart
and one adaptive step (A) and ETDRS with the same single chart repeated once (E)
from the second experiment. Also shown are the control experiment results with two
(B), three (C), and four (D) adaptive steps. Each y axis shows the VA test-retest
difference in logMAR, each x-axis depicts the mean VA for test and retest VA results.
Single circles represent data from individual participants, blue horizontal lines
represent limits of agreement for test-retest VA difference, orange lines represent the
linear function fits.
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The Bland-Altman plots in Figure 5 show repeatability for AIM-VA using one adaptive
step (2 charts; 5A) and ETDRS (5E) during the second experiment and the results
for the control experiment (5B-D) with two to four adaptive steps. Test-retest VA
differences (y axis) were larger for AIM-VA than ETDRS using one adaptive step, but
not for two or more adaptive steps. AIM's mean VA test-retest bias was -0.03 using
one adaptive step and +0.04 for ETDRS. The relative and absolute numbers of test-
retest difference for AIM-VA were ≤0.10logMAR were 68% (15/22) and ≤0.20logMAR
86% (19/22) and for ETDRS ≤0.10logMAR were 86% (19/22) and ≤0.20logMAR

100% (22/22). Limits of agreement in logMAR for test-retest difference were AIM-VA

[-0.36 0.41] and for ETDRS [-0.11 0.18].

As the variance of an AIM fit depends on the amount of data, its variance should

decrease with increasing trials. We ran a control experiment with four adaptive steps

(5 charts in total in a subset of twelve participants to test this hypothesis. As shown

in Figure 5B-D, the distance between upper and lower limits of agreement in the

Bland Altman plots significantly decreased with step sizes [t(4) =5.2, p<0.01] and

were comparable to that of ETDRS. We calculated whether and if so, how much the

test-retest difference improves across trials in absolute and relative numbers as well

as by determining the upper and lower limits of agreement (Table 1).

Table 1. Comparison of absolute and relative (%) test-retest differences across

levels of difference in logMAR as a function of number of adaptive steps and

number of charts during the control experiment. Bias is the mean of the

difference between test and retest results.

AIM-VA Chart/ Adaptive steps	Bias	≤ ±0.10 logMAR [% | n]	≤ ±0.20 logMAR [% | n]	≤ ±0.30 logMAR [% | n]	≤ ±0.40 logMAR [% | n]	Limits of agreement [logMAR]
1/0	0.039	25% | 3/12	58% | 7/12	92% |11/12	92% |11/12	0.44 / -0.37
2/1	0.007	42%| 5/12	92% |11/12	100% |12/12	100% |12/12	0.30 / -0.28
3/2	0.002	50% |6/12	92% |11/12	100% |12/12	100% |12/12	0.24 / -0.23
4/3	0.015	75% |9/12	100% |12/12	100% |12/12	100% |12/12	0.18 / -0.15
5/4	0.023	75% |9/12	100% |12/12	100% |12/12	100% |12/12	0.16 / -0.12

4. Discussion

4.1 AIM's ability to detect induced defocus and astigmatism

One of the objectives of the current study was to test AIM-VA's ability to detect

refractive errors and compare that ability with a standard letter chart. We show that

AIM detects defocus blur and that it is as sensitive to detect blur as the ETDRS chart

(Figure 2). A previous study tested the effect of spherical and astigmatic blur induced

with lenses on E and C using +1.50D (36), which is comparable with the loss of

vision shown in our study. We also show that oblique induced astigmatism has a

stronger blur effect than with or against the rule astigmatism (Figure 3), in good

agreement with a previous study (37). Interestingly, when comparing AIM and

ETDRS in regards of detecting astigmatic blur, AIM showed larger loss of VA.

Different optotypes vary in their spatial frequency contents (38, 39). letters other than

uniform stimuli like the C or O varying more with respect to orientation-specific

spatial frequency content. The combination of the continuous error report and the

more homogenous spatial frequency properties of the C used for AIM maybe more

suitable to detect astigmatic blur compared to optotypes in the EDTRS chart that are

more heterogenous regarding orientation and spatial frequency properties. AIM

showed overall more intra-individual variability for both defocus and astigmatism

than ETDRS, which may be due to the larger degrees of freedom due the continuous

indication task and 360-degree randomization of the C gaps compared to the five-

letters-per-line letter identification task.

Clinical standard chart tests deliver threshold values that are known as VAs.

Psychophysically, VA is one point on a psychometric function that describes the

probability that an observer can detect a signal, in this study gap orientation or letter

identity. The stimulus range over which an observer's performance transitions from

detecting to not being able to detect the signal is expressed by the slope of the

psychometric function and may vary between observers and may provide additional

important information about the health of the participant, the progression of eye

disease or their response to treatment. Practically speaking, the slope describes the

rate of transition from seeing to not seeing and can be either very steep or shallow.

Chart tests, such as the ETDRS and Snellen charts, do not take this parameter into

consideration whereas AIM-VA calculates a personalized slope value and generates

a personalized VA using a psychometric function fitting approach (Figure 1C).

Moreover, even though two observers may be able to accurately report a signal in a

forced choice paradigm, the quality of their vision may still be distinct between

observers or across time. For example, in terms of a conventional letter chart, one

observer may correctly name all five letters in all lines before not being able to detect

any single letter whereas a second observer may reach the same VA but make

mistakes in some of the lines with larger stimuli. Clinically those VAs are identical but

the suprathreshold performance differed. Unlike letter charts such as ETDRS, AIM-

VA provides an output for that observer-specific behavior, namely the minimum

angular error parameter for the most intense (here: largest) stimuli. A number of

sources can affect this parameter e.g., hand tremor, eye movements, or neurological

intrinsic variations (spiking variations, neuro-transmitter variation etc.) and structural

irregularities in the optics or retina. We termed these sources as *noise* that captures

all sources of those external and internal variability. We show that this noise

parameter significantly increased with increasing defocus and astigmatic blur.

Previous research demonstrated that the slope of the psychometric function is

affected by blur (29), specifically that the slope becomes more shallow with one

diopter defocus blur. Since threshold, slope, and noise interact, it remains unclear

what causes a change in an observer's psychophysical performance. Hence, we

introduce the *range of stimulus detectability improvement*, ROSDI (Figure 1B), which

takes all three parameters into account, ranging from the midpoint between random

performance and plateauing visual performance (noise) to 5% above an individual's

noise value. ROSDI is expressed here in logMAR and can be converted to all other

known VA units as well as across other signal intensity parameters. We found no

systematic effect for ROSDI due to induced refractive error in this sample of healthy

young observers, however, future research may consider ROSDI in pathological eye

diseases.

Astigmatism is a prevalent condition worldwide, ranging in prevalence between 8 to

62% (40) and may induce perceptual distortions that affect performance during an

acuity tasks. It has been shown that induced astigmatism generates neural response

biases (41) and that observers may show intrinsic bias preferences (41, 42). We

found that the induced error using astigmatic blur caused a significant effect of

amplitude, i.e., increasing blur increased the amplitude, on the sinusoidal pattern of

response errors, suggesting that the report error was systematically affected by

astigmatism (Figure 4, see also Supplementary Materials, Figure S15). Future

studies may use responses in a separate domain (e.g., verbal or manual) to estimate

the relationship between report error and amplitude and phase under astigmatism.

4.2 AIM test-retest results

In experiment two, we compared AIM-VA with ETDRS using a single chart as often

deployed in clinical practice. We show that AIM-VA using two charts (i.e., the initial

chart and one adaptive step) has greater test-retest variance (Figure 5A and 5E).

This may be due to three reasons, namely a memory bias for a single ETDRS chart,

fewer decision options for ETDRS (10 letters) compared with AIM (continuous 360°),

and AIM-VA's sampling of data along the slope from random report error (90°) to the

min. angular error. To address the latter point, we carried out a control experiment

whereby we increased the number of adaptive steps to sample more data along the

psychometric function to improve the fit. We hypothesized that within the time

window of the test to retest (within one day), the slope, noise, and threshold of an

observer should not significantly change. Indeed, increasing the number of charts

improved the repeatability of AIM-VA significantly and was comparable with ETDRS

using three or more adaptive steps charts with 4*4 cells (Figure 5B-D). The

repeatability of ETDRS has been reported and 87% of the standard ETDRS chart

generated VAs had a ≤0.10logMAR difference and 100% with a ≤0.30logMAR

difference (43). Other tests such as FrACT reported comparable limits of agreement

of approximately 0.20logMAR with a few outliers (44). AIM-VA showed comparable

results when using two or more adaptive steps (Table 1).

4.3 Future directions and study limitations

We acknowledge a number of limitations of the current study. We measured induced

refractive error in a relatively small cohort to proof-of-concept. Future studies may

explore the feasibility in a larger cohort, with and without visual impairments.

In an ongoing study concerning VA in amblyopic children, preliminary results suggest

that a near version of AIM-VA may be suitable to investigate vision with reliability

(45). The positions of AIM-VA's log-spaced stimuli sizes were randomized within

each chart whereas ETDRS used log-spaced stimuli sizes that were organized from

large to small down the chart. Future studies will examine whether an observer's

performance is significantly influenced by the randomization of the positions of letter

sizes within each chart. We introduced a number of AIM-specific outcome measures

including noise and ROSDI and show that noise was significantly increased with blur

levels whereas no effect for ROSDI was found. Building a personalized model and

using ROSDI longitudinal within observers is a future direction to investigate whether

only single aspects, i.e., threshold, slope, or min. error or the entire model of the

detectable range of visible stimuli may be affected, e.g., during amblyopia treatment.

With AIM, the number of cells per chart and the number of adaptive steps can be

independently varied. It is unclear how the repeatability changes by lowering the

number of cells-per-charts while increasing the number of adaptive steps. Future

research will investigate this question. In the current study, we applied AIM to

measure VA, however, due to its generalizability, future studies will exchange

stimulus size with other stimulus properties that are processed in distinct neural

sites, e.g., color, contrast, stereo disparity, motion etc.

5. Conclusions

The Angular Indication Measurement, or AIM, is a response-adaptive, self-

administered, and generalizable technique to interrogate visual functions. The

current study applied AIM to measure the visual acuities in healthy adults and

compared it with a standard clinical test in terms of its ability to detect induced

refractive errors. We show that AIM showed comparable sensitivity to a conventional

ETDRS letter chart. Using one adaptive step, AIM showed greater test-retest

variance than a single ETDRS chart, which is likely due to memory bias and fewer

degrees of variability for the ETDRS (10 letters) chart compared to AIM (continuous

360°). As hypothesized, increasing the number of adaptive steps improves the

repeatability to the level of a single ETDRS chart. Based on the findings of the

current study, at least 2 adaptive steps should be used to measure distance VA in

adults reliably. An AIM chart took approximately 30sec to administer. As it provides

additional information about an observer's principal error level (noise), error bias,

slope that considers an observer's personalized psychometric function, and the

range of stimulus detectability improvement, or ROSDI, in clinically useful VA units,

e.g., logMAR, AIM may be a useful tool for researchers and clinicians to measure

visual acuity.

Acknowledgment
668
669 Supported by NIH R01 EY029713. Portions of this study have been presented at the virtual Vision
670 Science Society Conference 2022 and Vision Science Society conference 2023.
671

672 **Disclosures**

673 AIM, including AIM Visual Acuity, is disclosed as provisional patented and held by Northeastern
674 University, Boston, USA. JS and PJB are founders and shareholders of PerZeption Inc. JH, JBS, NA,
675 and MF declare that they have no conflict of interest.
676

677 **Data availability**

678 Data underlying the results presented in this paper are not publicly available at this time but may be
679 obtained from the authors upon reasonable request.
680

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