

On the Power of One Goodness-of-Fit Test Based on Square Deviations between Chencov Type Estimators of Distribution Density in $p \geq 2$ Independent Samples

Petre Babilua*, Elizbar Nadaraya**,**

*Department of Mathematics, Faculty of Exact and Natural Sciences, Ivane Javakhishvili Tbilisi State University, Tbilisi, Georgia

**Academy Member, Georgian National Academy of Sciences, Tbilisi, Georgia

A goodness-of-fit test is constructed based on projection type estimates of distribution density. The limiting power of the constructed goodness-of-fit test is stated for Pitman types of “close” alternatives. © 2024 Bull. Georg. Natl. Acad. Sci.

Goodness-of-fit test, projection estimator, limit distribution, power

Let $X^{(i)} = (X_1^{(i)}, \dots, X_{n_i}^{(i)})$, $i = 1, \dots, p$, be independent samples of size n_1, n_2, \dots, n_p , from p ($p \geq 2$) general populations with distribution densities $f_1(x), \dots, f_p(x)$. Let, further, $L_2(r)$ be the space of functions with square-integrable measure μ , $d\mu = r(x)dx$ and $\{\varphi_j(x)\}$ be complete orthonormal system in this space.

Suppose that the desired density $f_i(x) \in L_2(r)$, $i = 1, \dots, p$. Based on independent samples $X^{(i)}$, $i = 1, \dots, p$, construct projection estimates for unknowns $f_i(x)$

$$\begin{aligned} \hat{f}_i(x) &= \sum_{j=1}^{\lambda_i(n_i)} \hat{\alpha}_j(i) \varphi_j(x), \quad \hat{\alpha}_j(i) = \frac{1}{n_i} \sum_{k=1}^{n_i} \alpha_j(X_k^{(i)}), \\ \alpha_j(x) &= \varphi_j(x) r(x), \quad \lambda_i(n_i) = o(n_i), \quad i = 1, \dots, p. \end{aligned} \tag{1}$$

Projection estimate of distribution density (1) was first introduced and studied by Chencov N. N. [1]. In the present paper, we consider the problem of testing the simple hypothesis, according to which

$$H_0 : f_1(x) = f_2(x) = \dots = f_p(x) \equiv f_0(x),$$

($f_0(x)$ is a given density function) against Pitman type “close” alternatives:

$$H_1 : f_i(x) = f_0(x) + \alpha(n_0) \psi_i(x), \tag{2}$$

$$\alpha(n_0) \rightarrow 0, \quad n_0 = \min(n_1, \dots, n_p) \rightarrow \infty, \quad \int \psi_i(x) dx = 0, \\ \psi_i(x) \in L_2(r), \quad i = 1, \dots, p.$$

For testing this hypothesis, we consider criterion of testing hypothesis based on statistic [2]:

$$T(n_1, \dots, n_p) = \sum_{i=1}^p N_i \int \left[\hat{f}_i(x) - \frac{1}{N} \sum_{j=1}^p N_j \hat{f}_j(x) \right]^2 r(x) dx, \quad (3) \\ N_i = \frac{n_i}{\lambda_i}, \quad i = 1, \dots, p, \quad N = N_1 + \dots + N_p,$$

describing the mutual deviation of estimates $\hat{f}_i(x)$, $i = 1, \dots, p$, from each other. In particular case, when $p = 2$ the statistic T takes more explicit form

$$T(n_1, n_2) = \frac{N_1 N_2}{N_1 + N_2} \int \left[\hat{f}_1(x) - \hat{f}_2(x) \right]^2 r(x) dx.$$

This particular case was considered in [3].

Let us consider the question about the limiting law of the distribution of statistic (3) for the hypothesis H_1 , when n_i tends to infinity so that $n_i = nk_i$, where $n \rightarrow \infty$ and k_i are constants. Let $\lambda_1 = \lambda_2 = \dots = \lambda_p = \lambda(n)$, where $\lambda(n) \rightarrow \infty$ as $n \rightarrow \infty$.

Assumptions: $r(x)\varphi_j(x)$, $j = 1, 2, \dots$, have bounded variations $V_j < \infty$, $r(x)f_0(x)$, $r(x)\psi_i(x)$ are bounded, and $r(x)$ is integrable.

Notations:

$$\Delta_n(f_0) = \frac{1}{\lambda_n} \sum_{j=1}^{\lambda_n} \int \alpha_j^2(x) f_0(x) dx, \quad \lambda_n \equiv \lambda(n),$$

$$\sigma_n^2(f_0) = \frac{2}{\lambda_n} \sum_{i=1}^{\lambda_n} \sum_{j=1}^{\lambda_n} \left(\int \alpha_i(x) \alpha_j(x) f_0(x) dx \right)^2,$$

$$K_n(x, y) = \sum_{i=1}^{\lambda_n} \varphi_i(x) \varphi_i(y) r(y),$$

$$d_n = \sum_{i=1}^{\lambda_n} \varphi_i V_i, \quad \varphi_i = \sup_x |\varphi_i(x)|,$$

$$S_n(m) = \lambda_n^{-m} \int \dots \int K_n(t_1, t_2) K_n(t_2, t_3) \dots K_n(t_m, t_1) \cdot f_0(t_1) \dots f_0(t_m) r(t_1) \dots r(t_m) dt_1 \dots dt_m.$$

The following is true.

Theorem. Let $\Delta_n(f_0) = \mu(f_0) + o(\lambda_n^{-1/2})$, $\sigma_n^2(f_0) = \sigma^2(f_0) + o(\lambda_n^{-1/2})$ as $n \rightarrow \infty$ and for all $m \geq 3$,

$$Q_n(m) \equiv \lambda_n^{m-1} S_n(m) = O(1), \quad n \rightarrow \infty.$$

Then if there is such a $0 < \delta_0 < 1$ that

$$n^{-1/2} d_n \ln n \rightarrow 0, \quad \alpha_n d_n^2 \rightarrow 0 \quad (\alpha_n = \alpha(n_0))$$

for $\lambda_n = n^\delta$, $0 < \delta \leq \delta_0$ and $\alpha_n = n^{(\delta-2)/4}$, then for the alternative H_1

$$\lambda_n^{1/2} (T_n - \mu_0) \xrightarrow{d} N(A(\psi), \sigma_0^2),$$

where

$$A(\psi) = \sum_{i=1}^p k_i \int \left[\psi_i(x) - \frac{1}{\bar{k}} \sum_{j=1}^p k_j \psi_j(x) \right]^2 r(x) dx,$$

$$\mu_0 = (p-1)\mu(f_0), \quad \sigma_0^2 = 2(p-1)\sigma^2(f_0),$$

$$\bar{k} = k_1 + \dots + k_p, \quad p \geq 2.$$

d denotes the convergence in distribution and $N(a, b^2)$ – normally distributed random variable with mathematical expectation a and variance b^2 .

Corollary 1. Let

$$\Delta_n(f_0) = \mu(f_0) + o(\lambda_n^{-1/2}),$$

$$\sigma_n^2(f_0) = \sigma^2(f_0) + o(\lambda_n^{-1/2}) \quad \text{as } n \rightarrow \infty,$$

and for all $m \geq 3$

$$Q_n(m) = \lambda_n^{m-1} S_n(m) = O(1).$$

If

$$\frac{d_n \ln n}{\sqrt{n}} \rightarrow 0,$$

then random variable $\lambda_n^{1/2} (T_n - \mu_0)$ for the hypothesis H_0 has normal distribution $N(0, \sigma_0^2)$ [2].

Based on corollary, we can construct criterion for testing hypothesis H_0 . Critical domain for testing this hypothesis can be established by the inequality

$$T_n \geq d_n(\alpha), \tag{4}$$

where $d_n(\alpha) = \mu_0 + \lambda_n^{-1/2} \sigma_0 \varepsilon_\alpha$, ε_α , ε_α is the quantile of the level $1 - \alpha$ ($0 < \alpha < 1$) of a standard normal distribution $\Phi(x)$.

Corollary 2. Under conditions of theorem local behavior of the power $P_{H_1}(T_n \geq d_n(\alpha))$ is as follows

$$P_{H_1}(T_n \geq d_n(\alpha)) \longrightarrow 1 - \Phi\left(\varepsilon_\alpha - \frac{A(\psi)}{\sigma_0}\right),$$

where

$$A(\psi) = \sum_{i=1}^p k_i \int \left[\psi_i(x) - \frac{1}{\bar{k}} \sum_{j=1}^p k_j \psi_j(x) \right]^2 r(x) dx,$$

$$\bar{k} = k_1 + k_2 + \dots + k_p, \quad p \geq 2.$$

It should be noted that criterion (4) for testing hypothesis H_0 against alternatives type (2) is asymptotically strictly unbiased, since $A(\psi) > 0$, and is equal to 0 if and only if

$$\psi_1(x) = \psi_2(x) = \dots = \psi_p(x).$$

Example. Let $-\pi \leq X_1^{(i)} \leq \pi$, $i = 1, \dots, p$ and $\varphi_j(x)$, $j = 1, 2, \dots$, – system of trigonometric functions on $[-\pi, \pi]$. It is easy to see $d_n = O(\lambda_n^2)$. The conditions $\frac{d_n \ln n}{\sqrt{n}} \rightarrow 0$, $\alpha_n d_n^2 \rightarrow 0$ for $\lambda_n = n^\delta$, $\alpha_n = n^{-1/2+\delta/4}$ are met if $0 < \delta < \delta_0 = \frac{2}{17}$.

Further, assuming that $f_0'(x)$ is bounded and use the method of proving of Theorem 3.9 from [4: 151], we get

$$\begin{aligned} \Delta(f_0) &= \frac{1}{2\pi} + o\left(\frac{1}{\sqrt{\lambda_n}}\right), \\ \sigma_n^2(f_0) &= \frac{1}{\pi} \int_{-\pi}^{\pi} f_0^2(x) dx + o\left(\frac{1}{\sqrt{\lambda_n}}\right), \\ |Q_n(m)| &\leq c_1 \lambda_n^{-1} (L_n)^m, \quad m \geq 3, \end{aligned}$$

and $L_n \square 4\pi^{-2} \ln \lambda_n$ – Lebesgue constant [4]. In this case critical domain (4) will be

$$T_n \geq (p-1) \cdot \frac{1}{2\pi} + [2(p-1)]^{1/2} \varepsilon_\alpha \cdot \lambda_n^{-1/2} \left(\frac{1}{\pi} \int_{-\pi}^{\pi} f_0^2(x) dx \right)^{1/2}.$$

მათემატიკა

განაწილების სიმკვრივის ჩენცოვის ტიპის შეფასებების კვადრატულ გადახრებზე დაფუძნებული ერთი თანხმობის კრიტერიუმის სიმძლავრის შესახებ $p \geq 2$ დამოუკიდებელ შერჩევაში

პ. ბაბილუა*, ე. ნადარაია**,**

*ივანე ჯავახიშვილის სახ. თბილისის სახელმწიფო უნივერსიტეტი, ზუსტ და საბუნებისმეტყველო მეცნიერებათა ფაკულტეტი, მათემატიკის დეპარტამენტი, თბილისი, საქართველო

**აკადემიის წევრი, საქართველოს მეცნიერებათა ეროვნული აკადემია, თბილისი, საქართველო

აგებულია განაწილების სიმკვრივის პროექციულ შეფასებებზე დაფუძნებული თანხმობის კრიტერიუმი. მოძებნილია აგებული თანხმობის კრიტერიუმის ზღვართი სიმძლავრე პიტმანის ტიპის დაახლოებადი ალტერნატივებისთვის.

REFERENCES

1. Chencov N. N. (1962) A bound for an unknown distribution density in terms of the observations. Dokl. Akad. Nauk SSSR **147**: 45-48.
2. Babilua P. K., Nadaraya E. A. (2023) On the limit distribution of integral square deviation between projection type estimators of distribution density in $p \geq 2$ independent samples. *Bull. Georg. Natl. Acad. Sci.*, (N.S.) 17, 4: 15-19.
3. Mirzakhmedov M. A. (1982) Nonparametric estimation of a probability density by orthonormal functions. *Izv. Akad. Nauk UzSSR Ser. Fiz.-Mat. Nauk*, 4: 21-25 (in Russian).
4. Zigmund A. (1965) *Trigonometricheskie serii (Trigonometric Series)*. Vols. I, II. Izdat. "Mir", M. (in Russian).

Received October, 2024