doi.org/10.26398/IJAS.0032-015

# THE VISUAL ACUITY MEASUREMENT AND ITS MULTIPLE TESTING FOR PATIENTS IMPLANTED WITH INTRAOCULAR LENSES DURING THE CATARACT SURGERY

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Abstract: In the present article we propose a statistical approach on the visual acuity measurement and its multiple testing for patients implanted with intraocular lenses (IOLs) during the cataract surgery. Main themes of our research are (i) obtaining psychometric functions, psychophysical thresholds and their variances of individual patients, (ii) testing the difference of visual acuities between monofocal IOL group and multifocal IOL group by adopting the Cochran-Mantel-Haenszel (CMH) test statistic (based on psychometric functions), (iii) multiple comparisons of visual acuities for individual patients by adopting the delta test statistic (based on psychophysical thresholds and their variances) and the false discovery rate (FDR-BH: Benjamini and Hochberg (1995)). Our 5-meter visual acuity test based on psychometric functions shows that the difference between the monofocal IOL group (Tecnis® 1-piece ZCB00) and the multifocal IOL group (Tecnis® Multifocal +4.0 ZMB00) is not significant under the 5% significance level (p-value = 0.2801 by CMH test statistic). We also see that FDR-BH procedure produces high average statistical power in the case of large number of null hypotheses (FDR-BH: 0.604, Holm: 0.392, Bonferroni: 0.370 for the number of null hypotheses = 210). It is very effective on the problem of multiple comparisons of individual visual acuities.

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*Keywords:* Visual acuity measurement, intraocular lenses (IOLs), monofocal IOL, multifocal IOL, false discovery rate (FDR).

#### **1. INTRODUCTION**

In cataract surgery, the intraocular lenses (IOLs) are implanted inside the eyes to replace the eyes' natural lenses. Monofocal IOL is designed to provide clear vision at a single focal point. Multifocal IOL, however, can help seeing things at different distances. It is important to obtain visual acuities of individual patients for deciding ophthalmological treatments after surgery. Yamauchi et al. (2013), Gundersen and Potvin (2013), Cardona et al. (2018) studied the comparative visual performance with monofocal and multifocal IOLs. In Yamauchi et al. (2013) the visual acuity was measured by using the decimal visual acuity chart (not based on psychometric functions) and the measured decimal values were converted to the logarithm of the minimum angle of resolution (LogMAR) scale.

Main themes of our research are (i) obtaining psychometric functions, psychophysical thresholds and their variances of individual patients, (ii) testing the difference of visual acuities between monofocal IOL group and multifocal IOL group by adopting the Cochran-Mantel-Haenszel (CMH) test statistic (based on psychometric functions), (iii) multiple comparisons of visual acuities for individual patients by adopting the delta test statistic (based on psychophysical thresholds and their variances) and the false discovery rate (FDR-BH: Benjamini and Hochberg (1995)).

The automated measurements of the visual acuity were proposed by Bach (1996); Beck et al. (2003); Schulze-Bonsel et al. (2006). However, their methods cannot calculate the variance of the visual acuity in one measurement. The testing of thresholds estimated by constant stimuli method was developed by Nagai et al. (2006), and the measurement of visual acuities and variances for individual patients by using psychometric functions was proposed by Mita et al. (2010). It has been reported that the psychometric function method has higher repeatability than the conventional methods (Tokutake et al., 2014). Therefore the new method can catch very few changes of visual acuities and variances which cannot been found by the conventional measurements.

For multiple comparison problems, several statistical approaches were proposed. Ryan, Einot-Gabriel and Welsch proposed some stepwise procedures for multiple comparison problems (Einot and Gabriel (1975); Ryan (1960); Welsch (1977)). Félix and Menezes (2018) reported comparisons of ten corrections meth-

ods for t-test in multiple comparisons via Monte Carlo study. By adopting the bootstrap logistic regression, Mita et al. (2014) developed an algorithm of multiple comparisons with a control. Mita et al. (2017, 2020) proposed an algorithm of closed multiple test based on the step-down procedure with Ryan-Einot-Gabriel-Welsch (REGW) significance levels. The closed multiple testing allows the rejection of any elementary null hypothesis ( $H_{ij}^0$ ) if all possible intersections of null hypotheses involving  $H_{ij}^0$  can be rejected using valid local level test. It controls the familywise error rate (FWER) for all the hypotheses in the strong sense. The number of elements in the hierarchical family of null hypotheses for closed multiple testing is given by  $M = \sum_{k=2}^{N} NC_k$ , where N is the number of all patients,  $NC_k$  is the notation of combination, and thus M becomes very large number as N increases.

Although the procedures of controlling FWER have been commonly used in multiple testing, Benjamini and Hochberg (1995) proposed a new point of view on the problem of multiplicity: the procedures of controlling the false discovery rate (FDR-BH) which is the expected proportion of errors among the rejected null hypotheses. In many applications the control of the FDR is the desirable control against errors originating from multiplicity. The power of the FDR controling method is uniformly larger than that of the other methods, and the loss of powers, as the number of null hypotheses increases, is relatively small for the FDR controlling method (Benjamini and Hochberg (1995)). The null hypotheses in FDR testings are  $H_{ij}^0$  ( $1 \le i < j \le N$ ). Therefore the number of FDR testings is given by  $M = {}_NC_2$ , and it is much smaller than that of closed multiple testing. Further Benjamini et al. (2006) studied two-stage procedure (FDR-TST). We applied the basic FDR-BH procedure to our visual acuity problem of multiple comparisons for many individual patients implanted with IOLs.

In the present article, (i) psychometric functions, psychophysical thresholds and their variances are obtained by adopting a guassing rate (Section 2), (ii) two different types of test statistics are computed based on psychometric functions and psychophysical thresholds (Section 3), (iii) multiple testings are performed by adopting FDR-BH procedure and also by adopting FWER procedures (Holm, Bonferroni) for reference (Sections 4, 5).

### 2. MODELLING OF THE VISUAL ACUITY FOR INDIVIDUAL PATIENT

We consider the logistic regression with a constant guessing rate (Mita et al. (2010, 2017); Tokutake et al. (2014)).

Let  $\pi(x_t; \alpha, \beta)$  be the probability that a patient answers the visual acuity test

 $t (t = 1, \dots, T)$  correctly:

$$\pi(x_t; \boldsymbol{\alpha}, \boldsymbol{\beta}) = (1 - c_0) \cdot \frac{\exp(\boldsymbol{\alpha} + \boldsymbol{\beta} x_t)}{1 + \exp(\boldsymbol{\alpha} + \boldsymbol{\beta} x_t)} + c_0 \quad (t = 1, \cdots, T),$$
(1)

where  $x_t$  is the visual target of test t (in our measurement,  $x_t$  is the logarithmic visual acuity: LogVA),  $\alpha$  and  $\beta$  are intercept and slope parameters, respectively, and  $c_0$  ( $0 \le c_0 < 1$ ) is a guassing rate constant which depends on the method of measurement.

We obtain optimum parameters  $\hat{\alpha}$  and  $\hat{\beta}$  for  $\alpha$  and  $\beta$ , respectively, by adopting the classical maximum likelihood estimates based on Fisher score method. Then we can obtain the psychometric function  $\varphi(x)$  which is a mathematical function from a stimulus level *x* to a response level  $\pi$  such that:

$$\varphi(x) = \pi(x; \hat{\alpha}, \hat{\beta}) \quad (-\infty < x < +\infty).$$
(2)

The psychophysical threshold  $\xi$  is the stimulus intensity that is detected exactly at the central value of  $\varphi(x)$  ( $-\infty < x < +\infty$ ). We see that

$$\lim_{x\to-\infty}\varphi(x)=1 \text{ and } \lim_{x\to+\infty}\varphi(x)=c_0.$$

Thus we define the psychophysical threshold  $\xi$  of the patient with guessing rate  $c_0$  such that

$$\xi = \varphi^{-1} \left( \frac{1+c_0}{2} \right). \tag{3}$$

The threshold  $\xi$  and its variance var( $\xi$ ) are obtained by

$$\xi = -\frac{\hat{\alpha}}{\hat{\beta}}, \quad \operatorname{var}(\xi) = \xi^2 \left( \frac{\operatorname{var}(\hat{\alpha})}{\hat{\alpha}^2} - 2 \frac{\operatorname{cov}(\hat{\alpha}, \hat{\beta})}{\hat{\alpha}\hat{\beta}} + \frac{\operatorname{var}(\hat{\beta})}{\hat{\beta}^2} \right), \tag{4}$$

where  $var(\hat{\alpha})$  and  $var(\hat{\beta})$  are variances of  $\hat{\alpha}$  and  $\hat{\beta}$ , respectively, and  $cov(\hat{\alpha}, \hat{\beta})$  is the covariance of  $\hat{\alpha}$  and  $\hat{\beta}$ .

# 3. TEST STATISTICS

#### 3.1. TEST STATISTIC BASED ON PSYCHOMETRIC FUNCTIONS

For the test of patient groups, we can use the Cochran-Mantel-Haenszel (CMH) test statistic based on psychometric functions (Landis et al. (1978); Penfield (2001); Somes (1986)).

Let r  $(r = 1, \dots, R)$  be strata of observations and let  $u_{g\ell}^r$   $(r = 1, \dots, R; g = 1, \dots, G; \ell = 1, \dots, L)$  be the number of observations at *r*-th stratum of the  $G \times L$  contingency table. For applying the CMH test statistic to the visual acuity results, we choose indices  $r, g, \ell$  in the contingency table such that

(i) strata  $r(r = 1, \dots, R)$ : visual acuity strata  $(x^1, \dots, x^R)$ ,

(ii) rows  $g(g = 1, \dots, G)$ : patient groups  $(\Gamma_1, \dots, \Gamma_G)$ ,

(iii) columns  $\ell$  ( $\ell = 1, \dots, L$ ): responses of visual tests ( $\ell = 1$  for "correct" and  $\ell = 2$  for "incorrect").

Then we fix the number of columns at L = 2 and choose the number of observations  $u_{g\ell}^r$   $(r = 1, \dots, R; g = 1, \dots, G; \ell = 1, 2)$  such that

$$u_{g1}^{r} = \sum_{i \in \Gamma_{g}} \varphi_{i}(x^{r}), \quad u_{g2}^{r} = N_{g}^{r} - u_{g1}^{r} \quad (r = 1, \cdots, R; g = 1, \cdots, G),$$
(5)

where  $N_g^r$  is the number of patients in group  $\Gamma_g$ , and  $\varphi_i(x^r)$  is the value of psychometric function of patient *i* at the visual acuity stratum  $x^r$ .

The null hypothesis is described such that there is no association between the row (patient group) and the column (response of visual test).

The Cochran-Mantel-Haenszel (CMH) test statistic  $\Psi$  for the set of G groups is given by Landis et al. (1978); Mita et al. (2020),

$$\Psi = \boldsymbol{u}^T \, \boldsymbol{W}^{-1} \boldsymbol{u},\tag{6}$$

where u is the vector of the number of observations and W is the variancecovariance matrix which can be determined by using the number of observations themselves (without using variances). This CMH test statistic  $\Psi$  has an asymptotic  $\chi^2$ -distribution with (G-1) degrees of freedom under the null hypothesis (Agresti, 2002).

### 3.2. TEST STATISTIC BASED ON PSYCHOPHYSICAL THRESHOLDS

We apply the delta test statistic to the problem of comparisons of visual acuities of individual patients (Mita et al., 2017; Nagai et al., 2006).

We shall test the null hypothesis for the thresholds of *n* patients:

$$H_{1...n}^0: \quad \xi_1 = \dots = \xi_n \quad (2 \le n \le N),$$
 (7)

where N is the number of all patients, n is the number of patients which are considered in the null hypotheses. If we define the following notations:

$$u_j = \xi_1 - \xi_{j+1} \quad (j = 1, \cdots, n-1),$$
(8)

then the null hypothesis can be rewritten such that  $u_1 = \cdots = u_{n-1} = 0$ .

Since the vector  $u = (u_j)^T$  has the normal distribution in multiple variables in our visual acuity testing, the delta test statistic  $\Delta$  for the set of *n* patients defined by

$$\Delta = \boldsymbol{u}^T \boldsymbol{W}^{-1} \boldsymbol{u} \tag{9}$$

has an asymptotic  $\chi^2$ -distribution with (n-1) degrees of freedom under the null hypothesis, where W is the variance-covariance matrix which is determined by using the variances of thresholds var $(\xi)$ .

In the cases of FDR-BH, Holm and Bonferroni procedures (refer to Sections 5.2, 5.3), we test the following null hypotheses for all pairs of patients:

$$H_{ij}^{0}: \quad \xi_{i} = \xi_{j} \quad (1 \le i < j \le N).$$
(10)

In these cases, the number of patients in considered is n = 2, and the delta test statistic  $\Delta_{ij}$  is reduced to the following simple form:

$$\Delta_{ij} = \frac{(\xi_i - \xi_j)^2}{\operatorname{var}(\xi_i) + \operatorname{var}(\xi_j)}.$$
(11)

### 4. FALSE DISCOVERY RATE

Let *Q* be the proportion of the rejected null hypotheses which are erroneously rejected:

$$Q = \frac{V}{V+S},\tag{12}$$

where V is the number of true null hypotheses which are declared significant, and S is the number of non-true null hypotheses which are declared significant. Benjamini and Hochberg (1995) defined the false discovery rate (FDR) E(Q) to be the expectation of Q.

The false discovery rate controlling procedure (FDR-BH) at  $q^*$  ( $q^*$  is usually chosen as  $q^* = 0.05$ ) is stated as follows.

Consider testing null hypotheses  $H_k^0$  ( $k = 1, \dots, M$ ) based on the corresponding *p*-values  $p_k$  ( $k = 1, \dots, M$ ), where *p*-values are assumed ordered such that  $p_1 \leq \dots \leq p_M$ . Define the following multiple testing procedure: let *K* be the largest *k* satisfying

$$p_k \le \frac{k}{M} q^* \equiv q_k; \tag{13}$$

then reject all  $H_k^0$  ( $k = 1, \dots, K$ ).

#### 5. APPLICATION TO THE VISUAL ACUITY TESTING

#### 5.1. 2-GROUP TEST BETWEEN MONOFOCAL IOL AND MULTIFOCAL IOL

Since we adopt the Landolt-C of four different orientations in our measurement, the guessing rate  $c_0$  is chosen as  $c_0 = 0.25$ , where the Landolt-C consists of a ring that has a gap. The gap can be at various positions; usually left, right, bottom and top. The explanatory variable *x* in our measurement is the logarithmic visual acuity (LogVA). We shall test the difference of 5-meter visual acuity between 2 groups:

 $\Gamma_1$ : patients implanted with monofocal IOL (Tecnis® 1-piece ZCB00),

 $\Gamma_2$ : patients implanted with multifocal IOL (Tecnis® Multifocal +4.0 ZMB00), where Tecnis® is a trade mark of Precisionlens Company, USA. Data are taken from patients underwent cataract surgery at the Cataract Unit of the Department of Ophthalmology, Tsukazaki Hospital, Himeji, Japan.

The number of Landolt-C targets  $T_g$  (g = 1, 2) and the number of patients  $N_g$  in groups  $\Gamma_g$  are chosen such that

$$T_1 = 7, T_2 = 7; N_1 = 21, N_2 = 12.$$

For adopting the constant stimuli method, we choose the number of trials  $v_{it}$  for patient *i* to the test *t* satisfying such that

$$v_{it} = 0$$
 or 20  $(i = 1, \dots, N_g; t = 1, \dots, T_g), \quad \sum_{t=1}^{T_g} v_{it} = 120 \quad (i = 1, \dots, N_g).$ 

For each patient i  $(i = 1, \dots, N_g)$  in group  $\Gamma_g$  (g = 1, 2), we obtain optimum values of parameters  $\hat{\alpha}_i$ ,  $\hat{\beta}_i$ , and psychophysical thresholds  $\xi_i$  and their variances var $(\xi_i)$ by adopting the Fisher score method. The psychometric functions  $\varphi_i(x)$   $(i = 1, \dots, N_g)$  for individual patients in groups  $\Gamma_g$  (g = 1, 2) are shown in Figures 1  $(\Gamma_1)$ , 2  $(\Gamma_2)$ .

The mean values of logistic regression results for each group  $\Gamma_g$  (g = 1,2) are shown in Table 1 (intercepts  $\overline{\alpha}_g$ , slopes  $\overline{\beta}_g$  and psychophysical thresholds  $\overline{\xi}_g$ ).

The 2-group test of difference between  $\Gamma_1$  and  $\Gamma_2$  is performed by adopting Cochran-Mantel-Haenszel (CMH) test statistic. The contingency table for computation of CMH test statistic is constructed by choosing strata  $x^r = -0.2 + 0.05(r - 1)$ ;  $(r = 1, \dots, 9)$ . The contingency table is shown in Table 2, where data are presented only for strata r = 1, 5, 9. The results of 2-group test are shown in Table 3.

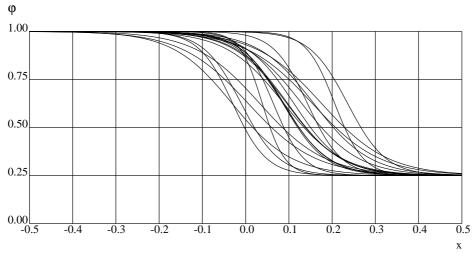
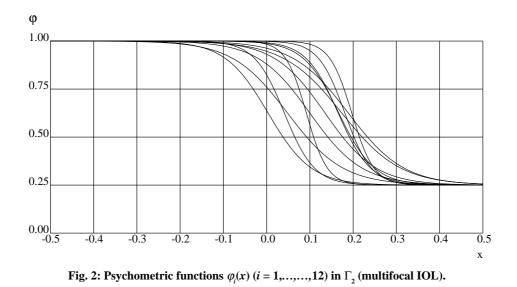


Fig. 1: Psychometric functions  $\varphi_i(x)$  (i = 1,...,21) in  $\Gamma_1$  (monofocal IOL).



Tab. 1: Mean values of logistic regression results						
	patients	intercepts	slopes	thresholds	variances	SD
groups	$N_g$	$\overline{\alpha}_{g}$	$\overline{\beta}_{g}$	$\overline{\xi}_{g}$	$\overline{\mathrm{var}}(\xi_g)$	$\overline{\mathrm{sd}}(\xi_g)$
$g = 1(\Gamma_1)$	21	1.80192	-19.7598	0.091760	0.000391788	0.0197936
$g = 2(\Gamma_2)$	12	3.10836	-24.3544	0.124381	0.000285412	0.0168941

		$\ell = 1$ (correct)	$\ell = 2$ (incorrect)	
strata	groups	$u_{g1}^r$	$u_{g2}^r$	total
$x^1 = -0.2$	$g = 1(\Gamma_1)$	20.7833	0.2167	$n_1^1 = 21.000$
	$g = 2(\Gamma_2)$	11.9592	0.0408	$n_2^1 = 12.000$
	total	$m_1^1 = 32.7426$	$m_2^1 = 0.2574$	$t^1 = 33.000$
$x^5 = 0.0$	$g = 1(\Gamma_1)$	17.5265	3.4735	$n_1^5 = 21.000$
	$g = 2(\Gamma_2)$	10.9021	1.0979	$n_2^5 = 12.000$
	total	$m_1^5 = 28.4286$	$m_2^5 = 4.5714$	$t^5 = 33.000$
$x^9 = 0.2$	$g = 1(\Gamma_1)$	8.2394	12.7606	$n_1^9 = 21.000$
	$g = 2(\Gamma_2)$	5.1059	6.8941	$n_2^9 = 12.000$
	total	$m_1^9 = 13.3453$	$m_2^9 = 19.6547$	$t^9 = 33.000$

Tab. 2: The  $2 \times 2$  contingency table based on psychometric functions

Tab. 3: Two-group test ( $\Gamma_1$ : monofocal IOL and  $\Gamma_2$ : multifocal IOL)

	monofocal	multifocal	strata	dof	CMH statistic	<i>p</i> -value
	$N_1$	$N_2$	R		Ψ	
$H_{12}^{0}$	21	12	9	1	1.1665	0.2801

### 5.2. MULTIPLE TEST OF INDIVIDUAL PATIENTS BY FDR-BH PROCEDURE

Hereafter to save the pages of the article, we show only the results of  $\Gamma_1$  (monofocal IOL): number of patients N = 21, number of null hypotheses  $M = {}_NC_2 = 210$ .

Let  $\xi_i$   $(i = 1, \dots, N)$  be psychophysical thresholds of patients *i* in the group  $\Gamma_1$ . We assume that

$$\xi_1 \leq \xi_2 \leq \cdots \leq \xi_N.$$

We shall test the following null hypotheses for all pair of patients:

null hypotheses for the pair of patients  $(i, j) H_{ij}^0$ :  $\xi_i = \xi_j \quad (1 \le i < j \le N).$ 

Let  $p_{ij}$   $(1 \le i < j \le N)$  be *p*-values obtained by delta test statistic for the null hypotheses  $H_{ij}^0$ . The *p*-values  $p_{ij}$  are shown in Figure 3, where *p*-values are presented by adopting the following symbols:

$$[*] \le 10^{-5} < [5] \le 10^{-4} < [4] \le 10^{-3} < [3] \le 10^{-2} < [2] \le 0.03 < [1] \le 0.05 < [0].$$

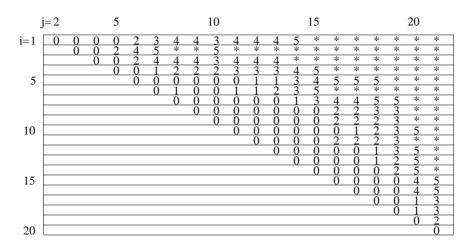


Fig. 3: The *p*-values  $p_{ij}$  (1  $\leq i < j \leq 21$ ) in  $\Gamma_1$  (monofocal IOL). Symbols: [\*]  $\leq 10^{-5} < [5] \leq 10^{-4} < [4] \leq 10^{-3} < [3] \leq 10^{-2} < [2] \leq 0.03 < [1] \leq 0.05 < [0]$ 

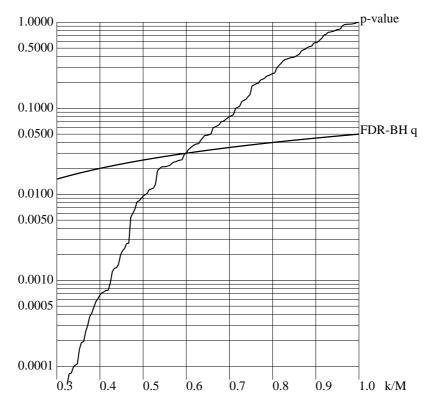


Fig. 4: FDR-BH  $q_k$  and p-values  $p_k$  (k = 1, ..., M) in  $\Gamma_1$  (monofocal IOL).

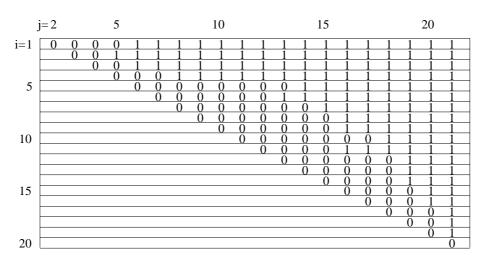


Fig. 5: The significance results of FDR-BH multiple test for null hypotheses  $H_{ij}^{0}$  ( $1 \le i < j \le 21$ ) in  $\Gamma_{1}$  (monofocal IOL). Symbols: [0] means not-significant, [1] means significant.

Let *M* be the number of null hypotheses (the number of patient pairs (i, j)  $(1 \le i < j \le N)$ ) in the group  $\Gamma_1$ , and let  $p_k$   $(k = 1, \dots, M)$  be the ordered *p*-values of  $p_{ij}$  such that

$$p_1 \leq p_2 \leq \cdots \leq p_M.$$

FDR-BH  $q_k$  ( $k = 1, \dots, M$ ) and p-values  $p_k$  are shown in Figure 4 ( $\Gamma_1$ ), where we adopt FDR at  $q^* = 0.05$ . The significance results of multiple test for null hypotheses  $H_{ij}^0$  ( $1 \le i < j \le N$ ) are shown in Figure 5, where [0] means not-significant and [1] means significant.

## 5.3. MULTIPLE TEST OF INDIVIDUAL PATIENTS BY HOLM AND BONFER-RONI PROCEDURES (FWER)

In the Holm procedure (Holm , 1979), the significance level at kth step is chosen such that

$$\alpha_k = \frac{\alpha_0}{M - k + 1} \qquad (k = 1, \cdots, M), \tag{14}$$

where  $\alpha_0$  is the type I familywise error rate (FWER).

In the Bonferroni procedure, the significance level at kth step is chosen such that

$$\alpha_k = \frac{\alpha_0}{M} \qquad (k = 1, \cdots, M), \tag{15}$$

where  $\alpha_0$  is the type I familywise error rate (FWER).

Holm and Bonferroni  $\alpha_k$  ( $k = 1, \dots, M$ ) and *p*-values  $p_k$  are shown in Figures 6 (Holm;  $\Gamma_1$ ), 7 (Bonferroni;  $\Gamma_1$ ), where we adopt FWER at  $\alpha_0 = 0.05$ .

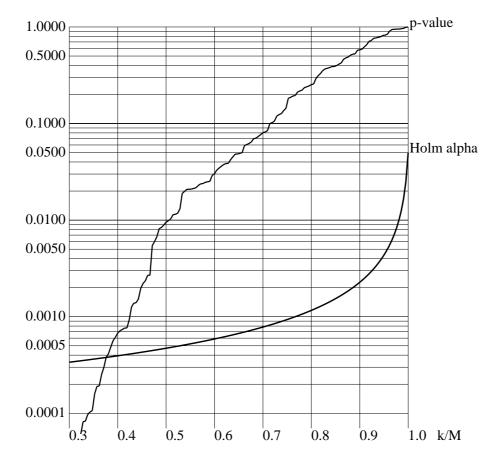


Fig. 6: Holm  $\alpha_k$  and *p*-values  $p_k$  (k = 1, ..., M) in  $\Gamma_1$  (monofocal IOL).

# 5.4. STATISTICAL POWERS

Let  $\chi_k^2$  be the critical value for  $q_k$  (or  $\alpha_k$  in the case of FWER) such that

$$\Pr[X^2 > \chi_k^2] = q_k \quad (k = 1, \cdots, M), \tag{16}$$

where  $\Pr[X^2 > \chi_k^2]$  is the upper probability of the chi-squared distribution  $X^2$  with 1 degree of freedom. Then, the statistical power  $\eta_k$  ( $k = 1, \dots, M$ ) can be obtained by

$$\eta_k = \Pr[\tilde{X}^2(\theta) > \chi_k^2] \quad (k = 1, \cdots, M), \tag{17}$$

where  $\Pr[\tilde{X}^2(\theta) > \chi_k^2]$  is the upper probability of the non-central chi-squared distribution  $\tilde{X}^2(\theta)$  with 1 degree of freedom, and the non-centrality parameter  $\theta$  is chosen as  $\theta = \Delta_k$  ( $\Delta_k$  is the delta test statistic which is referred to *p*-value  $p_k$ ).

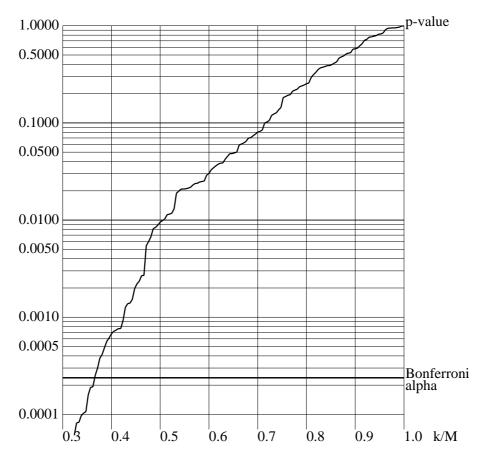


Fig. 7: Bonferroni  $\alpha_k$  and *p*-values  $p_k$  (k = 1,...,M) in  $\Gamma_1$  (monofocal IOL).

The average statistical power  $\overline{\eta}$  is given by

$$\overline{\eta} = \frac{1}{M} \sum_{k=1}^{M} \eta_k.$$
(18)

• ···

Table 4 shows the average statistical powers of multiple tests by FDR-BH, Holm and Bonferroni procedures.

	hypotheses	FDR-BH	Holm	Bonferroni
groups	M	$\overline{\eta} \; (q*=0.05)$	$\overline{\eta} (\alpha_0 = 0.05)$	$\overline{\eta} (\alpha_0 = 0.05)$
$\Gamma_1$	$_{21}C_2 = 210$	0.604	0.392	0.370
$\Gamma_2$	$_{12}C_2 = 66$	0.599	0.461	0.434

### 6. CONCLUSION

Novelty of the method in the context of visual acuity: One of the advantages of our method based on psychometric functions is that we can obtain not only the visual acuities (= psychophysical thresholds) but also their variances of individual patients. Thus we can assess changes in the visual acuity in an individual. Another characteristic of our method is that we adopt the delta test statistic (based on psychophysical thresholds and their variances) and the FDR-BH procedure for multiple testing in visual acuity problem. Therefore we can obtain the multiple testing for large number of patients implanted with intraocular lenses during cataract surgeries. For the test of patient groups, we can use the Cochran-Mantel-Haenszel (CMH) test statistic (based on psychometric functions) which requires only the numbers of observations (does not require variances). As we adopt two test statistics (delta and CMH) in our system, we can compute wide range of test problems in visual acuity field.

Monofocal IOL and multifocal IOL: Our 5-meter visual acuity test based on psychometric functions shows that the difference between the monofocal IOL group (Tecnis® 1-piece ZCB00) and the multifocal IOL group (Tecnis® Multifocal +4.0 ZMB00) is not significant under the 5% significance level (*p*-value = 0.2801 by CMH statistic  $\chi^2$ -test). Yamauchi et al. (2013) also measured visual acuity by using the decimal visual acuity chart (not based on psychometric functions) and the measured decimal values were converted to the logarithm of the minimum angle of resolution (LogMAR) scale. Yamauchi's result of uncorrected 5.0 meter visual acuity (UDVA) shows that the difference between monofocal and multifocal IOLs is not significant under the 5% significance level (*p*-value = 0.4976 by Student's *t*-test).

Multiple comparisons of the visual acuities of individual patients: The result of multiple testing is valuable for doctors to make a suitable plan of ophthalmological treatment for each patient after the cataract surgery. We see that FDR-BH procedure produces high average statistical power in the case of large number of null hypotheses (FDR-BH: 0.604, Holm: 0.392, Bonferroni: 0.370 for the number of null hypotheses = 210 in patients with Monofocal IOLs). It is very effective on the problem of multiple comparisons of individual visual acuities.

The potential of applications of this method: In the conventional visual acuity measurement, a specially trained inspector shows a visual chart (Snellen, Landolt-C or LogMAR chart) to an individual patient and checks whether the patient can recognize letters or symbols correctly. It takes a lot of time and it puts stress on patients. As we developed our new visual acuity system on a handy personal computer, all necessary steps in actual optometry site including automatic mea-

surement, data collection, logistic regression, computing test statistics and multiple testing are easily available without help of specially trained inspector. We believe that our new system works satisfactorily, however we must take more data from wide range of patients before this system becomes standard in real optometric site.

# ACKNOWLEDGEMENT

The authors are grateful to the editor and anonymous referees for their valuable comments and suggestions.

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